



State of Wildfires 2023-24

3 Authors:

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5 Matthew W. Jones^{1,*}, Douglas I. Kelley^{2,*}, Chantelle A. Burton^{3,*}, Francesca Di Giuseppe^{4,*}, Maria Lucia F. Barbosa^{5,6}, Esther Brambleby¹, Andrew J. Hartley³, Anna Lombardi⁷, Guilherme 6 Maria Lucia F. Barbosa^{3,6}, Esther Brambleby¹, Andrew J. Hartley³, Anna Lombardi⁷, Guilherme Mataveli^{8,1}, Joe R. McNorton⁴, Fiona R. Spuler⁹, Jakob B. Wessel^{10,11}, John T. Abatzoglou¹², Liana O. Anderson¹³, Niels Andela¹⁴, Sally Archibald¹⁵, Dolors Armenteras¹⁶, Eleanor Burke³, Rachel Carmenta¹⁷, Emilio Chuvieco¹⁸, Hamish Clarke¹⁹, Stefan H. Doerr²⁰, Paulo M. Fernandes²¹, Louis Giglio²², Douglas S. Hamilton²³, Stijn Hantson²⁴, Sarah Harris²⁵, Piyush Jain²⁶, Crystal A. Kolden²⁷, Tiina Kurvits²⁸, Seppe Lampe²⁹, Sarah Meier³⁰, Stacey New³, Mark Parrington³¹, Morgane M. G. Perron³², Yuquan Qu^{33,34}, Natasha S. Ribeiro ³⁵, Bambang H. Saharjo³⁶, Jesus San-Miguel-Ayanz³⁷, Jacquelyn K. Shuman³⁸, Veerachai Tanpipat³⁹, Guido R. van der Werf⁴⁰, Sander Veraverbeke^{33,1}, Gavriil Xanthopoulos⁴¹ 7 8 9 10 11 12 13 14

16 Institutions: 17

- 18 1. Tyndall Centre for Climate Change Research, School of Environmental Sciences, University of East Anglia,
- 19 Norwich Research Park, Norwich, UK, NR4 7TJ
- 2. Hydro-climate risks, UK Centre for Ecology & Hydrology, Wallingford OX10 8BB, U.K.
- 3. Hadley Centre, Met Office, Fitzroy Road, Exeter, UK, EX1 3PB
- 4. Earth System Modelling Section, Forecast Department, European Centre for Medium-range Weather Forecast, Shinfield Park, Reading RG29AX, United Kingdom
- 5. Department of Remote Sensing, National Institute for Space Research, Avenida dos Astronautas, 1758. Jd.
- Granja São José dos Campos São Paulo, Brazil, 12227-010
- 6. Natural Sciences Center, Federal University of São Carlos, Rodovia Lauri Simões de Barros, km 12 SP-189 -Aracaçu, Buri - São Paulo, Brazil, 18290-000
- 20 21 22 23 24 25 26 27 28 29 30 31 32 7. Climate Intelligence, Research Department, European Centre for Medium-range Weather Forecast, Shinfield Road, Reading, UK, RG29AX
- 8. Earth Observation and Geoinformatics Division, National Institute for Space Research, Avenida dos
- Astronautas, 1758. Jd. Granja São José dos Campos São Paulo, Brazil, 12227-010
- 9. Department of Meteorology, University of Reading, University of Reading, Earley Gate, Whiteknights Rd. 33 34 Reading RG6 6ET
- 10. Department of Mathematics and Statistics, University of Exeter, Harrison Building, University of Exeter, North 35 36 Park Road, Exeter, UK
- 11. The Alan Turing Institute, British Library, 96 Euston Road, London, UK
- 37 38 12. School of Engineering, University of California, Merced, 5200 N Lake Rd, Merced, CA, 95343, USA
- 13. Cemaden/MCTI, Estrada Doutor Altino Bondensan, 500 Distrito de Eugênio de Melo, São José dos Campos 39 40 - São Paulo, Brazil
- 14. BeZero Carbon, 25 Christopher Street, London, UK, EC2A 2BS
- 41 42 15. School of Animal Plant and Environmental Sciences, University of Witwatersrand Johannesburg, University Corner, Braamfontein, Johannesburg
- 43 16. Landscape Ecology and Ecosystem Modelling Group, Faculty of Sciences, Department of Biology,
- 44 Universidad Nacional de Colombia, Cra. 30 # 45-03, Bogotá D.C., CP 111321, Colombia
- 45 17. Tyndall Centre for Climate Change Research, School of Global Development, University of East Anglia,
- 46 47 Norwich Research Park, Norwich, UK, NR4 7TJ
- 18. Department of Geology, Geography and the Environment, Universidad de Alcalá, Colegios, 2 28801 Alcalá 48 de Henares
- 49 19. FLARE Wildfire Research, School of Agriculture, Food and Ecosystem Sciences, University of Melbourne,
- Grattan St, Parkville, Australia, 3010
- 50 51 20. Centre for Wildfire Research, Swansea University , Singleton Park Swansea SA2 8PP Wales, UK
- 21. ForestWISE—Collaborative Laboratory for Integrated Forest & Fire Management, Centre for the Research
- 52 53 and Technology of Agro-Environmental and Biological Sciences, Universidade de Trás-os-Montes e Alto Douro, 54 Quinta de Prados, Vila Real, Portugal, 5000-801
- 22. Department of Geographical Sciences, University of Maryland, College Park, MD 20742
- 55 56 57 23. Marine, Earth and Atmospheric Science, North Carolina State University, Raleigh, North Carolina, USA, 27695
- 58 24. Program in Earth System Sciences, Faculty of Natural Sciences, Universidad del Rosario, Bogotá, Colombia





- 59 25. Fire Risk, Research and Community Preparedness, Country Fire Authority, Burwood East, Victoria, Australia
- 60 26. Northern Forestry Centre, Canadian Forest Service, Natural Resources Canada, 5320 122 St NW,
- 61 Edmonton, AB T6H 3S5, Canada
- 62 27. Wildfire Resilience Center, School of Engineering, University of California, Merced, 5200 N Lake Rd, Merced,
- 63 CA, 95343, USA
- 64 28. GRID-Arendal, P.O Box 183, N-4802, Arendal, NORWAY
- 65 29. Department of Water and Climate, Vrije Universiteit Brussel, Pleinlaan 2, 1050 Brussel, Belgium
- 66 30. Land, Environment, Economics and Policy Institute, Department of Economics, University of Exeter, Rennes
- 67 Drive, Exeter, United Kingdom, EX4 4ST
- 68 31. Atmospheric Composition Section, Research Department, European Centre for Medium-range Weather
- Forecast, Robert-Schuman-Platz 3, 53175 Bonn, Germany
- 69 70 32. UMR 6539 CNRS/IRD/Ifremer/LEMAR, Institut Universitaire Européen de la Mer, University of Brest, F-29280 71 Plouzané France
- 72 33. Department of Earth Sciences, Faculty of Science, Vrije Universiteit Amsterdam, De Boelelaan 1105, 1081 HV Amsterdam, Netherlands
- 73 74 34. Institute of Bio- and Geosciences: Agrosphere (IBG-3), Forschungszentrum Jülich, Wilhelm-Johnen-Straße, 52428 Jülich, Germany
- 75 76 77 78 79 35. Faculty of Agronomy and Forest Engineering, Eduardo Mondlane University, 3453 Avenida Julius Nyerere, Maputo, Mozambique
- 36. Faculty of Forestry, Bogor Agricultural University, Kampus Ipb Darmaga, Bogor
- 37. European Commission Joint Research Center, European Commission, Rue du Champ de Mars 21, 1050 80 81 Brussels, Belgium
- 38. NASA Ames Research Center, PO Box 1 Moffett Field, CA 94035-1000
- 82 39. Upper ASEAN Wildland Fire Special Research Unit (WFSRU), Forestry Research Center, Faculty of Forestry,
- 83 Kasetsart University, 5th Floor, 72nd Anniversary of Faculty Forestry Building
- 84 40. Wageningen University, Droevendaalsesteeg 3, 6708PB Wageningen
- 85 41. Forest Fire Laboratory, Institute of Mediterranean Forest Ecosystems, Hellenic Agricultural Organization
- 86 (DIMITRA), Terma Alkmanos, Ilisia, 11528, Athens, Greece
- 87
- ★ These authors contributed equally to this work.
- 88 89
- 90 Correspondence to: 91
 - matthew.w.jones@uea.ac.uk doukel@ceh.ac.uk
- 92 93
 - chantelle.burton@metoffice.gov.uk
 - Francesca.DiGiuseppe@ecmwf.int
- 94 95 96

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Abstract 98

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100 Climate change is increasing the frequency and intensity of wildfires globally, with significant 101 impacts on society and the environment. However, our understanding of the global distribution of extreme fires remains skewed, primarily influenced by media coverage and regional 102 research concentration. This inaugural State of Wildfires report systematically analyses fire 103 104 activity worldwide, identifying extreme events from the March 2023-February 2024 fire season. 105 We assess the causes, predictability, and attribution of these events to climate change and 106 land use, and forecast future risks under different climate scenarios. During the 2023-24 fire 107 season, 3.9 million km² burned globally, slightly below the average of previous seasons, but fire carbon (C) emissions were 16% above average, totaling 2.4 Pg C. This was driven by 108 109 record emissions in Canadian boreal forests (over 9 times the average) and dampened by 110 reduced activity in African savannahs. Notable events included record-breaking wildfire extent 111 and emissions in Canada, the largest recorded wildfire in the European Union (Greece), 112 drought-driven fires in western Amazonia and northern parts of South America, and deadly fires in Hawai'i (100 deaths) and Chile (131 deaths). Over 232,000 people were evacuated in 113 114 Canada alone, highlighting the severity of human impact. Our analyses revealed that multiple drivers were needed to cause areas of extreme fire activity. In Canada and Greece a 115





116 combination of high fire weather and an abundance of dry fuels increased the probability of 117 fires by 4.5-fold and 1.9-4.1-fold, respectively, whereas fuel load and direct human suppression often modulated areas with anomalous burned area. The fire season in Canada 118 119 was predictable three months in advance based on the fire weather index, whereas events in 120 Greece and Amazonia had shorter predictability horizons. Formal attribution analyses 121 indicated that the probability of extreme events has increased significantly due to 122 anthropogenic climate change, with a 2.9-3.6-fold increase in likelihood of high fire weather in 123 Canada and a 20.0-28.5-fold increase in Amazonia. By the end of the century, events of similar 124 magnitude are projected to occur 2.22-9.58 times more frequently in Canada under high 125 emission scenarios. Without mitigation, regions like Western Amazonia could see up to a 2.9-126 fold increase in extreme fire events. For the 2024-25 fire season, seasonal forecasts highlight 127 moderate positive anomalies in fire weather for parts of western Canada and South America, 128 but no clear signal for extreme anomalies is present in the forecast. This report represents our 129 first annual effort to catalogue extreme wildfire events, explain their occurrence, and predict 130 future risks. By consolidating state-of-the-art wildfire science and delivering key insights 131 relevant to policymakers, disaster management services, firefighting agencies, and land 132 managers, we aim to enhance society's resilience to wildfires and promote advances in preparedness, mitigation, and adaptation. 133

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135 Short Summary

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137 This inaugural State of Wildfires report catalogues extreme fires of the 2023-24 fire season. 138 For key events, we analyse their predictability and drivers and attribute them to climate change 139 and land use. We provide a seasonal outlook and decadal projections. Key anomalies 140 occurred in Canada, Greece, and western Amazonia, with other high-impact events 141 catalogued worldwide. Climate change significantly increased the likelihood of extreme fires, 142 and mitigation is required to lessen future risk.

143 **1 Introduction**

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145 **1.1 Background** 146

147 The potential for wildfires is growing under climate change, with increases in the frequency 148 and intensity of drought and periods of fire-favourable weather driving reductions in vegetation 149 (fuel) moisture and priming landscapes to burn more regularly, severely, and intensely (Jones 150 et al., 2022; UNEP, 2022a; Abatzoglou et al., 2019). Additionally, human activities and land 151 use change can contribute to or exacerbate the risk of extreme fires, especially in tropical 152 forests where people are the primary cause of ignition and forest degradation (Lapola et al., 153 2023). Recent years have been marked by a series of extreme wildfire events spanning the 154 globe, with record levels of burned area (BA) occurring in the 2019-2020 Australian "Black 155 Summer" bushfires (Abram et al., 2021) and a series of high-ranking wildfire seasons 156 occurring in quick succession in the western US (2020 and 2021; Higuera & Abatzoglou, 157 2020), Siberia (2020 and 2021; Zheng et al., 2023), Europe (2022; European Commission 158 Joint Research Centre, 2023) and South America (2019, 2020; Kelley et al., 2021; Ferreira 159 Barbosa et al., 2022; Silveira et al., 2020). The 2023-24 fire season was marked by 160 unprecedented wildfire extent and emissions in Canada, deadly fast-moving fires in Hawaii 161 and Chile, the largest individual wildfires on record in the European Union and Canada, 162 widespread fires in northwestern South America including parts of Amazonia such as Brazil, 163 Bolivia, Colombia, and Venezuela (Mataveli et al., 2024; Kolden et al., 2024; European 164 Commission EU Science Hub, 2023).

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166 The prominence of recent extreme wildfires and wildfire seasons notably contrasts with overall 167 trends in the area burned by fires globally. Due mostly to a reduction in the global savannahs





168 tied to landscape fragmentation and changing rainfall patterns, global BA has fallen since the 169 beginning of this century by around one-quarter (Andela et al., 2017; Jones et al., 2022; Chen 170 et al., 2024). Critically, this decline in fire extent masks major shifts in the distribution of fires 171 globally, with regions such as eastern Siberia and the western US and Canada experiencing 172 a more than 40% increase in BA since 2000 (Jones et al., 2022) and regions such as southeast 173 Australia also showing significant increases over longer periods (Canadell et al., 2021). 174 Likewise, there have been shifts in the global distribution of BA and fire carbon (C) emissions 175 from non-forests to forests globally and from the tropics to the extratropics (Kelley et al., 2019). 176 Hence, focussing on global aggregated BA extent underplays the scale and magnitude of 177 changes in wildfire activity and impact on regional levels. An increase in forest and peatland 178 burning is particularly concerning due to the rich ecosystem services that these regions 179 provide, including C storage and biodiversity (UNEP, 2022b). The intensification of fire regimes in environments that are less fire-adapted is significant, as these ecosystems are 180 181 expected to be least resilient to such changes (Grau-Andrés et al., 2024).

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183 The extreme wildfire events of recent years have significantly impacted society and 184 ecosystems across the globe. Since 1990, wildfire disasters have directly killed or injured at 185 least ~18,000 people, a conservative measure based on incomplete records and reporting 186 biased to the global Northern countries (updated from Jones et al., 2022; Centre for Research on the Epidemiology of Disasters, 2024). In 2023, 232,000 people were evacuated due to 187 188 wildfires in Canada alone (Jain et al., 2024; Kolden et al., 2024). Also since 1990, fires are 189 estimated to have caused on the order of 10 million premature deaths globally through 190 degraded air quality (Johnston et al., 2012). Degraded air quality related to fires is experienced 191 most strongly in the tropics (Pai et al., 2022) and often disproportionately affects traditional 192 communities with poor public services or means of protection (Carmenta et al., 2021). Yet, 193 images of North America, including the infamous Wall Street, blanketed in smoke during the 194 2023 fire season highlight the global nature of this problem.

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196 As anthropogenic emissions of CO₂ remain persistently high, the world's natural C sinks in 197 forests, peatlands and other ecosystems are increasingly pivotal to moderating increases in 198 atmospheric CO₂ concentration and are often relied upon for delivering national plans for 199 reaching Net Zero (Smith et al., 2023) and offering sites for nature based solutions (NBS). Yet, 200 massive wildfire emissions from boreal forests and soils in Siberia and Canada across the 201 years 2020, 2021, and 2023 amount to over 1 billion tonnes of C, a gross flux comparable in 202 magnitude to annual CO₂ emissions from fossil fuel combustion in India, the EU27 or the USA 203 (Friedlingstein et al., 2023; Zheng et al., 2023). Ecosystem function is also impacted by 204 extreme wildfires through widespread mortality of forest stands that undermine the 205 regenerative capacity of forests (Nolan et al., 2021a) and the habitats of many endemic 206 species being degraded in biodiversity hotspots (Ward et al., 2020). Not only are these 207 emissions high, but they may represent a change in the fire-C cycle from sink to source, 208 undermining assumptions about post-fire regrowth offsetting fire emissions. The lands and 209 territories of traditional communities and Indigenous Peoples are degraded and transformed 210 through wildfires, raising climate justice issues (Garnett et al., 2018; Barlow et al., 2018; 211 Lapola et al., 2023). Further, conflating distinct fire types has also stigmatised intentional 212 small-scale intergenerational managed fire use and led to prohibitive fire governance that 213 affects local communities (Carmenta et al., 2021; Barlow et al., 2020).

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215 Mitigating and adapting to increases in wildfire potential are growing priorities of policymakers 216 and involve coordination with many other stakeholders. National and international disaster 217 management centres are seeking to enhance predictive capacity, while fire management 218 agencies are expanding or re-allocating their resources to rapidly suppress fires to avoid them 219 becoming too large, fast or intense. A number of international organisations such as the UN 220 Environment Programme (UNEP, 2022a), the World Bank (2020, 2024), and the Organisation 221 for Economic Co-operation and Development (OECD, 2023) and a range of other inter- or 222 non- governmental organisations are producing reports that consolidate evidence on the





223 changing risk of extreme fires and identify best practices for mitigating their impacts, including 224 through land management and urban/rural planning. Many land managers are developing and 225 implementing approaches such as fuel reduction, a process subject to permit systems issued 226 by regional fire management agencies in some countries (Fernandes and Botelho, 2003; 227 Stephens et al., 2012; Moreira et al., 2020; Chuvieco et al., 2023). Wildfire response agencies 228 are exploring innovative approaches to detecting and responding to fires, and there is rising 229 interest in the prospect of integrated fire management around the world (Food and Agriculture 230 Organization of the United Nations, 2024). Operators of C market projects and forest carbon-231 conservation initiatives, such as REDD+ are particularly wary of the risks that wildfires present 232 to the permanence of C offsets, seen as a key tool for achieving Net Zero emissions (Barlow 233 et al., 2012; Smith et al., 2023).

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Amidst extreme wildfires and wildfire seasons, stakeholders increasingly turn to scientists for answers to pressing questions that naturally arise. How extreme was this fire event in a historical context? Is climate change to blame? Will we see more wildfires like this in the future? Did land management exacerbate or ameliorate the problem? Can we predict events like this in future to improve early warning? What is the role of climate policy in reducing risk of extreme wildfires in future?

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242 While observational, statistical, and modelling tools for assessing extreme wildfire drivers and 243 predicting wildfire occurrence are advancing rapidly, their application to studying extreme 244 wildfire seasons or events on timescales relevant to public and political interest remains 245 limited. The State of Wildfires report embodies a structure for routinely cataloguing extreme 246 wildfire events at annual frequency and explaining their occurrence and relation to climate 247 change. We incorporate recent advances in disentangling the drivers of fire to fuel dryness, 248 fuel load, and weather, and ignition and suppression factors, and by applying these methods 249 in conjunction with models of global change we quantify the change in likelihood of the past 250 year's events under climate and land use changes. Observable fire metrics (e.g. burned area) 251 are the target variable of our causal inference and attribution work, which thereby advances 252 on more common climate attribution studies that attribute change in fire-favourable 253 meteorological conditions to climate change. Overall, this report capitalises on recent 254 advances in the study of extreme fire events and seasons to provide timely information about 255 shifting fire regimes and their causes. The findings of the report are relevant to policymakers, 256 the media, and the wider public.

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258 Due to the diversity of environmental settings in which fires occur and the range of ecological, 259 economic, or societal impacts caused by fires, defining an extreme fire or an extreme fire 260 season remains inherently challenging. To date, extreme fires have commonly been defined by their BA extent, by their feedback on the global climate compared to climatological data 261 262 and on large-scale ecosystem changes and by their socio-economic impact. While an extreme 263 fire event may result in significant anomalies against historical Earth Observations in the same 264 study region, the scientific community currently seeks to apply a more comprehensive 265 definition of extreme fire, including its impacts on societies, fire-prone regions and on 266 downwind society and ecosystems.

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268 1.2 Objectives of this Report

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This inaugural edition of the State of Wildfires report aims to stimulate development of tools for understanding and predicting extreme fires and to deliver actionable information to policy and practice stakeholders and society. In this edition we:

- 273
- Regionally identify extreme individual wildfires or extreme wildfire seasons of the period since March 2023, and place them in context of recent trends.





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 2. Shortlist a selective number of extremes (extreme individual wildfires or extreme
 wildfire seasons) with notable impacts on society or the environment, which we term
 the 'focal events' in this report.
 - Diagnose the contributions of fuel dryness, fuel load, ignitions and suppression to the occurrence of each focal event.
 - 4. Assess the capacity of operational predictive systems to predict each focal event.
 - 5. Attribute each focal event to anthropogenic factors including climate change and land use.
 - 6. Provide an outlook on the probability of extreme events in the coming fire season.
 - 7. Project future changes in the probability of each focal event under future climate scenarios.
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The State of Wildfires report will be an annually recurring report that can harness and adopt new methodologies brought forward by the global community of fire scientists between publications.

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292 To achieve our objectives, we integrate a variety of advanced methods developed in recent 293 years. For instance, we identify anomalies in past fire seasons using both regional BA and 294 emissions records and a global dataset of individual fires (Giglio et al., 2018; van der Werf et 295 al., 2017; Andela et al., 2019). This dual approach enables us to detect not only regional 296 anomalies but also individual fires that are exceptionally large or fast-moving. Additionally, we 297 leverage the latest advances in fire risk forecasting on seasonal to sub-seasonal timescales 298 provided by the ECMWF. Furthermore, we employ two state-of-the-art machine learning 299 techniques (Kelley et al., 2019; McNorton et al., 2024) to pinpoint the causes of the extreme 300 fire events of 2023-24. We also apply multiple attribution methods to quantify the influence of 301 climate change and land use change on these extreme events (e.g. Kelley et al., 2019, 2021; 302 Burton, Lampe, et al., 2023). Fire is an inherently stochastic process, and any attempt to 303 assess its causes needs to account for uncertainty. The strength of our approach in this report is in employing different techniques to address the complex questions of how unusual the fires 304 305 of the last season were, and what the drivers might have been. By using multiple tools to 306 measure different fire metrics and identify the factors driving fires, and by using different 307 techniques to attribute their causes to climate and human influences, we can either build up a 308 solid body of evidence or highlight aspects of extreme fires in recent years that we do not yet 309 understand. While the tools are diverse, they address the questions from different angles, 310 giving us a robust way to assess predictability or quantify uncertainty in their results - also vital 311 components in building confidence in the statements made in this report. We demonstrate the 312 range of tools we have available to tackle these important questions now, and highlight gaps 313 where we still need to develop the research further. The combination of these diverse 314 techniques marks a significant step toward the regular and comprehensive assessment of 315 past and future extreme wildfire events.

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317 Over the coming years and decades, we aim to develop and apply the tools presented in this 318 report in near-real time, thus enhancing our capacity to transfer key insights to decision-319 makers at the time they most need it. Ultimately, this report serves as a vehicle that 320 consolidates the novel advances and efforts of the fire science community to provide relevant 321 information to stakeholders, enabling society to adapt to emerging fire extremes.

322 2 Methods

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324 2.1 Extreme Wildfire Events of 2023-24

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We catalogued the extreme regional wildfire events or annual fire seasons in the period March
 2023-February 2024 based on a combination of anomalies evident in Earth Observations
 (Section 2.1.1), as well as expert assessment of wildfires that severely impacted people or





329 the environment (Section 2.1.4). This approach deliberately represents a broad definition of 330 extreme wildfire events by allowing our expert panel to flexibly integrate diverse forms of 331 wildfire impacts into their regional assessments of extreme wildfire events. In so doing, we 332 allow for the fact that it is challenging to define a consistent threshold for 'extreme' fires given 333 the diversity in the characteristics and impacts of fire across the many fire regimes (e.g. 334 Archibald et al., 2009; Cunningham et al. 2024). Metrics such as BA and emissions are a good 335 proxy for impact in some cases, yet they cannot fully account for the diverse relationships 336 between fire, ecosystems, people and societal values. Hence our broad definition of "extreme 337 fire" allows nuanced measures of fire impacts to be captured (e.g. fire severity or challenges 338 to suppression; Keeley 2009) and careful consideration of the complex interplay between 339 wildfire and the various forms of fire intentionally applied by humans and serving multiple 340 purposes. It also allows for elaboration of the links between wildfire and the various values 341 that benefit or are burdened by it, including ecosystem services, biodiversity, Indigenous 342 cultural values and biocultural landscapes, heritage, recreation, social capital, the built 343 environment, tourism, agriculture and other industries, and human health and wellbeing.

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345 2.1.1 Earth Observations of Fire346

347 2.1.1.1 Input Datasets

349 We assembled observations of global daily burned area (BA) for the period March 2001-350 February 2024 from NASA's MODIS BA product (MCD64A1, collection 6.1) at 500m resolution 351 (Giglio et al., 2018, 2021). We also produced a global record of individual fires for the period 352 March 2001-February 2024 by updating the Global Fire Atlas (Andela et al., 2019; Andela and 353 Jones, 2023) through February 2024, driven by the 500m MODIS BA data. The Global Fire 354 Atlas algorithm clusters burned cells into individual fires, tracks their daily progression, and 355 logs attributes such as fire size and mean daily rate of growth. Our updates are provided at 356 Andela and Jones (2023).

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358 In addition, we gathered estimates of fire carbon (C) emissions for the period March 2023-359 February 2024 from two models driven by Earth Observations of active fires or burned area: 360 firstly, the Global Fire Assimilation System (GFAS) product, provided operationally by the 361 Copernicus Atmospheric Services (CAMS) at 0.1 degree spatial resolution and daily temporal 362 resolution (Kaiser et al., 2012; European Centre for Medium-Range Weather Forecasts, 363 2024), and; secondly, the Global Fire Emissions Database (GFED; version 4.1s) product at 0.25 degree spatial resolution and daily temporal resolution (van der Werf et al., 2017). GFAS 364 365 is driven by the fire radiative power (FRP) retrievals in the MODIS active fire product 366 MCD14A1 and biome-level relationships between FRP and biomass consumed based on 367 GFED3 (Kaiser et al., 2012). For the 1997-2016 period, GFED4s is driven by MODIS BA data 368 (MCD64A1 collection 5) supplemented with small fire BA based on MODIS active fire data, 369 and a model for biomass productivity and fuel consumption (van der Werf et al., 2017). For 370 the post-2016 period, emissions are based on active fire detections scaled to emissions using 371 pixel-based scaling factors derived from the 2003-2016 overlapping period.

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373 We note that the MODIS MCD64A1 BA data used in various facets of our analyses are known 374 to be conservative due to the limitations to detecting small fires based on surface spectral 375 changes at 500m resolution. Recent work has shown that including detections of small active 376 fires increases global BA estimates by 93% (Chen et al., 2023). However, variability and trends 377 in regional BA totals using datasets that include small fires do not differ significantly from the 378 variability and trends present in the MODIS MCD64A1 BA data (Chen et al., 2023). In this 379 report, regional BA and fire C emissions totals are conservative and the context of missing 380 small fires must be considered when interpreting the threshold values of the upper quantiles 381 of the fire size distribution from the Global Fire Atlas. Nonetheless, there is consistency in our





approach across years such that anomalies are robust with respect to prior observations, and
 our focus on more extreme fires renders smaller fires obsolete in most of our analyses.

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385 **2.1.1.2 Regional Burned Area, Carbon Emissions and Fire Count Totals** 386

387 We calculated regional totals of BA and C emissions based on a variety of regional layers 388 defined in Table 1. The regional layers represent a range of biogeographical boundaries (e.g. biomes), geopolitical boundaries (e.g. countries), and values used in scientific reports (e.g. by 389 390 the Intergovernmental Panel on Climate Change; IPCC). We calculated monthly totals of BA 391 and fire C emissions for each region by aggregating monthly BA and daily C emissions data, 392 summing the data from the input datasets both spatially and temporally as required. In the 393 case of fire C emissions, we also calculated the mean estimate of fire C emissions from 394 GFED4.1s and GFAS, regionally.

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396 We adopt a March-February definition of the global fire season (e.g. the latest global fire 397 season spans March 2023-February 2024). Due to an annual lull in the global fire calendar in 398 the boreal spring months, fire season BA totals are least sensitive to the shifts in fire season 399 cutoffs of 1-2 months if the fire season centres on spring (Boschetti and Roy, 2008). This 400 makes the global fire season centred on spring a pragmatic option for the study of interannual 401 variability or trends in fire extent (Boschetti and Roy, 2008). The period March-February is 402 specifically oriented at the end of the austral fire season and before widespread fires have 403 begun in the boreal extratropics. The regions where this global definition of the fire season is 404 most problematic are: northern hemisphere South America, Southeast Asia, and Central 405 America (Giglio et al., 2013).

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In addition, we calculated totals of regional fire counts for each global fire season based on
the number of individual fire ignition points present within each region, using ignition point
vectors from the Global Fire Atlas. The resolution of the MODIS data supplied to the Global
Fire Atlas algorithm is 500 m and hence fires that are smaller in scale are omitted. Regional
or national systems may record greater fire counts due to the inclusion of smaller fires.

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413 Table 1: Regional layers to which global Earth Observations were disaggregated and used to 414 define regions with extreme wildfire seasons or extreme individual wildfire attributes. Regional 415 layers are available from Jones et al. (2024).

layers are available from Jones et al. (2024) Short Form Laver Source Notes Biomes NA Olson et al. (2001) NA Continents ArcGIS Hub (2024) **Continental Biomes** NA See above Spatial intersect of biomes and continents. Countries NA EU Eurostat (2020) UC Davis Global GADM-L1 UC Davis (2022) First sub-national Administrative Areas (GADM) administrative level such Level 1 as states of the US or provinces of China. Version 4.1. IPCC (2021); Intergovernmental Panel on IPCC AR6 WGI Climate Change Sixth Regions SantanderMetGroup Assessment Report (AR6) (2021) Working Group I (WGI) **Reference Regions** Global C Project Regional C RECCAP2 Ciais et al. (2022)



Cycle Assessment and Processes (RECCAP2) Reference Regions	Regions		
Global Fire Emissions Database (GFED) Basis Regions	GFED4.1s Regions	van der Werf et al. (2006)	

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418 **2.1.2** Identifying Extreme Fire Seasons and Events from Earth Observations

420 2.1.2.1 Regions with Extreme Wildfire Seasons

Anomalies in BA, fire C emissions, and fire counts in the latest global fire season (March 2023 February 2024) were calculated in several ways:

- 425 I. as relative anomalies (expressed in %) from the annual mean during all previous
 426 March-February periods since 2001;
- 427 II. as standardised anomalies (standard deviations) from the annual mean during all
 428 previous March-February periods since 2001;
- 429 III. as a rank amongst all March-February periods since 2001 (2003 for C emissions),
 430 March 2023-February 2024 inclusive, and;
 431
- In this report, anomalies in fire C emissions are reported based on the two-model mean
 estimate from GFED4.1s and GFAS, however anomalies based on the GFED4.1s or GFAS
 estimates individually are also available via Jones et al. (2024).
- 435

436 We identified regions in which the latest fire season was potentially classifiable as 'extreme' 437 based on the rank of BA, C emissions and fire count amongst all fire seasons. For visualisation 438 purposes, we identified regions in which the latest fire season ranked in the top 5 of all annual 439 fire seasons on record (see <u>Section 3.1</u>). The BA data for the period March 2001-February 440 2024 includes 23 fire seasons, while the C emissions data for the period March 2003-February 441 2024 includes 21 fire seasons. Hence, a top-5 ranking translates approximately to a fire 442 season in the upper quartile of those on record. While a high-ranking BA or C emissions is 443 indicative of an extreme fire season, we reiterate that our regional experts play a crucial role 444 in defining a broader range of events that could be considered extreme. 445

We further characterised the onset, peak, and cessation of anomalous monthly BA in March 2023-February 2024. First, we identified the month of the event's peak as the maximum difference between monthly BA values in March 2023-February 2024 and the climatological mean monthly values from the prior March-February periods. Thereafter, the event's onset and cessation were defined as the bounds of consecutive months with above-average BA prior to and following the peak (relative to the monthly climatology).

452

454

453 2.1.2.2 Regions with Extreme Individual Wildfire Attributes

We identified regions in which large or fast-moving fires occurred in the latest fire season based on records of individual fires from the Global Fire Atlas. For each region (**Table 1**) and year, we estimated the size of the largest fire, the daily rate of growth of the fire that spread most rapidly, the size of the 95th percentile fire, and the daily rate of growth of the 95th percentile fire. In the Global Fire Atlas, the daily rate of growth for any given fire is determined by calculating the average daily rate of growth at which the fire advanced across all its constituent cells. This includes cells burned by the head, flank, and backfire. This method





462 produces lower spread rates than if the calculation were based solely on the cells burned by463 the head fire.

464

Anomalies in each fire attribute were calculated using the same metrics as for BA (see *i-iii* above), and we identified regions in which the latest fire season featured fires with potentially extreme attributes based on the rank of BA and fire C emissions amongst all fire seasons.

468

469 **2.1.3 Updating Regional Trends in Burned Area and Carbon Emissions** 470

471 To place recent extremes in the context of fire trends of the past two decades, we update our 472 regional analyses of trends in annual BA from those published by Jones et al. (2022). In 473 contrast to Jones et al. (2022), we present trends that align more closely with global fire 474 seasons, spanning the period March 2001-February 2024 rather than trends over calendar 475 years. We quantified trends using the Theil-Sen slope estimator. The Theil-Sen approach is 476 useful when data may contain outliers or be non-normally distributed, as it calculates the 477 median of all possible pairwise slopes between data points. This makes it less sensitive to 478 outliers than a standard least squares regression slope. Changes were then assessed by multiplying by the trends (unit year-1) by the number of fire seasons in the period March 2001-479 February 2024 (23 fire seasons). Relative changes were calculated as the absolute changes 480 481 divided by the mean annual BA during the period, which is a more conservative approach than 482 adopting the BA total in the first fire season as the denominator. The significance of the slopes 483 estimated by the Theil-Sen method was evaluated using the Mann-Kendall test, with a 484 confidence level set at 95%. The Mann-Kendall test assesses whether there is a statistically 485 significant trend in the series of data. This test does not require the data to follow any specific 486 distribution and is robust against outliers.

487

In addition to reporting trends in *total* BA, we also present trends in *forest* BA as these regularly
diverge from total BA trends (see Section 3.1). Forest BA is calculated as described in Section
2.1.1.2 but after isolating burned cells in areas with tree cover exceeding 30% in NASA's
annual MODIS MOD44B collection 6.0 Continuous Vegetation Field product (250 m) (DiMiceli
et al., 2015). The 30% threshold is widely used amongst studies of forest cover change (e.g.
Li et al., 2017; Cunningham et al., 2020; Sexton et al., 2016).

494

495 2.1.4 Expert Consultation

496

497 We assembled a panel of regional experts (two from each continent, Table 2), to contribute to 498 the identification, description, and characterisation of extreme wildfire seasons or impactful 499 events in the latest fire season. A key role of the expert panel was to catalogue regional events 500 that significantly impacted society or the environment but which may not have been detected 501 by Earth-observing satellites due to issues such as scale, short duration, timing of overpass, 502 and cloud or canopy cover. This includes (but is not limited to) wildfires that impacted society 503 by causing fatalities, evacuations, displacement (e.g. homelessness), direct structure or 504 infrastructure loss or damage, degradation of air or water quality, loss of livelihood, cultural 505 practice or other ways of life, and loss of economic productivity. This definition also includes 506 (but is not limited to) wildfires that impact the environment via disturbance to vulnerable 507 ecosystems, biodiverse areas, or ecosystem services such as C storage. This approach 508 recognises that Earth Observations do not provide a complete record of all impactful fires. We 509 do not define ubiquitous quantitative thresholds of impact by any of the measures outlined 510 above, but rather invite in-region experts to identify events that triggered impacts that were 511 sufficient in magnitude to infiltrate public and political discourse. The sources of information 512 available for cataloguing regional events include national/regional fire records, fire service 513 reports, disaster management reports, news reports, and social media.





A second key role of our expert panel was to describe and contextualise the impacts of the fire seasons highlighted as extreme by Earth Observations or regional assessment (see **Section 3.1**). This includes the discussion of anomalies that rank highly in Earth Observations but nonetheless did not have major impacts in regions. Metrics such as percentage difference from the mean and standard deviations from the mean were also used by the expert panel to further characterise the magnitude of anomalies in regional BA, fire C emissions, fire counts and individual fire properties.

522

523

Table 2: Experts contributing to the identification of extreme events and characterisation of the global fire season during March 2023-February 2024.

Region	Expert	Country of Organisation / Nationality	Professional Background(s)	Others Consulted	
Africa	Natasha Ribeiro	Mozambique	Research	Robert Ang'ila, Karatina University;	
	Sally Archibald	South Africa	Research	Kebonyethata Dintwe, Botswana University; John Mendelsohn, Okavango Research Institute; Ronald Heath, Forestry South Africa; Helen De Klerk, Stellenbosch University	
Asia	Bambang Saharjo	Indonesia	Research, Litigation		
	Veerachai Tanpipat	Thailand	Research, Fire Control and Management Instructor and Consultant		
Europe	Paulo Fernandes	Portugal	Research	Davide Ascoli, University of Turin, IT; Hellenic Agricultural	
	Stefan Doerr	UK / Germany	Research	Organization "DIMITRA"; Institute of Mediterranear Forest Ecosystems; Niall MacLennan, Scottish Fire and Rescue Service	
	Gavriil Xanthopoulos	Greece	Research		
North America	Crystal Kolden	USA	Research, Firefighting		
	Jacquelyn Shuman	USA	Research		
	Piyush Jain	Canada	Research		
Oceania	Hamish Clarke	Australia	Research, Environmental Management	Grant Pearce, Fire and Emergency NZ; Simeon Telfer, SA Country Fire Service;	
	Sarah Harris	Australia	Research, Emergency Management	Agnes Kristina, Department of Fire and Emergency Services; Russell Stephens Peacock, QLD Fire and Emergency Services; Chris Collins, Tasmania Fire Service; David Field, NSW Rural	





Region	Expert	Country of Organisation / Nationality	Professional Background(s)	Others Consulted	
Africa	Natasha Ribeiro	Mozambique	Research	Robert Ang'ila, Karatina University;	
	Sally Archibald	South Africa	Research	University; Kebonyethata Dintwe, Botswana University; John Mendelsohn, Okavango Research Institute; Ronald Heath, Forestry South Africa; Helen De Klerk, Stellenbosch University	
				Fire Service.	
South America	Dolors Armenteras	Colombia	Research	The Chico Mendes Institute for Biodiversity Conservation (ICMBio),	
	Liana Anderson	Brazil	Research, Disaster risk reduction strategies	Brazil, Santarèm Office	

526

527 **2.1.5 Bespoke Analysis: Air Quality Impacts in Canada**

528

529 Media and grey literature reports highlighted that wildfires in Canada during 2023 led to 530 extreme degradation of air quality across North America (see Section 3.1.3.4). We evaluated 531 air quality impacts of Canadian wildfire emissions in 2023 in the context of the previous 20 532 years using PM2.5 surface concentrations from the CAMS reanalysis of global atmospheric 533 composition (Inness et al., 2019). The CAMS reanalysis provides a 3-dimensional time-534 consistent dataset of atmospheric composition including aerosols, chemical species and 535 greenhouse gases for the period 2003-2023). Estimated wildfire emissions for the reanalysis are provided by the GFAS v1.2 (Kaiser et al., 2012), based on fire radiative power 536 537 observations from the Terra and Aqua MODIS instruments, in addition to inventories of 538 anthropogenic and biogenic emissions, and natural emissions (such as desert dust). While 539 the reanalysis PM2.5 includes contributions from all emissions sources, the most extreme 540 values are strongly correlated with the estimated fire emissions. The extreme annual PM2.5 541 surface concentrations are calculated as the 95th, 97th and 99th percentiles of the daily 542 maximum concentrations for each year in the reanalysis, and presented as the area-weighted 543 average for Canada.

544

545 2.2 Shortlisting of Focal Events

546

547 In later sections of this report, we apply various models to understand the causes and 548 predictability of a selection of extreme wildfire seasons or events during March 2023-February 549 2024, to attribute the roles of climate change and land use change, and project the future 550 likelihood of events of a similar magnitude under various climate change scenarios (see 551 Methods below). We limited the number of analyses to three globally prominent focal events 552 of the 2023-24 global fire season in this report because our models are not currently 553 operational products and must be trained and optimised regionally, which is time-consuming. 554

555 Our models are structured and parameterised to predict anomalies in regional BA or fire count, 556 meaning that they can only be applied to events that feature in Earth Observations. In addition, 557 the models have not yet been applied to study fire C emissions or wildfires with extreme 558 attributes, and so we restrict their application to regions with anomalous BA in the latest fire 559 season. In discussion with our expert panel, we prioritised the three events studied in this





report by weighing up the anomalies in Earth Observations during the latest fire season as well as the impacts that these extremes had on people and the environment. The focal events are notable for their international significance, attracting attention from the media and policymakers both within and beyond their region.

565 2.3 Diagnosing Drivers and Predictability

566 567 568

2.3.1 Assessing Predictability of Extremes across Time Scales

569 We evaluated the time frame over which extreme events could have been forecasted using 570 existing forecasting products that produce the Canadian Fire Weather Index (FWI) as outputs 571 (van Wagner, 1987). Developed by the Canadian Forest Service as part of the Canadian 572 Forest Fire Danger Rating System (van Wagner, 1987), the FWI comprises various 573 components that consider the influence of dead fuel moisture and wind on fire ignition and 574 spread behaviour, with 2m temperature, 10m wind speed, precipitation, and 2m relative 575 humidity as prerequisite variables. The FWI specifically combines three fuel moisture codes 576 representing vegetation moisture state at different layers in the forest floor, as well as a spread 577 index influenced by fuel moisture state and wind speed (van Wagner, 1987). A higher FWI indicates fire weather conditions more conducive to wildfires in environments with sufficient 578 579 fuel load. Owing to its original design for use in forest ecosystems, the FWI is especially useful 580 for predicting the likelihood and severity of extreme events in ecosystems where weather is 581 the primary limitation to fire (i.e. those mainly limited by moisture or temperature); its accuracy 582 in forecasting BA in fuel-limited ecosystems is more limited (Carvalho et al., 2008; Bedia et al., 2015; Abatzoglou et al., 2018; Jones et al., 2022). Its applications encompass early 583 584 warning systems, pre-suppression and suppression planning, prescribed burn planning and 585 effectively alerting authorities and the public to abnormal fire danger conditions. FWI is 586 extensively used in operational global information platforms such as the European Forest Fire 587 Information System (EFFIS; https://forest-fire.emergency.copernicus.eu/, last access: 2 June 2024) and the Global Wildfire Information System (GWIS; https://gwis.jrc.ec.europa.eu/, last 588 access: 2 June 2024), and the Canadian Wildland Fire Information System (CWFIS; 589 590 http://cwfis.cfs.nrcan.gc.ca/, last access: 2 June 2024). 591

In addition to well established fire danger forecasts with lead times of a few days, skilful 592 593 predictions of fire danger can be made on sub-seasonal to seasonal time scale (S2S) for 594 Mediterranean Europe (Bedia et al., 2015), United States (Roads et al., 2010) and Asia 595 (Spessa et al., 2015). Drought and fire weather conditions throughout the world have been 596 found to correlate with large scale climate patterns such as the El Niño Southern Oscillation 597 (ENSO) (Field et al., 2016; Chen et al., 2017), the Indian Ocean Dipole (Cai et al., 2009) and 598 other climate modes such as the Atlantic Multidecadal Oscillation and the Pacific Decadal Oscillation implying predictability of fire-favourable weather conditions for various seasons and 599 600 regions (Aragão et al., 2018; Turco et al., 2018). Following the concept of seamless (Wetterhall 601 et al., 2018) prediction of fire weather on S2S timescales after Di Giuseppe et al. (2020). Di 602 Giuseppe et al. (2024) and Dowdy (2020), we collated FWI data from reanalyses and forecasts 603 designed to operate on S2S lead times of 10 days to 7 months. Specifically, the FWI is derived 604 from the operational monitoring provided by ERA5-land (Muñoz-Sabater et al., 2021), the forecasts from the operational high-resolution ECMWF weather system, and the seasonal 605 606 predictions from the ECMWF's long-range forecasting system, SEAS5 (Johnson et al., 2019; 607 Di Giuseppe et al., 2020; Di Giuseppe et al., 2024). The predictions are compared to recorded 608 peaks in fire activity, both in terms of burned areas and active fires as observed by the MODIS 609 satellites. These prediction systems vary in spatial and temporal resolutions. Short to mediumrange FWI forecasts (up to 10 days) are available daily at a resolution of 9 km, while the FWI 610 611 seasonal forecast is available monthly at a resolution of approximately 25 km, however 612 seasonal skill is limited to 2-3 months in normal conditions (Di Giuseppe et al., 2024).





614 Predictability is assessed qualitatively by visually examining how extreme FWI was as a 615 precursor of the event, replicating the use of this indicator by fire agencies during the fire 616 season. Most fire agencies would have local information on fuel conditions and would be able 617 to interpret FWI values in an educated manner, thus reducing intervention on high FWI values 618 in fuel conditions that are not hazardous. This fuel consideration would not be included in a 619 validation of the FWI using traditional skill scores, as these would be dominated by false 620 alarms. We retain that FWI is an index representing flammability and cannot fairly be validated 621 against fire activities.

622

623 **2.3.2** Identifying Key Drivers of Focal Events

2.3.2.1 Overview

624 625

626

627 We used two modelling systems with similar fire predictors to diagnose the direct drivers of each focal event that was shortlisted: the Probability of Fire (PoF) model (McNorton et al., 628 629 2024) and the Controls on Fire (ConFire) attribution framework (Kelley et al., 2019; Kelley et 630 al., 2021). The PoF model diagnoses the drivers of active fire (AF) observations from the 631 MODIS MCD14ML active fire product (collection 6.1; 1 km resolution; Giglio et al., 2016; NASA 632 FIRMS, 2020) using Shapley values (Lundberg and Lee, 2017), while ConFire diagnoses 633 drivers of BA from the MODIS MCD64A1 BA product (collection 6.1; 500 m resolution; Giglio 634 et al., 2018) regridded to 0.5°. Fires flagged as low confidence in the AF product were not 635 used. Although AF and BA have been used widely in global and regional scientific studies, 636 there are substantial differences between the two products (Roy et al., 2008; Di Giuseppe et 637 al., 2021; Chuvieco et al., 2019) and the strength of the relationship between them can vary 638 regionally (van der Werf et al., 2017; Hantson et al., 2013). Our use of two observational fire 639 products and two distinct model approaches provides a way to account for inherent 640 uncertainties in the definition of the fire events and the uncertainties in the methodologies.

641

642 Both methods use a number of predictors (we refer to the single variables as the drivers) of 643 AF or BA. The drivers are grouped into four main categories (we refer to grouped drivers as 644 the controls): weather; fuel abundance; fuel moisture, and; other drivers for PoF and human 645 drivers for ConFire (see Table 3). PoF drivers in "Other" include ignition and suppression 646 efforts as well as the residual error between forecast and observed fire activity. Grouping the 647 set of drivers between the four identified controls-weather, fuel moisture, fuel load, and 648 ignitions—is not always straightforward, as fuel moisture and weather variables are strongly 649 correlated, and fuel load is also related to weather conditions. Hence, some drivers can be 650 associated with more than one control. The categorisation stems mostly from the way the 651 driving datasets have been obtained and their underlying resolutions. We have also 652 considered the traditional approach of assessing fire weather in isolation within most fire 653 danger assessment metrics. We believe that grouping these metrics under the umbrella term 654 'control weather' offers a concise way to reference the drivers of the Fire Weather Index 655 (Matthews, 2009). Despite this, it is important to note that the techniques employed ensure 656 contribution from specific variables cannot be double counted between categories. Both 657 ConFire and PoF are also capable of quantifying the contribution of individual drivers (Kelley 658 et al., 2019; McNorton et al., 2024). However, we will focus our analysis on the impact of the 659 controls.

660

661 The selected weather parameters are consistent with each other and are available at a base 662 resolution of 9 km. Land cover maps are available at a higher resolution of 1 km. Additionally, 663 fuel load is constrained by external (to the weather model) observations from the ESA CCI 664 above ground biomass product and satellite inversion estimates of net ecosystem exchange 665 of C, while the fuel moisture is parameterized using land surface parameters constrained by 666 the weather variables.





668 2.3.2.2 Drivers and Controls Used in Fire Event Analysis

669

670 For our assessment of the contribution of weather and fuel moisture to the anomalous events 671 we take several predictors from ERA5-Land (9 km resolution; Muñoz-Sabater et al., 2021), 672 specifically variables that are known to correlate with AF or BA (Bistinas et al., 2014; Haas et 673 al., 2022; Haas et al., 2022). The drivers considered for each control are listed in Table 3. For 674 the weather component in isolation, we use 2m temperature, 2m dewpoint temperature, 10m 675 wind speed and daily total precipitation (note that these are the prerequisite variables used in 676 the formulation of the FWI; van Wagner, 1987). We use a fuel characteristic model to estimate 677 the fuel load and fuel moisture components following McNorton and Di Giuseppe (2024), with 678 model estimates of fuel moisture constrained by estimates of leaf area index (LAI) from the 679 ECMWF's Integrated Forecast System (IFS) and model estimates of fuel loads constrained 680 by aboveground biomass estimates from the ESA Climate Change Initiative (CCI; Santoro and 681 Cartus, 2021) and net ecosystem exchange estimates from the Copernicus Atmosphere 682 Monitoring Service (CAMS; Agustí-Panareda et al., 2019). Additional predictors regarding fuel 683 load and state include vegetation cover and type (Table 3). Proxies for ignition and 684 suppression controls, placed within the "Other" set of controls, are more challenging to 685 establish. Currently, we use population density, urban fraction, cropland fraction, pasture 686 fraction, lightning, orography (Table 3). For consistency all variables are interpolated to 9 km resolution. See "Modelling frameworks" in the Extended Methods Supplement for a 687 688 detailed description. PoF accurately captures the spatiotemporal patterns of fire activity in 689 Canada and Western Amazonia and effectively identifies areas of high fire danger in 690 Alexandroupolis, Greece. However, it underestimates the severity of fire events in that area. 691 See "Evaluation" in the Extended Methods Supplement for detailed evaluation.

692





694 Table 3: Drivers of fire and their parent control group included in the event fire analyses using ConFire and PoF. Drivers are individual explanatory variables, which are grouped into weather, fuel load, fuel moisture, and other controls. ** The 'Other' category includes proxies for ignition and suppression controls plus the missed prediction. +ive or -ive under "ConFire 695 696

697 controls" describes if a driver increases or decreases BA in ConFire. 698

Driver	POF control	ConFire	Frequen	Time	Source	Reference
		controls	су	Period		
2m Temperature	Weather	Weather	Daily	Jan 2014-	ERA5-Land	Muñoz-Sabater et
		+ive		NRT		al. 2021
2m Dewpoint	Weather	Weather	Daily	Jan 2014-	ERA5-Land	Muñoz-Sabater et
Temperature		-ive		NRT		al. 2021
10m Wind Speed+	Weather	Not used	Daily	Jan 2014- NRT	ERA5-Land	Muñoz-Sabater et al. 2021
Precipitation	Weather	Weather -ive	Daily	Jan 2014- NRT	ERA5-Land	Muñoz-Sabater et al. 2021
Live Leaf Fuel Load	Fuel Load	Not used	Daily	2014- 2020	Fuel Model	McNorton et al. 2024a
Live Wood Fuel Load	Fuel Load	Not used	Daily	2014- 2020	Fuel Model	McNorton et al. 2024a
Dead Foliage Fuel Load	Fuel Load	Not used	Daily	2014- 2020	Fuel Model	McNorton et al. 2024a
Dead Wood Fuel Load	Fuel Load	Not used	Daily	2014- 2020	Fuel Model	McNorton et al. 2024a
Mean & Max Vegetation Optical Depth (VOD) of the last 12 months	Not used	Fuel Load +ive	Monthly	2014-NRT	Satellite (SMOS)	Wigneron et al 2021
Vegetation Optical Depth (VOD)	Moisture	Moisture -ive	Monthly	2014-NRT	Satellite (SMOS)	Wigneron et al 2021
Low Vegetation (LAI)	Fuel Load/Moisture	Not used	Monthly	2002-2020 climatology	Satellite (multi-sensor)	Boussetta et al., 2021
High Vegetation (LAI)	Fuel Load/Moisture	Not used	Monthly	2002-2020 climatology	Satellite (multi-sensor)	Boussetta et al., 2021
Live Fuel Moisture Content	Fuel Moisture	Fuel Moisture -ive	Daily	2014-NRT	Fuel Model	McNorton et al. 2024a
Dead Fuel Moisture	Fuel Moisture	Fuel Moisture	Daily	2014-NRT	Fuel Model	McNorton et al.
Content		-ive				2024a
Snow cover	Fuel Moisture	Snow -ive	Daily	2014-NRT	ERA5-Land	Muñoz-Sabater et al. 2021
Pasture	Not used	Ignitions +ive Suppression -ive	Annual	2014-2023	HYDE	Klein Goldewijk et al., 2011
Cropland	Not used	Ignitions +ive Suppression -ive	Annual	2014-2023	HYDE	Klein Goldewijk et al., 2011
Urban population	Not used	Ignitions +ive Suppression -ive	Annual	2014-2023	HYDE	Klein Goldewijk et al., 2011
Rural populations	Not used	Ignitions +ive Suppression -ive	Annual	2014-2023	HYDE	Klein Goldewijk et al., 2011
Lightning	Not used	Ignitions +ive	Monthly	2000- 2020 climatology	LIS/OTD	Cecil et al., 2014
Type of Vegetation	Other**	Not used	Fixed	Jan 2014- NRT	ECLand	Boussetta et al., 2021
Urban Fraction	Other**	Not used	Fixed	Jan 2014- NRT	ECLand	McNorton et al. 2023
Orography	Other**	Not used	Fixed	Jan 2014- NRT	ECLand	Boussetta et al., 2021





701 2.3.2.3 Analysis of Fire Drivers

702

703 The PoF system uses gradient-boosted decision trees from the XGBoost library on detected 704 AF (McNorton et al., 2024). The training iteratively adds decision trees to an ensemble of 705 models to correct for errors made by previous iterations, resulting in a computationally efficient 706 optimization (Chen and Guestrin, 2016). The system training uses a classifier approach which defines a positive hit as an AF detection within the grid cell on a given day. The driver 707 708 attribution is performed using the SHapley Additive exPlanations (SHAP) method taken from 709 the SHAP library (Lundberg and Lee, 2017). These are then combined to provide overall 710 attribution to one of the four controls for AF predictions.

711

712 ConFire is an evidence-based uncertainty attribution framework that uses Bayesian Inference 713 to evaluate the likelihood of different control strengths and their influence on observed BA. 714 Rather than produce a single number output at a given location/time. ConFire produces a 715 probability distribution based on relationships found between driving data input into a simple 716 semi-empirical process-representation fire model, and BA observations. ConFire's 717 hierarchical design permits each variable representing a fire driver to contribute to multiple 718 controls (weather, fuel moisture, fuel abundance, or others) simultaneously. ConFire operates 719 on a monthly time step and all drivers are aggregated to monthly daily means using the FLAME 720 system (Barbosa, 2024). Each control operates as an optimised weighted linear combination 721 of its respective drivers. The influence of each control is then represented using a logistic 722 function, with BA calculated as the product of the four controls. Bayesian inference is used to 723 optimise the weighted contribution of drivers to each control, as well as the manner in which 724 each control affects BA. ConFire also includes a stochasticity term that represents the inherent 725 unpredictability in fire occurrence and resultant BA which can occur under similar bioclimatic 726 conditions (Kelley et al., 2021). We use the mean logit transformed BA distance between 727 distributions with (equation 8 in supplement) and without (equation 7 in supplement) this 728 stochasticity term provides a measure of uncertainty that has not been captured within the drivers considered. See "Modelling frameworks" in the Extended Methods Supplement 729 730 for a detailed description.

731

732 The model is trained and ran between 2014-2023, the common period of all driving datasets. 733 The model is trained on 50% of BA before being run in a predictive model for the full dataset. 734 We also perform a separate run, training on all data from 2014-2018 and evaluating against 735 2019-2023 following the protocol outlined in Barbosa (2024). This evaluation protocol aims to 736 test the ConFire model's ability to accurately represent the range of uncertainties and generate 737 observed distributions through its probabilistic framework. This tells us that the framework can 738 capture and reflect inherent uncertainties in fire processes during optimization. The model 739 consistently captures observations within its uncertainty range across all regions, strongly 740 aligning with real-world data and effectively representing BA anomalies and the stochastic 741 nature of fire in our regions. However, it consistently exhibits a low bias, often underestimating 742 BA, particularly in high-burn regions like deforestation areas in Western Amazonia and 743 northern Canada. It can also capture the anomalies in BA, particularly in 2023, though it 744 sometimes underestimates the anomaly's magnitude when not using stochasticity. Therefore 745 the size of BA anomalies may be underestimated. See "Evaluation" in the Extended 746 Methods Supplement for detailed evaluation.

747

748 **2.4 Attribution to Global Change** 749

750 2.4.1 Role of Global Change Factors

751

In addition to direct drivers of fire events as outlined in the previous section (e.g. weather, fuel,
 moisture, ignition and suppression), fires are also influenced by global change factors such as
 climate and land-use change which vary over longer timescales. Since the pre-industrial era,





755 global mean temperature has increased by nearly 1.3°C (Betts et al., 2023; Forster et al., 756 2023), with regional temperatures even higher in some areas, providing additional potential 757 for fuel drying and facilitating fire ignition and propagation. Climate change has also resulted 758 in altered precipitation patterns, with total rainfall and dry season length increasing or 759 decreasing variably across regions (Polade et al., 2014; Swain et al., 2018; IPCC, 2023a). 760 Meanwhile, changes to fuel load and ignition rates are driven by climate change and 761 anthropogenic land-use, with varying effects regionally (Finney et al., 2018; Romps, 2019). 762 For example, in fuel-limited savannah biomes land-use change can drive more fragmented 763 fuel loads and a reduction in fire (or an increase in fire resulting from land abandonment), 764 whereas in forest ecosystems fragmentation provides more potential for ignition and leads to 765 increases in fire occurrence (Andela et al., 2017; Rosan et al., 2022).

766 767

2.4.2 Overview of Attribution Approaches

768

769 In this report, we apply different modelling techniques with consistent meteorological and fire 770 observations to attribute (i) regional changes in the probability of high fire weather to 771 anthropogenic forcing (Section 2.4.3) and (ii) changes in monthly BA to total climate forcing, 772 socio-economic change, and all forcing (Section 2.4.4), specifically targeting our focal regions 773 in both cases. Our analyses require separate approaches to attributing change to each of 774 these global change factors. Our attribution to anthropogenic forcing explicitly targets the 775 changes driven by anthropogenic emissions from greenhouse gases and land-use change, 776 following the IPCC WGI definition (Hegerl et al., 2009; Mengel et al., 2021). We prescribe 777 these emissions in a model to specifically isolate human forcing from natural variability 778 (Section 2.4.3). Our attribution to total climate forcing considers changes driven by climate change since the pre-industrial period, including both anthropogenic forcing and natural 779 780 variability in line with the IPCC WGII definition of climate change impact attribution (IPCC, 781 2023b; IPCC 2023c). This involves comparing simulations driven with historical reanalysis to 782 a detrended counterfactual simulation with the historical warming signal removed (with both 783 simulations including historical transient land-use change) and therefore only the impacts of 784 climate change are attributed, not distinguishing between anthropogenic or natural causes 785 (Mengel et al., 2021; Burton, Lampe et al., 2023). Our attribution to socio-economic factors is 786 applied via the same set of simulations as our attribution to total climate forcing. The role of 787 socio-economic factors is isolated by comparing the early industrial period to the late industrial 788 period in the counterfactual scenario, in which only land-use and population density are 789 allowed to change (Burton, Lampe et al., 2023). Finally, all forcing compares the early 790 industrial period in the counterfactual scenario to the last industrial period in the factual 791 scenario, which gives the net effect of all forcings combined. These are summarised in the 792 Table 4 below.

- 793
- 794





Term	Definition	on Experiments compared Framework		Application				
Event attribution for fire weather								
Anthropogenic Forcing	Change in fire weather driven by anthropogenic emissions from greenhouse gases, land-use change and aerosols. As per (Ciavarella et al., 2018; Li et al., 2021)	human emissions NAT: Natural-only forcing from solar variability and volcanoes	HadGEM3-A attribution ensemble. 0.5 degree resolution	Fire weather (FWI)				
	Impact	s attribution for burned area						
Total climate forcing	Changes in BA due to climate change, irrespective of the cause of warming. As per (Mengel et al., 2021 and Frieler et al., 2024)	Factual (2003-2019): present-day climate (driven by GSWP3-W5E5 reanalysis), CO2, land-use and population Counterfactual (2003-2019): De-trended historical climate (warming signal removed), CO2 fixed at 1901 value, present-day land-use and population Development-day land-use and population Development-day land-use	ISIMIP3a impact attribution. 0.5 degree resolution	FireMIP ensemble and ConFire				
Socio-economic factors	Changes in BA due to land-use and population change. As per (Burton, Lampe et al., 2023)	Counterfactual (1901-1917): Warming trend removed, fixed 1901 CO2, limited land use and population change Counterfactual (2003-2019): Warming trend removed, fixed 1901 CO2, present-day land use and population	ISIMIP3a impact attribution	FireMIP ensemble and ConFire				
All forcing	Changes in BA due to climate, land-use and population change. As per (Burton, Lampe et al., 2023)	Counterfactual (1901-1917): Warming trend removed, fixed 1901 CO2, limited land use and population change Factual (2003-2019): Historical climate driven by reanalysis	ISIMIP3a impact attribution	FireMIP ensemble				

796 **Table 4**: Summary of the attribution approaches used in this report.

797

798 The tools described here enable us to assess the influence of climate and socio-economic 799 forcing on fire from 3 different perspectives. We use FWI to assess how the probability of 800 experiencing meteorological conditions conducive to high (>90th percentile) fire danger has 801 changed as a result of anthropogenic forcing. As climate change has a direct impact on fire weather, this approach enables us to isolate its effects without confounding factors of land-802 803 use change and ignitions, and tells us how a fire might develop once ignited. We use ConFIRE 804 to model high (>90th percentile) BA and its uncertainty as a result of climate change and socio-805 economic factors, which other fire models are unable to capture well. We use an ensemble of 806 7 fire models from FireMIP to put these specific extreme events into the context of changes in 807 global median BA resulting from climate change and socio-economics in the recent present-808 day period. The fire model ensemble takes into account biogeochemical feedbacks between 809 fire, climate, land-use and vegetation as a result of altered forcing across the 7 Dynamic Global 810 Vegetation Models (DGVMs). Each of these methods is described in more detail below.

811 2.4.3 Attributing Change in Likelihood of Fire Weather to Anthropogenic Forcing

We use an established approach to attribute change in probability of high (>90th percentile) fire weather conditions to anthropogenic forcing. The approach uses estimates of FWI, as used in previous studies from the World Weather Attribution (Barnes et al., 2023), using outputs from the HadGEM3-A large ensemble (Christidis et al., 2013; Ciavarella et al., 2018).

816 It follows the approach introduced by Stott et al. (2004) for attributing extreme weather events,





817 and it has been employed in other attribution studies targeting fire weather, such as Li et al. (2021).

819

As outlined in **Section 2.3.1**, the FWI is used operationally and in research contexts to rate fire danger based on meteorological conditions. Due to the availability of model output variables we use maximum daily temperature at 1.5 m as a proxy for noon values, total daily precipitation, mean daily relative humidity at 1.5 m, and mean daily wind speed at 10 m, following Perry et al. (2022). We calculate the daily FWI for the month of 2023-24 peak BA anomaly for each focus region, using the same month and region for validation over the historical timeseries (1960-2013) (**Supplementary Figures S41-S43**).

827

828 We validate and bias-correct the model estimates of high FWI for the period 1960-2013 by 829 comparing a 15-member HadGEM3-A ensemble with ERA5 reanalysis data (C3S, 2024) 830 representing "observed" FWI. The 0.25 degree resolution observed FWI from ERA5 was 831 coarsened by linear interpolation (calculated by extending the gradient of the closest two 832 points) to match the 0.5 degree model grid. We compare the timeseries of individual 833 components of the FWI (Figure S40), and the distribution of the modelled and observed FWI 834 (Supplementary Figures S41-S43), and apply a simple linear regression to find the bias 835 correction required for the 2023 model output. Before bias-correction, the modelled FWI is 836 generally higher than the observed FWI, and some regions (e.g. Greece) require a larger 837 correction than others. The correction adjusts the trend and absolute value while maintaining 838 variability, and the model successfully reproduces the observed distribution after applying the 839 correction in each region (see Extended Methods and Evaluation Supplement).

840

841 For the events occurring in the 2023-24 fire season, we calculate the FWI from the HadGEM3-842 A model simulations comprising 2 experiments of 525 members each, one driven by all 843 forcings including historical greenhouse gas emissions, aerosols, zonal-mean ozone 844 concentrations, land-use change and natural forcing (ALL), and a second counterfactual 845 simulation with natural-only forcing from solar variability and volcanic emissions, and 1850 846 land-use (NAT). By applying the bias correction from the previous step, and comparing the 847 fire weather in the two simulations to the 2023 observed FWI from ERA5, we calculate the 848 change in probability of high (>90th percentile) fire weather due to anthropogenic forcing.

8492.4.4Attributing Change in Regional BA to Total Climate Forcing, Socio-economic850Factors and All Forcing

851

852 2.4.4.1 Monthly High Burned Area

853

We use the ConFire attribution framework to attribute high (>95th percentile for Canada and Western Amazonia and >90th percentile for Greece) BA in the month of peak anomaly to total climate forcing and socio-economic factors using the ISIMIP3a protocol (see **Table 5**). We used 90th percentile for Greece as using the 95th would restrict our analysis to just three grid cells, and our uncertainty ranges are too large to make a clear statement on such a small subset of data.

860

861 We trained ConFire on observed monthly BA from MODIS MCD64A1 during 2003-2019 at 862 0.5° across the entire region. For model training, we drive ConFire with GSWP3-W5E5 forcings provided at a 0.5° spatial resolution by ISIMIP3a, as outlined in Table 5. The land 863 864 surface information (tree cover and non-tree vegetated cover) is derived from the JULES-ES 865 model's ISIMIP configuration described by Mathison et al. (2023) driven by ISIMIP3a GSWP3-866 W5E5 forcings. This model includes dynamic vegetation, i.e changing vegetation cover in 867 response to climate variables, growth, plant competition and mortality. So as not to double count the impact of fire, we turn the interactive vegetation-fire model off. The bias in this land 868 surface information is adjusted to the MODIS Vegetation Continuous Fields collection 6.1 869





870 remote sensed data for <60°N DiMiceli et al. (2022) and collection 6 for <60°N DiMiceli et al. (2015) using a trend-preserving empirical quantile mapping bias adjustment method 871 872 implemented with the ibicus software package Spuler et al. (2024). The bias adjustment 873 method calibrates a mapping between the empirical cumulative distribution function of each 874 surface cover type at each grid cell derived from the JULES-ES model output and the 875 corresponding quantiles in the MODIS remote sensed data at this grid cell over the reference 876 period (2003-2019). This method significantly reduces the model bias in the JULES-ES output 877 over the historical period for most regions and variables, ensuring accurate mean and 878 distribution shapes while preserving trends between historical and future periods (Figure 879 S22). See "Data and Data Processing" in Extended Methods Supplement for details.

880 881

882 ConFire was run in predictive mode on a monthly timestep and with a similar structure to that 883 used in Section 2.3.2.3, with variables representing specific drivers grouped into controls as 884 listed in Table 5. However, the specific driving variables used differed to Section 2.3.2.3 due 885 to the selection of variables available in the forcing data. Specifically, fuel load controls were 886 represented by total vegetation cover and tree cover as drivers; fuel moisture controls were 887 represented by mean consecutive dry days within each month, the fraction of dry days within 888 the month, daily mean precipitation, mean and maximum monthly temperature, mean and 889 maximum VPD as drivers; ignition controls were represented by climatological lightning, 890 pasture, crop and population density as drivers; and suppression controls were represented 891 by pasture, crop and population density drivers. The Bayesian Inference procedure used here 892 is useful when using correlated drivers, which are accounted for in the weights given to driver 893 contributions (Gelman et al., 2013; Kelley et al., 2023). ConFire simulations presented here 894 were performed on all data, with the three experiments outlined under "simulation framework". 895

For each of the experiments detailed above, we output the probability distribution BA for the
cells with the top 5% burned fractions for Canada and Western Amazonia, and 10% burned
fractions for Greece, to target the most extreme levels of burning.

899

900 To determine the impact of total climate forcing and socioeconomic factors on the increased 901 BA during our focal events, we conducted a paired sampling of monthly burned areas in our 902 target month(s) throughout the test period. This involved comparing the factual and 903 counterfactual scenarios for total climate forcing, as well as the counterfactual and early 904 industrial scenarios for socioeconomic factors (see Table 5). As there is no climate influence 905 in the Early Industrial simulation, we first adjusted the target event (a monthly regional BA 906 value) to that expected without climate change. For this adjustment, we find the percentile of 907 the observed BA in the factual and find the BA at the same percentile in the counterfactual. 908 We used paired samples to account for the uncertainty in the underlying mechanisms 909 relating our drivers to burned area, which would co-vary between experiments as per Kelley 910 et al. (2021). In total, we took 200 samples over the 17 years of each simulation, resulting in 911 3400 pairs. 912

913 The likelihood was then simply determined by the number of ensemble members in the 914 factual scenario that predicted greater BA than the counterfactual for total climate forcing, or 915 the counterfactual predicting greater BA than the early industrial scenario for socioeconomic 916 factors. The relative increase in BA extent is the BA in factual over counterfactual (total 917 climate forcing) or counterfactual over early industrial (socioeconomic).

918
919 As per, Section 2.3.2.3 we evaluated the model using the procedure outline in Barbosa
920 (2024). We do a separate training on 50% of the data between 2003-2011 and evaluation on
921 2012-2019. The framework using ISIMIP3a reanalysis data outperforms its near-real-time
922 counterpart in Section 2.3.2.3 at simulating burned areas, effectively representing high
923 burned areas and extremes across all regions. It shows a high probability of observations in





924 areas with extreme fires, indicating reliability for attribution analysis. Further details of the

- 925 method can be found in supplement **Extended Methods and Evaluation Supplement**.
- 926
- 927
- 928
- Table 5: Explanatory variables used for attributing extreme BA (Section 2.4.4) and decadal
 outlook (Section 2.5.2). The explanatory variables are forcing data from the Inter-Sectoral
 Impact Model Intercomparison Project (ISIMIP) protocols ISIMIP3a and ISIMIP3b (Frieler et
 al., 2024). +ive or -ive under "Controls" describes if a driver increases or decreases BA.

Variable	Control s	Construction	Source	Reference	
Max. consecutive dry days	Moisture +ive	Monthly max of running count of days since rainfall > 0.1mm/m	Based on precipitation from ISIMIP3a/3b	Frieler et al. (2024)	
No. dry days	Moisture +ive	fractional no. days of rainfall < 0.1mm/m	Based on precipitation from ISIMIP3a/3b	Frieler et al. (2024)	
Maximum monthly temperature	Moisture +ive	maximum of maximum daily temperature within the month	Daily temperature appriximated os ISIMIP3a/3b daily mean temperature + 0.5xdaily temperature range	Frieler et al. (2024)	
Mean monthly temperature	Moisture +ive		ISIMIP3a/3b	Frieler et al. (2024)	
Mean monthly VPD	Moisture +ive	mean of daily values constructed from specific humidity, surface pressure and max. temperature	ISIMIP3a/3b	Frieler et al. (2024); Barbosa (2024)	
Maximum monthly VPD	Moisture +ive	max of daily values	ISIMIP3a/3b	Frieler et al. (2024); Barbosa (2024	
Tree Cover	Moisture -ive & Fuel +ive	JULES-ISIMIP annual mean tree cover bias-corrected to VCF vs JULES-ISIMIP3a factual interpolated to monthly from annual values	JULES-ES-ISIMIP VCF	DiMiceli et al. (2017); Adzhar et al. (2022); Mathison et al. (2023)	
Total vegetation cover	Fuel +ive	tree cover plus none-tree vegetated cover simulated by JULES and bias-corrected as above	JULES-ES-ISIMIP VCF	DiMiceli et al. (2017); Adzhar et al. (2022); Mathison et al. (2023)	
Lightning	Ignitions +ive	Climatology	LIS taken from ISIMIP3a	Kelley et al. (2014); Frieler et al. (2024)	
Cropland	Ignitions +ive/ Suppression -ive	Interpolated from annual to monthly	ISIMIP3a/3b	Frieler et al. (2024))	
Pasture	Ignitions +ive/ Suppression -ive	Interpolated from annual to monthly	ISIMIP3a/3b	Frieler et al. (2024)	
Population Density	Ignitions +ive/ Suppression -ive	Interpolated from annual to monthly	ISIMIP3a/3b	Frieler et al. (2024)	

933 934

935 2.4.4.2 Monthly Median Burned Area

Finally, we report changes in present-day monthly median BA to total climate forcing, socioeconomic factors and all forcings using a novel attribution method developed using state-ofthe-art global fire models from the Fire Model Intercomparison Project (FireMIP) (Burton,
Lampe et al., 2023). This employs the same ISIMIP3a simulation framework outlined above,





with 7 fire-enabled DGVMs (see Table S1) for the period 1901-2019 for the factual and
counterfactual experiments. For clarity, ConFire was not used in this element of our attribution
approaches; rather, the native fire modelling scheme of each fire-enabled DGVM was
employed. Model fire schemes are described in Burton, Lampe et al. (2023).

944 We assess the recent period 2003-2019 in the historical simulation against the counterfactual 945 simulation to find how BA has changed in response to total climate forcing. To attribute 946 changes in BA to socioeconomic factors, the early industrial period (1901-1917) is compared 947 to the recent period (2003-2019) in the counterfactual only simulation. To assess the impact 948 of all forcings the early industrial period in the counterfactual is compared to the recent period 949 in the factual simulation.

950

951 A weighted ensemble of the monthly outputs of BA, based on regional performance against 952 observational data from GFED5 and FireCCI is used for the analysis. Due to large differences 953 in absolute values of BA between the GFED5 and FireCCI observational datasets and across 954 the models, the weightings in the ensemble are based on model capability to capture relative 955 anomalies present in the observational datasets on a regional basis, and all changes are 956 reported as relative anomalies. We focus on the change in monthly median BA, as the fire 957 models underpredict the high tails of the distribution. The weighted models are randomly 958 resampled to generate uncertainty estimates for each region. The method and results are 959 reported in full for all 43 IPCC AR6 regions in Burton, Lampe et al. (2023), and in the current 960 report we select the IPCC regions that align most closely with our focus regions defined in 961 Section 3.2.

962

963 2.5 Seasonal and Decadal Outlook

964 965 966

2.5.1 Seasonal Forecasts

967 Among the modes of variability in the climate system most relevant to wildfire activity globally 968 is the El Niño-Southern Oscillation (ENSO) (Chen et al., 2017; Mariani et al., 2016; Fuller and 969 Murphy, 2006; Cardil et al., 2023). ENSO is a complex, naturally occurring climate 970 phenomenon characterised by fluctuations in sea surface temperatures and atmospheric 971 pressure across the Pacific Ocean. ENSO has far-reaching consequences generating dry 972 conditions in vast parts of South and Central America, the mediterranean and African savanna. 973 It also exacerbates dry conditions on the west coasts of continents such as California in the 974 United States, southern Europe, and parts of Australia, which can become more susceptible 975 to fires. The geographic areas with such fire favourable teleconnections differ by ENSO phase 976 (Chen et al., 2017).

977

978 Another phenomenon demonstrably linked to global fire activity is the Indian Ocean Dipole 979 (IOD), which occurs in the Indian Ocean and is characterised by differences in sea surface 980 temperatures (SST) between its western and eastern parts. The IOD has a notable influence 981 on weather patterns in various regions of the Southern Hemisphere, particularly affecting the 982 southern and eastern parts of Australia (Harris and Lucas, 2019). Positive IOD events typically 983 lead to reduced rainfall in these areas, resulting in drier-than-normal conditions. However, 984 there is still ongoing debate regarding the direct influence of the IOD on Australian fires, as 985 the signal is often obscured by changes in land management practices (Harris and Lucas, 986 2019).

987

ENSO and the IOD have a planetary scale influence on weather anomalies through
teleconnections. Other atmospheric modes of variability in the Southern, Northern hemisphere
and in the arctic regions can also have strong influence on the seasonal trend of regional
burned areas. For example, Figure S1 shows the climate modes with strongest influence on
burned areas globally.





994 The large uncertainties in the analysis highlights that there are few regions in the word where 995 it is possible to establish statistically significant teleconnection between burned areas and 996 atmospheric modes. Hence, we utilised the Copernicus Climate Change Service (C3S) multi-997 model seasonal prediction system to evaluate large-scale climate modes with most proven 998 links to variation in fire activity: ENSO and IOD. The C3S multimodel incorporates forecasts 999 from several prominent meteorological agencies, including the ECMWF, UK Met Office, 1000 Météo-France, German Weather Service (DWD), Euro-Mediterranean Center on Climate 1001 Change (CMCC), US National Weather Service's National Centers for Environmental 1002 Prediction (NCEP), Japan Meteorological Agency (JMA), and Environment and Climate 1003 Change Canada (ECCC). This multi-model system aids in defining the expected evolution of 1004 these two modes of variability. These indices, calculated as Sea Surface Temperature (SST) 1005 anomalies over the ocean, offer a succinct means of understanding the expected strengths of 1006 atmospheric modes of variability in the coming months.

1007

993

1008 To look more closely at the impact that these two modes might have in landscape flammability 1009 we use seasonal outlooks of the Fire Weather Index (FWI) using one of the models from the 1010 aforementioned multi-model system, ECMWF-System 5. Leveraging seasonal predictions for 1011 the FWI we generated probabilities of exceedance using a 51-member forecast ensemble and 1012 a 24-year model climatological distribution (derived from a 25-member ensemble re-forecast) 1013 covering the period 1993-2016. The probability of exceedance is determined based on the 1014 proportion of forecast members meeting two anomaly criteria at any given geographical point. 1015 We consider the 75th percentile indicative of moderate anomalous conditions and the 95th 1016 percentile indicative of extreme anomalous conditions over a month for the next season.

1017

2.5.2 Decadal Projections of Burned Area

1018 1019

In order to project future changes in BA, we utilised the same modelling approach detailed in 1020 1021 Section 2.4.4.1, following a similar protocol to UNEP et al. (2022). We drive the model with 1022 ISIMIP3a and bias-corrected JULES-ES data. For predictive mode, we used bias-corrected 1023 GCM model outputs from ISIMIP3b. ISIMIP3b provides four sets of driving data for 5 bias-1024 corrected models, including historical data up to 2014 and future scenarios from 2015-2100 1025 (SSP126, SSP370, and SSP585). Each SSP represents future socio-economic pathways and includes GHG emissions that feed each of the 5 GCMS: GFDL-ESM4 (Held et al., 2019), 1026 1027 IPSL-CM6A-LR (Boucher et al., 2020), MPI-ESM1-2-HR (Mauritsen et al., 2019), MRI-ESM2-1028 0 (Yukimoto et al., 2019), and UKESM1-0-LL (Tang et al., 2019; Sellar et al., 2019). As part 1029 of ISIMIP3b, each GCM is bias-corrected as described in Lange (2019).

1030 At present, future projections (aligned with shared socio-economic pathways; SSPs) for land 1031 use and population density forcing were not available for ISIMIP3b, so we only considered the 1032 climate's influence on fire distribution and intensity as well as available fuel from vegetation 1033 cover, and not changes in ignitions or land use. We used JULES-ES land cover outputs as 1034 per the previous section, but with JULES driven by each scenario-GCM combination in turn. 1035 This was then bias-corrected in the same mapping procedure as Section 2.4.4.1, based on 1036 biases between JULES-ES driven by ISIMIP3a and VCF observations to keep consistent with 1037 GCM bias-correcting procedures. This mapping was subsequently applied to the surface 1038 information outputs from JULES-ES driven by historical (1994-2014) and future (2015-2099) 1039 model runs. To preserve the trend in the vegetation cover over the future periods, additive 1040 detrending of the mean was applied. The results in Section 3.5.3 are for the months June-August for Canada, July-September or Greece and August-October for Western Amazonia, 1041 1042 corresponding to those regions' fire seasons today. See "Data and Data Processing" in 1043 Extended Methods Supplement for details.

1044 Our approach provides a probability distribution representing the uncertainty in future 1045 emissions, climate, and vegetation responses for each scenario and year in the period 2010-





1046 2100. For the first 4 years (2010-2014), we joined the historical experiment to each SSP in turn.

For Western Amazonia, we additionally tested a 1-in-100 event under 2010-2020 climate, defined as the BA at the 99th percentile ConFires distribution. We then calculated decadal average likelihoods of the regions' event each decade up to 2100. Return times are simply 1/likelihood. The change in likelihood of an event occurring (of a given return time) was calculated relative to the ISIMIP3b 2010-2020 baseline period.

1053 We also calculated the integrated probability of a return of the 2023 Canadian event within the 1054 expected lifespan of an individual. According to UN population statistics, the average life 1055 expectancy of a Canadian citizen born today is 83 years (United Nations Population Division, 2022). In order to cover the 7-year period after 2100, we extrapolated the annual trend in 1057 probabilities. The integrated probability therefore is calculated as the product of the annual 1058 probability of not seeing a fire event like 2023, for each year between 2023 and 2106.

1059 **3 Results**

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1061 3.1 Extreme Wildfire Events of 2023-24

1063 **3.1.1 Highlights**

- 1065 **Global:** A total of 3.9 million km² burned globally during the 2023-24 fire season, 1066 slightly below the average of previous seasons (4.0 million km²) and ranking 13th since 1067 2001. Despite the lower BA, fire C emissions were 16% above average, totaling 2.4 1068 Pg C, ranking 7th highest since 2003. Global C emissions were pushed up by record 1069 emissions in Canadian boreal forests and pulled down by below-average fire activity 1070 in the African savannahs (the largest contributor to global mean annual totals). If global 1071 savannah fire emissions had been in line with their average in 2023-24, global fire C 1072 emissions would have been the highest on record.
- 1074 North America: Record fire activity in Canada's boreal forests, with BA reaching six 1075 times the average and fire C emissions over nine times the average, contributing 1076 significantly to global C emissions. Canada experienced extreme and widespread fires, 1077 with over 150,000 km² burned, prompting evacuations of 232,000 people and eight 1078 firefighters lost their lives. The United States saw generally below-average fire activity, 1079 but the Lahaina wildfire in Maui, Hawai'i, resulted in 100 civilian deaths, destroyed 1080 2,000 homes, and displaced 10,000 people. Texas recorded its largest ever single fire, 1081 which destroyed 130 homes, killed two civilians, and caused significant livestock 1082 losses. Mexico faced its highest BA in the last decade due to ongoing drought 1083 conditions, with 10,000 km² burned. 1084
- 1085 South America: South America experienced somewhat below-average fire extent • 1086 overall, but notable exceptions included significant anomalies in the northern and 1087 western regions, linked to extended drought and heatwave conditions. In Brazil's 1088 Amazonas state, fire counts reached record highs due to historic drought, severely 1089 impacting air guality in Manaus and leading to deforestation and habitat loss. In Chile, 1090 the Valparaíso wildfire in February 2024 resulted in at least 131 deaths and widespread 1091 property destruction, highlighting severe societal impacts. Fires in Bolivia's La Paz and 1092 Beni departments and Peru's Loreto region caused significant environmental damage 1093 including loss of protected forest and biodiversity, and disruption of indigenous communities' livelihoods. 1094 1095





 Europe: Low wildfire extent in general, however the Alexandroupolis fire in Greece set a new EU record for individual fire size. Individual fires in Greece, Spain, Italy, Portugal, France, and Scotland led to large-scale evacuations, significant suppression costs, disruption of water supplies, damage to infrastructure or agricultural lands, impacts on tourism and local economies, destruction of properties, or loss of life.

- Oceania: Above-average fire activity in northern Australia's savannahs, grasslands, and shrublands was linked to fuel growth from three years of La Niña. Although less impactful than the 2019-20 Black Summer fires, wildfires near Perth resulted in five homes lost and several injuries. The Tara and Mount Isa fires in Queensland destroyed 65 homes and claimed two lives. In Victoria, fires destroyed over 40 homes and injured five firefighters. In New South Wales, forest fires caused widespread smoke-related damages.
- 1110 Asia: The 2023-24 fire season in Asia saw generally low fire activity, with central Asia 1111 experiencing below-average burned areas. Lao PDR, Thailand, and Vietnam had high 1112 fire counts amidst heatwave conditions and a possible uptick in agricultural fire use, 1113 leading to severe regional haze and air quality issues. In Mongolia's Dornod Aimag province, wildfires burned large parts of the Daurian steppe and required extensive 1114 1115 firefighting efforts to stop fires from crossing into Russia. Additionally, wildfires in Russia's southern borders, including Tyumenskaya, Omskaya, and Amurskaya 1116 1117 Oblasts, caused evacuations, property damage, and at least one fatality. 1118
 - Africa: Low wildfire extent in general with BA 13% below average in the African grassland, savannah, and shrubland biome. However, extreme fires in Northern Africa, particularly Algeria and Tunisia, prompted significant emergency responses including assistance from the EU. In Algeria, wildfires occurring in temperatures around 50°C resulted in 34 deaths and over 1,500 evacuations, with over 8,000 personnel deployed to combat the fires. Tunisia faced similar challenges with strong winds exacerbating wildfires, leading to evacuations in the northwestern region. In coastal South Africa, fires in the Western Cape caused structure damage and evacuations.
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11283.1.2Global Perspective from Earth Observations1129

1130 3.1.2.1 Extreme Fire Seasons Evident in Earth Observations

According to the MODIS BA product, 3.9 million km² burned globally during the 2023-24 global fire season (March 2023-February 2024), slightly below the average of previous fire seasons (4.0 million km²) and overall ranking 13th of all fire seasons since 2001 (Jones et al., 2024).
Despite this, fire C emissions were 16% above average at 2.4 Pg C during the 2023-24 global fire season, which ranks 7th amongst all fire seasons since 2003 (based on annual averages of GFED4.1s and GFAS estimates; see methods; Jones et al., 2024).

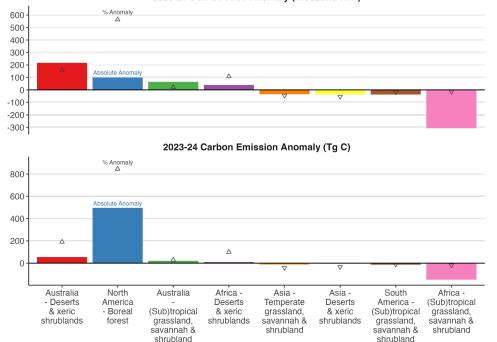
1139 Stark regional contrasts in the anomalies in BA, fire C emissions and individual fire properties 1140 are visible in the Earth Observations on various regional scales (Figure 1, Figure 2, Figure 1141 3). Figure 1 shows the strongest BA and fire C emissions anomalies of 2023-24 at continental 1142 biome scale versus previous fire seasons. BA was around 300 thousand km2 (13%) below 1143 the average of previous fire seasons in the African grassland, savannah and shrubland biome, 1144 which is the largest contributor to global mean annual BA totals. BA was below average in the 1145 South American grassland, savannah and shrubland biome in 2023-24 and in Asian non-forest 1146 biomes. The global pattern of lower BA in savannah-like systems in 2023-24 was not observed 1147 in Australia, where several non-forest biomes experienced well above-average BA.





1149 The North American boreal forests experienced a record-breaking fire season, with BA 1150 reaching six times the average since 2001 and fire C emissions reaching over nine times the average since 2003 (Figure 1; Jones et al., 2024). This strong regional signal primarily 1151 1152 explains the above-average global C emissions total of 2023-24, with the high rates of fire 1153 emissions per unit area in boreal forests aggregating to override the reduced emissions totals 1154 in African and South American savannahs. Record levels of fire C emissions were seen also 1155 across the global pan-boreal forest biome, with fire C emissions surpassing the pan-boreal 1156 record set in 2021 by more than 60%. This is despite a below-average fire season for BA and 1157 fire C emissions in boreal Asia during 2023-24, in contrast to the 2021-22 fire season when 1158 there was a synchronous peak in BA in both the Eurasian and North American boreal regions 1159 (Zheng et al., 2021). According to the Global Fire Atlas, new records for individual fire size 1160 and rate of spread were also set in the North American boreal forests during 2023-24, while 1161 the upper 5% of fires ranked by size and rate of growth in 2023-24 were in the top 2 and 3 1162 years on record since 2002, respectively.

1163



2023-24 Burned Area Anomaly (thousand km²)

1164

1165Figure 1: Anomalies in BA and C (C) emissions for selected continental biomes in the 2023-116624 global fire season (March 2023-February 2024), versus the average of prior fire seasons1167since 2001. The selected regions all contribute at least 0.1% towards global mean annual BA1168and experienced BA anomalies of over $\pm 30,000 \text{ km}^2$ in the 2023-24 global fire season. Relative1169changes (%) are also marked by triangular symbols and can be read on the same scale as1170the absolute values.

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1172

Anomalies in the African (sub-)tropical grasslands, savannahs and shrublands strongly drive
inter-annual variability in global fire C emissions because this biome contributes on average
58% towards total global BA and 40% towards total global fire C emissions. If fire C emissions
from African (sub-)tropical grasslands, savannahs and shrublands had been around average





1177 fire season in 2023-24, then global fire C emissions would have been the greatest of any fire 1178 season on record since 2003.

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Elsewhere at the biome scale, BA extent was in the top three years on record in the South
American broadleaf and mixed forests, the African xeric shrublands, and Australian xeric
shrublands, and the Australian (sub-)tropical grasslands, savannahs and shrublands (Figure
In contrast, BA or fire C emissions were the lowest on record in the European temperate
broadleaf and mixed forests and Asian xeric shrublands, and in the bottom three years on
record in the African savannahs, Asian montane grasslands and shrublands, and European
tropical grasslands and shrublands.

1187

1188 On national levels, the most prominent global anomaly of the 2023-24 fire season occurred in 1189 Canada where BA reached six times the average of previous fire seasons and fire C emissions 1190 reached nine times the average of previous fire seasons. Across the Canadian provinces and 1191 territories, the highest BA or fire C emissions on record were observed in Northwest 1192 Territories, British Columbia, Alberta, and Quebec while Yukon, New Brunswick, and Ontario 1193 also experienced high-ranking years (Figure 2, Figure 3). The positive BA anomalies in 1194 Canada were visible in the MODIS BA dataset from as early as April 2023 in most provinces 1195 and persisted throughout summer through to October and even through to December 2023 1196 and January 2024 in British Columbia and Alberta (Figure S2). Peak anomalies were 1197 observed in Eastern Canada in June 2023, arriving later in western Canada (August-1198 September). Data on individual fire characteristics from the Global Fire Atlas further reveals 1199 new record fire counts in many Canadian provinces, and high-ranking anomalies in fire count and daily rate of growth across Canada, as well as new records for fire size and rate of spread 1200 1201 in provinces of both eastern and western Canada (Figure 4; Jones et al., 2024). Section 1202 3.1.3.4 discusses the unprecedented Canadian fire season of 2023-24 in greater detail, 1203 including its impacts and regional context.

1204

1205 A second prominent regional feature of the 2023-24 global fire season, visible in Earth 1206 observations, is a cluster of administrative regions with positive BA and C emissions 1207 anomalies in the north and west of tropical South America (Figure 2, Figure 3). Bolivia, 1208 Guyana, Suriname and French Guiana, Honduras, Nicaragua and Belize all experienced a 1209 high-ranking fire season at national level in 2023-24. In addition, BA or fire emissions were 1210 ranked in the top three years in western parts of Amazonia including in Amazonas state of Brazil, the Loreto department of Peru, and the La Paz and Beni departments of Bolivia. 1211 1212 Anomalies in the western Amazon spanned June 2023-January 2024, peaking in August-1213 October 2023. In the north of South America, high-ranking fire seasons were seen in 1214 Venezuela, various subdivisions of Guyana, Suriname, and French Guiana, and in Amapá 1215 State in Brazil. The anomalies in northern South America spanned May 2023-February 2024, 1216 peaking in November 2023-February 2024 (Figure S2). The Global Fire Atlas data suggest 1217 that South American anomalies in BA during the 2023-24 fire season were principally driven 1218 by a large number of fires, whereas anomalies in fire size or rate of growth were uncommon 1219 in most of South America (Figure 4). Section 3.1.3.6 discusses the 2023-24 fire season of 1220 tropical South America and its impacts and regional context in greater detail.

1220

Several parts of South and Southeast Asia experienced high-ranking anomalies in BA or fire c emissions during the 2023-24 fire season, including various neighbouring administrative zones of Lao People's Democratic Republic (PDR), Thailand and Vietnam. The temporal peak of these anomalies was broadly in March-May 2023. Data on individual fire characteristics indicates that high-ranking fire counts, rather than anomalies in fire size, were the primary driver of the regional BA anomalies (**Figure 4**). **Section 3.1.3.2** discusses these anomalies and their impacts in greater detail.

1229

1230 The anomalies observed in xeric biomes of Oceania are also apparent as high-ranking BA or 1231 C emissions in the 2023-24 fire season in western parts of Australia, particularly in Western





Australia and the Northern Territory (Figure 2, Figure 3). Fires tended to affect more remote
areas and so the impacts on society were muted in comparison to the Black Summer events
affecting southeast Australia in 2019-20 (Abram et al., 2021); however, Section 3.1.3.5
discusses some notable exceptions.

1237 Other regional pockets of high-ranking BA anomalies or C emissions anomalies were 1238 observed in various dry zones of Africa and the Middle East, including the Sahel, Northern 1239 Africa and the Horn of Africa, Southern Africa (specifically South Africa and Botswana where 1240 three good rainfall years have resulted in grass fuel accumulation), parts of Iran, Iraq, parts of 1241 the Levant region, and parts of the Arabian Peninsula (Figure 2, Figure 3). Although various 1242 aspects of the fire season ranked highly in these regions, they are also fuel-limited with a 1243 generally a low baseline for BA and fire C emissions and the wildfire season. Nonetheless, 1244 regionally impactful wildfires were reported for Algeria, Tunisia and Morocco as well as coastal 1245 regions of South Africa and in Pakistan and are discussed further in Section 3.1.3.1 and 1246 Section 3.1.3.2.

1247 1248





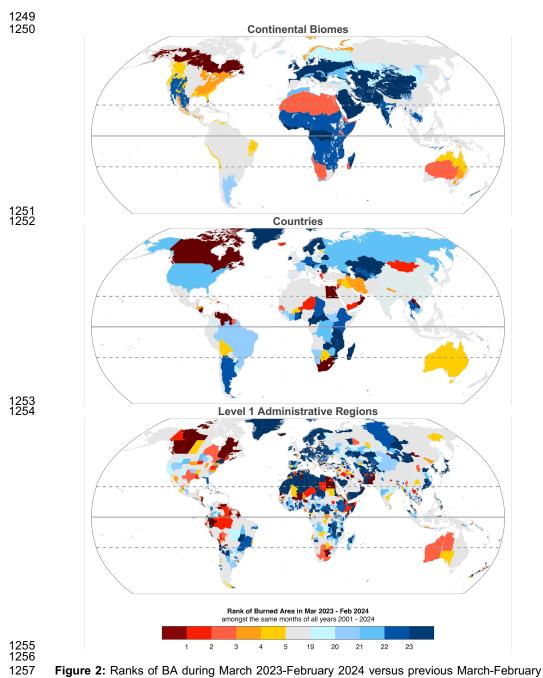


Figure 2: Ranks of BA during March 2023-February 2024 versus previous March-February periods (n = 23 global fire seasons) for three regional layers: (top panel) continental biomes; (middle panel) countries, and; (bottom panel) states or provinces. Results for regions with high-ranking (top 5 years) or low-ranking (bottom 5 years) events are highlighted. The timing of BA anomalies is shown in Supplementary Figure 2.





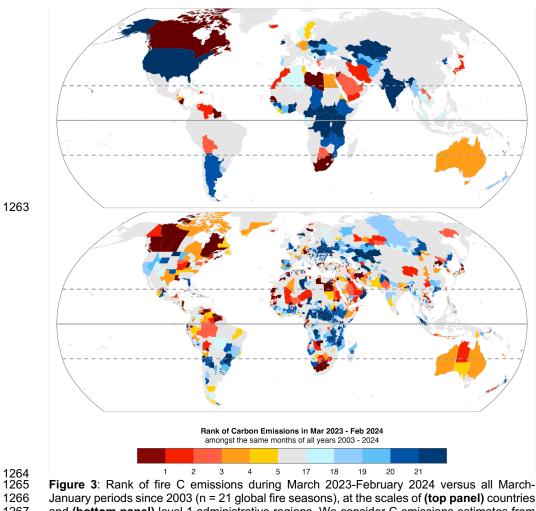


Figure 5. Rank of the C emissions during March 2023-rebruary 2024 versus an March-January periods since 2003 (n = 21 global fire seasons), at the scales of (top panel) countries and (bottom panel) level 1 administrative regions. We consider C emissions estimates from two products (GFAS and GFED), first calculating the mean emissions value from the two products, then ranking the values.





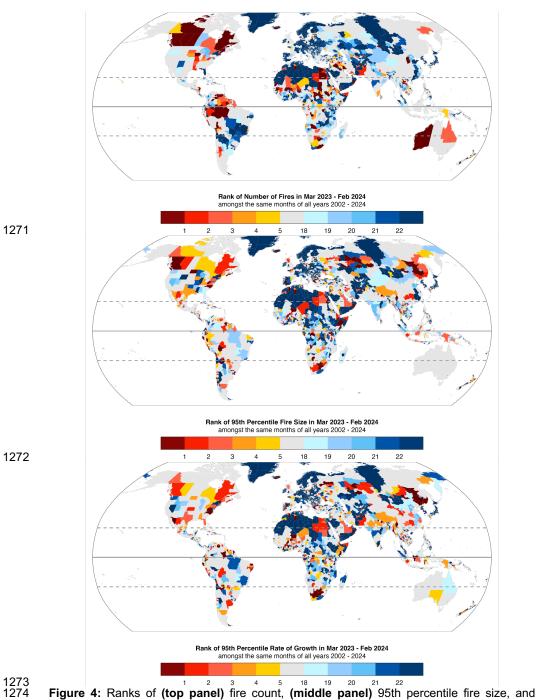


Figure 4: Ranks of (top panel) fire count, (middle panel) 95th percentile fire size, and
(bottom panel) 95th percentile daily rate of growth during March 2023-February 2024 versus
all March-February periods since 2002, at the scale of states or provinces (GADM administrative level 1 regions).





1279 3.1.2.2 Extreme Individual Fires in Earth Observations

1280

1281 The Global Fire Atlas (GFA) is one of several products tracking daily fire progression and 1282 identifying individual fires at global scale based on moderate resolution satellite data (Andela 1283 et al., 2019; Laurent et al., 2018; Artés et al., 2019). The product uses the MODIS BA product 1284 and the smallest unit of disaggregation is 500m and the shortest timestep on which the 1285 expansion of a fire can be observed is daily. The Global Fire Atlas is expected to represent 1286 the dynamics of large fires better than small, fast-moving fires.

1288 The GFA represents some of the most impactful individual fire events in 2023-24 with varying 1289 skill (Table 6; Figure S3, Figure S4, Figure S5). For example, the Alexandroupolis fire that 1290 occurred in the decentralised administration of Macedonia and Thrace, Greece, in late August 1291 was captured reasonably well. The GFA identifies two fires that ignited on 19th August and 1292 merged to form one contiguous burned unit with an area of approximately 900 km². Alignment 1293 of the fire's timing, size and perimeter with high-resolution satellite imagery (Figure S3) and 1294 detailed reports (Xanthopoulos et al., 2024) suggest an overall reliable representation of this 1295 event. The impacts of this fire are discussed in detail below in Section 3.1.3.3. Notably, 1296 observations from the Visible Infrared Imaging Radiometer Suite (VIIRS) sensor, which 1297 features higher resolution and higher overpass frequency, did not improve the characterisation 1298 of this particular event (Figure S3).

1299

A deadly fire near Valparaíso, Chile, is also captured with reasonable skill in the GFA (Figure S4). Around 90 km² was burned, with the fire skirting the city of Placilla de Peñuelas and encroaching upon Viña del Mar and Quilpué (Section 3.1.3.6). The timing, extent and perimeter of the fire as recorded by the GFA compares well with those reported by other sources (Table 6).

1305

Among the largest fires to occur in Canada during 2023-24 happened near the La Grande Reservoir in Quebec, Canada. According to both the Global Fire Atlas and a separate NASA fire tracking product based on the VIIRS sensor, the fire's extent was around 11 thousand km², whereas the National BA Composite (NBAC; Skakun et al. 2022) shows a slightly lower extent nearer 10 thousand km². The timing of the fire also showed high correspondence across the products.

1312

1313 The Lahaina wildfire in Maui, Hawai'i, is an example of an event that was captured poorly by 1314 the GFA. Issues relating to the small scale of this fire relative to the resolution of the MODIS 1315 BA data are evident in **Figure S4**. As the MODIS BA algorithm is focussed on the detection of 1316 wildland fire, its effectiveness in tracking fires at the wildland-urban interface is limited. In this 1317 case, burned areas were not detected in cells in urban areas or at the wildland-urban interface, 1318 and hence the size of the fire was under-estimated significantly (**Table 6**). The timing of the 1319 fire on vegetation land adjacent Lahaina was compatible with reference reports.

1320

1321 Another example of the challenges of defining individual fire extent and applying global 1322 algorithms to do so comes from Western Australia (Figure S5). Two fires recorded by the 1323 Department of Fire and Emergency Services, Western Australia (the Great Sandy Desert and 1324 Anna Plains fires) totalled 57 thousand km² in extent. In the Global Fire Atlas, the burned cells 1325 detected by MODIS were instead dissected into 53 separate fires with the largest unit burning 1326 560 km². The date ranges were also rather different with the first record of fires logged in 1327 agency data one month later than in the Global Fire Atlas and the final record logged one 1328 month earlier.

1329





Event	GFA Fire Size (km²)	GFA Dates	Reported Area (km²) (Reference)	Dates (Reference)	Reference	Comment
Alexandroupol is Wildfire, Greece	892	19/08/2023 to 30/08/2023	930	19/08/2023 to 31/08/2023	s et al.	Good characterisation of the event, with two fires merging in the date range and ultimate fire size comparable to reference.
Fire near La Grande Reservoir, Quebec, Canada	10,725	29/05/2023 to 23/07/2023	11,400 (VIIRS) 9,694 (NBAC)	01/06/2023 to 23/07/2023	NASA Earth Observatory (2023; VIIRS); Jain et al. (2024; NBAC)	Reasonable characterisatior of the fire's extent and timing
Valparaíso Wildfire, Chile	91.54	31/01/2024 to 10/02/2024	85	02/02/2024 to 05/02/2024	NASA Earth Observatory (2024); Copernicus Emergency Management Service (2023a)	Good characterisation of the scale of the event and its perimeters at various wildland-urban interfaces, versus reference data.
Lahaina Wildfire, Maui, Hawaii	1.50	08/08/2023 to 12/08/2023	8.49	08/08/2023 to 09/08/2023	Fire Safety Research Institute (2024)	MODIS data has coarse spatial resolution relative to scale of event. Spread into urban areas not captured.
Western Australia (Great Sandy Desert and Anna Plains fires)	45,544	31/08/2023 to 28/11/2023	56,561	27/09/2023 to 01/11/2023	Department of Fire and Emergency Services, Western Australia (Shapefile for the Great Sandy Desert and Anna Plains fires; Agnes Kristina, pers. comm.)	 GFA splits this event into 53 fires; we report their total combined area. Largest individual fire area ir GFA was 760 km² (ignited 06/09/2023). Great Sandy Desert and Anna Plains fires merged or 25th October 2023.

Table 6: Representation of selected individual fire events in the MODIS BA product (Giglio et al., 2018) and Global Fire Atlas (Andela et al., 2019).

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1334

1335 3.1.3 Year in Review by Continent

1338

1336

1339 South Africa and Botswana experienced higher than average BA and fire size (Figure 2, 1340 Figure 4) in 2023-24, linked to three consecutive years of above-average rainfall that increased grassy fuel loads in the arid (fuel-limited) savannas and grasslands. This has 1341 1342 potentially been exacerbated by a lack of prescribed burning and active fire suppression in the 1343 privately held land and conservancies in the region, that likewise would have resulted in fuel 1344 build-up (Atlas of Namibia, 2021). The socio-economic impacts of these large fires were 1345 minimal (extensive grassland fires linked to high rain years are expected due to periodic 7-20 1346 year wet-dry cycles in these ecosystems). However, some farmers in western Botswana

^{1337 3.1.3.1} Africa





reported increased livestock losses from predators such as lion, wild dog and hyena, with the
 reduced vegetation cover after the fires driving predators to shift from wildlife prey to livestock.

1350 In East Africa, the area burned was extremely low in 2023-24 due to the 2020-2023 triple la 1351 Nina experienced in this region, which causes droughts in East Africa (while increasing rainfall 1352 in southern Africa). This multi-year drought meant that there were no grass fuels to burn and 1353 reduced the likelihood of spread of accidental ignitions in many of the East African rangelands. 1354 However, the extreme fire weather enabled fires to burn through upland forests, which are not 1355 normally flammable. The Aberdare Forest, in Nyeri county Kenya had a fire incident on the 1356 17th of Feb 2023 that destroyed over 160 km², and was attributed to the dry conditions (*Citizen* 1357 Digital, 2023). This also occurred at a time when the country was undergoing a heatwave. 1358 This highlights the different responses of grassland and forest vegetation to the same climate 1359 conditions.

1360

1361 The 2023 heatwave in North Africa exacerbated fire behaviour in the region (AI Jazeera, 1362 2023a). Algeria recorded significant fires in the latter half of July, facilitated by high 1363 temperatures that reached upwards of 48°C (Al Jazeera, 2023b). Over 8,000 personnel, 1364 including firefighters and the military, were deployed to combat rapidly spreading fires across 1365 15 provinces (South African Broadcasting Corporation, 2023a). These efforts were critical in 1366 managing fires that forced over 1,500 people from their homes (euronews, 2023). Despite 1367 these efforts, the wildfires claimed the lives of at least 34 individuals, including 10 soldiers (Al 1368 Jazeera, 2023b).

1369

1370 Neighbouring Tunisia also faced wildfire outbreaks, exacerbated by strong winds that carried 1371 fires across the national border from Algeria, leading to the closure of two border crossings 1372 (Reuters, 2023a). The Tunisian wildfires prompted evacuations in the north-western region of 1373 Tabarka, affecting at least 300 people and extending firefighting efforts to Bizerte, Siliana, and 1374 Beja (Sullivan and Tondo, The Guardian, 2023). Resources such as firefighting aircraft and 1375 personnel were sent from EU nations to help tackle the fires, despite the challenging 1376 conditions imposed by near-record temperatures of 49°C (Gauldie, AirMed&Rescue, 2024). 1377 In August, forest fires in mountainous regions of Morocco were also fanned by strong winds 1378 and facilitated by protracted hot spring and summer temperatures (Erraji, Morocco World 1379 News, 2023; Copernicus Climate Change Service, 2024a).

1380

1381 During December 2023-January 2024, the Western Cape of South Africa experienced 1382 wildfires related to prolonged hot and windy conditions, causing substantial damage and 1383 prompting widespread evacuations. In the Overstrand local municipality, which includes 1384 coastal towns like Pringle Bay and Betty's Bay, multiple fires necessitated evacuations and 1385 destroyed properties. The Hangklip area between Pringle Bay and Betty's Bay was particularly 1386 affected with the fires destroying properties in the Sea Farm private nature reserve. On 1387 January 29, a "code red" status was declared, indicating a serious threat to residential areas, 1388 and evacuations were advised for communities including Silversands and Seafarms (Crisis24, 1389 2024; AfricaNews, 2024). A wildfire swept from Simonstown to Scarborough in Cape Town, 1390 necessitating large-scale evacuation (Hough, IOL News, 2023). This fire was challenging due 1391 to its rapid spread fueled by strong south-easterly winds and high temperatures. The 1392 firefighting efforts were supported by multiple helicopters and ground teams (South African 1393 Broadcasting Corporation, 2023a). The most extensive damage was reported from the Kluitjieskraal fire near Wolseley, where over 220 km² were burned, and more than 40 1394 1395 structures were destroyed. This fire also prompted evacuations and remained uncontained for 1396 several days due to its size and complex terrain that hindered ground access (Crisis24, 2024) 1397 . Despite these extreme wildfires, the plantation forestry industry was not affected, with 1398 relatively low losses due to fire.

- 1399
- 1400 3.1.3.2 Asia





1402 The 2023-24 fire season in Asia was generally not an extreme one, with much of central Asia 1403 experiencing low BA. Siberia, which has seen several record-breaking fire seasons since 2020 1404 resulting in globally significant fire emissions (Zheng et al., 2021), also experienced a 1405 somewhat typical year for BA and fire C emissions. Likewise most provinces of China and 1406 states of India experience a fairly typical fire season.

1408 Nonetheless, there were regional examples of high fire activity in the 2023-24 fire season. The 1409 Dornod Aimag province of eastern Mongolia, near the borders with Russia and China, 1410 experienced several extreme fires during April 2023 that are also visible as anomalies in the 1411 global fire observations (Figure 2, Figure 4). Over 15% of the area of Dornod Aimag burned 1412 in 2023-24 in contrast to the 23-year average of below 5%. The province includes the 1413 Mongolian part of the Daurian steppe, notable for being one of the last remaining undisturbed 1414 steppes in the world (UNESCO World Heritage Centre, 2017). Unusually dry and warm 1415 conditions in eastern Mongolia during spring led to severe wildfires. A notable wildfire spread 1416 into Dornod from neighbouring Sukhbaatar province, fanned by windy, dry conditions (Borneo 1417 Bulletin, 2023). The National Emergency Management Agency mobilised over 250 individuals, 1418 including firefighters and local residents, and helicopters were deployed to manage the fastspreading fires (Borneo Bulletin, 2023). The effects of these wildfires on herder and nomadic 1419 1420 populations compounded the losses of 2 million livestock over the harsh winter in Mongolia, 1421 and the Mongolian Red Cross has provided aid to 4,800 herder households (International 1422 Federation of Red Cross and Red Crescent Societies, 2023).

1423

1407

1424 Although BA extent and fire counts were overall below the 2001-2023 average along the 1425 southern border regions of Russia during 2023-24 (Figure 2, Figure 4), a number of disruptive 1426 wildfires fanned by strong winds broke out during April and May and affected regions bordering 1427 Kazakhstan, such as in the Tyumenskaya, Omskaya, and Amurskaya Oblasts where an 1428 emergency was declared, and Mongolia, such as in Irkutsk and Krasny Yar where at least one 1429 fatality was recorded (Le Monde, 2023). As well as detecting anomalies in fire size and rate of 1430 spread in these areas, the Global Fire Atlas also identified regionally large and fast-moving 1431 wildfires in the Russia-China border zone of Manchuria (Figure 4), however these were not 1432 widely reported on by media outlets or local authorities.

1433

1434 Lao People's Democratic Republic (PDR) experienced a notable fire season in 2023-24, 1435 marked by record-setting BA at national level since 2001 in the MODIS BA data (Figure 2; 1436 Figure S6). The fires were widespread, affecting various provinces from the south to the north, 1437 including Attapu, Khammouan, Louangphabang, Xaignabouli, and Bokeo. In Attapeu, BA in 1438 2023-24 was over twice the average of prior fire seasons since 2001. The fires in 2023-24 1439 were generally small in scale but anomalously numerous, consistent with the widespread use 1440 of slash-and-burn agricultural fires in these regions that have been problematic for regional air 1441 quality in this region during recent years (Meadley, Laotian Times, 2024). The uptick in fire 1442 counts in 2023-24 has been attributed in part to economic factors such as the high price of 1443 cassava and demand for greater corn supplies to supply animal feed, which act as incentives 1444 for farmers to clear forests for additional planting (Bhandari, Radio Free Asia, 2024). On top 1445 of economic factors, an enabling driver was an extreme heatwave that extended eastward 1446 from South Asia to Southeast Asia in April 2023 (Zachariah et al., 2023). The persistent smoke 1447 from these fires worsened air quality significantly in southeast Asia, where efforts to manage 1448 transboundary haze have been challenging during regional droughts, despite a new transboundary agreement being signed in 2023 (Antara News, 2023). Differences in fire 1449 1450 management between Thailand and Lao PDR were evident during the 2023 event, with 1451 authorities intensifying patrols and seeking to control forest fires and agricultural burning for 1452 improved air quality in Thailand (Meadley, Laotian Times, 2024). Conversely, deforestation 1453 remains a critical issue in Lao PDR, with the Laotian government facing challenges in gaining 1454 local community support for the prevention of agricultural expansion and logging.





1456 Earth Observations data showed high-ranking BA anomalies and fires with large size and rate 1457 of growth during 2023-24 in several regions of Pakistan, Iran, Iraq, and parts of the Levant 1458 region (Figure 4), consistent with reports of extreme drought-driven wildfires in some of these 1459 regions (Reuters, 2023b). 1460

1461 3.1.3.3 Europe

1462

1463 Overall, fires burned 8,400 km² in Europe from March 2023 to February 2024 according to the European Forest Fire Information System (EFFIS, 2024), of which 64.5% were from July to 1464 1465 September and 18.1% were in March and April. Large fires (>5 km²) amounted to 53.4% of 1466 the total BA, and those particularly large (>100 km², n=5) accounted for 17.7% of the total burned area. More than half (52.6%) of the BA corresponded to transitional woodland, with 1467 1468 forests, shrublands and grasslands, and agriculture respectively amounting to 19.1%, 13.2%, 1469 and 14.4%. At least 44 people died as a direct result of wildfires (Copernicus Climate Change 1470 Service, 2024; Centre for Research on the Epidemiology of Disasters, 2024).

1471

1472 Most countries in the Mediterranean Basin experienced mild to typical fire seasons in general, 1473 with variable timing but affecting mostly non-forest (open) vegetation types (EFFIS, 2024). In 1474 the Balkans, fire activity varied among countries, but was mostly very low by historical patterns 1475 such as in Croatia, however, a major exception was Greece described in more detail below 1476 (Figure 2, Figure 4, Figure 5). The other exceptions were North Macedonia, with a typical 1477 fire year and Bulgaria, the worst year in a decade, with fire activity extending into October in 1478 both countries; and Bosnia and Herzegovina, Serbia and Montenegro, where collectively ~270 1479 km² burned in January-February 2024.

1480 1481 Greece's 2023 fire season was reviewed at length by Xanthopoulos et al. (2024). It was the 1482 second worst on record regarding total area burned (1,727 km²), despite the recent efforts to 1483 strengthen the firefighting mechanism of the country with more aerial resources and new 1484 personnel, after another challenging fire season in 2021. The situation was kept under control 1485 until mid-July, but in the period July 13-27, maximum temperature in many parts of the country 1486 exceeded the average for the 2010-2019 period by as much as 10°C, according to the records 1487 of the National Observatory of Athens. This resulted in multiple fire starts pushing the limits of 1488 firefighting, which relies heavily on the aerial resources. The fires starting 18th of July on the 1489 tourist island of Rhodes grew rapidly on the second day, finally burning 207 km² and stopping 1490 at the sea. About 20,000 tourists had to be evacuated from hotels along the coast. While the 1491 fire on Rhodes was still burning, three forest fires started on 3rd of July started near the city 1492 of Aigio, in North Peloponnese, on the island of Corfu and near the town of Karystos in the 1493 south of Evia island. On July 25th a Canadair CL-215 crashed near the village of Platanistos 1494 while fighting this last fire. Then, on July 26th, the tail of a cold front that passed over Greece, 1495 with the characteristic wind direction change that accompanies it, created further challenges, 1496 as a number of fire starts in central Greece and Thessaly spread fast, burning mostly in light 1497 fuels, challenged firefighters and threatened inhabited areas. One of the fires entered an Air 1498 Force base on the 27th, causing a powerful explosion of ammunition that resulted in damages 1499 to the town of Nea Aghialos 5 km away. By the end of July, the BA across the country had 1500 reached 550 km².

1501

1502 The next wave of multiple challenging fires in Greece began on 19th August. A lightning-1503 caused fire that started before dawn NE of Alexandroupolis in the prefecture of Evros received 1504 limited attention at first and was destined to become the largest fire in recent European history. 1505 The fact that firefighting resources were focussed on evacuation of the villages in the path of 1506 the fire, rather than fire suppression may have contributed to its eventual size. On the 21st of 1507 August, a second fire started to the north of the first one, near the village of Dadia. Fanned by 1508 a strong NE wind, it spread quickly and within a few hours it reached the rear of the first one. 1509 On that day fire behaviour both in terms of spread and intensity was extreme (Athanasiou, 1510 2024). Nineteen immigrants who had illegally entered the country and were trekking through





the forest, were trapped by the flames and were found dead on the 22nd of August. Another
group was saved by the firefighters at the last moment. The authorities emphasised safety
and evacuated the hospital in the outskirts of Alexandroupolis.

1514

1515 On Aug. 22, while the Evros fire was the focus of attention, a fire originating at more than one 1516 point near the village of Phyli, south of mount Parnis in Attica, at the outskirts of Athens started 1517 growing against the strong NE wind. Once more, many settlements were evacuated and 1518 firefighting attention focused on protecting homes, as the fire moved slowly up the mountain slopes finally burning 62 km² in three days. The Evros fire kept growing at various rates for 1519 1520 the next 15 days, finally reaching 938 km² and becoming the largest on record in recent history 1521 in Europe. The simultaneous spread of the Evros fire, the fire in Attica and a number of smaller 1522 fires, is likely to have increased the growth rate statistics (km day⁻¹) for fires in the region 1523 (Figure 4 and Figure 5).

1524

1525 The BA of the Evros fire included 258 km² of deciduous oak forest and 218 km² of oak forest 1526 mixed with other species (Konstantinos Kaoukis, personal communication). The usually most 1527 challenging forest types regarding fire behaviour, contributed less to burned area: 128 km² 1528 forest and 152 km² of evergreen shrubs. The fire was mostly brought into control only when it 1529 reached agricultural areas and barren lands. The final size of the Evros fire may not be solely attributed to adverse meteorological conditions. One aggregating factor may have been the 1530 1531 recent shift in directing firefighting personnel more strongly from suppression towards evacuation, another the emphasis on aerial firefighting resources (Xanthopoulos et al., 2024). 1532 1533 The latter was not effective once the extreme fire behaviour commenced (21st to 23rd of August). Deep forest litter layers further hampered fire suppression in some areas, although 1534 1535 a group of local forest workers working with handtools, were credited by the local forest service 1536 officers with control of a large part of the fire perimeter to the north, saving an estimated 100 1537 ha of forest (Athanasiou, 2024).

1538

1539 Italy was the second most affected country after Greece, with continuous fire activity from July 1540 to October. More than 1,000 km² burned in the country, of which 69% were in Sicily (including 17 fires >10 km²), although the largest fire (31 km²) occurred in the nearby region of Calabria 1541 1542 (Istituto Superiore per la Protezione e la Ricerca Ambientale, 2023). A defining characteristic 1543 of these large fires was the importance (42% overall) of agricultural land in the BA composition. 1544 The outskirts of Palermo and the Madonie Natural Regional Park were impacted by multiple 1545 wildfires in late September, causing one fatality and affecting wildland-urban interfaces, farms, 1546 and tourism.

1547

1548 Fire activity was insignificant in France, except for benign mountain burning (175 km²) in 1549 March-April and then in January-February, mostly in the western Pyrenees. Like in France, 1550 the north of Spain (Asturias-Cantabria) experienced unusual Spring burning activity, 1551 amounting to 423 km² during late March and early April (Educación Forestal, 2023a). In 1552 particular, the Foyedo wildfire (27th March 2023) was the largest on record for Asturias, 1553 burning 101 km² across variable vegetation but with the predominance of conifer plantations, 1554 mostly Pinus pinaster. It was a wind- and spot-driven fire but its soil and overstorey burn 1555 severity were respectively low and mostly moderate, as slower-drying fuels were not available 1556 to burn (Cátedra Cambio Climático de la Universidad de Oviedo, 2023).

1557

1558 Only two other notable large wildfires occurred in 2023 in continental Spain, and again were 1559 unusual in that they happened in spring rather than summer. The Villanueva de Viver wildfire 1560 (23rd March 2023, Castellón and Teruel) burned around 50 km² and was driven by abnormal 1561 seasonally dry conditions, combined with a shift in wind direction. It mostly burned naturally-1562 regenerated continuous pine forest of Pinus halepensis. Canopy fire severity was 1563 heterogeneous, with 39% of the wildfire area being classified as high to very high severity 1564 (Mediterranean Center for Environmental Studies, 2023). The cost of fire suppression was 1565 2M€ and 1,800 people were evacuated (Las Provincias, 2023).





At over 100 km², the Pinofranqueado wildfire (17th May 2023, Cáceres) was the largest fire in the Iberian Peninsula in 2023 (Copernicus Emergency Management Service, 2023b). The BA was 90% forest, mostly pine (*Pinus pinaster*, *Juntaex.es*, 2023). It was a wind-driven fire and the Canadian FWI indexes indicate all fine fuel was available to burn and extreme fire behaviour (FWI>50). The fire significantly impacted the nesting of protected bird species and rainfall shortly after the wildfire caused important runoff, erosion, and disruption of water supply to the local population (Armero, *Hoy*, 2023).

1575 The two other significant fires in Spain happened in the Canary Islands, the Puntagorda 1576 wildfire (14th July 2023, La Palma, 32 km²) and the Arafo-Candelaria wildfire (15th August 1577 2023, Tenerife, 123 km²; Copernicus Emergency Management Service, 2023c). The latter 1578 spread for 9 days and 94% of its area was forest under conservation status, mostly of Canary 1579 pine (Pinus canariensis). While drought was advanced, atmospheric conditions were not 1580 particularly severe. The fire was topographically driven and highly heterogeneous in severity, 1581 but low to moderate severity prevailed (Educación Forestal, 2023b). Nonetheless, 26,000 1582 people were evacuated, 364 farms and 246 buildings (none residential) were affected, smoke 1583 impacts were substantial and damage was estimated at 80.4M€.

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1566

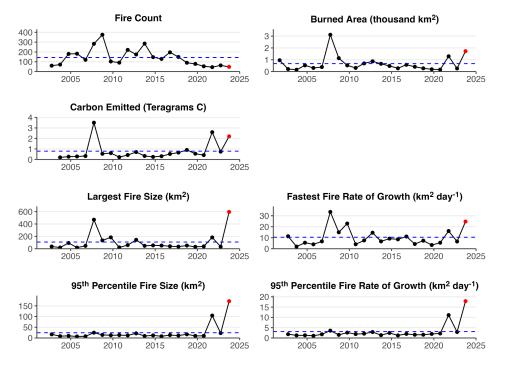
1585 Like in Spain, winter shrubland burning was relevant in continental Portugal (~50 km² in 1586 February) but subsequent significant wildfire activity was restricted to two fires. The Sarzedas 1587 (66 km² ha) and Baiona (75 km²) wildfires started on the 5th of August under extreme fire 1588 weather (FWI>50), and burned mostly (~70%) forest, respectively of pine (P. pinaster) and 1589 eucalypt (Eucalyptus globulus) stands (Direção Nacional de Gestão do Programa de Fogos 1590 Rurais, 2023). Prevailing burn severity was moderate and damage to infrastructure and 1591 emergency restoration amounted to 6.4 M€ cost for the Sarzedas fire and a forest value loss 1592 of 1.4 M€ (Instituto da Conservação da Natureza e das Florestas, 2023). The major run of the 1593 Baiona wildfire was on August 7, corresponding to 73% of the total BA, when it threatened 1594 wildland-urban interfaces and damaged several buildings; one camping park and 20 small 1595 villages were evacuated (Economia Online, 2023). Moderate to high burn severity classes were dominant and costs were estimated at 2.7 M€ (tourism) and 7 M€ (houses), plus 1.4 M€ 1596 1597 in forest values loss and 2.9 M€ for emergency stabilisation (Rádio e Televisão de Portugal, 1598 2023; SIC Notícias, 2023). Finally, on 12th October, and under anomalously extreme fire 1599 weather for the time of the year, the Ponta do Pargo wildfire burned 48 km² in Madeira island, 1600 with an estimated agriculture-related cost of 3 M€ (Rádio e Televisão de Portugal, 2023). 1601

1602 The year was also mild in other European countries where burned areas can be extensive, 1603 namely Romania, Hungary, and Poland, which collectively summed only ~210 km² burned. 1604 EFFIS recorded 2461 km² burned in Ukraine, the largest fire attaining 42 km², but these figures 1605 are far from those registered in recent years. In northern Europe, a notable fire occurred in 1606 Scotland near Cannich, during the spring, the primary fire season in the humid Atlantic climate 1607 of the UK (Belcher et al., 2021). It started on 19th of April and burned ~33 km² of mainly 1608 moorland making it one of the largest fires in the UK in recent history (Sabljak, The Herald, 1609 2023; personal communication Niall MacLennan, Scottish Fire and Rescue Service).

1610







1612

1613 Figure 5: Summary of the 2023-2024 fire season in Greece. Time series of annual fire count, 1614 BA, C emissions, PM2.5 emissions, 95th percentile fire size, fastest daily rate of growth, and 1615 95th percentile fire daily rate of growth. Black dots show annual values prior to the latest fire 1616 season, red dots the values during the latest fire season, and blue dashed lines the average 1617 values across all fire seasons.

1618

1619 *3.1.3.4 North America* 1620

1621 Wildfire across North America in 2023-2024 was characterised by record fire activity across 1622 Canada, lower than normal BA in the most flammable regions of the western US, above 1623 average fire activity across Mexico, and several extreme events that resulted in disastrous 1624 impacts to human communities. North America contributed the majority of the Northern 1625 Hemisphere area burned for the year, and two-thirds of the global C emissions from wildfire 1626 (Friedlingstein et al., 2023; Kolden et al., 2024). Canada burned over 150,000 km² in 2023, 1627 over twice the previous record and over seven times the annual average (Jain et al. 2024). The US burned 10,900 km² in 2023, well below the long-term annual average (National 1628 Interagency Fire Center [NIFC], 2024a). Mexico has a relatively short historical record of 1629 1630 wildfires, but March 2023-February 2024 saw the highest area burned in the last decade (10,000 km²) due to strong ongoing drought conditions (Comisión Nacional Forestal, 2024). 1631 1632

1633 The fire season began earlier than normal in Canada, driven by early snowmelt across much 1634 of the country and persistent drought conditions in the west. Abnormally high temperatures 1635 and lack of rainfall also saw forested regions of eastern Canada, including Quebec, transition 1636 rapidly to drought conditions at the end of May. British Columbia province recorded its first 1637 wildfire evacuation in mid-April, and in late May, over 16,000 people were evacuated from 1638 Halifax, the capital city of Nova Scotia, which saw its largest ever wildfire in 2023 (Jain et al. 1639 2024). In June, two lightning outbreaks in Quebec initiated several hundred new fires in what 1640 would eventually become a record BA year for the province (4,300 km²) (Boulanger et al.,





1641 2024). While the majority of the Quebec fires were in remote regions, the smoke they 1642 generated blanketed several major cities in eastern North America, including New York which 1643 experienced its worst air quality in half a century as the observed daily mean PM_{2.5} 1644 concentration rose to 148.3 µg m⁻³, over 4x the recommended daily limit (Wang et al. 2023). 1645 In total over 50 million people were exposed to high levels of PM2.5 for several days (Yu et 1646 al. 2024). This situation was further exacerbated in New York City by several wildfires in the 1647 nearby New Jersey pine barrens, a fire-prone dry pine forest that sees large fires during 1648 periods of drought.

1649

1650 The year started with low fire activity across the USA. In the high plains of the central US, an 1651 outbreak of large wildfires occurred coincident with dry conditions and strong winds in March-1652 April 2023. One wind-driven wildfire started by power lines in Oklahoma destroyed several 1653 dozen homes (Oklahoma Department of Emergency Management, 2023). Outside of the high 1654 plains, dry conditions also elevated fire activity across the Southern, Eastern, and New 1655 England regions of the USA. Mexico saw slightly above average fire activity during spring, 1656 which is the peak period of the fire season as debris burning and field clearing activities provide 1657 ignitions for predominantly shrubland and grassland wildfires.

1658

1659 As summer arrived in Canada, the western and boreal provinces and territories saw extreme 1660 and widespread fire activity, even as Quebec continued to burn. By the end of the year, record 1661 area had burned in British Columbia (2,300 km²), Alberta (2,700 km²), and the Northwest 1662 Territories (3,500 km²) accompanied evacuations of 232,000 people in numerous rural villages and large cities such as Yellowknife, NT and Kelowna/West Kelowna, BC, where a wildfire 1663 1664 jumped the 2 km-wide Okanagan Lake (Jain et al., 2024; CBC News, 2023). The extreme 1665 behaviour of these fires not only shrouded large swaths of North America in smoke, but also 1666 generated an unprecedented 140 pyrocumulonimbus clouds (Jain et al. 2024). Eight 1667 firefighters were killed during summer 2023 in Canada (Jain et al. 2024), but miraculously no 1668 civilians died directly in the fires. Canada was at the highest National Preparedness Level 5 1669 for an unprecedented 120 continuous days starting on May 11, indicative of the significant 1670 resource sharing required by fire management; in all, over 5500 international personnel from 1671 12 countries and the EU were deployed to Canada during the 2023 fire season (Canadian 1672 Interagency Forest Fire Centre, 2023).

1673

1674 In the US, a relatively low activity fire season became deadly in August owing to unusual 1675 weather conditions facilitating extreme fire behaviour in multiple areas around the country. On 1676 8th August, a pressure gradient-induced katabatic wind event fanned several small wildfires in 1677 Hawai'i, and 101 civilians died as the town of Lahaina on the island of Maui was consumed in 1678 the worst wildfire disaster in the US in a century (Pyne, 2017). Over 2,000 homes were 1679 destroyed and over 10,000 people were displaced as a result. Extreme heat and flash drought 1680 produced fire behaviour that killed five additional civilians in the US states of Washington (2 1681 fatalities), Louisiana (2 fatalities) and California (C. Kolden, unpublished data). These extreme 1682 events stood in contrast to overall low fire activity and were notable for where they occurred. 1683 Washington does not typically see many extreme, wind-driven wildfires, and Louisiana is one 1684 of the wettest states in the US. By the end of August, the US BA was only 40 percent of normal 1685 and the lowest since at least 2000 (NIFC, 2023; National Oceanic and Atmospheric 1686 Administration [NOAA], 2023).

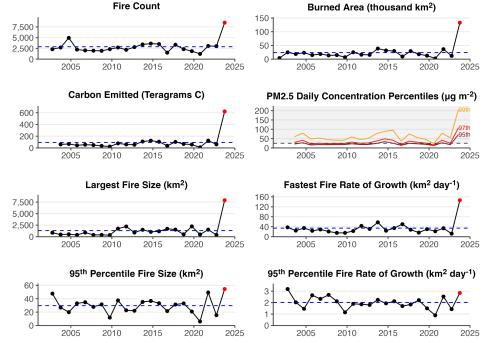
1687

As North America transitioned to fall and then winter, Canada continued to burn nearly a month longer than normal, with the last large fires not controlled until late October. On 22nd September, remarkably late in the fire season, Canada saw its largest ever one-day total for BA at approximately 4,400 km² (Jain et al., 2024). While many of the Canadian fires were fully extinguished by winter, others simply smouldered into the deep peat layers, aided by a warmer-than-normal winter with a reduced snowpack. At the end of February 2024, spaceborne thermal sensors detected several dozen fires in northern Alberta and British





1695 Columbia that were overwintering fires, likely sustained by peat smouldering (Shingler, CBC News, 2024; Scholten et al., 2021).
1697



1698

1699 Figure 6: Summary of the 2023-2024 fire season in Canada. Time series of annual fire count, 1700 BA, C emissions, PM2.5 daily concentration percentiles (95th, 97th, and 99th), 95th percentile fire size, fastest daily rate of growth, and 95th percentile fire daily rate of growth. Black dots 1701 1702 show annual values prior to the latest fire season, red dots the values during the latest fire 1703 season, and blue dashed lines the average values across all fire seasons. The PM2.5 daily 1704 concentration percentiles are based on area-weighted daily mean PM2.5 concentrations 1705 across Canada (see Methods). The 95th, 97th, and 99th percentiles correspond to PM2.5 1706 concentrations that are breached on 18, 10, and 3 days of the year, respectively. Grey shading 1707 represents concentrations that exceed the reference levels for 24-hour mean PM2.5 1708 concentration in Canada, statistically derived on the basis of several key epidemiological 1709 studies (Canadian Environmental Protection Act Federal-Provincial Working Group on Air 1710 Quality, 1998).

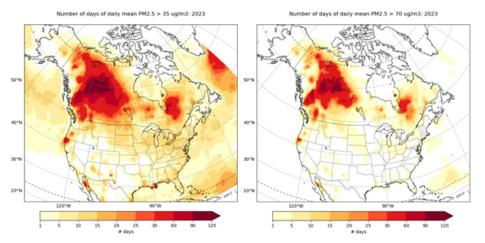
- 1711
- 1712

1713 US fire agencies recorded just over 10,900 km² burned in 2023, just over half of the 20-year 1714 mean of 29,000 km² (NIFC, 2024a). Notably, over half of the BA was associated with higher 1715 fire activity in the central plains grasslands and the southeastern US, while below normal fire 1716 activity characterised California and the western US throughout 2023 as the region exited a multi-year drought. However, the number of fires recorded was only slightly smaller than 1717 average. This quiet pattern broke in February 2024, however, when drought conditions from 1718 1719 the high plains region of the US down into north central Mexico coupled with strong winds to 1720 produce massive, fast moving wildfires across multiple states on both sides of the US-Mexico 1721 border. The US state of Texas recorded its largest ever single fire at over 4,000 km² 1722 (Smokehouse Creek fire) in late February and early March that destroyed 130 homes across 1723 the high plains region of the central US (NIFC, 2024b). Two civilians were killed by the flames





- in the relatively rural area dominated by ranching, over 10,000 head of cattle died, and damages are estimated to be at least \$4.6 million (NOAA, 2024).
- 1726
- 1727



1728

1729Figure 7: Impact of Canadian wildfires visible in air quality metrics for North America. Panels1730show the number of days in 2023 with mean PM2.5 concentration over a threshold of (left1731panel) 35 μ g m⁻³ and (right panel) 70 μ g m⁻³. Both the National Ambient Air Quality Standards1732(NAAQS) in the USA and the Canadian Ambient Air Quality Standards (CAAQS) have1733exposure targets of 35 μ g/m³ on average within a single day.

1734

1735 The impact of North American fires on air quality is significant with half of the PM2.5 in America 1736 suggested to originate from fires (O'Dell et al. 2019). An exceptional fire season, such as seen 1737 in Canada in 2023, therefore poses an elevated level of health risk. Canada's wildfires produced levels of PM2.5 across the Country that were well in excess of the last 20 years 1738 1739 (Figure 6). Additionally, long-range transport of pollution from Canada affected the Pacific 1740 Northwest, northern Midwest, and many eastern states (Figure 7). According to the National 1741 Ambient Air Quality Standards (NAAQS) in the USA, the threshold for PM2.5 exposure is not 1742 to exceed 35 µg/m³ on average within a single day. CAMS analysis suggests that people living 1743 within over half of US states experienced up to 2 weeks of exposure at or above this level. In 1744 Canada, the safe limit for PM2.5 exposure, as defined by the Canadian Ambient Air Quality 1745 Standards (CAAQS), is also 35 µg/m³ over a 24-hour period. However, the situation was worse 1746 in Canada due to closer proximity to fires, with many territories along the border experiencing 1747 up to a month of degraded air quality that exceeded national recommended exposure limits, 1748 British Columbia possibly facing twice the number of days at up to 2 months, and the Northern 1749 Territories potentially with 3 to 4 months of exposure.

1750 The scale of the impact becomes particularly evident when comparing the number of days at 1751 double the exceedance level (70 µg/m³) due to short-range versus long-range transport of 1752 pollutants. In Canada, where the pollution sources were more localised, the number of days 1753 above this higher level remains substantial, ranging from a week along the border to months 1754 still at the fire epicentres. In contrast, the USA, affected primarily by long-range transport from 1755 Canada, experienced approximately half the number of such high-pollution days. It is also 1756 important to consider the context of interannual variability in fire occurrence. Last year the 1757 USA experienced its lowest number of fires in 2 decades (Figure 4), so most of the pollution 1758 impacts came from Canada. Consequently, if the USA had also experienced a high fire year, 1759 the air quality would have worsened in many states.





1760 3.1.3.5 Oceania

As is commonly the case, there was a marked latitudinal difference in wildfire patterns in Oceania in 2023-24. Fire activity was above average in the savannahs, grasslands and shrublands of tropical, subtropical and arid northern Australia. In contrast, fire activity in the southern states of Australia was generally below average, and well below the levels seen during the high impacting 'Black Summer' fires of 2019-20. In New Zealand and the Pacific Islands, fire activity was relatively low compared to the preceding two decades. Given the vast scale of savannah fires, 2023-24 ranked among the top five years in BA for

Australia as a whole since 2001 (Fisher, 2024; **Figure 2**). Fire in tropical and arid areas is tightly linked to rainfall in the preceding season (Alvarado et al., 2020). The above average fire seasons in the Northern Territory and northern Western Australia were driven to a large extent by elevated fuel growth associated with the La Nina conditions of the previous three years. These fires represented the vast majority of areas burned across the country in 2023-24 (**Figure S7**).

1776

1761

1777 In the monsoonal north, savannah fires follow a strong seasonal pattern, with regular summer
1778 rain predictably followed by fire in the dry winter and spring months. In arid regions further
1779 south, fire remains tightly coupled to rain, but the seasonality is less pronounced. Anomalously
1780 large fires began as early as May and June in Western Australia and the Northern Territory
1781 respectively, continuing to as late as January.

1782

1783 The year was also marked by a series of early season, high impact fires in populated areas of 1784 southwestern Western Australia, southeast Queensland, NSW, Victoria and Tasmania. Hot, 1785 dry, windy conditions, and extended dry periods, are a major driver of forest, woodland and 1786 shrubland fires of the subtropical and temperate south of Australia (Collins et al., 2022). In 1787 addition to 2023 being the eighth-warmest year on record, the three months from August to 1788 October were the driest in over 100 years of records (Bureau of Meteorology).

From October to January a string of fire events led to loss of life, property and a range of other
human and environmental impacts throughout the country's southeast and southwest. In some
cases, significant fire activity was observed in areas impacted by the 2019-20 fire season.
Despite these impacts, average rain in southern and eastern parts of Australia tempered fire
activity for the austral summer. In Queensland and NSW, large fires in remote areas pushed
the total BA in line with the long term mean, but this figure was well below average in Victoria,
the Australian Capital Territory and Tasmania.

1797

1798 In the southwest of Australia, a volunteer firefighter was killed while responding to a fire near 1799 Esperance. The Kings Park fire in October occurred in a popular tourist area containing 1800 vulnerable flora and threatened Perth Children's Hospital. Perth was again affected by fires in 1801 December, with several injured and five homes lost. A further two homes were lost in the 1802 region in mid-January from fires that burned 60 km². A similar sized fire burned through rugged 1803 terrain in the Gammon Ranges 600 km north of Adelaide, threatening highly significant cultural 1804 and ecological values.

1805

1806 A large number of significant fires affected the eastern States of Australia in October. The 1807 Tara and Mount Isa fires in Queensland burned well over 400 km² combined, destroying 65 1808 homes and claiming the lives of two people. International and interstate support were deployed 1809 from New Zealand and Victoria to respond to the fires. Further south in New South Wales, 1810 significant fires included the Coolagolite Rd fire in Bega (over 70 km², 2 houses destroyed), the Willi Willi Rd fire in Kempsey (over 290 km², 8 houses destroyed, one person killed) 1811 1812 and.large fires around Tenterfield (approximately 300 km², four homes destroyed). In 1813 November the Hudson Fire burned 228 km², destroyed 4 properties and led to the death of a 1814 volunteer fire fighter, who was killed by a falling tree while fighting the fire. In December the





1815 Duck Creek Pilliga Forest fire burned 1,385 km² and initiated 3 documented fire-generated 1816 thunderstorms, with smoke impacts extending 500 km away and reaching Sydney.

1817

1818 In neighbouring Victoria the fire season was bookended by high impact events in October and 1819 then in February and March. Fires in Gippsland during October totalled 120 km² and exhibited 1820 some overnight fire runs that were regarded as unusual (Mills et al., 2022). An extended dry 1821 period saw fires impacting towns in central and western parts of the state in late February and 1822 early March. Over 40 homes were lost and five firefighters were injured fighting two fires that 1823 originated in the Grampians National Park and burned 60 km². Interactions with the 1824 atmosphere and topography were suggested to explain extreme behaviour that was reported. 1825 This fire was followed by another near Ballarat affecting grass, forest and a pine plantation. 1826 Despite several extreme fire weather days and evacuation advice, a significant suppression 1827 effort aided by interstate deployments minimised impacts. The fire burned over 200 km².

1828

1829 In the island state of Tasmania the fire season began early with the Coles Bay Bushfire burning 1830 27 km² of both private land and national park in September and then fires on Flinders Island 1831 in October. Other impactful fires that occurred during the season include the Dolphin Sands 1832 fire on the east coast of Tasmania that destroyed two homes and burned 2.5 km² and the 1833 multiple fires in the Brady lake area (Tasmania's central highlands) in February that destroyed 1834 two homes and burned up to 100 km² and a fire in the Waterhouse Conservation areas that 1835 required campers to evacuate.

1836

New Zealand experienced a normal fire season after three well below average seasons prior under La Nina conditions. The fire season began early with a relatively large fire in September on the western side of Lake Pukaki in the central South Island in wilding pines. This fire totalled 29 km² and this was the third major wildfire event in recent years in this area at an earlier than normal stage of the fire season, following the 2020 Pukaki (Aug.) and Ohau (Oct.) fires. New Zealand then experienced a spate of fires around Canterbury, on the South Island between late Jan. and mid Feb. 2024 with several houses burned and farmlands affected.

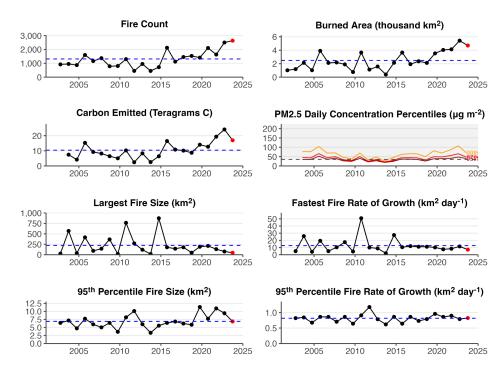
1845 3.1.3.6 South America

1846 1847 The 2023-24 fire season in South America was characterised by a moderately below average 1848 fire activity but with positive wildfire anomalies in specific regions, which were exacerbated by 1849 extended periods of drought and heatwave across the continent (Clarke et al., 2024; Figure 1850 2, Figure 3, Figure 4). In the Brazilian State of Amazonas, which features the largest extent 1851 of preserved old growth forests in Amazonia, June and October 2023 saw the highest fire 1852 counts since records began in 1998 (National Institute for Space Research, 2024; see also 1853 Figure 8). This continues a recent trend towards record-setting months for fire in Amazonas 1854 state, with new maxima being set in 7 months of the year since 2019 (National Institute for 1855 Space Research, 2024). Recent changes in deforestation and land use patterns are 1856 contributing to elevated fire ignitions in the state, amplified by climatic extremes in 2023 by a 1857 historic drought and heatwave driven by El Niño (Espinoza et al., 2024, Clarke et al., 2024). 1858 Due to emissions of wildfire smoke, many areas of Amazonas experienced poor air quality 1859 from September to December 2023, including in the state capital, Manaus, where over 2 1860 million people were exposed to the second-worst air quality in the world in October (Ministério 1861 Público Federal, 2023). The event was so severe that, in November 2023, the Federal Public 1862 Ministry opened a Civil Action case against the State of Amazonas, demanding evidence that 1863 the State was investing in fire prevention and combat in line with the Plan for the Prevention 1864 and Control of Deforestation and Fires (Estado do Amazonas, 2020). This procedure 1865 evaluates whether Amazonas authorities are accountable for environmental damage causing 1866 severe air pollution, reflecting the Public Ministry's growing involvement at both federal and 1867 state levels in monitoring environmental degradation and seeking to make authority figures 1868 accountable (Ministério Público Federal, 2023).





1870



1871

Figure 8: Summary of the 2023-2024 fire season in the Brazilian state of Amazonas. Time series of annual fire count, BA, C emissions, PM2.5 emissions, 95th percentile fire size, fastest daily rate of growth, and 95th percentile fire daily rate of growth. Black dots show annual values prior to the latest fire season, red dots the values during the latest fire season, and blue dashed lines the average values across all fire seasons.

1877 1878

1879 National fire monitoring systems in Brazil indicate that some areas of Amazonia experienced 1880 anomalies in BA at the sub-state level. For example, BA in the municipality of Santarém in the 1881 State of Pará rose from an average of 70 km² in 2019 to 2022 to over 1,000 km² in 2023, and has already exceeded 250 km² in 2024 (MapBiomas Brasil, 2024). Similarly, in neighbouring 1882 1883 Belterra municipality, BA extent was more than 3 times greater during the year 2023 than in 1884 2019-2022 (MapBiomas Brasil, 2024). In Floresta Nacional do Tapajós, one of the most 1885 studied forest sites in the Amazon which spans Satarém and Belterra, forest fires accounted for more than 60% of the burned areas (MapBiomas Brasil, 2024). 4 thousand people live in 1886 1887 24 communities of traditional and Indigenous populations in the region and depend on 1888 protected forest resources for their cultural heritage, food security, economy and livelihood in 1889 Floresta Nacional do Tapajós and Reserva Extrativista Tapajós-Arapiuns (Instituto Chico 1890 Mendes de Conservação da Biodiversidade, 2019). Fires in 2023 compounded the challenges 1891 faced by these communities, who were already isolated by low river levels resulting from the 1892 drought that severely reduced their mobility and fishing, impacting on food security and 1893 enhancing socio-economic vulnerabilities.

1894

1895 In Chile, the 2023-24 fire season was marked by a significant escalation in both the number 1896 and size of wildfires, especially in the central and southern regions (**Figure 4**). Chile 1897 experienced its second-highest BA in the past 20 years (>4,000 km²; Jones et al., 2024). The 1898 peaks in BA at national scale were accompanied by peaks in the 95th percentile of fire size





1899 and daily rate of growth in highly-populated regions such as Valparaíso (Figure 2, Figure 4), 1900 indicating unusually large and fast-moving fires. These fires drew international attention due 1901 to their deadly impacts on society. In February 2024, severe wildfires struck the Valparaíso 1902 region in Chile, particularly affecting Viña del Mar and other surrounding areas (NASA Earth 1903 Observatory, 2024). These fires, fueled by a prolonged heatwave, resulted in the deaths of at 1904 least 131 people and destroyed thousands of homes, leaving at least 1,600 people homeless 1905 (UN Resident Coordinator in Chile, 2024; Al Jazeera, 2024; El Disconcierto, 2024). The 1906 emergency impacted the Lago Peñuelas National Reserve where more than 60 km² of forest 1907 were affected (Oberholtz, Fox Weather, 2024). The National System for Disaster Prevention, 1908 Mitigation and Attention (SENAPRED) issued a red alert and ordered the evacuation of over 1909 18 nearby towns (Oberholtz, Fox Weather, 2024). The February 2024 wildfires in Valparaíso 1910 followed other major disruptive wildfires in February 2023, which affected nearby regions of 1911 central Chile including Maule, Nüble, Bio bío, La Araucanía and Los Rios.

1912

1913 Several countries with land in the west of Amazonia experienced anomalies in BA and fire 1914 behaviour during the 2023-24 fire season, which correlated with specific regional climatic and 1915 environmental conditions. Bolivia, Peru, Ecuador and Colombia did not experience particularly 1916 extreme levels of BA at national scale, possibly influenced by above-average rainfall over three consecutive years of La Niña between 2020 and 2022 (iMMAP Inc., 2023). However, 1917 1918 Peru's Loreto region, which neighbours the Brazilian state of Amazonas, faced its highest BA 1919 on record in the 2023-2024 fire season signalling the wider impacts of drought conditions in 1920 western Amazonia (Figure 2). The timing of peak anomalies in BA also coincided with those 1921 in Amazonas, around September-October 2023 (Figure S2). The northern Bolivian departments of La Paz and Beni experienced similarly-timed anomalies in BA. Remote parts 1922 1923 of the Colombian Amazon also saw a significant uptick in BA since November 2023 which 1924 peaked in January 2024 (Mongabay, 2024). As a result of months of record-high temperatures 1925 and drought conditions since the beginning of El Niño, the region recorded higher C emissions, 1926 reflecting the severity of the burning at the end of the studied fire season. While the direct 1927 impacts on society throughout these regions was not as pointed as in the case of fires in Chile, 1928 the events are likely to have contributed to reductions in regional air quality and also impacted 1929 forest ecosystems and raised C emissions from fires in South America. At the end of January there was a wildfire in the mountains surrounding Bogota that affected the air quality that 1930 1931 affected thousands of citizens of the capital of Colombia (France-Presse, VOA News, 2024). 1932

- 1933 In 2023-2024, Venezuela experienced its highest level of fire activity on record, particularly in 1934 January and February 2024 (ALER, 2024; Tiempo, 2024; Figure 4, Figure S2, Figure S8), 1935 notably affecting the states of Anzoátegui, Cojedes, Guárico, and Monagas, areas which 1936 dominant land cover primarily consists of savannas and extensive grasslands. This surge in 1937 fires during the dry season was intensified by unusually warm and dry conditions in the 1938 preceding months. These conditions, likely a result of global warming and changes in 1939 circulation and rainfall patterns associated with El Niño, making the landscapes more 1940 vulnerable to fires.
- 1941

1942 3.1.4 Context of Recent Extremes

1943

1944 The anomalies of 2023-24 occur against a backdrop of trends in BA this century that point 1945 towards shifts in fire regime. **Figure 9** shows significant trends in BA and forest BA across the 1946 fire seasons in the period March 2001-February 2024 derived from MODIS BA data (see 1947 Methods). While many world regions are experiencing declines in total BA, increases in forest 1948 BA are far more prevalent than declines at the scale of continental biomes, countries, and 1949 administrative regions.

1950

Northern hemisphere extratropical biomes in North America and Asia show a clear signal
towards increased forest BA since 2001, which are also visible on national and provincial
scales in Canada, the US and Russia and on provincial scales in various states of western





1954 and eastern Canada, the western US, and Siberia. These trends occasionally propagate to 1955 trends in total BA, for example in western and northern Canada and in the Sakha Republic 1956 (eastern Siberia). The large 2023-24 anomalies in BA in Canada align with the doubling of 1957 forest BA in Canada across fire seasons since 2001 (a significant trend, p < 0.05) and a 23% 1958 increase in total BA in Canada (marginally significant at p < 0.1). Three Canadian provinces 1959 showed significant increases in both total and forest BA this century: British Columbia (+35-1960 42%); Northwest Territories (+55-68%), and; Yukon (+60-135%). No Canadian provinces 1961 experienced a significant decline in forest BA or total BA. More widely, there was a 58% 1962 increase in forest BA in the North American boreal forest biome since 2001, and a 134% 1963 increase across the pan-boreal forest biome of North America and Eurasia. The succession 1964 of events affecting boreal forests in Canada in 2023, Siberia in 2020, and both North America 1965 and Asia during 2021 appear to be part of a continued trend towards rising fire extent in the 1966 high latitudes this century.

1967

Elsewhere in the southern hemisphere extratropics, significant rises in forest BA are seen in
Chile since 2001 (+87%), including in the central regions of Araucanía, Bio bío, Maule, Ñuble,
and Valparaíso ranging from 35 to 109%. Extreme fires in Valparaíso during 2023-24 and in
Araucanía, Biobío and Ñuble in the 2022-23 fire season Maule, Nüble, Bio bío, Araucanía
(Section 3.1.3.6) are also consistent with a longer-term rise BA in central Chile (Figure 9).
Increases in BA are not generally significant in fire-prone parts of the southern hemisphere
extratropics, such as Africa or Australia.

1975

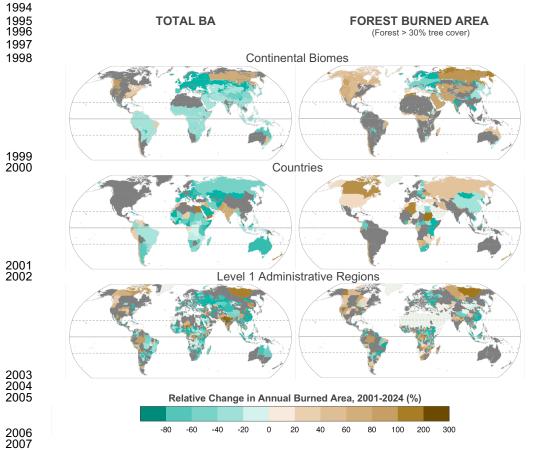
1976 In the tropics, trends in total and forest BA show a variety of patterns. While total BA has 1977 reduced across much of the savannah-occupied regions of South America. Africa and northern 1978 Australia, trends in forest BA (>30% tree cover) are far more varied (Figure 9). Hence, fires 1979 in woody tropical vegetation show a less consistent global trend. In addition, exceptions to the 1980 general decline in total BA across the tropics are seen in the Brazilian state of Amazonas, the 1981 Congo basin, and across much of India (Figure 9). The trend in Amazonas, among the most 1982 pristine parts of Amazonia, contrasts with other states of Brazil such as Mato Grosso and 1983 Pará, where declines in deforestation rates and deforestation-related fires have fallen since their peak during the early 2000s (Silva Junior, 2020). The anomalous fire activity and C 1984 1985 emissions in Amazonas state during the 2023-24 fire season (but not other states of Brazil) 1986 thus appear to be consistent with the emerging pattern of increased fire in the region. 1987 Meanwhile, the 2023-24 anomalies in BA and other fire properties in the Bolivian, Peruvian, 1988 Colombian and Venezuelan parts of Amazonia (Section 3.1.3.6) typically occurred against a 1989 backdrop of reduced BA or no significant trend in recent decades (Figure 9).

1990 1991

1991







2008 Figure 9: Relative changes (%) in (left panels) total annual BA and (right panels) forest BA 2009 across March-February fire seasons during 2001-2024 for three regional layers: (top panels) 2010 continental biomes; (middle panels) countries, and; (bottom panels) level 1 administrative regions (e.g. states or provinces). Forest BA considers only areas with tree cover over 30% at 2011 2012 the native (500 m) resolution of the BA observations. Relative changes are calculated as the 2013 trend in BA across fire seasons March 2001-February 2022 through March 2023-February 2014 2024 multiplied by the number of years in the time series and divided by the mean annual BA 2015 during the period. Trends in BA are derived using the Theil-Sen slope estimator. Only 2016 significant trends in BA are shown (dark grey fill signifies no significant trend).

2018 **3.2 Focal Events of this Report** 2019

2020 **3.2.1.1 Canada** 2021

In this year's report, the extreme wildfire season in Canada is selected as one of our focal
 events. It emerges as a major event of global relevance for the following reasons:

 Record-Breaking Burned Area: The North American boreal forests, particularly in Canada, experienced an unprecedented fire season. The BA reached six times the average since 2001.

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2028	• High C Emissions: Fire C emissions in Canada were over nine times the average			
2029	since 2003, contributing significantly to global C emission totals for the year.			
2030	• Early and Persistent Fires: Positive BA anomalies were visible from April 2023			
2031	(Figure S9) and persisted through to October, with some regions experiencing fires			
2032	until January 2024. The fire season lasted nearly a month longer than normal, with the			
2033	largest one-day BA total ever recorded in Canada occurring on 22nd September 2023.			
2034	• Regional Anomalies: Peak fire anomalies were observed in Eastern Canada in June			
2035	2023 and later in Western Canada (August-September), indicating widespread and			
2036	prolonged fire activity across the country.			
2037	• Record Fire Size and Spread: New records for individual fire size and rate of spread			
2038	were set, with many provinces experiencing high-ranking anomalies in fire count and			
2039	daily growth rates.			
2040	• Extensive Impact Across Provinces: The highest BA or fire C emissions on record			
2041	were observed in Northwest Territories, British Columbia, Alberta, and Quebec, with			
2042	other provinces like Yukon, New Brunswick, and Ontario also experiencing significant			
2043	fire activity.			
2044	Air Quality Impact: Smoke from these fires led to severe air quality issues, affecting			
2045	major cities in North America, including New York, which experienced its worst air			
2046	quality in half a century.			
2047	• Firefighting Challenges: Canada was at its highest National Preparedness Level for			
2048	an unprecedented 120 continuous days, indicating the significant resource sharing and			
2049	international assistance required to manage the fires.			
2050	• Human and Economic Toll: Over 232,000 people were evacuated across various			
2051	regions, and despite the extreme fire activity, no civilian deaths were directly attributed			
2052	to the fires, showcasing the effective, albeit strained, emergency response efforts.			
2053	• Global Significance: The extreme fire season in Canada was a major contributor to			
2054	the overall above-average global fire C emissions in 2023-24, highlighting its global			
2055	environmental impact.			
2056	To appear the source of appeific regional PA anomalies, four anomalous PA regions/month			
2057	To assess the causes of specific regional BA anomalies, four anomalous BA regions/month			
2058 2059	combinations were chosen across Canada: Western Taiga Shield and Taiga Boreal Plains for			
2059	May and June (includes Alberta and British Columbia Boreal plains, and the Mountain			
2060	Cordillera); and Eastern Taiga Shield in Quebec for June and July. Figure 10 maps the			
2001	magnitude of anomalies in these regions and months. Though note the size and long period			

2065 3.2.1.2 Greece

BA over Canada in 2023 (Figure S9).

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In this year's report, the extreme wildfire season in Greece is selected as one of our focal events. It emerges as a major event of global relevance for the following reasons:

this protracted event means that even these regions/month to not capture all the anomalous

- Second-Highest BA on Record: Greece experienced its second worst fire season in terms of total area burned, with 1,727 km² affected, despite recent efforts to strengthen firefighting mechanisms. The 2023 fire season was notably more severe than typical years, with the total BA significantly exceeding the country's historical averages and recent challenging fire seasons.
- Multiple Large Fires: From mid-July to late August, Greece faced numerous large fires that overwhelmed firefighting capabilities. Key fires included those on the island of Rhodes, which burned 207 km², and the massive Evros fire, which reached 938 km².
- Evros Fire Disaster: The Evros fire became the largest on record in recent European history, significantly impacting both forested and agricultural areas. It also led to the tragic deaths of 19 immigrants who were trapped by the flames.
- Urban and Infrastructural Impact: Fires near populated areas necessitated large scale evacuations, including 20,000 tourists on Rhodes and multiple settlements



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around Mount Parnis in Attica. The Evros fire also caused a powerful explosion at an Air Force base, resulting in damage to the town of Nea Aghialos.

- Significant Evacuations: Numerous evacuations took place, highlighting the severity
 of the situation. These included evacuations in Alexandroupolis and its surrounding
 villages due to the Evros fire.
- Economic and Environmental Damage: The fires caused extensive damage to properties, infrastructure, and natural reserves, with significant impacts on biodiversity and local economies.
- **Firefighting Challenges:** The simultaneous spread of multiple fires stretched firefighting resources to their limits, with a notable focus on evacuations rather than fire suppression in some instances.

Abnormally high burned areas were reported around Alexandroupolis in August and extended further west across the administrative region of Macedonia and Thrace. Anomalies were also present in central Greece and around Athens in July and August (**Figure S10**). **Figure 10** maps the magnitude of the anomalies for August.

2100 3.2.1.3 Western Amazonia

2101 Our final focal event of 2023-24 is a box drawn in western Amazonia with bounding 2102 coordinates 2.25° N, -56.00° E, -9.75° S, -77.75° W. It includes Amazonas (Brazil), Loreto 2103 (Peru), and La Paz and Beni (Bolivia) where peak fire anomalies occurred simultaneously and 2104 coincided with a historic drought and heatwave. It emerges as a major event of global 2105 relevance for the following reasons:

- Record-Setting Fire Activity: The 2023 fire season in Western Amazonia saw unprecedented fire counts, with new records set across Amazonas state in Brazil, Loreto department in Peru, and La Paz and Beni departments in Bolivia.
- Severe Air Quality Degradation: Smoke from widespread fires led to significantly degraded air quality across the region, impacting millions of people and posing serious public health risks.
- Broad Socio-Economic and Health Impacts: The fires caused extensive socioeconomic disruptions, including health issues from poor air quality, legal actions for inadequate fire prevention, and impacts on livelihoods, particularly for Indigenous and traditional communities.
- Widespread Environmental Degradation: The fires contributed to significant C
 emissions and environmental degradation, affecting forest ecosystems and increasing
 the region's vulnerability to future climatic extremes. Western Amazonia has global
 significance due to its critical role in C storage and biodiversity and relatively low levels
 of disturbance.
- Impact on Indigenous and Traditional Communities: Fires in key areas disrupted
 the lives of Indigenous and traditional populations, exacerbating their vulnerabilities
 due to isolation and reduced access to resources.

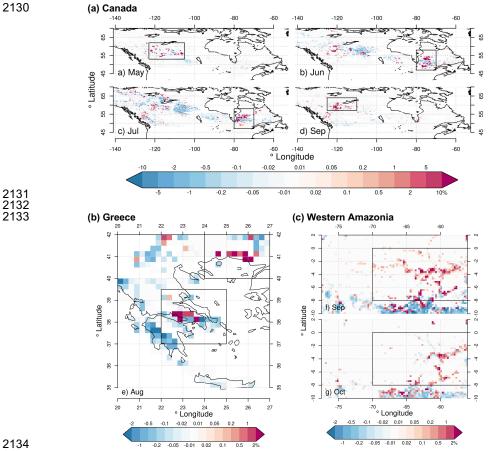
Abnormally high burned areas were reported in Western Amazonia during September and October. **Figure 10** maps the magnitude of these anomalies. Some anomalous BA starting in August and extending to November (**Figure S11**).

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2136 Figure 10: Spatially explicit anomalies in BA fraction (%) during key months for focal events in (a) Canada, (b) Greece and (c) Western Amazonia. Plotted data are the absolute change 2137 2138 from the climatological mean BA fraction for the month (%), based on MODIS MCD64A1 2139 aggregated to 0.25°. Red indicates higher BA in that month of 2023 vs the 2002-2022 climatological average for that month. Boxes indicate focus events for our analyses in this 2140 2141 report. The top panels show anomalies in Canada for various months, the lower-left panel 2142 shows anomalies in Greece for August, and the lower-right panels show anomalies in Western 2143 Amazonia during September and August.

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Diagnosing Drivers and Assessing Predictability 2146 3.3

2148 Highlights 3.3.1 2149

2150 In Canada, fire weather conditions were the primary drivers of the extent of most fire 2151 activity. However, instances of intense burning events, particularly in mid-July, early August, 2152 and late September were missed by the driver attribution analysis, resulting in 'others'. This, 2153 highlighted potential inadequacies in predicting certain ignition sources or accurately 2154 representing fire propagation across vast landscapes in current forecasting systems.





The exceptional nature of the event in Alexandroupolis, Greece, could not have been predicted using fire weather forecasts. While there were discernible indications in the Fire Weather Index (FWI) records that the days around the event were more extreme than most days in the fire season, similar conditions were also observed in the last 10 days of July without resulting in the same catastrophic impacts. This underscores the importance of expanding early warning beyond fire weather, considering fuel availability and ignition variability, to enhance reliability in fire risk assessment.

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2164The extreme fire season in South America was driven by prolonged drought conditions2165linked to the strong El Niño. Many fires developed, triggered by lightning ignitions early in2166the season amidst high FWI values. Other than weather conditions that acted as persistent2167controls for fire activity, several intense active fire periods were not predicted in late August2168and throughout September likely due to the result of unrepresented ignition sources.

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Extreme burned areas were driven by anomalies in multiple controls, with weather, fuel abundance, and moisture being the most relevant factors. The synchrony of all three factors resulted in the most severe BA anomalies in the three highlighted events. This underscores that no single factor can explain the most severe fires. Instead, multiple contributing controls must often coincide for the most extreme events to occur.

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Fuel load is a critical modulator for burned extent. Higher and or dries fuel loads combined with high fire weather conditions determined the unprecedented extent of burning in Canada and Western Amazonia. Conversely, the boundaries of extreme fires in Canada and Greece often corresponded to areas with lower fuel loads, demonstrating that discontinuity in fuel availability influenced fire behaviour and may have created firebreaks.

2182 3.3.2 Predictability of Focal Events using Fire Weather Forecasts

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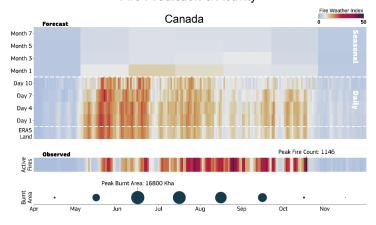
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2184 **3.3.2.1 Canada** 2185

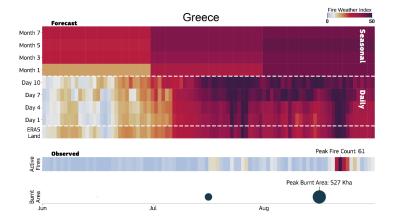
2186 The early establishment of fire weather conditions as well as the late cessation of the fire 2187 season are well captured in the FWI reanalysis in Canada (Figure 11). The FWI also captures 2188 the intermittent pattern of fire danger and its correlation to actual fire activity. However, at the 2189 seasonal time scale, the signal is weakened, and there were no prior indications that the 2190 upcoming season would have been as extreme as it was with respect to fire activity (Figure 2191 11). For most part of 2023 Canada was in drought. The analysis of the other subindices of the 2192 FWI system, notably the drought code shows more persistent patterns in fire weather with 2193 longer predictability lead times (not-shown). The weather-limited nature of the Canadian fire 2194 season means that the FWI modelling framework serves as an essential indicator of 2195 anomalous conditions, acting as a prerequisite for the intensity and spread of fires. It provides 2196 valuable insights into the sequence and extent of extreme fire weather days during such 2197 events.

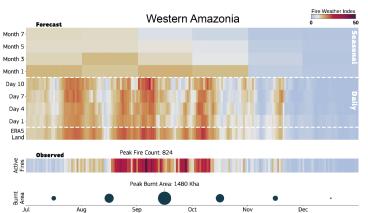






Fire Prediction & Activity





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Figure 11: Chicklet plots displaying seamless FWI predictions over time from various 2200 forecasting systems of the ECMWF (see Methods). The x-axis corresponds to specific dates 2201 throughout the year, while the y-axis denotes either observations or the time leading up to the





date when a forecast was generated. The vertical colour coherence allows for quick
 identification of the time windows of predictability associated to the observed fire activity both
 provided in terms of number of detected active in a day fires and total burned areas in a month
 (circles).

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2208 **3.3.2.2 Greece** 2209

2210 The establishment of fire-prone conditions in the Mediterranean, particularly in Greece, is part 2211 of the region's seasonal weather cycle (Figure 11). In 2023, this pattern persisted, and 2212 extreme landscape flammability could be forecasted well in advance. In arid and semi-arid 2213 regions fire occurrence is driven not solely by weather but also by fuel availability and its 2214 intermittent short-term drying. In these regions the FWI often reaches extreme levels for much 2215 of the summer. However, fires do not always occur even when the FWI is extreme, as ignition 2216 and early suppression play a crucial role. The anomalous event in Alexandroupolis, highlights 2217 the limitations of relying solely on fire weather indices in these areas. There were no 2218 discernible indications in the FWI records that the particular day was more extreme than the 2219 day before or the one afterward, emphasising the need for a more holistic approach to fire risk 2220 assessment in regions where fuel is a limiting factor or live fuel moisture plays a crucial role 2221 in the extent of the burnings (Di Giuseppe, 2023).

2223 3.3.2.3 Western Amazonia

The correlation between FWI and fire activity in the South America region at the shorter lead times is generally poor, primarily due to the strong dependency on either lightning or human ignitions (Kelley et al., 2021). In 2023, this pattern persisted (**Figure 11**), with the onset of fire weather following the seasonal pattern well ahead of the time where fires were triggered. Seasonal predictions indicated high fire danger during the summer period probably driven by El Niño conditions.

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2 **3.3.3** Identifying Key Drivers of Focal Events

Weather, fuel moisture, fuel abundance, and ignitions are the four primary controls identified as influencing the occurrence and intensity of the focal fire events. Anomalies in individual drivers of these controls, such as temperature or soil moisture, are calculated by comparing regional daily 2023 averages with the average for 2003-2022.

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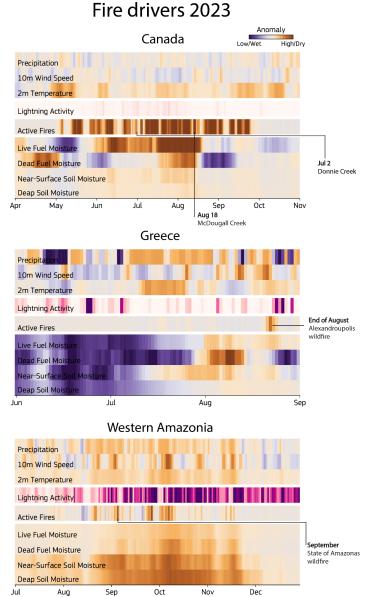
2239 Analysing the time series of a few key drivers helps contextualise the conditions under which 2240 the examined events took place (see Figure 12). However, it is by leveraging the infrastructure 2241 provided by PoF and ConFire models that we can perform a statistical causality attribution of 2242 the four controls for the observed fire occurrence (see Figure 13). By design, the PoF and 2243 ConFire models provide control attribution even if a fire event is not recorded. In such cases, 2244 a low probability across all controls indicates an accurate prediction. However, even in 2245 instances where no fire event is recorded, the models enable us to gain insight into the 2246 complex interactions among weather conditions, fuel characteristics, ignition sources, and 2247 other environmental factors. For example, a high probability of fire not associated with any AF 2248 could indicate successful suppression or fire-prevention policies implemented at a local level. 2249 Moreover, as our driver selection may not fully account for human influence, it is important to 2250 note that the category of "other" also encompasses unaccounted-for variables, as well as a 2251 measure of observed active fires and partially burned areas that were not forecasted by the models. Therefore, this analysis not only enhances our ability to understand the main controls 2252 2253 on fire activity, but it also helps identify missing pieces of information needed to fully 2254 comprehend the events. It is important to note there is not a one-to-one correspondence





between observed fire activity and the attribution analysis, as the analysis is based on model predictions alone.

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2259JulAugSepOctNovDec2260Figure 12: Anomaly driver stripes for the three focal events. The drivers are selected to2261contextualise the conditions under which the examined events took place. All values are2262expressed as anomalies compared to the 2003-2021 climatology with the exception of lighting2263activity which is expressed as absolute flash density.





2266 3.3.4 Drivers of Active Fire Extremes

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2268 3.3.4.1 Canada

2269 Persistent fire-favourable weather conditions played a crucial role in controlling the extent of 2270 active fires in Canada during the summer of 2023. Dry weather contributed to extensive drying 2271 of both live and dead vegetation, further exacerbating fire risk (Figure 12, 13). Most of the 2272 explainability of the Canada event comes from anomalous weather conditions. Increased 2273 lightning activity often coincides with or precede significant fire periods, indicating lightning as 2274 a key source of ignitions in the region, in agreement with the attribution of 59% of wildfires and 2275 93% of total BA to lightning ignition sources in Canada during 2023 (Jain et al., 2024). Adverse 2276 weather conditions in mid-May in western Canada were identified as influential factors in 2277 shaping fire events. However, multiple instances of intense burning events, notably in mid-2278 July, early August, and late September, fall into the 'Other' category, heavily contributing to 2279 the total number of events for which there is no attribution among the controls. The fact that 2280 clusters of events were not predicted, suggests potential inadequacies in accounting for some 2281 ignition sources or accurately representing fire propagation across vast, densely vegetated 2282 landscapes.

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2284 **3.3.4.2 Greece** 2285

2286 The driver anomalies (Figure 12) and control attribution (Figure 13) do not suggest an 2287 abnormally fire-prone year in Greece, failing to justify the extent and severity of the 2288 Alexandroupolis fire. An anomalously wet spring may have led to increased foliage and 2289 subsequently quick drying of plant material. A sustained dry period in late July and August 2290 further dried out new foliage, creating favourable conditions for fire activity, as indicated by the 2291 anomalously dry live and dead fuel moisture content in August. Despite these conditions, the 2292 unexpected extent and severity of the Alexandroupolis fire were not predicted, highlighting the 2293 intrinsic difficulties in forecasting isolated extreme events even when most predictors are 2294 included. Additionally, the high wind speeds at the time partially contributed to the extensive 2295 BA during the fire.

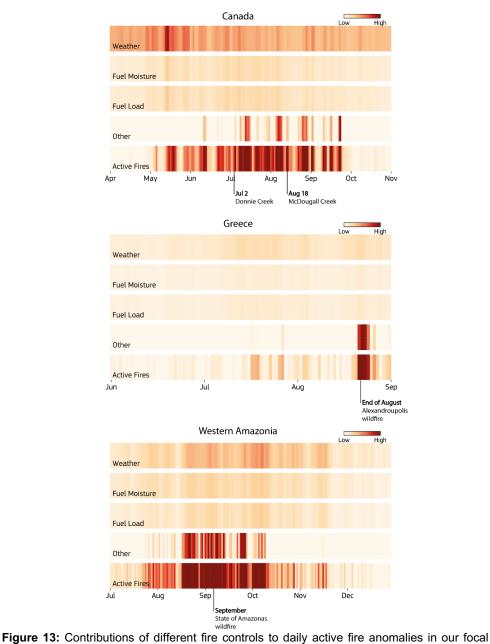
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2297 3.3.4.3 Western Amazonia 2298

2299 Prolonged drought conditions, stemming from anomalously low rainfall and high temperatures, 2300 created favourable conditions for an active fire season in Western Amazonia (Figure 12, 2301 Figure 13). These conditions had a significant impact on the typically wet ecosystem, affecting 2302 soil moisture as well as live and dead fuel moisture. Despite weather conditions serving as a 2303 persistent control for fire activity, several intense active fire periods in late August and 2304 throughout September were not predicted, possibly due to unrepresented ignition sources. 2305 Additionally, fire activity from September onwards was intensified by intense lightning activity, 2306 characteristic of the region, which substantially contributed to ignitions.







Fire controls

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events. All values are scaled to the observed daily fire anomaly such that the sum of the 4

daily control values matches the total observed anomaly.





2315 **3.3.5 Drivers of Regional Burned Area Extremes** 2316

- 2317 3.3.5.1 Canada
- 2318

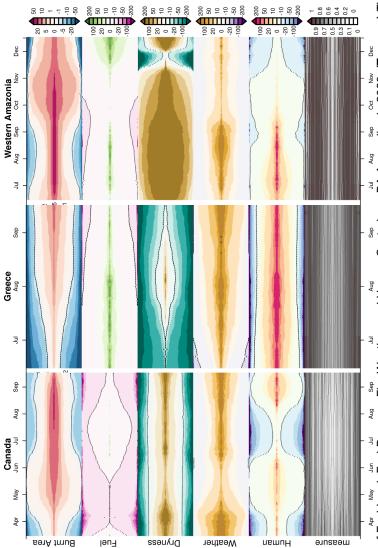
2319 ConFire detected a significant anomaly in BA starting in late April, as seen in the observations 2320 (Figure S9) with very high confidence between May (99.2% likelihood) and June 2023 (>99.9% likelihood; Figure 14). While less confident in a positive anomaly in September and 2321 2322 August (71% likelihood), ConFire detects the possibility of much higher burning in August and 2323 September, corresponding with the increase in burning in the Western Shield (Figure S9). 2324 Figure 14 shows the controls that contribute to these anomalies. Our analysis indicates a 2325 >99.9% likelihood that elevated fire weather conditions persisting throughout the 2023 fire 2326 season, led to a notable increase in burning, explaining between 4.6-45% (based on 10 to 2327 90th percentile confidence range) of the BA anomaly in May and 1-110% in June. If fire 2328 weather did explain more than 100% of the BA anomaly, then other controls must have acted 2329 to suppress BA. However, the anomalous weather conditions subsided later in the fire season. 2330 Drier fuel conditions could have contributed significantly to the increase in BA (up to 65% of 2331 Mays brunt area anomaly and 45% of Junes), though with low confidence (60.5% in May and 2332 61.3% in June), and wetter fuels exerting a suppressive effect on fire spread was also 2333 possible, suggesting their potential role in mitigating fire severity. A small but confident 2334 suppressive effect (100% likelihood in May, 64.8% likelihood in June) from fuel load was 2335 observed, reducing relative increases in BA between 1.7-7.1% in May. Direct human-induced 2336 landscape changes exhibited a small impact on the extent of burned areas (likelihood of 97.4% 2337 in May), explaining between 1.2-22% of Mays increased burning and 0.6-24% of Junes. The 2338 analysis revealed relatively low noise levels in the results ("uncertainty measure" in Figure 2339 14), indicating a robust signal despite uncertainties associated with the controls. This is shows 2340 that the higher burnt arae anomalies have the most confident signal. This robustness is partly 2341 attributed to the large spatial coverage of Canada's grid cells, enhancing the signal-to-noise 2342 ratio. 2343

2344 3.3.5.2 Greece

2345 2346 The analysis reveals an anomalously high BA, particularly from mid-August onwards, though 2347 with a lower confidence level compared to the Canadian case (69.9% likelihood; Figure 14). 2348 Figure 14 shows the controls that contribute to these anomalies. Our findings demonstrate 2349 with very high confidence the presence of anomalously high fire weather conditions during the 2350 2023 fire season in Greece (98.3% in July, and >99.9% in August and September). In August, 2351 these conditions explained between 4.3-140% of the increased BA, escalating to 5.6-170% by 2352 September. Assessing the impact of fuel moisture on BA, our analysis shows a wide range of 2353 possibilities, with confidence ranging from a 21% increase to 180% decrease in relative BA 2354 extent, which would have offset some of the increases from fire weather. This uncertainty 2355 underscores the complexity of the interactions between fuel moisture and fire behaviour in 2356 Greece. Similar to the findings in Canada, direct human-induced landscape changes exerted 2357 minimal influence on the extent of burned areas in Greece during the analysed period. While 2358 the analysis indicates a slightly higher than normal fuel load







explained by each control. At each timestep, the gradient of colours along the y-axis represent the range of modelled outcomes: high diversity in the colours indicates high uncertainty, whereas a consistent colour (e.g., red for positive BA anomalies) indicates greater confidence in the Uncertainty measure representing potential variations around our simulated BA anomaly as a result of factors not considered by the modelling Figure 14: Influence of Fuel Load, Fuel Dryness, Fire Weather, and Human Controls on BA Anomalies in 2023. (Top row) The colour indicates the modelled BA anomalies as relative increases/decreased in BA extent (% relative to the climatology). (Rows 2-4) The % of the anomaly projected direction and magnitude of change. Dotted black line indicated threshold between positive and negative anomaly. (Bottom row) ramework, as a fraction of land area. Lower numbers (lighter colours) indicate higher confidence in the BA anomaly in the top row. 2359 2360 2361 2362 2363 2364 2365 2365 2365

Uncertainty





2367 3.3.5.3 Western Amazonia

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2369 Our analysis indicates a reasonably high confidence that the considered drivers contributed 2370 to anomalous high burning during September (73.8% likelihood), October (94.9% likelihood) 2371 and November (96.1% likelihood; Figure 14). Figure 14 shows the controls that contribute to 2372 these anomalies. The primary driver of the observed BA anomaly appears to be dry fuel 2373 conditions, with a very high likelihood (99.7%) of drier-than-normal conditions persisting 2374 through November. This led to a substantial increase in BA, explaining at least 57% of the 2375 increase in BA. While fire weather conditions were also elevated (likelihood >99.9%), their 2376 impact on BA was comparatively lower, resulting in at least 2% increase in BA during October 2377 and November, though with a small probability of contributing much more. Direct human 2378 influence was identified as a contributing factor, with a high likelihood (92.9% in September, 2379 94.5% in October) of increasing burned area. However, the magnitude of this influence was 2380 approximately one-tenth of that attributed to fuel dryness. The influence of fuel load on BA 2381 was found to be small and statistically insignificant, with likely influences ranging from a small 2382 suppressive effect to explaining a maximum of the increased BA in September, to virtually no 2383 influence in October, suggesting that fuel dynamics played a minor role in driving the observed 2384 fire activity. The analysis reveals a higher confidence in the simulations indicating positive 2385 anomalies, indicating a robust signal in the attribution of drivers to observed BA anomalies. 2386

2387 **3.3.6 Spatial Variation in Drivers of Burned Area Extremes**

2389 3.3.6.1 Canada

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2391 In Canada, most BA anomalies were linked to widespread high fire weather, with 95% of the 2392 country being influenced by higher-than-normal fire weather (Figure S12). There was a 2393 tendency for fuels to dry out, although this was not as widespread. Fuel load anomalies were 2394 more scattered, but areas of low fuel anomaly did correspond to some boundaries in fire 2395 extremes (Figure 15, S12). Increased human influence may have had some influence at 2396 suppressing fires, but this is not significant, and in some places, the model indicates a small 2397 possibility of increased fire from human activity. In the Eastern Shield, fire extremes in some 2398 areas were driven by high fire weather and dry fuel, compounded with more vegetation cover 2399 and hence higher-than-normal fuel load in some places (Figure 15). However, the borders of 2400 extreme fires corresponded to a suppressive effect from decreased fuel load. In the Western 2401 Shield, dry fuel and high fire weather drove fire incidents, with high fire weather dominating in 2402 some areas. Increased suppression may have had some influence at suppressing fires, but 2403 this is not significant, and in some places, the model indicates a small possibility of increased 2404 fire from human activity.

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2406 In June, there were high anomalous burned areas in Quebec's Eastern Shield, which were 2407 divided into two major fire components with a slightly reduced BA in between (Figure 15). 2408 Both components were associated with high fire weather, but in areas where high fire weather 2409 occurred without the contribution of other controls, it tended not to cause high levels of burning. 2410 The highest burned areas were mainly found in the northern component and were associated 2411 with anomalously low levels of moisture and high fire weather. Some cells with the very highest 2412 BA also showed anomalously high fuel load (Figure 15). In a region further north, around 56-2413 57 degrees north and 72-80 degrees west, there was high fuel load, dry conditions, and high 2414 fire weather, but fire in the area was found to be highly fuel limited and largely insensitive to 2415 even large changes in controls (Kelley et al. 2019). The southern component of high burning 2416 corresponded with high fire weather and either fire fuel or high fuel load. Additionally, any 2417 boundaries between higher and normal/lower levels of BA also saw lower than average fuel 2418 loads, which may have inhibited fire spread (Figure 15).





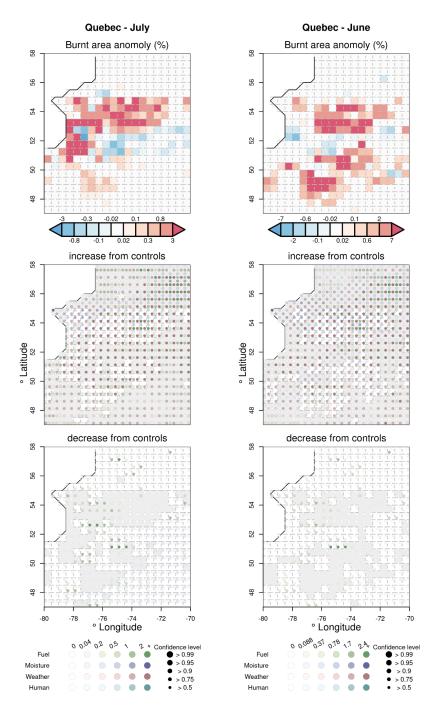




Figure 15: Anomalies in controls during months and regions of high BA in Quebec. The top 2421 left map of each region shows BA anomalies at 0.5 degrees for that month in 2023 versus the 2422 2014-2023 monthly average. The left of the other two maps looks at anomalies in controls that 2423 would cause higher BA, with areas not greyed out representing regions with greater than





2424 monthly average BA in 2023. The right map shows drivers that would have led to lower than 2425 normal levels of burning, with areas not greyed out showing lower or non-change from monthly 2426 average BA in 2023. Each grid cell has four points: green points show anomalies in fuel load, 2427 purple in fuel moisture, red in fire weather, and cyan in humans. This way, we can see if 2428 controls acted in unison to cause extreme levels of burning or prevent extreme fires from 2429 extending further. The shade of the point shows the most likely expected level of anomaly in 2430 that control, while the size shows how confident we are in the direction of the anomaly.

2431 2432

2433 In May, the Western Shield and Taiga/Boreal plains experienced higher-than-normal fire 2434 weather across the region (Figure S13). The increased burned areas in the west were due to 2435 extreme low fuel moisture and high fire weather, as well as higher fuel loads. In contrast, areas 2436 to the south experienced high fire weather without the extreme burned areas. Additionally, 2437 anomalies in fuel loads and burning levels became more evident in September, with some areas displaying lower-than-average fuel and burning (Figure S13). These anomalies 2438 2439 persisted, with regions still experiencing high fire weather and variations in fuel moisture 2440 levels. Furthermore, the eastern areas with higher fire weather also showed higher fuel loads, 2441 while drier fuel moisture was observed in less extreme regions to the east. Additionally, 2442 specific locations saw higher fire weather and above-average fuel moisture, while areas just 2443 north of the extreme fires experienced wetter-than-normal fuel moisture (Figure S13).

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2445 3.3.6.2 Greece

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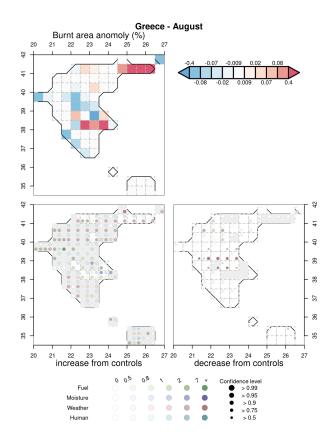
2447 Interestingly, most of Greece showed a tendency towards less suppression from people 2448 (Figure S12). However, the dominant driver over most (73%) of Greece was high fire weather, 2449 with some areas in central Greece showing notably low fire weather. These areas do not 2450 correspond to a fire anomaly, though higher fuel loads were detected. Except for central 2451 Greece, other areas of lower than average BA correspond to lower than average fuel loads 2452 (Figure 16). There was no significant anomaly in fuel moisture across Greece, though the 2453 northeastern fire extreme does correspond to a joint increase in fire weather and decrease in 2454 fuel moisture. Extremes in Eastern Coastal Greece correspond to anomalies in fuel load, fuel 2455 dryness, and heightened fire weather (Figure 16).

2456

2457 In August, Northern Greece experienced high fire weather and low fuel moisture, particularly 2458 around the Alexandroupolis fires and Macedonia and Thrace (Figure 16). The region 2459 extended further east, experiencing extreme fires that reached into Central Macedonia. Unlike 2460 Canada, the framework in North Greece did not show the same level of detail in the boundaries 2461 around extreme levels of burning. However, the transition to less burning in the South of West 2462 Macedonia did correspond to reduced fuel load. Areas around Athens and Central Greece 2463 that saw unusually high levels of burning also experienced decreased fuel moisture and 2464 increased fuel load (Figure 16). In contrast, areas in Southern Thessaly and Central Greece 2465 that did not experience high burned areas saw lower than normal fire weather. In the 2466 Peloponnese region, there was either high fire weather or reduced fuel moisture, but these 2467 conditions rarely occurred together, which might explain the lack of increased BA throughout 2468 the region. (Figure 16)







2469

2470 Figure 16: Same as Figure 15 but for Greece.

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2472 **3.3.6.3 Western Amazonia** 2473

2474 High fire weather anomalies were almost universal across our Western Amazonia region 2475 (Figure S12). However, anomalies in other controls varied widely across the region, which appears to modulate the occurrence of areas with higher than average burning (Figure S12, 2476 2477 S14). In September, the regions of high burning in the Western Amazon all exhibited higher 2478 than average fire weather, with the highest BA anomalies associated with the highest fire 2479 weathers. The areas with the highest BA anomalies along the Amazon River were located 2480 around Manaus and also showed lower than average fuel moisture and higher than average 2481 fuel load. The very highest pixels exhibited the anomaly in increased fuel loads. In the southern 2482 and central part of the region (65 to 62 west/7/8 south), near Porto Velho, BA anomalies were 2483 associated with extremely high fire weather and higher fuel loads. The areas of highest burning also showed an increase in human-driven burning. Further east (59 west, 7 south), the highest 2484 2485 fire weather and decreased fuel moisture occurred alongside higher burning. Areas of higher 2486 BA in the north (63 west, 0 north) were associated with extreme fire weather, though offset by 2487 lower than normal fuel loads. This may explain why they were not as extreme as the fires 2488 around Manaus, though the area is not generally considered fuel-limited (Kelley et al. 2019), 2489 and fuel had little overall impact across the region Figure 14).





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2492 3.4 Attribution to Global Change

2494 **3.4.1 Highlights** 2495

In Canada, anthropogenic forcing increased the chance of high fire weather in 2023, total climate forcing has led to higher present-day BA and socio-economic factors have likely decreased burning. Human influence at least doubled the probability of experiencing high fire weather in June 2023. It is virtually certain that total climate forcing increased recent high BA in the region by up to 38%, but it is less certain that socioeconomic factors decreased high burning. All forcings combined have led to an overall reduction in today's average BA across Canada.

2503

In Greece, anthropogenic forcing increased the chance of high fire weather in 2023. total climate forcing has likely led to higher present-day BA and socio-economic factors have likely decreased burning. Human influence increased the probability of experiencing high fire weather in August 2023 by 1.9-4.1 times. It is likely that total climate forcing increased recent high BA in the region by up to 30%, but socio-economic factors could have led to an increase or decrease. Climate change has increased today's average BA in the Mediterranean region, but this has been mainly offset by socio-economic factors.

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2521

In western Amazonia, anthropogenic forcing has greatly increased the chance of high fire weather in 2023. It is virtually certain that total climate forcing has led to higher BA, and very likely that socio-economic factors also contributed. Human influence increased the probability of experiencing high fire weather by 20.0-28.5 times. It is virtually certain that total climate forcing increased recent high BA in the region by up to 47%, and very likely that socio-economic factors exacerbated the increase. Climate change has increased today's average BA in the Northwest region, and all forcings have led to an overall increase in burning.

2520 3.4.2 Change in Likelihood of High Fire Weather due to Anthropogenic Forcing

2522 3.4.2.1 Canada

2523 2524 The fire weather conditions we saw in Canada during June 2023 were 2.9-3.6 times more 2525 likely due to anthropogenic forcing. Here we assess the 95th percentile of FWI over the country 2526 during the month of peak anomaly in BA (June) in the ALL and NAT HadGEM3 simulations. More of the ALL distribution lies above the observed 95th percentile of FWI from ERA5 2527 2528 compared to the NAT distribution (Figure 17), and we therefore conclude that the probability 2529 of experiencing the high fire weather observed during June 2023 is more likely in a climate 2530 forced with anthropogenic emissions. 2531

2532 3.4.2.2 Greece

2534 The high fire weather conditions experienced during the peak anomaly in BA in August 2023 2535 were 1.9-4.1 times more likely due to anthropogenic forcing (Figure 17). In this case the 95th 2536 percentile of FWI is outside of the distribution so we instead assess the 90th percentile of FWI 2537 over the country. This is likely a result of our linear inference of 2023 for the bias correction 2538 based on the 1960-2013 period, where in fact 2023 was so anomalous that it doesn't fit this 2539 trend. The 2023 event threshold here also lies at the very high end of simulated fire weather, 2540 meaning it was very unusual in the model simulations. The result range here is also larger 2541 than for Canada, meaning there is less certainty about how much human influence has 2542 increased the probability, although it does highlight at least a 50% increase in likelihood of 2543 high fire weather.

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2545 3.4.2.3 Western Amazonia

High fire weather in western Amazonia during Sept-Oct 2023 was 20.0-28.5 times more likely
due to anthropogenic forcing (Figure 17). In this region there is a large shift in the ALL forcing
distribution compared to the NAT only forcing for the 95th percentile of FWI, and the high risk
ratio shows a strong anthropogenic signal in driving the meteorological conditions that led to
high fires over this period.

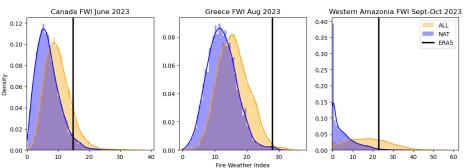


Figure 17: 95th percentile of FWI for June 2023 over (left) Canada, (middle) Greece and (right) western Amazonia, in the HadGEM3 ensemble of ALL (anthropogenic plus natural forcing, orange) and NAT (natural-only forcing, blue) bias-corrected simulations, and ERA5 reanalysis (black line).

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25603.4.3Change in Likelihood of High Burned Area in 2023 due to Total Climate Forcing2561and Socioeconomic factors

2563 3.4.3.1 Canada

2565 We show that total climate forcing was virtually certain (99.9% likelihood) to have led to more 2566 burning in events such as those observed in Canada in June 2023 (Figure 18). In the cells of Canada that burn most regularly (95th percentile burned area), our attribution results indicate 2567 2568 that total climate forcing increased the extent of area burned in Canada by between 0.7 -2569 6.22% during the month of June in the period 2003-2019. This estimate is almost certainly 2570 conservative, as we are testing the influence of climate change on the likelihood of equivalent 2571 levels of burning in years preceding 2023, and these years have experienced less warming. 2572 Additionally in Figure 14, considering the anomalies of 2023, we show between 1-30% 2573 additional increase in BA extent caused by anomalous weather conditions during this 2574 heatwave on top of the 2003-2019 period. The impact of socio-economic factors is less certain, 2575 with only a 54.77% likelihood of decreasing burning, affecting BA by between -10.12 and 2576 3.03% during 2003-2019.

2577

Overall, we estimate that BA in Canada in June 2023 was 0.8-38.0% greater due to total climate forcing in the 2003-2019 period combined with this year's anomaly in the climatic variables. As a caveat, we note that this is not a formal attribution of the 2023 anomaly because no counterfactual exists for the year, but rather an attribution of the change in BA in 2003-2019 with the additional influence of climate factors on BA in 2023 superimposed.





2583 3.4.3.2 Greece

2584

2585 We show total climate forcing caused a change in high BA in Greece of between -0.66 and 2586 18.58% in the period 2003-2019, with a likelihood of an increase of 84.92% (Figure 18). In 2587 this case we use the 90th percentile (gridcell with the 10% highest burned areas) to represent high BA over the region for the month of August, over 2003-2019. Again this increase is likely 2588 2589 a conservative figure given the additional warming since 2019, and estimating the additional 2590 burning that might have been experienced during the anomalous conditions of 2023, we find 2591 an additional change of -9.3-11%. Socioeconomic factors likely (79.9%) caused a decrease in 2592 burning, though could have caused an increase, affecting BA extent by -36.58 to 5.4%. 2593

2594 Overall we estimate that BA in Greece in August 2023 was influenced by -15.2-29.6% due to 2595 total climate forcing in the 2003-2019 period combined with this year's anomaly in the climatic 2596 variables. As per the results for Canada, we note that this is not a formal attribution of the 2597 2023 anomaly, but an attribution of the change in BA in 2003-2019 with the additional influence 2598 of climate factors on BA in 2023 superimposed. In the case of Greece, uncertainties around 2599 whether total climate forcing and socioeconomic factors caused an increase or decrease in 2600 BA are higher, and the smaller region size makes detecting a strong signal of change more 2601 challenging. 2602

2603 3.4.3.3 Western Amazonia

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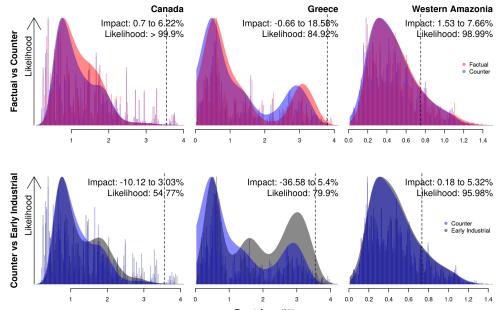
2605 Over the period 2003-2019, total climate forcing was virtually certain to have caused an 2606 increase in burned areas like the one experienced in western Amazonia in September and 2607 Oct 2023 (99% likelihood), with a likely range of increase in extent of 1.53-7.66% (Figure 18). 2608 Here we assess the change in BA due to total climate forcing in the cells that burn most 2609 regularly (95th percentile) of our defined region of western Amazonia over Sept and Oct 2003-2610 2019. Extending our analysis to the 2023 anomaly, we estimate that additional burning could 2611 have been up to 0.8-36.2% on top of the 2003-2019 levels. Despite finding little influence of 2612 humans specifically in 2023 compared to the previous 10 years in (Figure 14), since early-2613 industrial, we show socioeconomic factors have had a large influence on the occurrence of 2614 extreme levels of burning. Events similar to 2023 were very likely exacerbated by 2615 socioeconomic conditions (95.98% likelihood) increasing BA by 0.18-5.32%.

2616

Overall we estimate that BA in western Amazonia in September-October 2023 was increased by 2.3-46.7% due to total climate forcing in the 2003-2019 period combined with this year's anomaly in the climatic variables. Though again, we note that we determine this by convoluting the change in BA in 2003-2019 with the additional influence of climate factors on BA in 2023 superimposed, and therefore it is not a formal attribution statement of the 2023 anomaly.







Burnt Area (%)

2622 2623 Figure 18: Likelihood of high BA from ConFire ISIMIP3a. Panels show (top row) the likelihood 2624 (probability) of an extreme event such as experienced in the month of peak burning in 2023 2625 with (red) and without (blue) total climate forcing, and (bottom row) likelihood of the same 2626 extreme event with (blue) and without (grey) socioeconomic forcing for (left) Canada, (middle) 2627 Greece (Centre) and (right) Western Amazonia. Likelihood shown on each plot, and impact 2628 refers to the increase in BA due to the forcing.

2629 2630

2631 3.4.4 Change in Burned Area Anomaly due to Total Climate Forcing, Socioeconomic 2632 Factors, and All Forcing

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2634 3.4.4.1 Canada 2635

2636 As reported in Burton, Lampe et al. (2023), we also show how median BA for months of interest 2637 has changed overall in the region today due to total climate forcing (Figure 19), socioeconomic 2638 forcing (Figure S15), and all forcings (Figure S16). Using AR6 regions that best match our focus areas, we show that BA has increased by 1.9% [0.1, 3.6] in North West North America 2639 2640 (NWN) due to total climate forcing, but reduced by -0.2% [-1.7, 1.3] in North East North 2641 America (NEN) (Figure 19). In these regions, socioeconomic forcing has dampened the 2642 effects of climate change, by reducing BA by -9.5% [-13.6, -6.3] in NWN, and -8.5% [-12.5, -2643 5.7] in NEN (see Supplement). All forcings combined have led to an overall reduction in BA of 2644 -8.3% [-12.5, -4.9] in NWN and -8.7% [-12.8, -5.8] in NEN (Figure S16). 2645

2646 3.4.4.2 Greece

2647 2648 Burton, Lampe et al. (2023) find a larger increase in median BA for months of interest due to 2649 total climate forcing in the Mediterranean region (MED), with an increase of 14.5% [11.5, 18.1] 2650 today compared to the counterfactual (Figure 19). This is particularly the case for the high 2651 burned areas, where the increase is larger compared to the lower end of the distribution. 2652 However, socioeconomic factors have largely offset this by reducing BA by -10.2% [-13.6, -





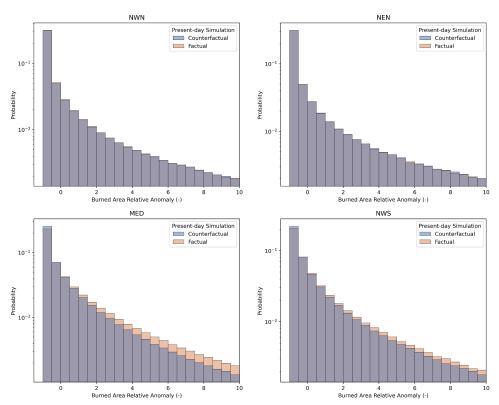
6.6]. However all forcings combined have led to an overall regional increase in BA of 0.5% [3.5, 5.5] (see Figure S16).

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2656 3.4.4.3 Western Amazonia

As per Burton, Lampe et al. (2023), total climate forcing has increased median BA for months of interest by 11.5% [5.4, 18.4] in Northwest South America (NWS) today compared to the counterfactual (**Figure 19**). Again, this increase is mostly impacting the BA at the higher end of the distribution. This is mostly offset by socioeconomic factors (-9.0% [-18.9, 1.2]), although all forcings combined have still led to an overall increase in BA of 1.5% [-6.9, 10.5] in the region (see Supplement).



2665 2666

Figure 19: Change in median BA due to total climate forcing from FireMIP. Present day BA
 (2003-2019) for factual (historical forcing, orange) and counterfactual (detrended climate,
 blue), for AR6 regions. Panels show (top left) North West North America (NWN), (top right)
 North East North America NEN, (bottom left) Mediterranean (MED), and (bottom right) North
 West South America (NWS). Probability is shown on a log scale.

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2674 3.5 Seasonal and Decadal Outlook

20/0 3.5.1 Fighlights	2676	3.5.1	Highlights
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- Predictions see the Indian Ocean Dipole persisting in its positive phase well into 2024. The 2023–2024 El Niño event emerged as the fourth-most powerful on record. As 2024 comes to an end most simulations forecast a transition to ENSO-neutral conditions. Positive IOD phases are associated with elevated BA Amazonia, Indonesia, and parts of Australia though these tend to depend on interactions with ENSO.
- Seasonal predictions of fire weather through August 2024 highlight moderate positive anomalies in Western Canada, South America (including southern Amazonia) but no clear signal for extreme anomalies is present in the forecast. These forecasts have limited skills beyond month 1-2 (e.g., Section 3.3.2).
- 2688 Effective strong mitigation efforts could limit increases in future fire extremes. • Canada in particular is very likely to see substantial increases in BA under high 2689 emissions scenarios (SSP370, SSP585) compared to high mitigation (SSP126), driven 2690 2691 by dry conditions and climate and CO2 driven increase in fuel load. However the lack 2692 of significant difference between middle-of-the-road (some mitigation e.g. SSP370) 2693 and no mitigation (SSP585) scenarios in this region highlights the need to constrain 2694 uncertainties in subsequent studies. In Western Amazonia, while SSP585 predicts 2695 significant increases in fire likelihood by 2100, the SSP126 scenario suggests that 2696 proactive climate strategies could maintain conditions that largely remain unchanged, 2697 underscoring the importance of mitigation actions for this region.
- However, even with mitigation, we will still see an increase of extremes in Canada. This necessitates robust adaptation strategies to manage and prepare for the expected rise in extreme fire events as global mean temperature continues to rise towards 1.5C above PI, ensuring that communities and ecosystems are better equipped to handle future challenges.
- 2703 There is still a lot of uncertainty regarding how extreme fire events will change • 2704 with current efforts to reduce emissions. Our projections show a high level of 2705 uncertainty in all regions for the expected increase in extreme events under moderate 2706 mitigation efforts (SSP370 scenario) and without any mitigation (SSP585 scenario). 2707 The wide range of projections means that we cannot determine if current mitigation 2708 efforts are effective, and more research is needed to reduce uncertainties in future 2709 projections. However, even in the best-case scenarios, both SSP370 and SSP585 2710 show significant increases in the frequency of extreme events in Canada and Western 2711 Amazonia. While all regions show, with confidence, that strong mitigation (SSP126) 2712 will reduce the occurrence of extreme events, there is still a large plausible set of future 2713 increases in extremes. Combined with the lack of understanding in future land use and 2714 direct human influence on fire, this makes it hard to inform adaptation efforts.
- Greece's smaller domain size means that Global Climate Models (GCMs)
 introduce biases and uncertainties making future projections of extremes
 difficult. This highlights the need to improve model projections of future climate or
 incorporate other information, observations or model/statistical approaches to
 constrain uncertainties on small scales.

2721 **3.5.2 Seasonal Outlook for 2024** 2722

The 2023–2024 El Niño event emerged as the fourth-most powerful on record, unleashing widespread droughts, as well as floods and other anomalous conditions worldwide. Officially declared by the World Meteorological Organization (WMO) on July 4, 2023, its meteorological impacts have unfolded between November 2023 and April 2024 (Joshi, 2023). Climate scientists have found that the 2023–24 El Niño event, compounded by the climate change





crisis, has elevated global temperatures beyond the records set during the 2016 El Niño event.
This has caused a 1.5°C increase in global temperature compared to pre-industrial times for
most of the year and established new temperature records in 2024 (McCulloch et al., 2024).

2732 As of mid-April 2024, El Niño conditions persist in the central-eastern equatorial Pacific, with 2733 important oceanic and atmospheric indicators aligning with an ongoing El Niño event that is 2734 gradually diminishing (Figure 20). Most simulation prediction plumes forecast a transition of 2735 the El Niño event to ENSO-neutral in Apr-Jun, 2024, which then persists during the boreal 2736 summer. The Indian Ocean Dipole (IOD) index is currently positive with the most recent value 2737 of +0.95 °C and forecasts indicate that it will remain in a positive state for the next season. 2738 Connections are established between a positive IOD phase and fire risk in Indonesia and parts 2739 of Australia, though outcomes generally depend on interactions with the ENSO phase (Pan et al., 2018; Ren et al., 2024; Abram et al., 2021). Similarly, the positive phase of the IOD is 2740 2741 linked with heightened fire risk in South America, particularly in the Amazon basin, where it 2742 can interact with other climate teleconnections to exacerbate droughts and increase wildfire 2743 risk (Cardil et al., 2023; Dong et al., 2021). 2744

2745 Figure 21 shows the predicted probabilities of the monthly average FWI exceeding moderate 2746 (75th percentile) or high (95th percentile) thresholds of the monthly climatology. As the strength of a positive ENSO diminishes, most areas in Southeast Asia and South America are 2747 2748 predicted to experience a decrease in the likelihood of anomalous conditions over the next 2749 three months. Parts of Western Canada are predicted to reach moderate anomalous 2750 conditions once again in early summer, and this combined with overwintering fires could promote a second consecutive high fire season as has already been reported in the media 2751 2752 (BBC News, 2024, New York Times, 2024). Predictions also suggest that the moderate FWI 2753 threshold will be exceeded in southeast, central, and western Brazil with the high threshold 2754 exceeded in southern and western parts. In parts of Africa, moderate FWI anomalies may be experienced throughout June-August.

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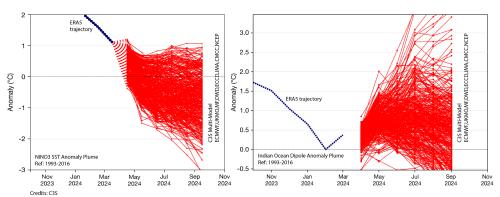


Figure 20: Anomaly plume for the ENSO and IOD as forecasted by the multimodal C3S
 ensemble system for April 2024. Ongoing positive ENSO event is gradually diminishing while
 Indian Ocean dipole is likely to remain in a positive state until September 2024 (Copernicus
 Climate Change Service, 2024b).

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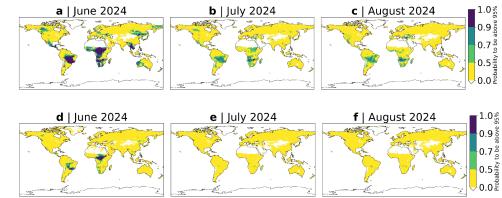


Figure 21: Probability for the monthly average FWI exceeding the (top row) 75th percentile
threshold (anomalous conditions) and (bottom row) 95th percentile threshold (extremely
anomalous conditions) of the monthly climatological distributions. The probability is calculated
using the 51 ensemble member realisation from ECMWF's long-range forecasting system,
SEAS5 FWI, and comparing them with the 1991-2016 climatology (Copernicus Emergency
Management Service, 2019).

2776 3.5.3 Future Changes in Likelihood of Extreme Fire Events

2778 3.5.3.1 Canada

2780 The probability of Canada experiencing a fire event similar to June 2023 (measured in terms 2781 of BA in the cells with the highest 5% burned areas) is estimated to be 0.14% under the climate 2782 conditions of 2010-2020 (Table 7; Figure 22). This corresponds to a return time of 2783 approximately once every 700 years. Even though the GCMs were bias-corrected to 1979-2784 2014 values as part of the ISIMIP3b protocol, internal variability and trends were maintained. 2785 This difference can lead to discrepancies between models due to the non-linear response of 2786 fire to small changes in climate variables and simulated vegetation between ISIMIP3a and 2787 ISIMIP3b, which substantially altered the results over our baseline periods. When using bias-2788 corrected GCM model output instead of reanalysis data, the likelihood ranges from 0.08% to 0.31%, indicating a slight bias of the probability of the event introduced by using GCM 2789 2790 historical simulations. We describe future changes as significant if the range across GCM 2791 projects does not overlap with this range.

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2793 Over the next 10 years, the likelihood of a similar fire event occurring increases significantly 2794 to between 0.14% and 0.48%, with no significant difference across scenarios. This suggests 2795 that a 2023 fire event could be up to three times more likely by the 2030s than in the last 2796 decade. By the middle of the century, the likelihood of a 2023 fire event recurrence increases 2797 to 0.31% - 0.9% - two to almost four times as likely as today (Table 7; Figure 22). The different SSP scenarios diverge significantly by 2070. Under SSP126, the likelihood of a 2023 fire event 2798 2799 recurring stabilises and remains largely unchanged until 2100, with a range of 3.07% to 3.5%. 2800 This means the recurrence of a 2023-like event would be around 2.8% to 8.3% in the last 2801 decade. Integrating between today and 2100, under SSP126, the probability of at least one 2802 event similar to or worse than the 2023 event occurring again is estimated to be between 16% 2803 and 25%.

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1ts today using reanalysis 'factual' and today a the BA extreme identified in Section 3.4.3. 1-ir ay climate conditions, following the definition of fferent time horizons based on each models like its intervent in the intervent of the total of the total of the intervent of the total of	Table 7 : Summary of the likelihood of extreme even for our three focal regions. '2023' events focuses on at the likelihood of a 1-in-100 event under present di how much more frequent the events will be at two di during 2010-2020. "indicates non-significant result.
cant result. I traithood 1%/vaar)	during 2010-2020. *indicates non-signifi Barion Event ISSP Reviewents
oe at two different time horizons based on each models likelihood in the future projections over likeliho	how much more frequent the events will
r present day climate conditions, following the definition of extreme in (UNEP, 2024). We also determ	at the likelihood of a 1-in-100 event unde
focuses on the BA extreme identified in Section 3.4.3. 1-in-100 for Western Amazonia additionally loc	for our three focal regions. '2023' events
dreme events today using reanalysis 'factual' and today and into the future using bias-corrected GC	Table 7: Summary of the likelihood of e





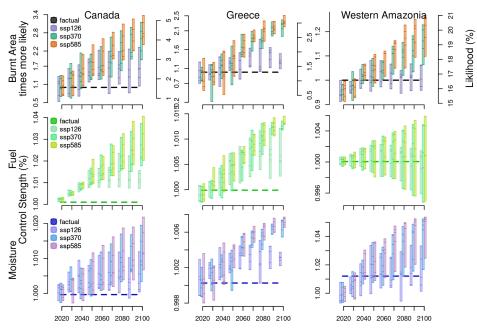




2810 Mitigation efforts under SSP370 have some impact, lowering the likelihood to 0.93% to 1.91% 2811 by 2100 compared to 1.15% to 2.2% in SSP585 (no mitigation; Table 7; Figure 22). The 2812 probability of an event like 2023 occurring at least once by 2100 is estimated to be between 2813 30% and 52% under SSP370 and between 35% and 56% under SSP585. Someone born in 2814 Canada in the current decade, with a life expectancy of 83 years (United Nations Population 2815 Division, 2022), has a 43-66% probability of seeing a similar event in their lifetimes under 2816 SSP585, compared with only an 11% likelihood of someone who was 83 years old in the 2817 2010s. Someone born in Canada today would also have a 18-45% probability of seeing an 2818 event of similar magnitude twice under SSP585. This reduces substantially to 20-31% for one occurrence and 4-10% for two occurrences under SSP126. 2819

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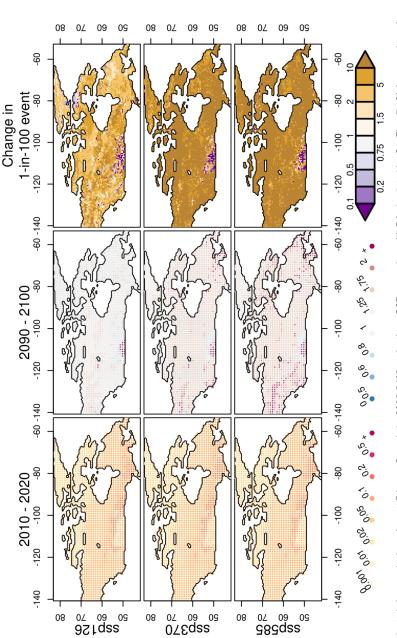
The divergence between shared socioeconomic pathways (SSPs; **Table 7**; **Figure 22**) is primarily driven by variations in the increase in fuel load, although an increase in fuel dryness is a major driver of increased burning across SSPs, particularly in SSP585. The worst-case scenario within SSP585 sees a substantial increase in both and changes in both controls are needed to explain divergence between SSPs. 2826



2827

2828 Figure 22: Future projections from ConFire of the likelihood of BA extremes and their 2829 corresponding fuel and moisture controls. Each set of bars shows changes in each decade, 2830 with each bar representing a different SSP scenario and the spread of bars indicating the variation across GCMs. The BA data indicates (right axis) the likelihood of the defined event 2831 2832 occurring in any given year for future decades and (left axis) the frequency of occurrence for 2833 relative to reanalysis-driven simulations (factual ISIMIP3a - dashed line) of the years 2010-2834 2020. The fuel and moisture rows illustrate the change in the maximum burning allowed by 2835 that control relative to reanalysis-driven simulations for the years 2010-2020.





cell indicates the likelihood (larger = higher likelihood) of a BA fraction (or being greater than the threshold indicated by the coloured dot (see legend at the base). Likewise, the size of the dot varies with likelihood that the BA fraction exceeds the threshold indicated by the coloured dot (see legend at the base). Likewise, the the left column indicates a high likelihood of the BA fraction exceeding 0.05%, whereas a small dark red dot indicates a small (but non-zero) likelihood of the BA fraction exceeding 0.05%, whereas a small dark red dot indicates a small (but non-zero) likelihood of the BA fraction exceeding 0.05%, whereas a small dark red dot indicates a small (but non-zero) likelihood of the BA fraction exceeding 0.05%. requency in 2090-2100 of a 1-in-100 year event defined for the period 2010-2020, expressed as a ratio of 2010-2020 values. In the left column, the size of the dot in each grid -igure 23: Projected changes in June-August BA over Canada by 2090-2100 under three SSP scenarios, with BA simulated by ConFire. (Left) Average June-August BA fraction (%) for 2010-2020. (Middle) Relative change in June-August BA extent projected for 2090-2100 expressed as a ratio of 2010-2020 values. (Right) Increased (or decreased)



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2844 While some areas of Canada see increases in BA and fire extremes (spatially measured as 2845 the proportional change in probability of a 2010-2020 1-in-100 annual event, as per UNEP 2846 (2024) almost everywhere (Figure 23), increases in BA are likely to be particularly strong in 2847 the critical regions of agriculture in Southern Alberta and Saskatchewan, starting from 2030 2848 (Figure S17) and becoming particularly strong by 2090s (Figure 23). However, some regions 2849 just to the North of this region may see a decrease in future extremes. While Yukon and 2850 Northwest Territories will see an increase in BA and fire extremes from 2040 (Figure S18), 2851 the increase in extremes remains similar to the rest of Canada, at around a doubling in 2852 occurrence. By the end of the century, however, SSP370 and SSP585 diverged from the high 2853 mitigation SSP126, and both scenarios see a much larger increase in burned areas and some 2854 areas with extreme occurrences becoming 5 times more common. SSP585 sees these 2855 increase in fire risk extending down through British Columbia.

2857 **3.5.3.2 Greece** 2858

2859 Using the ISIMIP3b GCMs for Greece introduced a strong bias when using the 95% of burned 2860 areas. We therefore used the 90% percentile (i.e BA in the 10% cells that see the most BA) to 2861 measure areas showing extreme burning. Here, using GCMs instead of reanalysis changed 2862 the likelihood of Greece's August 2023 event from 1.3% to 0.67-1.78% for 2010-2020 (Table 2863 7; Figure 22). While there is an indication of increased likelihood of extensive BA by the middle 2864 of the century, the projected range of changes in future extremes remains smaller than the 2865 introduced historical bias and is, therefore, insignificant. For SSP126, this small but 2866 insignificant increase remains until 2100, though at times, SSP126 shows the possibility of a 2867 substantial increase - up to 2.3 times more likely than today. This shows that Greece may still 2868 see large increases in the occurrence of extreme BAs even under strong mitigation, though 2869 uncertainties are too large to state this confidently. This may be due to Greece's smaller region 2870 and the need for a larger sample of points to overcome the inherent noise in simulating BA 2871 extremes. Our uncertainty quantification was necessary to determine these significant/non-2872 significant results. SSP350 and SSP585 show significant increases by 2070, and significantly 2873 diverge from SSP126 by 2080. There is no significant difference between the scenarios by 2874 2100, which indicate a likelihood of seeing an annual recurrence of 2023's event of between 2875 2-54-3.26% by 2100 - a 1.83-to-4.28-fold increase today and an average return time of every 2876 39-31 years. While fuel loads show a larger increase than fuel moisture in future projections, 2877 the regions are less sensitive to fuel (Kelley et al. 2019; Figure 14), and differential increases 2878 in fuel dryness dominate the variations between SSPs by 2100.

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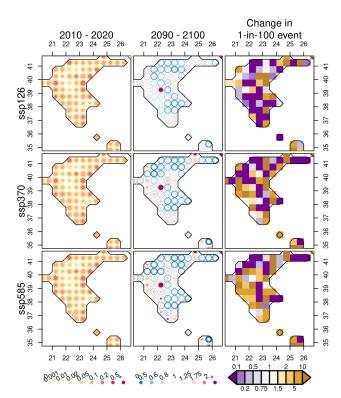
2856

While it is hard to ascertain a clearer climate signal on area extremes across Greece, there are some spatial patterns in likelihood of extreme occurrence with Greece (**Figure 24**). The middle of Greece may see a decrease in extremes in the future as early as the 2030s and 2040s (**Figure S20, S21**), with Central Greece and Macedonia seeing that large decreases by 2090 (**Figure 24**), though some areas of Central Greece, West Greece and Peloponnese seeing a potential rise in extremes. Changes in BA are less pronounced and future projections suggest either decreases or increases in summer burning is possible.

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Figure 24: Same as **Figure 23**, but for Greece and months July-September.

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2892 3.5.3.3 Western Amazonia

2893 The analysis of the South American region suggests a 16.58% chance of experiencing high 2894 2895 burned areas (measured in terms of BA in the cells with the highest 5% burned areas) in 2896 Sep/Oct 2023, based on data from 2010-2020 using ISIMIP3a (Table 7; Figure 22). Replacing 2897 this with ISIMIP3b introduces a small bias, with a likelihood of 15.13-16.57%. These events, 2898 while higher than average, are not as extreme as those in Canada and Greece, with a likely 2899 return time of 6 years in ISIMIP3a and 6-7 years in ISIMIP3a. However, this may hide the 2900 spatial variations and shift in locations of extremes, particularly along the deforestation front, 2901 as highlighted in Figure S14.

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2903 There are only small increases in projected future extremes, noting that we do not consider 2904 land use change, which can be a major driver of BA in some regions (Lapola et al., 2023; 2905 Kelley et al., 2021; Silveira et al., 2020). Under the SSP585 scenario, there is a notable and 2906 significant increase in future burning, estimated to be 1.2-1.27 times more likely than current 2907 levels by 2100. This increase is primarily attributed to drier fuel moisture, in line with the trend 2908 suggested by the FWI analysis in Section 3.4.2. There is potential for mitigation, particularly 2909 under the SSP126 scenario, which converges with the other two SSPs. This suggests that the 2910 levels of extreme fire activity may largely remain unchanged by 2100.

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While fuel load could potentially have a significant impact, it remains uncertain whether it would lead to an increase or decrease in burning. **Section 3.3** emphasises the positive influence of land use on the 2023 burning levels in some locations. However, despite the likely





2915 substantial impact inferred from results in these sections, future land use projections were not 2916 available at the time of writing and should be prioritised for next year's report.

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2925

2918 The likelihood of the occurrence of the 2023 event increases substantially in the centre of the 2919 region by the 2030s (Figure S21) and is almost universal across the region by 2024 (Figure 2920 S22), though small pockets around Manaus and roads in the south may see either not much 2921 change or the chance of a decrease. Perhaps more important for the region's fire-sensitive 2922 forests is the projected increase in BA in SSP370 and SSP585 across forests in the north of 2923 the region could have a significant impact on local vital ecosystems within the next few 2924 decades (Figure S23).

2926 When testing more extreme levels of burning of a 1% likelihood in ISIMIP3a, there is a 2927 noticeable divergence compared to present conditions and between different SSP scenarios 2928 in future projections (Table 7). Introducing a minimal bias using ISIMIP3b changes the 2929 likelihood to between 0.8-1.53. Across all scenarios, there is an increase in extreme events 2930 throughout the 21st century, although the significance is only reached in the last couple of 2931 decades in SSP370 and SSP585. Without mitigation, there is a likelihood of 2.32-3.34 of a 1-2932 in-100 event by 2100, representing an increase of 2.24-2.9 times. However, stringent 2933 mitigation measures show a decrease in extremes, with a 1-in-100 event only increasing to 2934 1.31-1.5. This suggests that strong mitigation measures may offset the risk of extreme burned 2935 areas in the region in the future.

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2937 4 **Discussion and Conclusions**

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2939 4.1 Summary of the State of Wildfires in 2023-24

2941 Objective 1: Regionally identify extreme individual wildfires or extreme wildfire seasons 2942 of the past 12 months, and place them in context of recent trends

2944 In this report, we used a combination of data from global Earth Observation products and data-2945 driven models (Jones et al., 2024; Giglio et al., 2018; van der Werf et al. 2017; Kaiser et al. 2946 2012), regional products (e.g. EFFIS), and insights from regional experts to identify key events 2947 of the period March 2023-February 2024. Highlighted results are provided in Section 3.1.1 2948 and summarised here. The 2023-24 fire season saw 3.9 million km² burned globally, slightly 2949 below the previous average, yet carbon emissions were 16% above average due to severe 2950 fires in Canada's boreal forests and despite low fire activity in African savannahs. In North 2951 America, Canada experienced record-breaking fires, with extensive damage and significant 2952 evacuations, while there were also localised yet severe incidents in the US (notably in Lahaina, 2953 Hawai'i, where 100 people lost their lives). South America saw below-average fire activity 2954 overall, but with an anomalous high fire season in Western Amazonia in 2023 and northern 2955 parts of the continent so far in 2024. Severe impacts of individual fires were also seen, notably 2956 in Paraíso, Chile, where 131 people lost their lives. Europe had generally low fire activity, 2957 except for significant incidents in Greece which included the largest fire on record in the EU. 2958 Oceania saw above-average fire activity focussed on Western and Northern parts of Australia. 2959 Asia experienced low fire activity overall, but with some significant regional impacts in 2960 Southeast Asia and parts of Russia. Africa's fire extent was low, although notable fire-related 2961 emergencies occurred in Northern and Southern Africa.

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2963 Objective 2: Shortlist a selective number of focal events (extreme individual wildfires 2964 of extreme wildfire seasons) with notable impacts on society or the environment.

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2966 While identifying events to become foci of the report is not trivial due to the diversity of impacts 2967 that wildfires present, we selected a pragmatic set of three events for further analysis. In



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Canada, the 2023-24 fire season was marked by an unprecedented burned area, high carbon emissions, and a prolonged fire season with severe air quality impacts and significant human and economic tolls. Greece experienced its second-worst fire season on record, with extensive damage, large-scale evacuations, and significant urban and infrastructural impacts, particularly from the massive Evros fire. Western Amazonia faced unprecedented fire counts, severe air quality degradation, and broad socio-economic and environmental impacts, highlighting the critical global significance of the region's fire activity.

2976 Objective 3: Diagnose the contributions of fuel dryness, fuel load, ignitions and 2977 suppression to the occurrence of each focal event.

2979 Using Bayesian inference and data-driven methods, we attributed observed fire activity and 2980 resulting burned areas to four controls: weather, fuel abundance and status, and others. The 2981 "others" category includes human influences, ignition sources like lightning, and residuals of 2982 all missing predictions. For BA, we can still include long-term variations in "other" human 2983 drivers to test anomalies in direct human influences on fire. We found that fire weather 2984 conditions alone do not entirely drive the occurrence of significant fire events in 2023-24. Even 2985 in Canada, where fires are primarily weather-driven, the most intense burning events in mid-2986 July, early August, and late September were not predicted, highlighting potential shortcomings 2987 in forecasting specific ignition sources and accurately depicting fire spread across extensive 2988 landscapes. By incorporating variables beyond fire weather, however, we successfully 2989 attributed drivers of much of the fire activity and BA with reanalysis meteorology and remote 2990 sensed land data. The event in Greece could not be uniquely attributed to one source; instead, 2991 all elements, including human influence, strongly contributed to the prediction. Availability of 2992 input data, related to the complexity of fire at the urban interface, limited the ability of the 2993 modelling frameworks to accurately diagnose contributions to the Alexandroupolis fire. Future 2994 developments aim to further expand on the variables related to manmade ignition sources and 2995 suppression efforts (e.g. urbanisation). A similar situation occurred in the Amazonian domain, 2996 where a combination of extended drought, human-induced practices, and weather explained 2997 fire activity. Additionally, changing patterns of human activity and fuel loads explained fire 2998 anomaly edges. It was found that the most severe burning could not be explained by one 2999 single control and often multiple factors were concomitant. Across all three regions, the extent 3000 of fire anomalies and the boundaries between extreme and non-extreme burned areas were 3001 often influenced by lack of anomalous control or by lower-than-average fuel anomalies.

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3003 Objective 4: Assess the capacity of operational predictive systems to forecast each 3004 focal event.

3006 Employing fire weather indices, we have analysed current capability to pre-alert for the 3007 extreme events highlighted in this report in line with fire agency practices. Fire agencies 3008 typically use local fuel condition data to interpret FWI values, which reduces intervention on 3009 high FWI values when fuel conditions aren't hazardous. The analysis mostly concentrated on 3010 the ability of metrics such as the FWI to indicate anomalous conditions rather than their 3011 reliability in prediction. In Canada, conditions were well predicted, and FWI could even capture 3012 the timing of the release break. However, this metric is less successful in the Amazon domain, 3013 where weather is not the primary driver of fire activity. Of particular concern is our current poor 3014 capacity to pre-alert for extreme fires such as the one in Alexandroupolis, Greece. Despite 3015 indications of extreme conditions in FWI records surrounding the event, similar circumstances 3016 in late July did not yield catastrophic impacts. Comparable findings in South America further 3017 emphasise that despite weather conditions serving as persistent controls, several intense fire 3018 periods in late August and September went unpredicted, likely due to unrepresented ignition 3019 sources. This highlights the necessity of broadening early warning systems beyond fire weather, taking into account fuel availability, fuel types, and ignition variability, to enhance the 3020 3021 reliability of fire risk assessment. The primary goal and philosophy of fire danger rating is to





indicate potential fire behaviour in presence of ignitions, and so our results are consistent withand emphasise the caveats of FWI.

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3025 Objective 5: Attribute each focal event to anthropogenic factors including climate 3026 change and land use.

3027 Using three methods of fire attribution, we find that the probability of high fire weather has 3028 been increased in all of the focal regions due to anthropogenic forcing, extreme fires have 3029 increased due to climate change, but average BA has been largely offset by socioeconomic 3030 factors. The highest increase in FWI has been in western Amazonia, where the probability of 3031 high fire weather has increased by 20.0-28.5 times due to human influence. In Canada we 3032 find that the probability of high FWI at least doubled as a result of anthropogenic forcing, in 3033 line with findings from the World Weather Attribution (Barnes et al., 2023). In both Canada and 3034 western Amazonia it is virtually certain that climate change has increased the extreme BA 3035 experienced in the region (by up to an estimated 38% and 47% respectively), such as the 3036 events witnessed in 2023. The influence of socioeconomic factors on extreme burning is less 3037 certain for each region, except for western Amazonia where it is very likely that they 3038 exacerbated the fires. When considering how median BA has changed in the present day 3039 period, results show that climate change has led to an increase in most of the AR6 regions 3040 related to our case study areas. However in all three areas, socioeconomic factors have 3041 reduced BA, dampening the effects of climate change that would otherwise have been 3042 experienced. When the impact of all forcings is assessed, we find there is a net increase in 3043 mean BA in Northwest South America and the Mediterranean, but an overall reduction in mean 3044 BA in North West and East North America, signalling that socioeconomic factors outweigh the 3045 increase from climate change in these areas.

3046Objective 6: Provide an outlook on the probability of extreme events in the coming fire3047season

3048 With the weakening of a positive ENSO, it is anticipated that most regions in Southeast Asia 3049 and South America will witness a decrease in the likelihood of anomalous conditions persisting 3050 over the next three months. However, the ongoing Indian Ocean Dipole (IOD) is expected to 3051 elevate landscape flammability across South America and West Africa.We provide the 3052 expected anomaly in the next 3 months for the FWI, highlighting the likelihood of moderate 3053 anomalous conditions in western Canada in early summer. Given the severity of the previous 3054 fire season and the possibility of overwintering fires, this could once again lead to an early 3055 start of the Canadian fire season. At extended lead times (greater than 2 months), forecast 3056 skills are limited, and the model is likely to represent climatological values, so no discernible 3057 signal about moderate or anomalous conditions is identified (Di Giuseppe et al., 2024).

3058 Objective 7: Project future changes in the probability of each focal event under future 3059 climate scenarios.

3060 We projected future changes in the annual likelihood of each focal event under different 3061 climate scenarios. In Canada, we project significant increases in the likelihood of burned areas 3062 over the coming decades, driven by dry conditions and a climate- and CO2-driven increase in 3063 fuel load. Climate-change mitigation efforts, particularly low emissions scenarios such as 3064 SSP126, offer potential to limit future fire extremes. Similarly, projections for Western 3065 Amazonia indicate a notable increase in future burning, primarily attributed to drier fuel trends. 3066 Using a more extreme baseline than the observed event - a 1-in-100 year event under recent 3067 climate, shows that likelihood of extremes increases with the magnitude of the extreme. 3068 Effective mitigation actions are crucial in keeping existing conditions largely unchanged, 3069 emphasising the importance of limiting temperature rise to well below 2.0°C. However, even 3070 with mitigation efforts, an increase in extremes is expected in Canada as global temperatures 3071 continue to rise towards 1.5°C, necessitating robust adaptation strategies to manage and





3072 prepare for the rise in extreme fire events. Meanwhile, Greece's smaller domain size makes 3073 confident projections difficult, given remaining biases and uncertainties in future projections, 3074 highlighting the necessity of improving model projections or incorporating additional 3075 information to constrain uncertainties on smaller scales.

3076 4.2 Frontiers in Observing and Modelling Extreme Fire Occurrence

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As part of this inaugural edition of the State of Wildfires report, we review current capabilities in the observation and modelling of extreme fire. This section will not be revised annually but will be revisited on longer intervals at times that are appropriate given progress in the observational and modelling fields. The review serves to define the current challenges that studies such as our face and identify pathways to overcoming those challenges harnessing new capabilities and technologies.

3085 4.2.1 Defining of Extreme Fire Events

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While the study of impactful fires or unusual fire seasons is critical for understanding changes
in the exposure and vulnerability of people and the environment to fire, a number of challenges
hinder our ability to arrive at a clear and universally-accepted definition of "extreme" event.

3091 Key challenges to the data-oriented identification of extreme events include:

- Lack of Consensus on Quantitative Criteria: There is considerable variability in the
 measurable criteria used to define "extreme fire" across different studies, such as fire
 size and area affected. Additionally, there is no well-defined statistical threshold
 established for what constitutes an outlier event, which is crucial for consistent data
 analysis and classification.
- Geographic Variability: Definitions of what constitutes an extreme fire vary significantly by region, complicating the establishment of a universally accepted definition. The size threshold can range from >100 hectares to >100,000 hectares depending on geographic location, influenced by local fire regimes.
- Evolving Definitions: Over time, the definition of extreme fire has evolved and expanded, leading to ambiguity. This evolution means the term now encompasses a wider range of fire types and behaviours. Climate change is imposing an increasing trend in fire severity in many regions, suggesting that definitions should have the flexibility to evolve.
- Context Dependence: Definitions of extreme fire are often context-dependent, varying with ecosystem types and fire histories. The baseline for defining what is considered extreme is not standardised, such as whether it should be based on fire return intervals or on the quantified damage to ecosystems and societies.

3112 Key challenges to the knowledge-oriented identification of extreme events include:

- Lack of Consensus on Qualitative Criteria: There is significant variability in how criteria such as fire behaviour, resistance to control, and ecosystem and socio-economic impacts are interpreted and integrated into definitions of "extreme fire." This variability reflects differing expert opinions and knowledge bases. The challenges here principally relate to the subjective nature of what constitutes significant impact, whether ecological, economic, or societal, further complicating the establishment of a clear and universally accepted definition.
- Terminological Overlap and Redundancy: The term "extreme fire" overlaps with
 other fire-related terms such as "catastrophic fire" and "megafire", which may also lack
 clear definitions or be used interchangeably, leading to confusion.



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- Influence of Language and Culture: The interpretation and usage of the term can differ across languages and cultures, influencing how extreme fires are defined and reported in the global scientific community.
 - **Societal Influence on Scientific Terminology:** Scientific terminology often evolves in response to its usage in broader societal contexts, such as media or public discourse. As the societal challenges arising from wildfires ultimately motivate scientific inquiry, the language used in scientific communication must be adaptable and responsive to ensure it remains relevant and accessible to non-scientific audiences.
 - **Scientific Rigour and Clarity:** There is a need for a definition that is clear, consistent, and scientifically rigorous to allow for standardised and repeatable measurements across studies. Existing definitions often fail to meet these criteria comprehensively.

3136 3137 Defining the term 'extreme fire' and how to address it requires a significant and inclusive effort 3138 across the fire science community. We actively avoid a strict definition of 'extreme fire' here due to the risk that the wider fire science community rejects our proposed solution, prompting 3139 3140 confusion. Instead, we adopt a broad definition of 'extreme fire' as discussed in Section 2.1.2. 3141 Both of the challenges above can be addressed by developing standardised criteria and 3142 protocols for defining extreme fires (Chu et al., 2023). This task has been approached with 3143 respect to the term "megafire" although it remains to be seen if that intervention will be effective 3144 (Linley et al., 2022). Doing so requires a transdisciplinary approach, including input from the 3145 science community, fire practitioners, legislators, and impacted communities at a minimum 3146 (Shuman et al., 2022).

- 3147 3148 **4.2.2 Observing Extreme Fire Events**
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3150 Global-scale data for characterising extreme fire events are primarily sourced from satellite 3151 observations, notably active fire detections, BA maps, and tracking of smoke plumes. 3152 However, to accurately define how extreme a fire event is, it is crucial to contextualise present-3153 day observations within historical data. Unfortunately, the historical records of satellite-derived 3154 active fire and BA products are relatively short. The longest coherent observations on a global 3155 scale are derived from the MODIS instruments onboard the Aqua and Terra satellites, 3156 launched in 1999 and 2002 respectively. Various global BA products, such as MCD64 product 3157 family (Giglio et al., 2018) and FireCCI51 (Lizundia-Loiola et al., 2020), as well as active fire 3158 data like the MCD14 product family (Giglio et al., 2016), have been generated based on 3159 imagery acquired by MODIS. Although these time-series now span over two decades, they 3160 are still relatively short when compared to the decadal to centurial fire return intervals observed 3161 in many ecosystems.

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3163 Pre-MODIS satellite data, like that from the AVHRR program, exists and provides a continuous 3164 imagery archive from 1982 onwards. Although efforts are ongoing to generate a coherent BA 3165 product from AVHRR data (Otón et al., 2021), there are limits to the global applicability of 3166 these products. For example, unresolved challenges stemming from coherence issues 3167 between imagery from different AVHRR sensors result in artefacts and spurious trends in 3168 various regions worldwide (Giglio and Roy, 2022), although this has been debated by other 3169 authors (Pullabhotla et al., 2023). Other pre-MODIS data available are the Along-Track 3170 Scanning Radiometer (ATSR) and Visible and Infrared Scanner (VIRS) active fire data (Giglio 3171 et al., 2013; Arino et al., 1999), but with important limitations concerning detection sensitivity, 3172 missing data and only allow for the expansion of existing time-series by 3 years (start in 1997). 3173

3174 Efforts are ongoing to extend the MODIS time-series by incorporating active fire data from 3175 ATSR and VIRS with BA data (e.g. Chen et al. 2023). However, due to the different 3176 characteristics of these data, creating a coherent, multi-satellite time-series of active fire data 3177 and/or BA is not straightforward. Concerns also arise with the impending decommissioning of





3178 MODIS-Terra, raising doubts about the continuity of existing long-term fire records. However, 3179 operational satellite sensors such as VIIRS onboard NOAA's series of satellites and SLSTR 3180 onboard the Sentinel-3 satellites offer promising capabilities for medium-resolution BA 3181 mapping (e.g. Román et al., 2024; Lizundia-Loiola et al., 2022). Urgent attention should be 3182 directed towards developing methodologies to integrate these new datasets into a coherent, 3183 long-term BA dataset. Furthermore, advancements in medium-resolution satellite data 3184 availability and revisit times, particularly from Landsat and Sentinel-2, now enable global BA 3185 mapping at spatial resolutions as fine as 20-30m (e.g., Roteta et al., 2021; Chuvieco et al., 3186 2022), suggesting a potential future direction for coherent long-term global BA monitoring. 3187

3188 While BA is a key variable to characterise extreme fire occurrence, multiple other aspects of 3189 extreme fires can be characterised using satellite data. BA products can be used to cluster 3190 burned pixels in burned patches to obtain the number and size of individual fires (Archibald 3191 and Roy, 2009). Furthermore, the daily fire rate-of -spread, length of the active fire line and 3192 spread direction can be extracted from the daily fire expansion (Andela et al., 2019). These 3193 algorithms, such as the Global Fire Atlas used in this study, give global scale, coherent 3194 estimates of patterns and trends in fire number, fire size and rate of spread. However, these 3195 algorithms are sensitive to the temporal accuracy of the per-pixel burned date detection, the 3196 spatial resolution of the BA product, and any errors within each product. Recent advances 3197 focusing on clustering VIIRS thermal anomalies and extracting fire rate of spread, fire 3198 expansion and length of the active fire line show promising results (Andela et al., 2022, 3199 Hantson et al., 2022, Chen et al., 2022) but have so far not been developed globally. Future 3200 development towards a global product should allow for a more detailed characterization of fire 3201 characteristics in near real-time, well-suited for detection and quantification of fire extremes.

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3203 Active fire detections also record the amount of radiation emitted by the fire at the moment of 3204 satellite overpass (Fire Radiative Power, FRP), within the pixel detected by the satellite. While 3205 this information is related to the intensity of the fire, the usage of FRP has been difficult as a 3206 low-intensity fire burning a large extent of the pixel can have a higher FRP than a high-intensity 3207 fire burning a small fraction of a pixel. These complications have limited a more standardised 3208 and operational usage of FRP for quantifying fire extremes. Advances in active fire detection 3209 from higher resolution sensors may allow for a more comprehensive estimate of fire intensity 3210 when combined with FRP estimates from coarse resolution sensors (Schroeder et al., 2014, 3211 2016).

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3213 4.2.3 Predicting Extreme Fire Events

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3215 Since the 1970s, fire predictions have relied on empirical fire behaviour models tailored to 3216 specific ecosystems (Bradshaw et al., 1983; Noble et al., 1980; Stocks et al., 1980; van 3217 Wagner, 1987), becoming pivotal tools for fire management agencies (San-Miguel-Ayanz et 3218 al., 2013). The ease of implementation and the availability of weather data have contributed 3219 significantly to their widespread adoption. However, despite their utility, several studies have 3220 highlighted the limited effectiveness of the FWI and similar metrics in fuel-limited ecosystems, 3221 where fires are driven by the short-term superficial drying of intermittently available biomass 3222 (Yebra et al., 2013). The absence of consideration for actual fuel availability presents a 3223 constraint to the meaningful application of the FWI in savanna-type ecosystems. Beyond 3224 weather conditions, the remaining prerequisites for fire activity-namely, fuel and ignitions-3225 are intricately linked to vegetation state, lightning activity, and human behaviours. Improving 3226 fire forecasts beyond solely considering fire weather could be achievable by accurately 3227 describing these components. This has been widely recognised in the global vegetation-fire 3228 community for several decades (Hantson et al., 2016), and consequently great advances have 3229 been made to address this through the development of fire-enabled DGVMs, as used in this 3230 report. However, explicit representation of these processes introduces biases and instabilities 3231 that, when used in isolation, limits their utilage for assessing climate and human drivers of BA 3232 extremes (Hantson et al., 2020; Burton, Lampe et al., 2023).





3234 The availability of remote observations for fuel, either independently (Yebra et al., 2018) or 3235 supported by modelling frameworks (McNorton and Di Giuseppe, 2024), has demonstrated 3236 potential in aiding the development of new fire models and indices that partially incorporate 3237 fuel considerations into their formulation (Di Giuseppe, 2023, Hantson et al., 2016). However, 3238 it is the emergence of the data-driven revolution that holds the promise of significantly 3239 enhancing our predictive capabilities for extreme fires (McNorton et al., 2024), and has driven 3240 the development of the semi-empirical tools used in this report. Not only can these tools 3241 enhance predictive capability for extreme fires, they also present an opportunity to disentangle 3242 the drivers of the prediction, giving us the capability to address or at least understand the 3243 causes of the event, as demonstrated by the POF and ConFire frameworks used here. The 3244 coupling of FireMIP models with observational data (Burton, Lampe et al., 2023 and used 3245 here), also showcases the potential to bridge the advanced modelling capabilities of FireMIP 3246 with application-specific approaches such as ConFire and POF.

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3248 Despite these technological advancements, widespread adoption is unlikely to occur 3249 suddenly, as there typically exists a delay between the creation of new indices, their 3250 operationalization, and global acceptance by those responsible for fire prevention and control.

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Another emerging element is the recent availability of fire danger predictive systems at the seasonal and subseasonal timescales (Di Giuseppe et al., 2024). Currently, there is limited evidence on how these longer-range tools could contribute to prevention planning and adaptation strategies. While they exhibit minimal skill beyond two months, they may offer valuable pre-seasonal warnings under specific conditions established during important atmospheric modes of variability. Certainly, this aspect will need to be further investigated in the following issues of this report.

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4.2.4 Attributing Extreme Fires to Global Change

3262 The prediction and management of extreme fire events have become increasingly complex 3263 due to the multifaceted impacts of global change. Climate change exacerbates fire risks 3264 through rising temperatures, altered precipitation patterns, and more frequent and severe 3265 droughts, as shown in Canada and NW Amazon in this report. These climatic shifts affect 3266 vegetation productivity, with elevated CO₂ levels potentially increasing biomass and thereby 3267 providing more fuel for fires. Nutrient deposition and other environmental changes influence 3268 ecosystem responses, further altering fire potential. Land use changes and management 3269 practices also significantly influence fire dynamics. For example, human activities such as 3270 deforestation, urban expansion, and agricultural practices can both mitigate and exacerbate 3271 fire risks, with socioeconomic factors shown to have a strong influence on overall extreme fire 3272 likelihood in NW Amazon, and potentially contributing to increases in BA in 2023 in some 3273 areas of Greece. Effective land management strategies, including prescribed burns and forest 3274 thinning, are crucial for reducing fuel loads and minimising fire impacts. While climate driven 3275 estimates of extreme behaviour are plentiful, few modelling frameworks take into account most 3276 of these dynamic factors and their interactions (Rabin et al., 2017).

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We have used model-data fusion techniques that account for these factors in this report, and have been able to attribute some of their influences in certain places. This report utilises semiempirical models that blend empirical data with process-based understanding to better predict fire behaviour. Quantifying uncertainty in these models is essential, especially when dealing with extremes. By generating probability distributions, researchers can better understand the likelihood of various fire scenarios, informing more effective management and policy decisions.

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Thanks to the uncertainty quantification techniques, we have been able to ascertain where we are confident in our attributions. However, uncertainties still remain, many from not considering





3288 the complex interactions and feedback onto fire, some from fire itself as it consumes fuel, and 3289 affects from weather. Coupled vegetation-fire models explicitly represent many of these 3290 feedbacks. However, current FireMIP models struggle to accurately simulate extreme fire events (Hantson et al., 2020). One key factor hampering improvements in model development 3291 3292 is our limited understanding of factors driving fire extinguishers in a natural setting. While much 3293 process based knowledge exists on the factors influencing fire start and fire spread, only 3294 limited knowledge exists on the myriad of factors that can stop a fire, from changes in fuel 3295 moisture, structure and heterogeneity to landscape fragmentation and fire fighting, and how 3296 these interact (e.g. Finney et al., 2012). Without a strong theoretical understanding of these 3297 factors, process based modelling of extremes at a global scale might be limited in the near 3298 future. For the 2023 focal events, we have shown that low fuel loads and variations in human 3299 modification of the landscape can limit fuel spread (Figure 15, Figure 16, Figure S13, Figure 3300 **S14**). However, we only look at a handful of events and further examples are required at larger 3301 scales to inform improvement of process-based rates of spread in fire models. 3302

- 3303 To move forward, we need to combine these concepts in attribution techniques and quantifying 3304 uncertainty with coupled vegetation fire models, such as in FireMIP. Early attempts of this are 3305 promising - ConFire (used in Section 3.3, Section 3.4 and Section 3.5) borrows many of the 3306 FireMIP-style model concepts, though it still lacks many feedbacks from fire itself. We have 3307 also used the latest FireMIP models coupled to an uncertainty framework for broad-scale, 3308 uncertainty-based attribution (obtained from Burton, Lampe et al., 2024). But they struggle at 3309 reproducing the tails of distributions where extreme events are found. Another way to develop 3310 these techniques is to move towards an integrated system that would inform both attribution 3311 and future projections in a seamless way. We make some progress in this direction here using 3312 tools such as ConFire, by using the information gained from fire drivers to build future 3313 projections, however there is more work to do to link statistical approaches for today's fires to 3314 future projections.
- 3315

3316 The human role in driving fire and extremes is hard to represent. Despite the often reported 3317 influences people have on both increasing extreme burning or causing the observed decline 3318 in global BA, the role of humans in the landscape remains hard to capture and on the whole, 3319 remains one of the most uncertain aspects of this report. Agent-based modelling (ABM) is 3320 trying to address this by simulating the behaviours and interactions of individual human entities 3321 (e.g., deforestation, crop residue burning, suppression, etc.) within a given environment (Ford 3322 et al., 2021). This approach provides a dynamic representation of how different factors 3323 contribute to fire risk and links well with subsequent sections of the report. These approaches 3324 could be a major contributor to subsequent issues of this report. However, the integration of 3325 these advanced modelling techniques into operational use faces challenges, as there is often 3326 a delay between the development of new approaches and their widespread adoption by fire 3327 management agencies. This underscores the need for continuous improvement and adoption 3328 of innovative modelling approaches to address the growing threat of extreme fire events 3329 effectively.

3330

3331 In addition to fire weather index (FWI) and BA projections, it is crucial to go beyond these 3332 metrics to consider wider impacts such as intensity, and emissions. Understanding the intensity of fires helps in assessing their destructive potential and the severity of their 3333 3334 ecological and societal impacts. Emissions from fires, including C dioxide and other 3335 greenhouse gases, contribute to climate change and air quality issues. Finally, evaluating the 3336 broader impacts of fires, such as on biodiversity, human health, and economic stability, is 3337 essential for developing comprehensive adaptation and mitigation strategies. Quantifying and 3338 understanding the uncertainty in these projections is crucial for developing adaptive strategies 3339 that can effectively respond to the evolving fire risks posed by global change.

3340

3341 **4.2.5 Modelling Future Projections of Fire Extremes**

3342





3343 Projections of extreme fire events under future climate scenarios indicate a significant increase 3344 in their frequency and intensity. Semi-empirical models used in this report project that extreme 3345 BA events, currently rare, are likely to become more common by the end of the century. These 3346 projections highlight the urgent need for robust fire management strategies and policies to 3347 mitigate the impacts of these increasingly severe fire events on ecosystems, communities, 3348 and global C dynamics. Quantifying and understanding the uncertainty in these projections is 3349 crucial for developing adaptive strategies that can effectively respond to the evolving fire risks 3350 posed by global change. In our ConFire uncertainty quantification framework, we have been 3351 able to make some confident inferences about the potential state of wildfires in the coming 3352 decades. However, we have also identified that there is a significant amount of crucial 3353 information that is currently beyond our reach due to the uncertainties involved. Our ability to 3354 forecast for the upcoming season, as well as for the next 2-3 decades, requires further 3355 refinement as we are observing mixed and uncertain responses. Beyond that, we still show 3356 similar uncertainties in responses of Canada and Western Amazonia under different scenarios 3357 as highlighted by UNEP (2022), which uses the previous generation of climate models - shows 3358 a large overlap in the potential range of changes in the occurrence of fire extremes between 3359 SSP370 and SSP585. This does not imply that mitigation efforts for one scenario will be 3360 ineffective compared to another, but rather indicates a lack of understanding regarding the 3361 response of extremes to these scenarios. By narrowing down the uncertainty ranges, we can 3362 better target adaptation efforts and evaluate the effectiveness of mitigation strategies. The 3363 reduced likelihood of extreme event recurrence in our high mitigation, however, does show 3364 that we can start separating out how mitigation efforts might affect fire extremism, though not 3365 in the level of detail needed for policy.

3366

3367 There are three main ways we may be able to constrain uncertainties in the coming reports. 3368 The first is development of the underlying GCMs that project future change in the drivers of 3369 BA. For individual models, this is a slow process and, beyond informing CMIP model 3370 development, is outside what the State of Wildfires can contribute to. However, bringing in more models, including having another model to incorporate any remaining biases in 3371 3372 simulated fire from the correct models into our uncertainty projection, will help us constrain 3373 uncertainties more (Kelley et al., 2023) The second is obtaining more information and understanding of how fire drivers relate to fire extreme's as outlined in the previous section. 3374

3375

3376 Better ways of describing the statistical relationship between observed and modelled climate. 3377 land surface and fire today is a third approach. Investigating the dynamical climatic drivers of 3378 extreme fire conditions in different regions can help to physically disentangle and potentially 3379 constrain sources of uncertainty in future climate projections, for example by constructing 3380 physical storylines (Shepherd et al., 2018; Mindlin et al., 2023). These storylines of plausible 3381 future change, or other similar approaches to quantify and explain uncertainty in projections, 3382 provide critical information for communities to develop robust adaptation strategies (Lemos et 3383 al., 2012) and prepare for future losses and damage caused by evolving fire risks posed by 3384 global change. Next to understanding future uncertainty, further insights into these dynamical drivers can support the development of improved physics-informed bias adjustment of climate 3385 3386 models (Maraun et al. 2017). Currently available methods to bias adjust climate models for 3387 their use in fire models, such as the ISIMIP3BASD method, have been shown to modify the 3388 climate change trend, particularly in extreme threshold indices (Casanueva et al 2020, Spuler 3389 et al 2024) or increase spread in climate model projections (Lafferty and Sriver, 2023). Bias 3390 adjustment methods should therefore be evaluated carefully and leave scope for future 3391 method development that physically links present-day biases to future uncertainty. 3392

3393 4.3 Frontiers in Studying Extreme Fire Impacts

3394





As above, this section will not be revised annually but will be revisited on longer intervals at times consistent with progress in the fields of extreme fire impact assessment. This section identifies current challenges and emerging opportunities for overcoming them.

3398

3399 **4.3.1 Direct Exposure of People and the Built Environment** 3400

3401 The wildland urban interface (WUI) has been the focus of much of the direct exposure of 3402 populations to fire (i.e., populations residing within the footprint of a fire). While extreme fires 3403 can become urban conflagrations - including the Lahaina fire in August 2023 - there has been 3404 a focus on reducing fire losses in WUI through fire prevention, fuel reduction, and other 3405 mitigation efforts (Calkin et al., 2023). The dual roles of increased population living in the WUI in some areas (Radeloff et al., 2018) and increased fire potential has contributed to 3406 3407 accelerated community impacts (Higuera et al., 2023). Recent studies in the US have shown 3408 a doubling in the direct population exposure to large fires during 2000-2019 (Modaresi Rad et 3409 al., 2023), yet with a majority of the increase attributable to fires encroaching on the WUI rather 3410 than recent increases in human settlement in such landscapes. They showed similar increases 3411 in wildfires intersecting roads and energy infrastructure, thereby causing additional 3412 complexities to transportation and energy reliability. In the US, most of the structures lost in 3413 fires occurred in grasslands and shrublands, rather than forested environments, underscoring 3414 the importance of looking beyond traditional forest-centric fire risk assessments. Recently 3415 developed efforts to map the WUI at global scales (Schug et al., 2023) present opportunities 3416 for identifying areas most vulnerable to direct fire impacts, and characterise trends in fire 3417 exposure more broadly (Chen et al., 2024; Tang et al., 2024).

3418

3419 Moreover, efforts are needed to better understand characteristics of fires that result in direct 3420 exposure of people, loss of structures, and loss of life. For example, Abatzoglou et al. (2023) 3421 showed that fires coincident with strong downslope winds in the western US were responsible 3422 for a majority of structure loss and fatalities during 1999-2020 despite accounting for only 12% 3423 of all fires and burned areas. Downslope wind driven fires defy typical terrain-driven fire 3424 behaviour by pushing fire downhill from wildlands and into communities in the WUI - often 3425 overwhelming both fire suppression efforts and fuel treatments. During the 2023 fire year, 3426 many of the large loss fires in Hawaii, Greece, and Chile were driven by such conditions 3427 (Synolakis and Karagiannis, 2024). Focused efforts to diagnose characteristics accompanying 3428 other extreme fires, including meteorological conditions such as dry cold frontal passage (Van 3429 Wagtendonk, 2006) and pyrocumulonimbus (Lareau et al., 2018), as well as pre-existing 3430 vegetation conditions such as extensive tree mortality (Stephens et al., 2022), can lead to 3431 additional insight on fires most likely to result in impacts. Highlighting geographic areas prone 3432 to such conditions and how they may change into the future can help prioritise mitigation 3433 efforts.

3434

3435 One of the current gaps inhibiting the characterisation of extreme fires and understanding 3436 drivers of fire impacts on humans, such as fatalities, evacuations, structure loss, secondary 3437 morbidity from toxic smoke, economic losses, and impacts to food, water, energy, and 3438 transportation systems, is the lack of comprehensive data at national-to-global scales 3439 quantifying such outcomes. For example, wildfire morbidity data are collected in only a few 3440 countries due to the relatively infrequent occurrence of wildfires that incur fatalities and the 3441 lack of clear responsibility for cataloguing such information, particularly for civilians (Haynes 3442 et al. 2019). Estimates of smoke-induced morbidities rely on identifying spikes in hospital visits 3443 above a baseline and then extrapolating, nearly always associated with a specific event 3444 (Johnston et al. 2021). The US state of California pioneered a systematic cataloguing of 3445 structure losses to wildfire beginning in 2013 (California Department of Forestry and Fire 3446 Protection, 2024). However, this initiative required a commitment of fiscal and personnel 3447 resources from the state wildfire agency and has not been adopted elsewhere in the US, 3448 despite the programme's success in yielding substantial additional insights about wildfire 3449 structure loss mechanisms (Kolden and Henson, 2019; Syphard and Keeley, 2019). Canada





now catalogues wildfire evacuations, but there is considerable complexity in characterising
initiation responsibility, duration, and number of people impacted (Beverly and Bothell, 2011).
Global insurance companies document insured losses as one mechanism of economic
impacts from wildfires, but such records are neither representative of the broader losses due
to high rates of uninsured or under-insured property ownership (Hazra and Gallagher, 2022)
nor are they publicly available from private companies, leaving researchers to attempt
imperfect estimations from websites that model property prices (e.g., Conlisk et al., 2024).

3457

3458 While the global disaster database EM-DAT and others like it have attempted to rectify this 3459 gap in recent years, such data often overgeneralize both the spatial extent and the impacts of 3460 wildfire disasters (e.g., reporting all of the fires across a country for an entire summer as a 3461 single event) because they rely on news reporting that varies widely in accuracy, rather than 3462 official record-keeping. Expanding and improving quantification of wildfire impacts on humans 3463 is critical to overcoming the BA fallacy that fire size is a proxy for impact and aids in developing 3464 models that can point to the most effective mitigation solutions (Kolden, 2020). This is 3465 challenging for many metrics, but remote-sensing based documenting of structure loss and 3466 fire incursion into the WUI is now possible with the availability of high-resolution sensor arrays 3467 (Frazier and Hemingway, 2021), and the widespread distribution of air quality monitoring 3468 sensors should enable an increase in more interdisciplinary research on relationships between 3469 wildfire smoke events and medical morbidity outcomes (Liang et al., 2021).

3470

3471 4.3.2 Air Quality and Health Impacts

3472

3473 Exposure to outdoor pollution is recognized as a major global risk to human health (Murray et 3474 al., 2020). The abundance of fine-sized particles with a diameter less than 2.5 µm (PM2.5) are 3475 of particular concern due to their ability to cause a range of cardiovascular diseases. A growing 3476 concern is the contribution of fire smoke to air quality, not only due to increasingly large and 3477 severe fire events across the globe degrading air quality (Figure 7) but also because fire has 3478 been suggested to be more toxic per unit of PM exposure than other pollution sources 3479 (Aguilera et al., 2021). The World Health Organization (WHO) sets global Air Quality Guidelines 3480 on the threshold for PM2.5 exposure. Thresholds are continually revised and in September 2021 3481 they reduced guidelines for an annual mean exposure limit for PM 2.5 from 10 µg/m3 to 5 µg/m3. 3482 Given that 95% of the world's population is estimated to be exposed to annual mean PM2.5 3483 concentrations of at least 10 µg/m3 (Shaddick et al., 2018), additional pollution exposure due to 3484 increasing or severe wildfire pollution events creates an elevated health risk that also goes beyond 3485 both current and past guideline limits. 3486

Here we discuss three of the major challenges in quantifying the impact of extreme fire pollution on human health. The first is accurately measuring the amount of pollution that a wide variety of communities are exposed to and then attributing the contribution of a wildfire event, that could be 100s or 1000s of km away, to the measured concentration. The second is that PM2.5 is not the only pollutant of concern, the EPA regulates six pollutants of concern for American citizens and a wildfire produces them all. The third is accurately linking exposure to a wide range of pollutants to their associated short-term and long-term health impacts.

3494

3495 Tools to assess air quality primarily consist of ground-based measurements and modelling. 3496 Ground based observations provide an accurate measurement of pollution at their location. 3497 However, measurement locations are spatially sparse. Ongoing efforts to increase spatial 3498 coverage include deployment of small relatively affordable particulate matter sensors, such as 3499 the PurpleAir network, by a wide range of communities (not just scientists), and efforts to relate 3500 surface PM concentrations to measured aerosol optical depth from satellites (Li et al., 2021). 3501 One additional constraint of observations is that they cannot differentiate pollution sources, 3502 but modelling can. Dispersion modelling uses emission estimates, reanalysis meteorology, 3503 and topography to provide estimates of ambient pollutant concentrations at varying 3504 spatiotemporal scales.





3505

3506 A challenge for models currently lies in the large uncertainty in fire emissions (Reddington et 3507 al., 2016; Carter et al., 2020; Pan et al., 2020). Emission data will therefore require calibration 3508 against observations and adjusting before the contribution of fire to pollution can be quantified. 3509 Improved emission datasets will increase confidence in these assessments. Given the 3510 complexity of smoke composition, other pollutants, such as gases and metals, are also present 3511 and have deleterious health implications. Increased study of the chemical composition of fire 3512 smoke plumes is therefore needed alongside monitoring of short and long range transport of 3513 these pollutants.

3514

3515 Environmental and personal factors both influence cardiovascular health, making it 3516 challenging to isolate the effects of fire smoke. Impacts may also not be immediate; some 3517 effects can be acute, such as exacerbation of asthma, while others emerge over a longer 3518 period, like the development of cardiovascular disease, and are thus much more difficult to 3519 directly connect to a specific extreme fire event. Conducting epidemiological studies that link 3520 fire smoke exposure to specific health outcomes requires comprehensive data collection and 3521 follow-up. These studies are resource-intensive, time consuming, and subject to potential 3522 limitations in data. In the shorter-term development of effective response planning that 3523 includes coordinated efforts between public health officials, environmental agencies, and 3524 emergency services could help in mitigating some of the health impacts of fire smoke.

3525

3526 **4.3.3** Impacts on Indigenous and Traditional Communities

3527

3528 Indigenous peoples and local communities (IP&LC) are disproportionately exposed to extreme 3529 fire impacts because of their proximity to the land and resources from which their cultures, 3530 livelihoods and often food and medicines derive. Once landscapes are degraded through fire 3531 the access and abundance of various resources can be shifted. At the same time these 3532 communities are often less supported by the state due to access and their political and 3533 economic marginalisation linked to systemic socioeconomic disadvantages. These 3534 communities suffer not only post-fire impacts, but can be disincentivized from particular land 3535 and resource use activities because of the increasing threat of forthcoming wildfires. Whilst 3536 the multiple important values (e.g. instrumental, intrinsic and relational) associated with 3537 landscapes by IP&LCs are threatened by fire, there is a lack of systematic pre- and post-fire 3538 assessment of these impacts (van Leeuwen and Miller-Sabbioni, 2023).

3539

3540 Historically fire governance has added an additional burden to IP&LC, often prohibiting cultural 3541 fire use and management to the detriment of local knowledge and values, and in some cases 3542 also increasing the propensity of wildfire in tropical systems as well as savannahs (Carmenta 3543 et al., 2019; Daeli et al., 2021; Croker et al., 2023). In some contexts there is a shift towards 3544 correcting these issues and renewed interest in and support for cultural burning and 3545 Indigenous approaches to land management. For example, integrated fire management is 3546 gaining traction globally and sits at the heart of a number of interventions and international 3547 policy efforts (e.g. Fire Hub, 2023). The premise of IFM is to maximise the 'good' fire and 3548 minimise the occurrence of wildfire often through an approach of connecting knowledges (i.e. 3549 expert and place-based or Indigenous). Whilst nations sit at different stages of development 3550 in respect to IFM, and many IP&LC feel there is a long way to go, the growing interest is 3551 promising (Bilbao et al., 2019; Luque et al., 2020; Rodríguez-Trejo et al., 2022). Research is 3552 needed to better understand the effectiveness of IFM and what mechanisms and processes 3553 work best for rebalancing the influence of various forms of knowledge on fire management. 3554 For instance in North America where fire use is considered a form of medicine to the land, and 3555 anointed to particular patches of the landscape for care (Palmer, 2021). These approaches 3556 are perceived as potentially more just as they allow for the many meanings and uses of fire to exist and persist. For example, In Australia, the term 'Country' is used to convey the cultural 3557 3558 and spiritual connection of Aboriginal peoples to the land and water in specific regions. This





link profoundly colours Indigenous peoples' experience of extreme fire events such as the
 Black Summer fires of 2019-20 (Nolan et al., 2021b).

3561

3562 **4.3.4 Economic Impacts** 3563

3564 Extreme wildfires cause economic disturbance worldwide, with the nature of their impacts 3565 varying across regions due to distinct economic structures, environmental conditions, and 3566 mitigation and response strategies. These economic impacts arise through a myriad of 3567 channels including property and infrastructure loss, business downtimes, supply chain 3568 disruptions, decreased tourism revenues, health costs, reduced economic productivity, and 3569 damages to ecosystem services. While tangible costs (e.g., insured losses such as property 3570 or infrastructure) are relatively straightforward to measure, intangible costs (e.g., lives lost, 3571 shortened lifespans and impaired performance/productivity due to smoke exposure, damage 3572 to species and habitats) are more challenging to quantify owing to issues of data availability 3573 and their occurrence over varying temporal and spatial (geopolitical) scales. Consequently, 3574 assessing the true economic costs of an extreme wildfire event or season is inherently challenging. Furthermore, while wildfires are predominantly destructive, certain economic 3575 3576 sectors may experience positive effects such as those arising from reconstruction and 3577 rehabilitation in the aftermath, or even from the fire suppression efforts during the event 3578 (Nielsen-Pincus et al., 2014, Meier et al., 2023a).

3579

3580 Building upon these challenges, existing work on quantifying the economic impacts of wildfires 3581 has largely focussed on developed economies in Europe (World Bank, 2023), the US, and 3582 Australia. For example, modelling extreme wildfires in Mediterranean Europe, Meier et al. 3583 (2023b) estimate that economic losses caused by a single 1-in-10 year extreme wildfire event 3584 amount to €162–439 million in Portugal, €81–219 million in Spain, €41–290 million in Greece, 3585 and €18–78 million in Italy. The unprecedented 2018 and 2020 wildfire seasons in California 3586 are estimated to have resulted in approximately \$150 billion (including indirect costs beyond 3587 the immediate burn zone) and \$19 billion in economic damages, respectively (Wang et al., 3588 2021; Safford et al. 2022). Similarly, Australia's 2019-2020 fire season has been documented 3589 as the country's costliest natural disaster, with economic damages, measured as a decrease 3590 in GDP, estimated at \$10 billion (Wittwer and Waschik, 2021).

3591

While proxies of economic data can be publicly accessible from satellite data shortly after events, this is often not the case for more traditional economic indicators such as sectoral GDP (or alternative wealth measures), employment, hospitalisations, fire prevention and suppression spending, house prices, or tourism revenues. These data are often not publicly available, not harmonised across larger geographic areas, or unable to capture medium and long-term effects. Consequently, the full economic costs of such events may only become apparent years, or even decades, after they occur.

3599

3600 In addition to these data challenges, the econometric analysis of wildfires is arguably more 3601 challenging than for other natural hazards such as earthquakes or hurricanes, because wildfire 3602 occurrence is largely shaped by human activity through land-use choices, fire prevention and 3603 suppression policies, and socioeconomic factors. While earthquakes can be seen as non-3604 human induced and therefore it is easier to find a counterfactual for a causal analysis, wildfire 3605 occurrence correlates with many economic outcome variables of interest. Despite these 3606 drawbacks, carefully constructed counterfactual analyses or other econometric approaches, 3607 such as instrumental variables, can provide reliable causal estimates of different aspects of 3608 economic impact estimations of extreme wildfires. These methods hold significant promise for 3609 future research, offering the potential to better understand and mitigate the economic 3610 consequences of wildfires through more accurate and comprehensive analyses, while also helping to target suppression policies and allocate resources effectively. 3611

3612





3613 4.3.5 Loss of Biodiversity, Ecosystem Function and Carbon Storage

3614

3615 The reshaping of fire regimes and ecosystem functioning by climate change and altered 3616 ignition and suppression patterns has a range of consequences for biodiversity, ecosystem 3617 function, and ecosystem services such as C storage. Mechanisms such as reduced seed 3618 quantity or quality, resprouting exhaustion, organic soil burning, and misalignment of key 3619 reproductive processes with optimal conditions under intensified fire regimes can hamper 3620 post-fire regeneration (Burell et al., 2022; Nolan et al., 2021a; Johnstone et al., 2016), 3621 Interaction with other disturbance mechanisms (e.g. drought and insect infestation) can further 3622 compound these effects. For example, in boreal forests, rising fire activity has reduced the 3623 prevalence of fire-adapted species like black spruce, with more frequent and severe fires 3624 hindering their regeneration and favouring deciduous species such as birch and aspen, thus 3625 altering forest structure and C dynamics (Baltzer et al., 2021). In Western North America, the 3626 growing frequency of high-severity fires coupled with warmer and drier post-fire conditions has 3627 been linked with poor forest recovery and transition to non-forest vegetation types such as 3628 shrublands or grasslands, impacting ecosystem services like habitat provision, water 3629 regulation, and C storage (Coop et al., 2020). Likewise, in transitional systems where tropical 3630 savannas and tropical forests can exist under the same climatic conditions depending on fire 3631 history, increased frequency can favour savannah establishment and reduced frequency can 3632 promote forest establishment (Staver et al., 2011).

3633

3634 There are a range of significant challenges to observing and modelling the impacts of shifting 3635 fire regimes on ecosystems and C storage. Fire regimes vary widely, leading to 3636 inconsistencies in historical data and complicating model development. Short-term 3637 observations often miss long-term trends, especially in areas with episodic recruitment (Nolan 3638 et al., 2021a; Coop et al., 2020). Ecosystem responses are complex, with species-specific 3639 reactions influenced by fire-adaptive traits and genetic variability (Grau-Andrés et al., 2024). 3640 Fire regimes also respond to multiple disturbances that can be difficult to disentangle, such as 3641 climate change and extreme events alongside land use change or fire suppression (Nolan et 3642 al., 2021a; Coop et al., 2020).

3643

3644 To overcome these challenges, researchers are leveraging new technologies. Advanced 3645 remote sensing, including satellite imagery and drones, provides detailed data on fire extent 3646 and severity, enabling real-time and long-term monitoring of outcomes (Burell et al., 2022; 3647 Baltzer et al., 2021). Improved modelling techniques, such as process-based simulation 3648 models and the integration of big data with machine learning, enhance predictive capabilities 3649 (Coop et al., 2020; Nolan et al., 2021a). Enhanced field data collection through long-term 3650 monitoring networks and citizen science initiatives also contributes to comprehensive datasets 3651 (Baltzer et al., 2021; Coop et al., 2020). Other opportunities include advances in genomic tools 3652 for understanding species-specific fire adaptations, climate-adaptive management 3653 frameworks incorporating real-time data and predictive models, and the use of AI and machine 3654 learning for better fire impact predictions (Nolan et al., 2021a; Coop et al., 2020; Grau-Andrés 3655 et al., 2024).

3656

3657 A good example of emerging opportunities stemming from the assemblage of large datasets 3658 and application of AI is the comprehensive meta-analysis by Grau-Andrés et al. (2024). By 3659 analysing data from published studies, the meta-analysis detected systematic tendencies for 3660 intensified fire regimes to reduce plant abundance, diversity, and overall health globally, with 3661 stronger effects from increased fire severity than frequency. Forests, particularly conifer and 3662 mixed forests, are more negatively impacted than open-canopy ecosystems. Woody plants 3663 are more susceptible than herbs to frequent fires due to their slower growth and recovery 3664 rates. These impacts are more severe in ecosystems with arid and cold climates, where plant 3665 recovery is further hindered by limited water and lower temperatures. Additionally, step-3666 changes from surface to crown fires in forests can significantly reduce plant abundance, 3667 diversity, and health.





3668

3669 4.3.6 Nature-based Solutions and Net Zero 3670

3671 Terrestrial ecosystems remove about 30% of annual anthropogenic C emissions through 3672 enhanced growth and ecosystem recovery, but this is partly offset by land use change that 3673 increases annual C emissions by about 12% (Friedlingstein et al., 2023). Emissions removal 3674 or avoidance through nature-based solutions has therefore emerged as a key component of 3675 regional, national and corporate net zero strategies (Seddon et al., 2020). Forestry projects, 3676 focused on planting, restoration, or conservation of forests, are one sector of nature based 3677 solutions that are at particular risk of fire, which can result in inaccurate C accounting and 3678 reversal risks. Spatial clustering of C projects in specific geographic areas may compound risk 3679 from climate variability or change, like the impacts of El Niño. In C markets, projects often 3680 account for reversal risks by contributing a proportion of their credits to a buffer pool. 3681 Nevertheless, such allocations may not be adequate in regions exposed to increasing wildfire 3682 extremes (Badgley et al., 2022; Anderegg et al., 2024).

3683

3684 Despite these challenges, C markets also provide opportunities to unlock the resources 3685 required to improve management of fire prone ecosystems, like savannas (Russel-Smith et 3686 al., 2015) and temperate forests (Nikolakis et al., 2022) with benefits for ecosystems, climate, 3687 and local communities. For these markets to be effective and scale, they require accurate and 3688 transparent monitoring systems, enabling science based fire management, accurate 3689 accounting of C losses from fire, and assessment of reversal risk from fire. Novel satellite 3690 based monitoring systems can provide early warning and response to wildfire activity and 3691 improved estimates of ecosystem C stocks. Nevertheless, field based studies remain critical 3692 for understanding both immediate and long-term impacts of fire on forest ecosystems (Silva 3693 et al., 2020) and novel models forecasting evolving fire risk at decadal time scales are required 3694 to improve management strategies and making credible C claims.

3695

3696 4.3.7 Water Quality and other Aquatic Impacts

Fire impacts freshwater ecosystems mainly through (i) the loss of vegetation and litter cover and (ii) the enhanced input of soil, sediment and residual ash. The former leads to a reduction in rainfall interception and evapotranspiration from plants, and thus a greater proportion of rainfall reaching streams, lakes and reservoirs (Smith et al., 2011). This increased runoff from burned hillslopes can lead to accelerated erosion, debris flows and localised flooding (Shakesby and Doerr, 2006).

3703 Wildfire ash is typically enriched in nutrients and contaminants (e.g. metals, polycyclic 3704 aromatic hydrocarbons) compared to vegetation and soil (Bodi et al., 2014; Sánchez-García 3705 et al., 2023) and together with soil and organic debris can directly and indirectly affect water 3706 quality. These include enhanced turbidity, temperature, nutrient and toxin content, and 3707 decreased dissolved oxygen. Associated effects can be increased mortality in freshwater 3708 ecosystems, algal blooms, as well as water quality impacts for water supply catchments (Smith 3709 et al., 2011). For example, increased organic matter in water can be costly for water treatment 3710 plants to remove. The 2016 Horse River Fire affecting Fort McMurray, Canada, caused \$9 M 3711 in additional water treatment expenditures (Pomeroy et al., 2019). The magnitude of such 3712 impacts will depend on the type of ecosystem burned (including the presence of any legacy 3713 contamination), the size and severity of the fire, the rate of vegetation recovery and critically 3714 also the intensity and patterns of rainfall events following the fire (Shakesby and Doerr, 2006; 3715 Nunes et al., 2018). The direct hydrological impacts of fire are relatively well understood, and 3716 models have been developed to support catchment managers in assessing risks to water 3717 supply (Nunes et al., 2018; Neris et al., 2021), however, given the wide range of potential post-3718 fire rainfall events, large uncertainties remain in predicting actual outcomes.





3719 Fire emissions to the atmosphere also contain compounds that can be either nutrient-enriching 3720 or toxic in aquatic ecosystems at high concentrations (Hamilton et al. 2022; Perron et al., 3721 2022). While the impact of 2023 fire events on marine ecosystems has not yet been 3722 extensively evaluated in the literature, past extreme fire events were linked to disturbance of 3723 the open ocean productivity, sometimes thousands of kilometres down the fire plume pathway. 3724 In the Northern Hemisphere large fires across Siberian forests and peatlands during the 3725 summer of 2014 were suggested to contribute significant reactive nitrogen inputs to the 3726 surface Arctic ocean, helping boost and sustain an unusually large phytoplankton bloom 3727 (Ardyna et al., 2022). A similar increase in East Siberian Sea productivity (by over 200%) was 3728 attributed to direct nutrient (nitrogen) deposition and black carbon-induced fast-ice melting 3729 from increased fires on the Eurasian continent in 2019-2020 (Seok et al., 2024). Coastal 3730 community shifts were observed in the Santa Barbara Channel (California, USA) during the 3731 2017 record-breaking Thomas fires (Kramer et al., 2020; Ladd et al., 2023). In the Southern 3732 Hemisphere, the unprecedented Black Summer fires which occurred in southeastern Australia 3733 in 2019/2020 fueled a continent-sized phytoplankton bloom in the iron-depleted southern 3734 Pacific Ocean (Tang et al., 2021). The C uptaken and sequestered to ocean depths by this 3735 bloom was reported to have offset the C emitted by the fire itself (Tang et al., 2021), although 3736 this enhanced oceanic C uptake is likely to be temporary as changes in oceanic productivity and C sequestration efficiency, as a function of wildfire nutrient inputs, fluctuates each fire 3737 3738 season (Hamilton et al., 2022; Wang et al., 2022).

- 3739
- 3740 4.4 Roadmap for Advancing the State of Wildfires Report
- 3741

The fire science community is currently navigating several research frontiers to support better societal and environmental outcomes related to wildfire. By addressing these frontiers, the field aims to provide new observations and tools that enhance preparedness, response, mitigation, and adaptation strategies. The field is advancing its ability to observe individual fires, assess conditions leading to extreme fires, and predict their occurrence on timescales ranging from hours to decades. Additionally, there is increasing focus on monitoring and modelling the diverse impacts of extreme fires on society, the environment, and the economy.

The State of Wildfires report presents key opportunities to harness new advances and
highlight key priorities. Here, we outline the roadmap for the State of Wildfires report in
facilitating improvements in the observation, prediction, and modelling of extreme fires and
their impacts.

- 3755 4.4.1 Observability
- 3756 3757

3754

3757 The State of Wildfires report will advocate for and utilise new harmonised fire 3758 observation products.

3759

3760 Consistent, long-term records of fire extent and properties are fundamental for studying 3761 extremes, which cannot be characterised without reference to historical ranges. On a global 3762 scale, the MODIS instrument has provided valuable data on BA and active fires, underpinning 3763 key advances in tracking daily fire progression and fire emissions models. Over two decades, 3764 MODIS data has enabled the detection of trends and patterns in fire activity and properties. 3765 However, the consistency of these records is now at risk as the Terra satellite, which houses 3766 one of the two MODIS sensors, nears decommissioning.

3767

While fire observations from other moderate-resolution datasets (such as VIIRS) and highresolution datasets (e.g. Landsat and Sentinel sensors) are increasingly available, there is an urgent need to harmonise these records with MODIS data to ensure consistent observation across datasets. The State of Wildfires report further underscores the critical strategic need for a continuous and harmonised dataset of fire observations beyond the MODIS era. To





3773 support future iterations of the report, we will advocate for the provision of harmonised 3774 products within the Earth Observations communities.

3775 3776

3778

The State of Wildfires Report will facilitate a community effort on a protocol for defining 3777 extreme fire events or fire seasons.

3779 This report emphasises important issues in the identification of fire extremes, consistent with 3780 those raised elsewhere in similar contexts such as the definition of the term 'megafire' (Linley 3781 et al., 2022). In future years, regional experts would benefit from a protocol or guidelines that 3782 can be used for categorising extreme fire events. To support future iterations of the State of 3783 Wildfires report, we will coordinate workshops with broader sections of the fire science 3784 community with the aim to produce guidance for future years. Central to this task is the 3785 inclusion of communities from broad geographies so that any output respects fire impacts that 3786 are considered to be regionally significant.

3787

3788 The State of Wildfires Report will stimulate progress on combining multiple fire 3789 observation streams to better identify and characterise extreme fire.

3790

3791 This report highlights the need to advance our capacity to observe fires that are impactful in 3792 diverse ways at large scales. In particular, there is a growing need to move 'beyond burned 3793 area' and towards a wider set of intensity, severity and behaviour metrics that often relate 3794 more strongly to impacts on society and the environment than BA. The integration of individual 3795 fire data from the Global Fire Atlas in this iteration of the State of Wildfires Report is one 3796 example of including wider fire parameters such as size and rate of growth, which are 3797 expected to correlate more strongly. Applications of the dataset could be explored further; for 3798 example, with the Global Fire Atlas it would be feasible to assess the occurrence of days with 3799 a large number of synchronous large fires which are known to present among the greatest 3800 challenges to fire management and suppression attempts (e.g. Abatzoglou et al., 2021). 3801 Impactful fires might also be better identified if their intersection with exposure layers, such as 3802 population centres (e.g. Modaresi Rad et al., 2023), is considered in such assessments. 3803 Similarly, combining datasets that describe individual fire behaviour with observations of fire 3804 radiative power and estimates of biomass combustion from data-driven models might further 3805 inform the identification of intense or severe fires that are often tied to the greatest 3806 consequences for ecosystems and society (e.g. Nolan et al., 2021a). The State of Wildfires 3807 report must stimulate progress on moving 'beyond burned area' and combining diverse 3808 observational capacities to better identify and characterise extreme fire events, and we intend 3809 to expand our use of such insights in future iterations.

3810

3811 4.4.2 Predictability

3812 3813 The State of Wildfire will advocate for the use of extended range forecast to identify 3814 early onset of fire weather conditions

3815 Currently, global systems for monitoring fire danger primarily rely on short to medium-range 3816 weather forecasts, typically up to 10 days. However, evidence suggests that anomalous 3817 conditions conducive to fire weather can be confidently predicted up to one month in advance 3818 using state-of-the-art seasonal forecasting systems. In certain regions, this prediction window 3819 can extend to two months. Generally, beyond this period, forecasts do not outperform 3820 climatology.

3821 However, an extended predictability window of up to 6-7 months is possible when anomalous 3822 fire weather results from large-scale phenomena such as the El Niño-Southern Oscillation 3823 (ENSO) and the Indian Ocean Dipole (IOD), which have teleconnections affecting regions like 3824 Indonesia and Australia.





The State of Wildfires report further emphasises the benefits of extending the forecast range to the subseasonal to seasonal timescale. This extension provides an outlook for the establishment of anomalous fire weather conditions, offering valuable advance warning and preparation time.

The State of Wildfire will stimulate progress on the use of AI and informatics methodsto aid the forecast of fire activity globally.

3831 Despite the increasing importance of fire management, prediction systems still rely on 3832 methodologies developed (and ongoing) since the 1960s, utilising empirical models of 3833 landscape flammability tailored to specific biomes. The Fire Weather Index (FWI), widely 3834 employed by fire management agencies due to its ease of implementation, has limited skills 3835 in fuel-limited ecosystems like savannahs. Tropical savannahs globally account for over two-3836 thirds of mean annual BA (Jones et al., 2024). Additionally, while FWI metrics express the 3837 potential for various aspects of fire behaviour if ignited, they do not consider human or natural 3838 sources of ignition, making their correlation with fire activity limited in many parts of the planet.

Recently, there has been a notable surge in the potential of data-driven applications. Employing machine learning techniques to accurately predict fire occurrence globally at high resolution, up to ten days in advance, could become a reality. Data-driven methods could attribute a probability of fire occurrence with increasing precision, learning from available data where and when human behaviour will trigger an ignition. A notable advantage of these techniques is the facility to ingest novel input data, such as road density, without needing to directly explore the complex physical relationships that govern the occurrence of fires.

3846 While not technically AI, ConFire uses Bayesian statistics within an informatics framework to 3847 combine a data-driven approach with process-based modelling, advancing our understanding 3848 of uncertainty around fire drivers and future projections. Current ongoing developments in 3849 ConFire, aimed at a more flexible modelling structure, could be utilised in future reports and 3850 would further reduce uncertainty, providing higher accuracy in predicting fire event drivers. For 3851 example, the controls in ConFire are currently assumed to be linearly related to burned area, 3852 and this is in the process of being updated to more accurate relationship curves that are 3853 informed and optimised by the model, based on observed data (Barbosa 2024). Another 3854 development in progress is updating the framework to enable one probability distribution to be 3855 applied to both the mean or the tails of the distribution of burned area, instead of having to 3856 select between them as is currently the case.

3857 A priority for these developments is obtaining more data, including near real time 3858 counterfactual data, improved fuel information, and more detailed data on human/fire 3859 interactions, especially over deforestation areas such as Amazonia (See Evaluation section 3860 in the Supplement).

More generally, we also need to find better ways to represent extremes in the data, which includes an improved understanding of what classes as an extreme. A data-driven approach could be used to inform this; for example, machine-learning could be applied to large observation datasets to find any clustering of 'normal' fires, and pick out unusual fires as extremes. This would still need to be used in combination with expert solicitation though, to understand where fires have had extreme impacts on the ground that may not show up in the data.

3868 The State of Wildfires report underscores the benefits of exploring the emergence of these 3869 new technologies to understand their potential and assess prediction skills.





3870 **4.4.3 Attribution**

3871 Fire attribution techniques are still relatively new compared with more established approaches 3872 for extremes such as heatwaves. Part of the challenge in attributing fire is that it is a complex 3873 hazard comprising multiple compound risks across both meteorological and human drivers, 3874 as shown in this report. A wide variety of techniques therefore exist across the attribution 3875 literature to represent fire, from fire weather indices, drought indices, and fire weather proxies 3876 such as vapour pressure deficit or individual driving variables (temperature, precipitation, 3877 relative humidity), to BA from fire models or data-driven techniques. We have applied 3 3878 different approaches here to give a robust answer to the questions of how anthropogenic 3879 forcing and climate change are driving changes in the fires we experience today. We address 3880 changes in both fire weather and burned area. Further work could use other fire weather 3881 indices, multiple CMIP Global Climate Models, or ensembles from other climate models as 3882 another approach.

3883 Using the ISIMIP impacts attribution framework enables us to tackle the questions of how BA 3884 has changed in response to total climate forcing, socioeconomic factors, and all forcing across 3885 multiple fire models for the first time. There is the opportunity to extend this to more fire models, 3886 other reanalysis driving datasets, and ideally to extend this up to the period covered in this 3887 report. Currently the driving data and simulations go up to 2019 for example, and we can only 3888 estimate the additional changes in BA due to climate change covering 2019 to 2023 through 3889 extrapolation. However our analysis suggests that this could make an important difference in 3890 the results, for example ConFire results for Canada show a 0.7-6.22% increase in BA over 3891 2000-2019, whereas if the analysis is extended to include the 2023 anomaly this increases to 3892 0.7-38.0%. Achieving near-real-time data and analysis has been one of the driving ambitions 3893 of this report, and can help us better address policy-relevant impact questions more quickly 3894 after an event has happened. This is becoming increasingly possible, with operational 3895 attribution tools coming online, and initiatives such as the World Weather Attribution tackling 3896 extreme events quickly after their occurrence. However only one of these studies has included 3897 fire so far, and a challenge for the community is therefore to be able to address these extreme 3898 fires from an attribution perspective quickly and more frequently.

3899

3900 **4.4.4 Future Projections** 3901

3902 In the coming few years, more data will become available through projects such as FireMIP 3903 and ISIMIP3b, which will provide multi-fire model projections of future BA for the first time. 3904 Using fire-enabled DGVMs enables us to more accurately model the interactions of fire and 3905 vegetation, as well as some representations of human processes of ignition and suppression. 3906 On a slightly longer time horizon, CMIP7 will also provide more Earth System Models with 3907 integrated fire feedbacks, and the community should take advantage of these modelling 3908 advances to reassess our current understanding of how fire will change in response to climate 3909 change over the next century. Currently only one model is used for future projections in this 3910 report, and in future years a more robust approach would be to include more models, as we 3911 have done in other sections of this report.

3912

3913 However, addressing the current limitations in wildfire modelling is crucial for this 3914 development. Models like FireMIP and ensemble approaches provide valuable insights into 3915 BA and feedbacks within the carbon cycle. To improve the utility and relevance of these 3916 models for policy and impact assessment, immediate goals include enhancing the predictions 3917 of fire beyond burned area, to consider intensity, spread and combustion rates for example. 3918 The models also provide information on emissions from fires, which could be explored in future 3919 reports using a similar weighted-multi-model ensemble to inform air quality impacts. However 3920 their performance in predicting extreme fire events is currently limited, which is where other 3921 tools such as ConFire and POF are useful to fill this gap. Addressing the critical feedback 3922 mechanisms and uncertainties in models like ConFire and POF is essential for a more





accurate representation of fire impacts, and immediate goals for development of these
 methods includes integrating more comprehensive fuel data to improve predictions of burned
 area, expanding the uncertainty schemes to include intensity, fire spread and emissions, and
 better constraining uncertainties between SSP scenarios through improved representation of
 BA drivers and their trends.

Over the next couple of years, the focus should be on the rapid integration of near-real-time
(NRT) data and leveraging advancements in remote sensing and machine learning. These
innovations will enhance the accuracy and timeliness of fire models for future projections. By
focusing on these targeted actions over the next few years, we can significantly advance the
state of wildfire modelling, making it more accurate, reliable, and relevant for policy and impact
assessment.

3936 4.5 Competing Interests Statement

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3935

3938 SV is a member of the editorial board of Earth System Science Data. The authors declare no 3939 further conflict of interest.

3940

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3963 4.7 Author Contributions

39643965 Conceptualization: Jones, M. W., Kelley, D. I., Burton, C. A., Di Giuseppe, F.

3966 Project Administration: Jones, M. W., Kelley, D. I., Burton, C. A., Di Giuseppe, F.

Data Curation: Jones, M. W., Kelley, D. I., Burton, C. A., Di Giuseppe, F., Barbosa, M. L. F.,

3968 Brambleby, E., Lampe, S., Mataveli, G., McNorton, J., Spuler, F., Wessel, J., Parrington, M.

Formal Analysis: Jones, M. W., Kelley, D. I., Burton, C. A., Di Giuseppe, F., Brambleby, E.,

3970 Lampe, S., Mataveli, G., McNorton, J., Spuler, F., Wessel, J., Parrington, M.

- 3971 Jones, M. W., Kelley, D. I., Burton, C. A., Di Giuseppe, F., Barbosa, M. L. F., Brambleby, E.,
- Hartley, A., Lampe, S., McNorton, J., Spuler, F., Wessel, J., Parrington, M., Hamilton, D. S.
- 3973 Resources/Software: Andela, N., Giglio, L., Parrington, M., van der Werf, G. R., Barbosa, M.
- L. F., Brambleby, E., Hartley, A., Lampe, S., McNorton, J., Spuler, F., Wessel, J., Burke, E., San-Miguel-Ayanz, J., Qu, Y.





3976 **Visualisation:** Jones, M. W., Kelley, D. I., Burton, C. A., Di Giuseppe, F., Mataveli, G., 3977 McNorton, J., Qu, Y., Lombardi, A.

Writing – Original Draft Preparation: Jones, M. W., Kelley, D. I., Burton, C. A., Di Giuseppe,
F., McNorton, J., Anderson, L., Archibald, S., Armenteras, D., Clarke, H., Doerr, S.,
Fernandes, P., Harris, S., Jain, P., Kolden, C., Ribeiro, N., Saharjo, B., Shuman, J., Tanpipat,
V., Xanthopoulos, G., Carmenta, R., Hamilton, D. S., Hantson, S., Meier, S., Perron, M. M. G.,
Parrington, M., Hartley, A., J., Spuler, F., Wessel

3983 Writing – Review & Editing: All authors. 3984

3985 4.8 Data Availability

3986

3987 BA data from NASA's MODIS BA product (MCD64A1) are extended from Giglio et al. (2018) 3988 and are available at Giglio et al. (2021, https://lpdaac.usgs.gov/products/mcd64a1v061/, last 3989 access: 2 June 2024). GFED4.1s fire C emissions data are extended from van der Werf and 3990 are available at https://globalfiredata.org/ (last access: 2 June 2024). GFAS fire C emissions 3991 are extended from Kaiser et al. (2012) and are available data at https://confluence.ecmwf.int/display/CKB/CAMS+global+biomass+burning+emissions+base 3992 d+on+fire+radiative+power+%28GFAS%29%3A+data+documentation (last access: 2 June 3993 3994 2024). Global Fire Atlas are extended from Andela et al. (2019) and are available at Andela and Jones (2024, https://doi.org/10.5281/zenodo.11400062, last access: 2 June 2024). 3995 3996 Regional summaries of the MODIS BA, GFED4.1s, GFAS, and the Global Fire Atlas are 3997 presented here are available at Jones et al. (2024, https://doi.org/10.5281/zenodo.11400540, 3998 last access: 2 June 2024). Studies utilising our regional summaries should cite both the current 3999 article and the primary reference for the variable(s) of interest: Giglio et al. (2018) for BA; van 4000 der Werf et al. (2017) for GFED4.1s fire C emissions; Kaiser et al. (2012) for GFAS fire C 4001 emissions; Andela et al. (2019) for the Global Fire Atlas.

4002

4003 Driving data, re-gridded BA target data for ConFire and ConFire outputs are available at Kelley 4004 et al. (2024, https://doi.org/10.5281/zenodo.11420743, last access: 2 June 2024). Historical 4005 (1960-2013) HadGEM3-A are available through the Centre for Environmental Data Analysis 4006 (CEDA) archive of the NERC's Environmental Data Service (EDS) at 4007 http://catalogue.ceda.ac.uk/uuid/99b29b4bfeae470599fb96243e90cde3 (last access, last 4008 access: 2 June 2024). FireMIP / ISIMIP driving and output data is available from the Inter-4009 Sectoral Impact Model Intercomparison Project (ISIMIP) repository at https://data.ISIMIP.org/ 4010 (last access: 2 June 2024).

4012 4.9 Code Availability

4013

4011

4014 ConFire attribution framework code (Kelley et al., 2021; Barbosa, 2024), was incorporated into 4015 the FLAME repository (<u>https://github.com/douglask3/Bayesian_fire_models/tree/ConFire</u>) and 4016 will be archived with a doi upon publication. Configuration settings for **Section 3.3** are in 4017 namelists/nrt.ini, while **Section 3.4** and **Section 3.5** are in namelists/ISIMIP.ini. Scripts for 4018 reproducing plots can be found in State of Wildfire GitHub repo (currently available at 4019 <u>https://github.com/douglask3/State_of_Wildfires_report</u>, with an archived doi available upon 4020 final publication).

4021

4022 The code used to produce the FWI attribution results is available in State of Wildfire GitHub 4023 repo (currently available at https://github.com/douglask3/State of Wildfires report, with an 4024 archived doi available upon final publication). FWI code can be accessed via the ECMWF 4025 GitHub (https://github.com/ecmwf-projects/geff). Met Office implementation of the FWI is 4026 based on this code. and can be accessed at 4027 https://github.com/MetOffice/impactstoolbox/tree/main with registration. All details of the data 4028 and code used for the FireMIP attribution results is documented in Burton, Lampe et al. (2023). 4029





4030 The current version of ibicus, used for JULES-ES bias correction, is available from PyPI 4031 (https://pypi.org/project/ibicus/, last access: 2 June 2024) under the Apache License version 4032 2.0, and is described in detail in https://ibicus.readthedocs.io/en/latest/ (last access: 2 June 4033 2024). The source code is available via GitHub (https://github.com/ecmwf-projects/ibicus, last 4034 access: 2 June 2024). Ibicus is also archived on Zenodo by Spuler and Wessel (2023; 4035 https://doi.org/10.5281/zenodo.8101898, last access: 2 June). Model code and evaluation for 4036 bias-correction of JULES-ES model output can be found at the State of Wildfire GitHub repo 4037 (currently available at https://github.com/jakobwes/State-of-Wildfires---Bias-Adjustment, with 4038 an archived doi available upon final publication).

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