



# 1 Quantifying Greenhouse Gas Emissions from Woodfuel used in 2 Households

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8

9 **Abstract.** The combustion of woodfuel for residential use is often not considered to be a source of greenhouse gas  
10 (GHG) emissions in households since emissions from woodfuel combustion can be offset by the CO<sub>2</sub> absorbed by  
11 the growth of the forest as a carbon sink (IPCC, 2006). However, this only applies to wood that is harvested in a  
12 renewable way, i.e., at a rate not exceeding the regrowth rate of the forest from which it is harvested (Drigo et al.,  
13 2002). This paper estimates the share of GHG emissions attributable to non-renewable woodfuel harvesting for  
14 use in residential food activities. It adds to a growing research base estimating GHG emissions from across the  
15 entire agri-food value chain, from the manufacture of farm inputs, through food supply chains, and finally to waste  
16 disposal (Tubiello et al., 2021). Country-level information is generated from United Nations Statistics Division  
17 (UNSD) and International Energy Agency (IEA) data on woodfuel use in households. We find that, in 2019, annual  
18 emissions from non-renewable woodfuel use in household food consumption were about 745 million tonnes  
19 (Mt CO<sub>2</sub>eq yr<sup>-1</sup>), with uncertainty ranging from -20 % to + 22 %, having increased 6% from 1990. Overall, global  
20 trends were a result of counterbalancing effects: the emission increases were largely fuelled from countries in Sub-  
21 Saharan Africa, Southern Asia, and Latin America while significant decreases were seen in countries in Eastern  
22 Asia and South-eastern Asia. The Food and Agriculture Organisation of the United Nations (FAO) has developed  
23 and regularly maintains a database covering GHG emissions from the various components of the agri-food sector,  
24 including pre- and post-production activities, by country and world regions. The dataset is developed according to  
25 International Panel on Climate Change guidelines (IPCC, 2006), which avoids overlaps across AFOLU and energy  
26 components. It relies mainly on UNSD Energy Statistics data, which are used as activity data for the calculation  
27 of the GHG emissions (Tubiello et al., 2022). The information used in this work is available as open data with  
28 DOI <https://doi.org/10.5281/zenodo.7310932> (Flammini et al., 2022).

29 **Keywords:** Agri-food systems, GHG emissions, sustainable woodfuel, household, food consumption

30



## 1 1. Introduction

2 In 2019, about 27% of the global population relied on traditional biomass (wood, crop residues, animal dung, etc)  
3 to meet household energy needs (IEA, 2020). The dependence on woodfuel is greatest in developing countries  
4 where it provides about one-third of total energy and is commonly used for cooking and residential heating (FAO,  
5 2010). Approximately 70% of households in Sub-Saharan Africa depend on wood-based biomass as their primary  
6 cooking fuel. That figure is roughly 44% in South-East Asia (World Bank, 2011).

7 Woodfuel for domestic purposes is obtained from many supply sources, not only from forestlands. These sources  
8 include trees outside forests (such as scrubs, bush fallow, dead wood, dry branches, twigs), trees planted with  
9 agricultural crops (agroforestry or forest plantations), residues of wood harvesting, by-products of land cover  
10 change, and salvage harvesting (FAO, 2010). Several studies have examined the impact of woodfuel use in  
11 households on deforestation and human health. For the former, extensive research was conducted as a response to  
12 the 1970s and 1980s “fuelwood crisis”, where conclusions were made that harvesting of fuelwood for energy is  
13 not the primary source of forest depletion (Arnold et al., 2006; Dewees, 1989; World Bank, 2011).

14 In terms of impact on human health, around 3.2 million premature deaths are caused due to the inhalation of  
15 polluted air in households, sourced mainly from the traditional use of biomass for heating and cooking. The  
16 pollution comes in the form of small particles that are absorbed into the lungs and enter the bloodstream. Air is  
17 considered polluted when the mean concentration of particulate matter (PM<sub>10</sub> and PM<sub>2.5</sub>) and other combustion-  
18 derived indoor pollutants such as Carbon Monoxide are beyond WHO air quality guideline values (WHO, 2014).  
19 Another study pointed at an estimation of 3 million deaths per year from indoor air pollution by open fires and  
20 smoky stoves (IEA, 2021; WHO, 2021). However, very few studies have examined the climate impact of woodfuel  
21 consumption for residential use, except in the context of carbon offsets for carbon financing (e.g., using improved  
22 cookstoves). For example, one report estimated that the global potential for GHG emission reductions for improved  
23 cookstoves (ICS) is estimated at 1 Gt CO<sub>2</sub> per year (Lee et al., 2013).

24 This paper strives to quantify the GHG emissions attributable to household food systems consumption of woodfuel.  
25 Previous reports have set the CO<sub>2</sub> emissions associated with woodfuel consumed in households to 0 which is in  
26 line with International Panel on Climate Change (IPCC) guidelines (IPCC, 2006). Such emissions are in fact  
27 covered by the ‘forestry’ and ‘land use’ components of the AFOLU sector, while the limited emissions of CH<sub>4</sub> and  
28 N<sub>2</sub>O from woodfuel burning are reported under the Energy sector. This is based on two assumptions: i. combustion  
29 of biomass is considered renewable and has no net CO<sub>2</sub> emissions impact (the CO<sub>2</sub> absorbed by the tree during its  
30 growth is equivalent to the amount released during burning or decomposition process); ii. all CO<sub>2</sub> that is  
31 sequestered over the years by trees is released during burning. Therefore, the wood removed by land-cover change  
32 (net forest conversion), or forestland degradation will eventuate, at some point, into a release of CO<sub>2</sub>. Following  
33 the IPCC approach, it is not possible to single out the amount of CO<sub>2</sub> associated with woodfuel burning at the  
34 household used for cooking.

35 In a renewable biomass harvesting scenario, the expectation is that the wood removed will fully regrow. New trees  
36 take up the carbon that is produced by the combustion the carbon balance in the atmosphere remains neutral. On  
37 the other hand, woody biomass is non-renewable if its extraction results in a long-term loss in carbon stocks, i.e.,  
38 if the extraction rate does not allow the biomass to regrow (Drigo et al., 2014). At the same time, to estimate the  
39 real emissions associated with woodfuel, it is not possible to simply apply an emissions factor to the amount of



1 woodfuel burned, since some part of wood harvested as woodfuel can be considered sustainable. This is determined  
2 if the rate of extraction is at or below the annual increment.

3 In FAOSTAT, emissions associated with the ‘unsustainable’ share of woodfuel burned are already covered under  
4 emissions on ‘forestland’ (forest degradation) and ‘net forest conversion’ (deforestation) therefore, adding  
5 emissions from woodfuel for cooking to total agri-food emissions would result in a double-counting.

6 In this publication, we define the amount harvested more than the annual increment as non-renewable biomass  
7 (NRB). Obtaining accurate information about NRB fractions has historically been a challenging exercise (Lee et  
8 al., 2013). A milestone approach on the assessment of the fractional NRB was through the use of a spatial model  
9 called the Woodfuel Integrated Supply/Demand Overview Mapping (WISDOM) which was first applied in Mexico  
10 by FAO in 2002 (Drigo et al., 2002). The WISDOM model has over the years been subsequently applied to other  
11 countries and world regions. The paper published in 2015 by Robert Bailis, Rudi Drigo, Adrian Ghilardi and Omar  
12 Masera presents non-renewable biomass fraction (NRBf) by applying an evolution of the WISDOM model to a  
13 number of countries across Asia, Africa and Latin America (Bailis et al., 2015).

14 This paper presents a methodology to apply the NRBf to woodfuel consumption used for food in the household,  
15 based on data from the UNSD Energy Statistics database. Our methodology does not distinguish between woodfuel  
16 emissions associated with the deforestation component and the degradation component. However, previous  
17 research estimated that emissions from forest degradation were one-fourth of those from deforestation in 2001–  
18 2010, and increased to one-third in more recent years (2011–2015) (Federici et al., 2015).

19 The results are presented consistently to FAOSTAT countries and regions, in an effort to further expand FAOSTAT  
20 work on disseminating data on GHG emissions from agri-food systems at the country-level. Accounting for GHG  
21 emissions across all agri-food systems activities will help researchers, policymakers, and businesses uncover novel  
22 climate mitigation opportunities through food system interventions.

## 23 **2. Materials and methods**

### 24 **2.1 Gap filling**

25 The UNSD fuelwood data used herein are gap filled to improve the quality of the available timeseries and to  
26 estimate data for missing countries. Notably, the original UNSD energy dataset had missing data for China for the  
27 entire time series, and this gap was filled by complementing it with IEA Energy data for primary solid biofuels  
28 (defined as any plant matter used directly as fuel or converted into other forms before combustion). The NRBf data  
29 was available for most countries in regions such as Africa, Asia, and Latin America. For countries with no data  
30 available on their NRBf values, sub-regional and regional NRBf averages were used and applied accordingly.

### 31 **2.2 Emissions estimates**

32 For FAO, biofuel is defined as “any fuel produced directly or indirectly from biomass”, while woodfuel is  
33 described as all types of biofuels derived directly or indirectly from woody biomass (grown on either forest or non-  
34 forest land) (FAO, 2004). GHG emissions are calculated according to the IPCC guidelines, at Tier 1 (IPCC, 2006),  
35 by applying the following formula:

$$36 \quad E_{i,g} = A_{i,y} * f_w * NRBf_i * EF_g \quad (1)$$

37 where



1 E = GHG emissions by gas (g) in select country or region I, for select inventory year, y, kilotonnes of CO<sub>2</sub>  
2 equivalent (kt CO<sub>2</sub>e yr<sup>-1</sup>)

3 A = volume of woodfuel consumed in the household (activity data) for select country or territory i, for select  
4 inventory year y, reported in cubic metres (m<sup>3</sup>)

5 f<sub>w</sub> = share of woodfuel used for cooking for select country y,

6 NRBf = non-renewable biomass fraction for select country y, based on FAO WISDOM,

7 EF = emission factor of woodfuel, by gas, based on IPCC (2006) default values,

8 The volume of woodfuel consumed in the household is extracted from UNSD Energy Statistics database (Flow  
9 1231): Consumption by households and converted to energy by applying a representative calorific value of 11.2.  
10 The calorific value is calculated by multiplying the average heating value of air-dried wood fuel and completely  
11 dry wood and its average density. This heating value is estimated from the heating value of woods typically used  
12 as woodfuel, as reported in the IEA Energy Statistics Manual (IEA, 2004). The average density of the woodfuel is  
13 estimated by taking the density of woods typically available in tropical countries (FAO, 2007). This assumes that  
14 most of wood harvesting for food preparation takes place in pan-tropical countries. The share of woodfuel used  
15 for cooking is set to unity for all tropical countries concerned (i.e., in tropical countries all woodfuel used in the  
16 household is for cooking), while countries with little-to-no tropical coverage would have its share set as 0.847. The  
17 rest is used for heating (Daioglou et al., 2012; Morgan, 2011).

18 The NRBf fraction is obtained from the ‘expected’ NRBf, where suboptimal harvesting of woodfuel is assumed.  
19 The NRBf referenced is taken from the average of the low plantation productivity variant and the high plantation  
20 productivity variant of NRB<sub>B1</sub> + NRB<sub>B2</sub>. The latter is generated with the assumption that woodfuel users can meet  
21 their woodfuel demand from both land cover change by-products and from other sources (Bailis et al., 2015). The  
22 same NRBf is assumed for all years and countries reported.

23 The calculation was run using R software for all countries and world regions, mapping UNSD to FAOSTAT  
24 countries for the application of subregional and regional statistics.

25

### 26 **2.3 Data uncertainty and limitations**

27 There are limitations and uncertainties associated with the estimates presented herein. First, we note that, although  
28 we assume that all fuelwood in tropical households (and the vast majority in the non-tropical countries of the  
29 regions concerned) is used for food preparation, the input data refer to fuelwood energy use in households without  
30 further breakdown. Second, the underlying data on energy use have gaps, especially in China and Africa. For  
31 countries with no data, these were imputed from the IEA Energy Database instead. Thirdly, for the underlying  
32 NRB fractions, out of 90 countries and territories, 6 were imputed based on regional averages. The uncertainty in  
33 the original woodfuel consumption data is much smaller for some countries than for others, depending on whether  
34 the activity data are collected using specific surveys where a sense of the uncertainty can be measured, or whether  
35 national statistical offices use proxies and/or assumptions. In our case, using the level of uncertainty for stationary  
36 non-energy intensive industries and “well-developed statistical systems” (such as energy statistics), an uncertainty  
37 of ±5 % can be assumed for activity data (IPCC, 2006, volume 2, chap. 2, Table 2.6; IPCC, Estimating



1 Uncertainties in GHG Emissions from Fuel Combustion, Table 3; Flammini et al., 2022). Uncertainty in activity  
2 data was then combined with uncertainty in fuel emission factors ( $-15\%$  to  $18\%$ ), computed by taking the IPCC  
3 lower and upper values of emissions factors of wood/wood waste. Uncertainty on the conversion factor is  
4 calculated as  $\pm 12\%$ . Lastly, the uncertainty for the NRBf values was computed to be  $\pm 1.3\%$ . The resulting overall  
5 uncertainty from the energy statistics and emission factors was obtained by applying the IPCC (IPCC, 2006) default  
6 error propagation method, resulting in the range  $-20\%$  to  $+22\%$ .

7 An additional limitation of this methodology is that, although unsustainable woodfuel extraction could be  
8 associated with both with deforestation and forest degradation, our methodology does not single out the emissions  
9 that are attributable to each of them.

10

### 11 3 Results

12 The results show that, globally, for the year 2019, the GHG emissions associated with the unsustainable (or non-  
13 renewable) fraction of woodfuel used in households were 741,652 kt for CO<sub>2</sub> emissions, 1,987 kt for CH<sub>4</sub> and 26.5  
14 kt for N<sub>2</sub>O. Therefore, the CO<sub>2</sub>eq emissions were above 0.7 Gt in 2019, 6% greater than in 1990. These emissions  
15 can be compared with the total 4.84 Gt CO<sub>2</sub> yr<sup>-1</sup> from deforestation (4.04 Gt CO<sub>2</sub>) and forest degradation (0.80 Gt  
16 CO<sub>2</sub>) estimated at global level by Federici et al. (2015) over 1991-2015. This amount of emissions associated with  
17 unsustainable woodfuel use in the household should be added to the 1.3 Gt CO<sub>2</sub>eq yr<sup>-1</sup> associated with household  
18 food consumption (excluding bioenergy) in the year 2019 reported in FAOSTAT (Tubiello et al., 2022). Therefore  
19 2 billion tonnes are a more precise figure of the emissions associated with human activity at this important step of  
20 the agri-food chain.

21 For comparison, the GHG emissions of the whole agri-food sector amount to 16 Gt CO<sub>2</sub>eq yr<sup>-1</sup> and 6 Gt CO<sub>2</sub>eq  
22 yr<sup>-1</sup> alone from post-agricultural production activities (including food processing, transport, retail and household  
23 consumption). Household woodfuel emissions correspond to 4.7% and 12.5% respectively.

24 The top 10 countries (out of 90 countries covered by the dataset) are responsible for 69% of global GHG emissions  
25 attributable to woodfuel use for household food systems in 2019. No country from Latin America and the  
26 Caribbean were among the top 10 GHG emitters. However, in terms of GHG emission per person (based on  
27 population data from FAOSTAT), higher values can be seen in African countries: out of the 10 top emitters, five  
28 are from Sub-Saharan Africa, three are from Southern Asia, one from Eastern Asia and one from South-eastern  
29 Asia.

30 Nigeria and India were the largest emitters in 2019 in absolute terms. It is to note that China saw the biggest  
31 reductions in CO<sub>2</sub> emissions over the time frame of 1990 to 2019, and China was the highest emitter from 1990 to  
32 2006.

33 Sub-Saharan Africa, South Asia and Eastern Asia were the largest emitters among subregions, although with  
34 different trends over 2005-2019. Eastern Asia decreased over the whole period, from 138 Mt CO<sub>2</sub>eq yr<sup>-1</sup> in 1990  
35 to 135 Mt CO<sub>2</sub>eq yr<sup>-1</sup> in 2000 and further decreased to 51 Mt CO<sub>2</sub>eq yr<sup>-1</sup> in 2019, while emissions in Sub-Saharan  
36 Africa nearly doubled from 202 in 1990 to 379 Mt CO<sub>2</sub>eq yr<sup>-1</sup> in 2019. Southern Asia was a significant emission  
37 source in 2019 but has increased only slightly (around 3% over a 19-year period) since 1990, from 197 to 203



1 Mt CO<sub>2</sub>eq yr<sup>-1</sup> in 2019. Emissions increased only by 1% in Latin America and South-eastern Asia decreased by  
2 more than 60% over the same period.

3 We also compared estimates of emissions from woodfuel use in household food consumption with the estimates  
4 from net forestland conversion in FAOSTAT. As discussed in the Materials and methods section, FAO estimates  
5 of emissions from net forestland conversion are proxies for deforestation emissions data. It is also important to  
6 note that there are various sources of woodfuel use in households as described in the Introduction section, and net  
7 forestland conversion is just one of them. On a global scale, woodfuel household food CO<sub>2</sub> emissions were between  
8 15% to 23% of the global net forest conversion CO<sub>2</sub> emissions.

9

#### 10 4 Discussion

11 36.3% of overall household food emissions can be attributed to unsustainable harvest of woodfuel used in the  
12 household for cooking. Although these GHG emissions are covered in the AFOLU section, according to the IPCC  
13 guidelines, as part of the ‘deforestation’ activity, these emissions are strictly related to the cooking, which happens  
14 towards the end of the agri-food chain. This paper presents an estimation of the emissions from ‘unsustainable’  
15 woodfuel use for cooking in the households. It is important to understand the magnitude of these emissions versus  
16 total deforestation emissions and total household emissions because any mitigation action of these emissions  
17 cannot be enacted without addressing cooking systems. In other words, to reduce this important share of agri-food  
18 system emissions, any mitigation action should focus on, or at least consider, providing alternative and/or more  
19 efficient cookstoves to the users of unsustainable woodfuel for cooking. An intervention aimed only at halting  
20 deforestation or reducing household emissions will be partial or ineffective.

21 The high proportion of non-renewable woodfuel consumption in regions such as Sub-saharan Africa is reflective  
22 of the population where low-income households have a higher dependency on biomass for their energy needs  
23 (Dutschke et al., 2006) and energy use is less varied than their middle- and upper- income counter parts (the only  
24 two primary services are cooking and lighting) (Sovacool, 2011). The massive reduction in non-renewable  
25 woodfuel emissions from China over the period can be attributed to the exponential income of rural farmers with  
26 strong rural energy policies which supported the development of other energy sources (most notably, electricity)  
27 (Yao et al., 2012).

28 The updated assessment of total agri-food system emissions as supplemented by the data in this work still reaffirms  
29 previous findings and works by the IPCC (2019), Crippa et al. (2021a, b) and Tubiello et al. (2022). However, the  
30 most significant difference with previous work was observed in relation to household consumption emissions. Our  
31 updated value estimates were bigger than our previous estimates of 1.2 Gt CO<sub>2</sub>eq. fuelled mostly by woodfuel  
32 combustion in household food systems. FAOSTAT estimates in this work, 1.9 Gt CO<sub>2</sub>eq., were more than 4 times  
33 those of EDGAR-FOOD (with reference to 2015, the last year for which EDGAR data were available).

34 A notable trend with the incorporation of non-renewable woodfuel emissions into the overall household food  
35 system emissions is the amplification of country-level emissions in countries/territories with high dependence on  
36 woodfuel as their source of energy (Schilman et al., 2021; World Bank, 2011). Our refined assessments of  
37 emissions contributions highlight the importance of non-renewable woodfuel into the overall food systems  
38 emissions. Regarding the three major components of the food system (on-farm production, land use change and



1 pre- and post-production activities), our analysis highlights that in 2019, pre- and post-production emissions have  
2 almost the same share of emission contributions to farm-gate activities (38.0% vs 41.7%) at the global level, while  
3 land use change provided a smaller contribution (20.3%). For the same year, household food systems took the  
4 biggest share of pre- and post-production emissions (31.5%) while non-renewable woodfuel combustion was  
5 36.3% of household food systems.

6

## 7 **5 Data availability**

8 The GHG emission data presented herein cover the period 1990–2019 at the country level. They are available as  
9 open data, with DOI <https://doi.org/10.5281/zenodo.7310932> (Flammini et al., 2022).

10

## 11 **6 Conclusions**

12 This paper provides updated details of the FAOSTAT database on GHG emissions along the entire agri-food  
13 systems chain (Tubiello et al., 2022), with a focus on improving the estimates of the household consumption  
14 emissions.

15 The data are provided in open-access mode to users worldwide and are available by country over the period 1990–  
16 2019, with plans for annual updates. The major trends in non-renewable woodfuel consumption within household  
17 food-systems that were identified in this work can help locate emissions hotspots in agri-food systems and inform  
18 the adoption/effectiveness of policies on cooking fuel switches on the country, regional and global level. This work  
19 also emphasizes the increasingly important role that pre- and post-production processes along supply chains play  
20 in the overall GHG footprint of agri-food systems, in a regional and global level.

21 This paper also helps to expand the impacts of woodfuel use beyond just health measures but to also highlight the  
22 climate impact attached to using non-renewable woodfuel as a source of cooking fuel. Finally, the methodological  
23 work underlying these efforts complements and extends recent pioneering efforts by FAO and other groups in  
24 characterizing technical coefficients to enable quantifying the weight of agri-food systems within countries’  
25 emissions profiles.

26

## 27 **7 Competing interests**

28 At least one of the (co-)authors is a member of the editorial board of *Earth System Science Data*. The peer-review  
29 process was guided by an independent editor, and the authors also have no other competing interests to declare.

30

## 31 **8 Disclaimer**

32 The views expressed in this paper are the authors’ only and do not necessarily reflect those of FAO or UNIDO.

33

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1  
 2 TABLES

3  
 4  
 5 **Table 1.** Typical heating values of woods used as fuelwood

Wood Type	Heating value
Air-dried wood (10% to 20% moisture content)	16 MJ/kg
Completely dry wood (oven-dried)	18 MJ/kg
Average	17 MJ/kg

6 Source: IEA Biofuel Energy Statistics (Page 174 of the IEA Energy Statistics Manual) (IEA, 2004)

7  
 8 **Table 2.** Typical densities of woods used as fuelwood

Wood Type	Density
Air-dried wood	725 kg/m <sup>3</sup>
Oven-dried wood	593 kg/m <sup>3</sup>
Average	659 kg/m <sup>3</sup>

9 Source: Wood-energy supply/demand scenarios in the context of poverty mapping (Table A2.4) (FAO, 2007)

10  
 11 **Table 3.** List of tropical countries

Algeria	Benin	Lao People's Democratic Republic	Guinea-Bissau
Angola	Dominica	Liberia	Timor-Leste
Antigua and Barbuda	Dominican Republic	Libya	Rwanda
Argentina	Ecuador	Madagascar	Saint Helena
Bolivia (Plurinational State of)	El Salvador	Malawi	Saint Kitts and Nevis
Botswana	Equatorial Guinea	Malaysia	Anguilla
Brazil	Ethiopia	Maldives	Saint Lucia
Belize	Eritrea	Mali	Saint Vincent and the Grenadines
Solomon Islands	Falkland Islands (Malvinas)	Mauritania	Sao Tome and Principe



British Virgin Islands	Fiji	Mauritius	Senegal
Brunei Darussalam	French Guiana	Mexico	Seychelles
Myanmar	Djibouti	Mozambique	Sierra Leone
Burundi	Gabon	Namibia	Viet Nam
Cambodia	Gambia	Nepal	Somalia
Cameroon	Ghana	Curacao	South Africa
Cabo Verde	Grenada	Aruba	Zimbabwe
Central African Republic	Guadeloupe	Saint Maarten (Dutch part)	South Sudan
Sri Lanka	Guatemala	Bonaire, Sint Eustatius and Saba	Sudan
Chad	Guinea	New Caledonia	Suriname
Chile	Guyana	Vanuatu	Togo
Colombia	Haiti	Nicaragua	Trinidad and Tobago
Comoros	Honduras	Niger	Turks and Caicos Islands
Mayotte	India	Nigeria	Uganda
Congo	Indonesia	Panama	Egypt
Democratic Republic of the Congo	Cote d'Ivoire	Papua New Guinea	United Republic of Tanzania
Costa Rica	Jamaica	Paraguay	Venezuela (Bolivarian Republic of)
Cuba	Kenya	Peru	Guinea-Bissau

1 Source: Journal of Tropical Psychology, Volume 1 (Morgan, 2011)

2

3 **Table 4.** Non-renewable fractions (NRBf) based on country and region

Country/region	NRB fraction	Country/region	NRB fraction
Angola	0.350	Malawi	0.371
Argentina	0.283	Malaysia	0.465
Bangladesh	0.510	Mali	0.291
Belize	0.993	Mauritania	0.348
Benin	0.217	Mexico	0.268



Bhutan	0.559	Mozambique	0.397
Bolivia (Plurinational State of)	0.325	Myanmar	0.085
Botswana	0.895	Namibia	0.476
Brazil	0.238	Nepal	0.524
Brunei Darussalam	0.872	Nicaragua	0.579
Burkina Faso	0.476	Niger	0.235
Burundi	0.570	Nigeria	0.511
Cambodia	0.384	Pakistan	0.836
Cameroon	0.758	Panama	0.496
Central African Republic	0.264	Papua New Guinea	0.403
Chad	0.237	Paraguay	0.384
Chile	0.138	Peru	0.309
China	0.16	Philippines	0.214
Colombia	0.344	Rwanda	0.585
Congo	0.099	Senegal	0.361
Costa Rica	0.18	Sierra Leone	0.219
Côte d'Ivoire	0.163	Singapore	0.755
Democratic Republic of the Congo	0.24	Solomon Islands	1
Dominican Republic	0.33	Somalia	0.524
Ecuador	0.99	South Africa	0.238
El Salvador	0.372	Sri Lanka	0.244
Equatorial Guinea	0.94	Sudan	0.411
Eritrea	0.679	Suriname	0.181
Ethiopia	0.613	Thailand	0.03
French Guiana	0.165	Timor-Leste	1
Gambia	0.412	Togo	0.44
Ghana	0.286	Trinidad and Tobago	0.554
Guatemala	0.334	Uganda	0.613



Guinea	0.297	United Republic of Tanzania	0.235
Guinea-Bissau	0.279	Venezuela (Bolivarian Republic of)	0.527
Guyana	0.039	Viet Nam	0.115
Haiti	0.666	Zambia	0.340
Honduras	0.637	Zimbabwe	0.377
India	0.231	Eastern Asia	0.16
Indonesia	0.434	Latin America and the Caribbean	0.396
Jamaica	0.185	Melanesia	0.702
Kenya	0.635	Northern Africa	0.369
Lao People's Democratic Republic	0.273	South-eastern Asia	0.421
Lesotho	0.525	Southern Asia	0.484
Liberia	0.283	Sub-Saharan Africa	0.415
Libya	0.327		

1 *Source: Authors' own elaboration.*

2



1 FIGURE LEGENDS

2

3 **Figure 1.** Global GHG emissions from woodfuel use in households for cooking from 1990 to 2019 (Mt), including  
4 uncertainty ranges. Source: Authors, based on data from IEA and UNSD (2022)

5

6 **Figure 2.** GHG emissions trends from the top 10 emitters of 2019 from 1990 to 2019 (Mt CO<sub>2</sub>eq). Source: Authors.

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8 **Figure 3.** Share of global GHG emissions from household woodfuel use in food sector stratified according to sub-  
9 region. Source: Authors.

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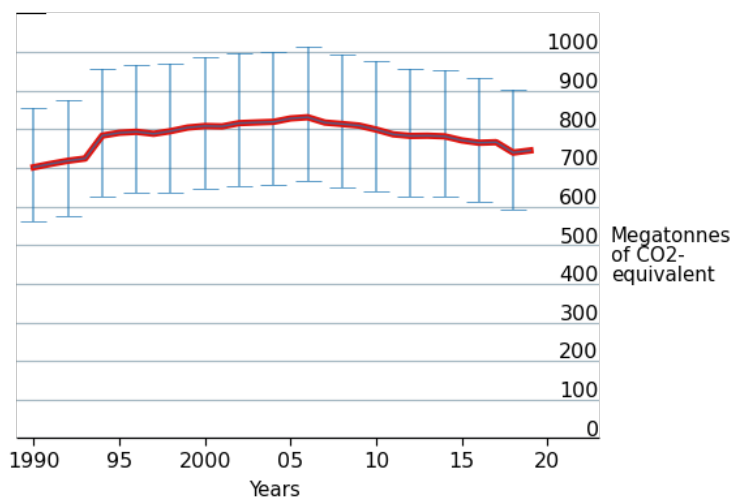
11 **Figure 4.** Trends in CO<sub>2</sub> emissions from net forestland conversion and woodfuel use in household food  
12 consumption from 1990 to 2019. Source: FAOSTAT, 2022

13

14 **Figure 5.** Proportion of emissions for non-renewable woodfuel use in household food systems in comparison to i)  
15 the overall food system (Pie Chart 1), ii) pre- and post-production (Pie Chart 2) and iii) household food systems  
16 (Pie Chart 3), for the year 2019. Source: FAOSTAT, 2022



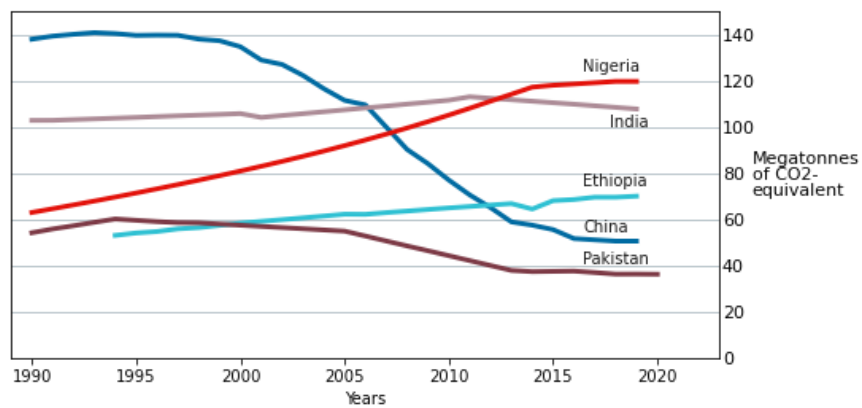
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4 **Figure 1.** Global GHG emissions from woodfuel use in households for cooking from 1990 to 2019 (Mt), including  
5 uncertainty ranges. Source: Authors, based on data from IEA and UNSD (2022)

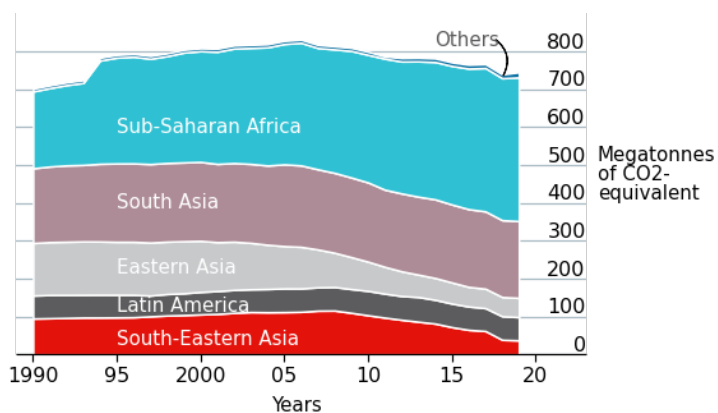
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8 **Figure 2.** GHG emissions trends from the top 10 emitters of 2019 from 1990 to 2019 (Mt CO2eq). Source: Authors.

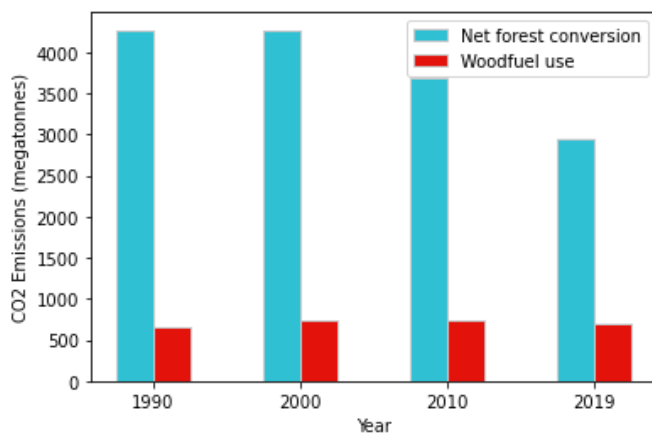
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2 **Figure 3.** Share of global GHG emissions from household woodfuel use in food sector stratified according to sub-region.  
 3 Source: Authors.

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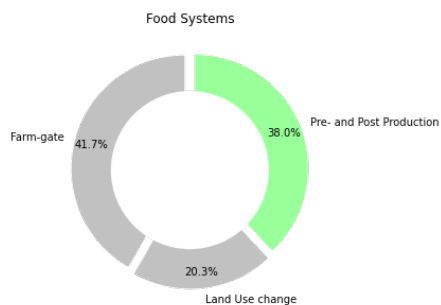
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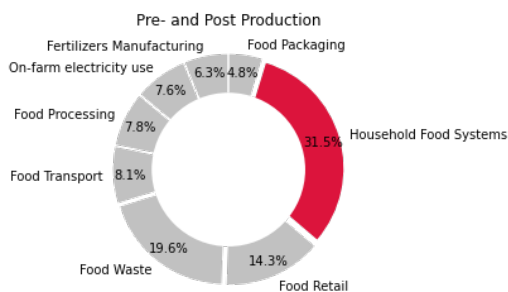
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Pie Chart 1



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Pie Chart 2

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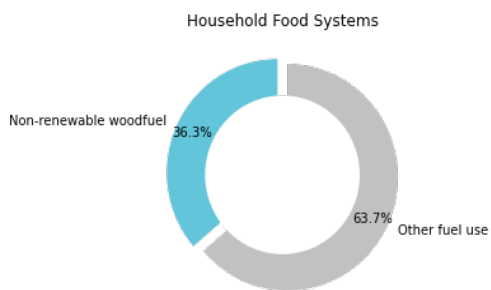
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Figure 5. Proportion of emissions for non-renewable woodfuel use in household food systems in comparison to the overall food system (Pie Chart 1), pre- and post-production (Pie Chart 2) and household food systems (Pie Chart 3) for the year 2019. Source: FAOSTAT, 2022