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CAMELS-GB: Hydrometeorological time series and landscape attributes for 671 catchments in Great Britain

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Abstract

We present the first large-sample catchment hydrology dataset for Great Britain, CAMELS-GB (Catchment Attributes and MEteorology for Large-sample Studies). CAMELS-GB collates river flows, catchment attributes and catchment boundaries from the UK National River Flow Archive together with a suite of new meteorological timeseries and catchment attributes. These data are provided for 671 catchments that cover a wide range of climatic, hydrological, landscape and human management characteristics across Great Britain. Daily timeseries covering 1970-2015 (a period including several hydrological extreme episodes) are provided for a range of hydro-meteorological variables including rainfall, potential evapotranspiration, temperature, radiation, humidity and river flow. A comprehensive set of catchment attributes are quantified including topography, climate, hydrology, land cover, soils and hydrogeology. Importantly, we also derive human management attributes (including attributes summarising abstractions, returns and reservoir capacity in each catchment), as well as attributes describing the quality of the flow data including the first set of discharge uncertainty estimates for Great Britain. CAMELS-GB (Coxon et al, 2020; available at https://doi.org/10.5285/8344e4f3-d2ea-44f5-8afa-86d2987543a9) is intended for the community as a publicly available, easily accessible dataset to use in a wide range of environmental and modelling analyses.

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1 Introduction

Data underpin our knowledge of the hydrological system. They advance our understanding of water dynamics over a wide range of spatial and temporal scales and are the foundation for water resource planning and regulation. With the emergence of new digital technologies and increased monitoring of the earth system via satellites and sensors, we now have greater access to data than ever before. This proliferation of data has been reflected in recent projects where there has been a focus on sharing data and collaborative research (SWITCH-ON; Ceola et al., 2015), collecting new datasets through the creation of terrestrial environmental observatories (TERENO; Zacharias et al., 2011) or the Critical Zone Observatories (CZO; Brantley et al., 2017), and cloud based resources for modelling and visualising large datasets such as the Environmental Virtual Observatory (EVO; Emmett et al., 2014) and the CUASHI hydrodesktop (Ames et al., 2012).

To synthesize hydrologically relevant data and learn from differences between catchments, several large-sample hydrological datasets have been produced over the last decades. These datasets rely on complementary data sources to provide the community with hydrometeorological time series and landscape attributes enabling the characterisation of dozens to thousands of catchments (see Addor et al., 2019 for a review). Many studies have demonstrated the importance of large sample catchment datasets for understanding regional variability in model performance (Coxon et al., 2019; Kollat et al., 2012; Lane et al., 2019; Newman et al., 2015; Perrin et al., 2003), testing model behaviour and robustness under changing climate conditions (Coron et al., 2012; Fowler et al., 2016; Werkhoven et al., 2008), understanding variability in catchment behaviour including hydrologic signatures and classification (Sawicz et al., 2011; Yadav et al., 2007), assessing trends in hydro-climatic extremes (Berghuijs et al., 2017; Blöschl et al., 2017; Gudmundsson et al., 2019; Hannaford and Buys, 2012; Stahl et al., 2010), exploring model and data uncertainty (Coxon et al., 2014; Westerberg et al., 2016) and regionalising model structures and parameters (Lee et al., 2005; Merz and Blöschl, 2004; Mizukami et al., 2017; Parajka et al., 2005; Pool et al., 2019; Singh et al., 2014).

However, while the number of studies involving data from large samples of catchments is rapidly increasing, publicly available large sample catchment datasets are still rare. Researchers spend considerable time and effort compiling large sample catchment datasets, yet these datasets are rarely made available to the community due to data licensing restrictions, strict access policies or because of the time required to make these datasets readily usable (Addor et al., 2019; Hannah et al., 2011; Nelson, 2009; Viglione et al., 2010). Notable exceptions of open-source, large-sample, catchment datasets include the MOPEX dataset that includes hydro-meteorological timeseries and catchment attributes for 438 US catchments (Duan et al., 2006), the CAMELS dataset that covers 671 US catchments (Catchment Attributes and MEteorology for Large-Sample studies, Addor et al., 2017; Newman et al., 2015), the CAMELS-CL dataset that contains data for 516 catchments across Chile (Alvarez-Garreton et al., 2018) and the Canadian model parameter experiment (CANOPEX) database (Arsenault et al., 2016). Because daily streamflow records often cannot be redistributed, researchers have computed streamflow indices (hydrological signatures) and made them publicly available together with catchment attributes. This is the approach selected for the Global Streamflow Indices and Metadata Archive (Do et al., 2018; Gudmundsson et al., 2018), which includes >35,000 catchments globally, and the dataset produced by Kuentz et al., (2017) which includes data for >30,000 catchments across Europe. Overall, datasets for large samples of catchments are vital to advance knowledge on hydrological processes (Falkenmark and Chapman, 1989; Gupta et al., 2014; McDonnell et al., 2007; Wagener et al., 2010), to underpin common frameworks for model evaluation across complex domains (Ceola et al., 2015) and ensure hydrological research is reusable and reproducible through the use of common datasets and code (Buytaert et al., 2008; Hutton et al., 2016).

In Great Britain, there is a wide availability of gridded, open source datasets and free access to quality-controlled river flow data via the UK National River Flow Archive (NRFA). While this is a

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large resource of open data by international standards, these datasets have not yet been combined and processed over a consistent set of catchments and made publicly available in a single location. Further these are dynamic datasets subject to change which cannot support consistent repeatable analysis. Finally, the range of variables and catchment attributes is more limited than other large-sample datasets such as CAMELS.

To address this data gap, we produced the CAMELS-GB dataset (Coxon et al., 2020). CAMELS-GB collates river flows, catchment attributes and catchment boundaries from the NRFA together with a suite of new meteorological timeseries and catchment attributes for 671 catchments across Great Britain. In the following sections we describe the key objectives behind CAMELS-GB and how they have shaped the content of the dataset. We also provide a comprehensive description of all data contained within CAMELS-GB including 1) its source data, 2) how the timeseries and attributes were produced and 3) a discussion of the associated limitations.

2 Objectives

CAMELS (Catchment Attributes and MEteorology for Large-sample Studies) began as an initiative to provide hydro-meteorological timeseries (Newman et al., 2015) and catchment attributes covering climatic indices, hydrologic signatures, land cover, soil and geology (Addor et al., 2017) for the contiguous United States. Since then, the dataset has been used widely in other studies (e.g. Addor et al., 2018; Gnann et al., 2019; Pool et al., 2019; Tyralis et al., 2019) and has provided the framework for the production of similar datasets. CAMELS for Chile (CAMELS-CL, Alvarez-Garreton et al., 2018) was released and CAMELS datasets for other countries are in production (Brazil and Australia). While each CAMELS dataset has unique features (for example CAMELS-CL provides snow water equivalent estimates and CAMELS-GB characterises uncertainties in streamflow timeseries), all the CAMELS datasets consistently apply the same core objective; make hydrometeorological time series and landscape attributes for a large-sample of catchments publicly available. They strive to use the same open-source code, variable names and datasets in order to increase the comparability and reproducibility of hydrological studies. In creating the CAMELS-GB dataset, we wanted to build on the successful CAMELS blueprint to provide a large-sample catchment dataset for Great Britain based on four core objectives.

Firstly, we wanted to build on the wealth of data already available for GB catchments but synthesize the diverse range of data into a single, consistent, up-to-date dataset. The UK has a rich history of leading research in catchment hydrology and integrating large samples of data for many catchments. For example, the Flood Studies Report (NERC, 1975) extracted high rainfall events, peak flows and catchment characteristics for 138 catchments to support flood estimation using catchment characteristics. The UK NRFA contains a wealth of data (including flow timeseries, catchment attributes, catchment masks) for the UK gauging station network which contains approximately 1,500 gauging stations as summarised in the UK Hydrometric Register (Marsh and Hannaford, 2008). Where possible, we have made use of the existing data available on the NRFA in CAMELS-GB to ensure consistency and to avoid duplicating efforts. We also build on these existing datasets by providing new catchment attributes and timeseries that are currently not available on the NRFA (e.g. potential-evapotranspiration, temperature, soils and human impacts).

Secondly, we wanted to provide a large-sample catchment dataset for Great Britain based on information that i) are sufficiently detailed to enable the exploration of hydrological processes at the catchment scale, ii) are well documented (ideally in open-access peer-reviewed journals), iii) rely on state-of-the-art methods and iv) include recent observations. Consequently, some catchment attributes currently available on the NRFA have been re-calculated for CAMELS-GB as better quality or higher spatial resolution datasets are now available (e.g. to derive land cover and hydrogeological attributes). This also means that we have primarily used the best available national datasets for the derivation of

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the catchment timeseries and attributes. These timeseries and attributes can be compared at a later stage to estimates to be derived from global datasets.

Thirdly, we wanted to provide qualitative and quantitative estimates of the limitations/uncertainties of the data provided in CAMELS-GB. Characterising data uncertainties is crucial as different data collection techniques or quality standards can bias comparisons between catchments. By providing quantitative estimates of uncertainty (including the first set of national discharge uncertainty estimates), we hope to raise awareness and encourage users of the dataset to consider these uncertainties in their analyses.

Finally, where possible, we have ensured that the underlying datasets (such as gridded geophysical and meteorological data) are publicly available to allow reproducibility and reusability.

3 Catchments

The catchments included in the CAMELS-GB dataset were selected from the UK NRFA Service Level Agreement (SLA) Network. Approximately half of the NRFA gauging stations are designated as SLA stations in collaboration with measuring authorities (as described in Dixon et al., 2013; Hannaford, 2004), embracing catchments which are considered to contribute most to the overall strategic utility of the gauging network. Selection criteria include hydrometric performance, representativeness of the catchment, length of record and degree of artificial disturbance to the natural flow regime. The flow records for these SLA stations are subject to an additional level of validation on the NRFA and are also used to calculate performance metrics that quantify completeness and quality (see the methods and metrics outlined in Dixon et al., 2013 and Muchan and Dixon, 2014). This process focuses on the credibility of flows in the extreme ranges and the need to maintain sensibly complete time series, thus providing good quality and long time series for CAMELS-GB. All gauges from the UK SLA network are included in CAMELS-GB except catchments from Northern Ireland (due to a lack of consistent meteorological datasets across the UK) and two gauges where no suitable surface area catchment could be derived. This results in a total of 671 catchments covering a wide range of climatic and hydrologic diversity across GB that is representative of the wider gauging network (see Supplement Fig S1 for a comparison of key attributes for the CAMELS-GB catchments and all GB gauged catchments).

In keeping with the CAMELS-CL dataset (Alvarez-Garreton et al., 2018), we chose to include both non-impacted and human impacted catchments in the dataset complemented with catchment attributes on the size and type of human impacts these catchments experience. Human impacted catchments are provided to support the current IAHS Panta Rhei decade which is focused on how the water cycle is impacted by human activities (McMillan et al., 2016; Montanari et al., 2013) and also enable national scale hydrological modelling and analyses across catchments that are impacted by reservoirs, abstractions and land use change.

4 Catchment Masks

Catchment masks are provided in the dataset to allow other users to create their own catchment hydro-meteorological timeseries and attributes from gridded datasets not used in this study. The catchment masks were derived from CEH’s Integrated Hydrological Digital Terrain Model (IHDTM; Morris and Flavin, 1990) and a set of 50m flow direction grids. The flow direction grids are based on a Digital Elevation Model and contours from the UK Ordnance Survey Land-Form Panorama dataset (now withdrawn and superseded by OS Terrain 50) and hydrologically corrected by “burning in” rivers using CEH’s 1:50K digital river network (Moore et al., 2000). The catchment boundaries were created using bespoke code for identifying all IHDTM cells upstream of the most appropriate grid cell to represent the gauging station location and generating a meaningful “real-world” boundary around

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these cells. Catchment masks are provided as shapefiles in the OSGB 1936 co-ordinate system (British National Grid).

ASCII files were generated from the shapefiles by converting the shapefile onto a 50m raster grid and then exporting the rasters to individual ascii files. These files are used to calculate all catchment averaged time series and attributes in CAMELS-GB. To calculate the catchment average timeseries/attribute for each dataset, the 50m grid cells in each catchment mask were assigned a value from the respective dataset grid cell (determined by which dataset grid cell the lower left hand corner of the mask grid cell lay within) and an arithmetic mean of these values were calculated (unless specified otherwise). This ensures a weighted average is calculated that accounts for the differences in grid cell sizes between the catchment mask (on a 50m grid) and any other datasets (often on a 1km grid). This is particularly important for smaller catchments in areas of highly variable data.

It is important for users to note that as the topographical boundaries are used throughout the study to quantify the hydrometeorological timeseries and attributes, this could mean significant errors where the catchment area is poorly defined.

5 Time Series Data

Daily meteorological and hydrological time series data are provided for the 671 CAMELS-GB catchments including flow, rainfall, potential evapotranspiration, temperature, short-wave radiation, long-wave radiation, specific humidity and wind speed (summarised in Table 1). These datasets were chosen for inclusion in CAMELS-GB to cover the common forcing and evaluation data needed for catchment hydrological modelling, to allow users to derive different estimates of potential evapotranspiration and to provide the key hydro-meteorological data for catchment characterisation.

Hydro-meteorological timeseries data for the 671 catchments were obtained from a number of datasets for a 45 year time period from the 1st October 1970 – 30th September 2015. These long time series enable the dataset's use in trend-analysis, provide a valuable dataset for model forcing and evaluation and ensures the robust calculation of hydro-climatic signatures. These long time series also cover a wide range of nationally important climatic events such as the 1976 drought and 2007 floods (see summaries of UK drought and flood episodes for a more extensive review including Folland et al., 2015; Marsh et al., 2007; Stevens et al., 2016). From previous analyses, it is important to note that there are key known non-stationarities over this period in hydro-meteorological data for GB. For example, seasonal changes in precipitation have been well documented (Jenkins et al., 2009) and linked to changes in river flow (Hannaford and Buys, 2012; Harrigan et al., 2018).

5.1 Meteorological Timeseries

Meteorological timeseries were derived from high-quality national gridded products chosen for their high spatial resolution (1km^2), long time series availability and basis on UK observational networks. For each of the meteorological datasets, daily time series of catchment areal averages were calculated using the catchment masks and methods described in Section 3. These timeseries are available for all CAMELS-GB catchments with no missing data.

Daily rainfall timeseries were derived from the CEH Gridded Estimates of Areal Rainfall dataset (CEH-GEAR) (Keller et al., 2015; Tanguy et al., 2016). This dataset consists of 1km^2 gridded estimates of daily rainfall for Great Britain and Northern Ireland from 1st January 1961 – 31st December 2015. The daily rainfall grids are derived using natural neighbour interpolation of a national database of quality-controlled, observed precipitations from the Met Office UK rain gauge network. It should be noted that the rainfall timeseries available in CAMELS-GB use the same underlying data but are not identical to catchment average rainfall series available from the NRFA which are derived using only 1km grid cells with >50% of their area within the catchment boundary.

Daily meteorological timeseries were derived from the Climate Hydrology and Ecology research Support System meteorology dataset (CHESS-met; Robinson et al., 2017a). The CHESS-met dataset consists of daily 1km^2 gridded estimates for Great Britain from 1st January 1961 – 31st December 2015 and includes several meteorological variables derived from observational data (see Table 1). CHESS-met was derived from the observation-based MORECS, which is a 40 km resolution gridded dataset, derived by interpolating daily station data (Hough and Jones, 1997; Thompson et al., 1981). The CHESS-met variables are obtained by downscaling MORECS variables to 1 km resolution and adjusting for local topography using lapse rates, modelled wind speeds and empirical relationships. CHESS-met air temperature and wind speed were directly downscaled from MORECS, specific humidity was calculated from MORECS vapour pressure, downward short-wave radiation was calculated from MORECS sunshine hours while long-wave radiation was calculated from the downscaled temperature, vapour pressure and sunshine hours (see Robinson et al 2017b for details).

Daily potential evapotranspiration timeseries were derived from the Climate Hydrology and Ecology research Support System Potential Evapotranspiration dataset (CHESS-PE; Robinson et al., 2016). The CHESS-PE dataset consists of daily 1km^2 gridded estimates of potential-evapotranspiration for Great Britain from 1st January 1961 – 31st December 2015. Potential evapotranspiration is calculated using the Penman-Monteith equation and CHESS-met datasets (see Robinson et al., 2017b). In recognition of the uncertainty in PET estimates, we provide two estimates of potential evapotranspiration available from CHESS-PE. The first estimate (PET) is calculated using the Penman-Monteith equation for FAO-defined well-watered grass (Allen et al., 1998) and is used to calculate all subsequent PET catchment attributes provided in CAMELS-GB. This estimate only accounts for transpiration and doesn't allow for canopy interception. The second estimate (PETI) uses the same meteorological data and the Penman-Monteith equation for well-watered grass but a correction is added for interception on days where rainfall has occurred (Robinson et al., 2017b). The seasonal differences between these two data products can be seen in Figure S10b (supplementary information). Generally, the PETI estimate with the interception correction is higher because interception is a more effective flux than transpiration under the same meteorological conditions. CHESS PETI can be between 5%-25% higher than CHESS PET at the grid-box level, whereas at a regional level, CHESS PETI is 7% higher than PET in England and 11% higher than PET in Scotland overall (Robinson et al., 2017b). In comparison to other PET products commonly used in GB, the CHESS PETI estimate is similar to grass-only MORECS (the United Kingdom Meteorological Office rainfall and evaporation calculation system; Hough and Jones, 1997) which has its own interception correction.

5.2 Hydrological Timeseries

Daily streamflow data for the 671 gauges were obtained from the UK NRFA on the 27th March 2019 using the NRFA API (https://nrfaapps.ceh.ac.uk/nrfa/nrfa-api.html, last access 11 December 2019). This data is collected by measuring authorities including the Environment Agency (EA), Natural Resources Wales (NRW) and Scottish Environmental Protection Agency (SEPA) and then quality controlled, on an ongoing annual cycle, before being uploaded to the NRFA site. Data are provided in $m^3 s^{-1}$ and mm day^{-1} calculated using catchment areas derived from the catchment boundaries described in Section 4.

Figure 1a shows the flow data availability for all gauges contained in the CAMELS-GB dataset covering different time periods. Over the 46 year time period (1970 – 2015), 60% (401) of the gauges have 5% missing flow data or less and 81% (542) of the gauges have 20% missing flow data or less. 97% (654) of the gauges have at least 20 years of data and 70% (468) of the gauges have at least 40 years of data. Figure 1b shows the number of years of available flow data for each CAMELS-GB gauge across Great Britain. Overall there is good spatial coverage of long flow timeseries across Great Britain, with slightly shorter timeseries concentrated in Scotland and in central GB.

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6 Catchment Attributes

6.1 Location, Area and Topographic Data

Catchment attributes describing the location and topography were extracted for each catchment from the NRFA (see Table 2). Catchment areas are calculated from the catchment masks described in Section 4. Catchment elevation is extracted from CEH's Integrated Hydrological Digital Terrain Model and the minimum, mean, maximum catchment elevation is provided alongside different percentiles (10th, 50th and 90th). Mean drainage path slope is also provided. This catchment attribute was developed for the Flood Estimation Handbook (Bayliss, 1999) and provides an index of overall catchment steepness by calculating the mean of all inter-nodal slopes from the IHDTM for the catchment.

6.2 Climatic Indices

Climatic indices were derived using the catchment daily rainfall, potential evapotranspiration and temperature time series described in section 5.1 (see Table 2). The Penman-Monteith formulation without correction for interception is used to calculate all PET catchment attributes provided in CAMELS-GB as it has more consistency with other global and national PET products. To provide consistency with previous CAMELS datasets, we compute the same climatic indices for all catchments in CAMELS-GB. However, it is important to note that in CAMELS-GB climatic indices are calculated for the full meteorological timeseries available in CAMELS-GB (water years from 1st Oct 1970 to 30th Sept 2015), whereas CAMELS and CAMELS-CL both use the water years from 1990 to 2009. The meteorological timeseries and code (https://github.com/naddor/camels, last access: 11 December 2019) are provided for users to calculate indices over different time periods if required.

6.3 Hydrologic Signatures

Hydrologic signatures were derived using the catchment daily discharge and rainfall time series described in section 5.1 and 5.2 (see Table 2). To provide consistency with the previous CAMELS datasets, we compute the same hydrologic signatures for all catchments in CAMELS-GB but add an additional formulation of baseflow index developed by the UK Centre for Ecology & Hydrology and commonly used in Great Britain (Gustard et al., 1992; see Appendix A and Figure S10a). Hydrologic signatures are calculated for the flow timeseries available during water years from 1st Oct 1970 to 30th Sept 2015 (previous CAMELS datasets calculated these metrics during water years from 1990 to 2009) using code available on github (https://github.com/naddor/camels, last access: 11 December 2019). We advise users to take the length of the flow timeseries and percentage of missing data (available in the hydrometry catchment attributes – see section 6.7) into account when comparing hydrologic signatures across catchments.

6.4 Land Cover Attributes

Land cover attributes for each catchment were derived from the UK Land Cover Map 2015 (LCM2015) produced by CEH (Rowland et al., 2017). While other land cover maps are available from CEH for 1990, 2000 and 2007, attributes are only provided for LCM2015 as different methods have been used to derive each of the land cover maps preventing straightforward analysis of changes in land cover over time. LCM2015 was chosen as it contains the most up-to-date data and methodology used to derive the land cover. LCM2015 uses a random forest classification of Landsat-8 satellite images based on the Joint Nature Conservation Committee (JNCC) Broad Habitats, encompassing the range of UK habitats.

In this study, the 1km percentage target class is used from the LCM2015 products, consisting of a 1km raster with 21 bands relating to the percentage cover value of different target classes that represent Broad Habitats. This is a significant number of land cover classes and so the 21 target classes were mapped to eight land cover classes; deciduous woodland, evergreen woodland, grass and pasture, shrubs, crops, suburban and urban, inland water, bare soil and rocks (see Appendix B). These are the same as the eight land cover classes used when running the JULES model with the CHESS meteorological driving data, and so provide consistency with other national scale efforts across Great Britain (Best et al., 2011; Blyth et al., 2019; Clark et al., 2011). For each catchment the percentage of the catchment covered by each of the eight land cover types was calculated and is provided in CAMELS-GB, alongside the most dominant land cover type (see Table 2).

Key limitations of this dataset are that the land cover attributes reflect a snapshot of the land cover in time and are subject to uncertainties in the Landsat-8 satellite images and the random forest classification. It is important to note that the land cover attributes provided in CAMELS-GB are different to those provided on the NRFA website which use LCM2000 and different land use groupings.

6.5 Soil Attributes

Soil attributes for each catchment were derived from the European Soil Database Derived Data product (Hiederer, 2013a, 2013b), and the Pelletier et al., (2016) modelled depth to bedrock global product. The European Soil Database (ESDB; European Commission Joint Research Centre, 2003) is the most detailed and comprehensive soils dataset available for Europe. It was selected for CAMELS-GB as no national soils datasets exist for GB that are both freely available and cover the same comprehensive range of soil descriptors.

As this dataset only characterises the top soil layers, we also used the Pelletier et al., (2016) modelled soil depth dataset to give an indication of the depth to unweathered bedrock extending up to 50m depth. Soil attributes for depth available to roots, percentage sand, silt and clay content, organic carbon content, bulk density and total available water content were calculated from the ESDB. We additionally estimated the saturated hydraulic conductivity and porosity (saturated volumetric water content) using two pedo-transfer functions, with the aim of providing one estimate consistent with CAMELS and a best estimate for European soil types. These were, (1) the widely-applied regressions based on sand and clay fractions first proposed by Cosby et al., (1984) based on soil samples across the United States, and (2) the HYPRES continuous pedotransfer functions using silt and clay fractions, bulk density and organic matter content developed using a large database of European soils (Wösten et al., 1999, 2001; Wösten, 2000) (see Appendix C for equations).

To estimate average values of all soil properties with depth, we calculated a weighted mean of the topsoil and subsoil data for each 1km grid cell. Weights were assigned based on the topsoil/subsoil proportion of the overall soil depth for that cell. Catchment average soil properties were calculated by taking the arithmetic mean (or harmonic mean for saturated hydraulic conductivity as advised in Samaniego et al., 2010) of all 1km grid cells that fell within the catchment boundaries. To give an indication of the distribution of soil properties across the catchment, the 5th, 50th and 95th percentile values of all grid cell values falling within the catchment boundaries was also calculated for all soil attributes apart from percentage sand, silt and clay. There were some grid cells where no soil data was available. Rather than set default values for these grid cells, we chose to exclude them from the calculations of catchment-average properties and provide the percentage of no-data cells within a catchment as an indication of the data availability of the catchment-average properties.

There are some key limitations associated with these datasets. Firstly, the soils information given on a 1km grid is only representative of the dominant soil typological class within that area. This means that much of the soil information is not represented in the soil maps, and the variation of soil properties within the 1km grid is lost. The high spatial heterogeneity of soils data means that correlations between soil property values given in the soil product and ground soil measurements are likely to be low (Hiederer, 2013a, 2013b). Secondly, as can be seen from Figure S10c-d in the

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supplement, there are large uncertainties relating to the choice of pedotransfer function. Care should be taken when interpreting results for saturated hydraulic conductivity, as the HYPRES equation is relatively inaccurate with a low R2 value of 0.19, and application of the single continuous pedotransfer function may result in poor results for some soil types (Wösten et al., 2001). Finally, it is important to be aware that measured soils data was unavailable for some urban areas including London, and these areas had been gap-filled (Hiederer, 2013a, 2013b).

6.6 Hydrogeological Attributes

Hydrogeological attributes for each catchment were derived from the UK bedrock hydrogeological map (BGS, 2019) and a new superficial deposits productivity map, both developed by the British Geological Survey. The UK bedrock hydrogeological map is an open source dataset that provides detailed information (at 1:625,000 scale) on the aquifer potential based on an attribution of lithology with seven classes of primary and secondary permeability and productivity (see Appendix D). The superficial deposits productivity map is a new dataset of similarly attributed superficial deposits aquifer potential across Great Britain (at 1:625,000 scale). These two datasets were chosen as they are the only two spatially continuous, consistently attributed hydrogeological maps of the bedrock and superficial deposits at the national scale for GB.

These two datasets were combined by superimposing the superficial deposits layer on top of the bedrock layer to provide catchment attributes for CAMELS-GB that characterise the uppermost geological layer (i.e. superficial deposits where present and bedrock where superficial deposits are absent). Combining the two datasets gave a total of nine hydrogeological productivity classes (see Appendix D). For each catchment, the percentage of the nine hydrogeological classes was calculated and is provided in CAMELS-GB (see Table 2). These nine classes indicate the influence of hydrogeology on river flow behaviour and describe the proportion of the catchment covered by deposits of high, moderate or low productivity and whether this is predominantly via fracture or intergranular flow (see Table 2). Such classifications have previously been used to enable correlations between catchment hydrogeology and measures of baseflow (Bloomfield et al., 2009).

Users should be aware that the aquifer productivity dataset is heuristic, based on hydrogeological inference that are based on mapped lithologies rather than on statistical analysis of borehole yields. It can be used for comparison between catchments at the regional to national scales. It should not be used at the sub-catchment scale where more refined hydrogeological information would be required to understand groundwater-surface water interactions. The hydrogeological attributes provided in CAMELS-GB will differ to those available on the NRFA website as CAMELS-GB uses the latest geological data.

6.7 Hydrometry and Discharge Uncertainty

Several attributes are provided in CAMELS-GB describing the gauging station type (i.e the type of weir, structure or measurement device used to measure flows) as listed on the NRFA, period of flow data available, gauging station discharge uncertainty and channel characteristics such as bankfull (see Table 2). The catchment attributes for discharge uncertainty are described in more detail below.

6.7.1 Discharge Uncertainty Estimates

Discharge uncertainty estimates for CAMELS-GB were calculated from a large data set of rating curves and stage-discharge measurements using a generalized framework designed to estimate place-specific discharge uncertainties outlined in Coxon et al, (2015). This framework estimates discharge uncertainties using a nonparametric locally weighted regression (LOWESS), where subsets of the stage-discharge data contained within a moving window are used to calculate the mean and variance at every stage point, which then define the LOWESS fitted rating curve and discharge uncertainty, respectively. Stage and discharge gauging uncertainties are incorporated into the framework by

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randomly sampling from estimated measurement error distributions to fit multiple LOWESS curves and then combining the multiple fitted LOWESS curves and variances in a Gaussian Mixture Model. Time-varying discharge uncertainties are accounted for by an automatic procedure where differences in historical rating curves are used to separate the stage-discharge rating data into subsets for which discharge uncertainty is estimated separately. The framework has been shown to provide robust discharge uncertainty estimates for 500 gauging stations across England and Wales (see Coxon et al., 2015 for more details).

For CAMELS-GB we extended the application of the framework to Scottish gauging stations to provide discharge uncertainty estimates across Great Britain. Discharge uncertainty estimates for CAMELS-GB catchments are provided for several flow percentiles (Q95, Q75, Q50, Q25, Q5 and Q1 derived from the flow timeseries provided in CAMELS-GB described in Section 5.2) for the most recent rating curve to allow users to evaluate discharge uncertainty across the flow range. The upper and lower bound of the discharge uncertainty prediction interval is provided as a percentage of the flow percentile for each catchment and flow percentile where available. In total discharge uncertainty estimates are available for 503 (75%) CAMELS-GB gauges. As the method is data based, the rating curve and its uncertainty interval cannot be computed for gauging stations where there are fewer than 20 stage-discharge measurements, or for flows above (below) the highest (lowest) stage-discharge measurement. This means that for some (or all) flow percentiles (particularly Q95 and Q1) there may be no discharge uncertainty estimate as indicated by 'NaN'. There are 45 stations where stage-discharge data were available, but discharge uncertainty estimates are not provided as the resulting uncertainty bounds were deemed to not accurately reflect the discharge uncertainty at that gauging station or because there was no sensible relationship between stage and discharge.

Users are advised that the CAMELS-GB discharge uncertainty estimates (1) are dependent on the types of error included in and underlying assumptions of the discharge uncertainty estimation method (see Kiang et al., 2018 for a comparison of seven discharge uncertainty estimation methods) and (2) may not be applicable to the whole flow timeseries (as they cover the most recent rating curve) or for stations where flow is measured directly (i.e. at ultrasonic or electromagnetic stations).

6.8 Human Influences

Providing information on the impact of humans in each catchment is a vital part of CAMELS-GB. To account for the degree of human intervention in each catchment we compiled data on reservoirs, abstraction and discharge returns provided by national agencies.

6.8.1 Benchmark Catchments

The UK Benchmark Network consists of 146 gauging stations that have been identified by the NRFA as suitable for the identification and interpretation of long-term hydrological variability and change against several criteria including length of record, quality of flow data, known impacts within the catchment and expert consultation (for a full description see Harrigan et al, 2018). Consequently, these gauging stations can be treated as relatively 'near-natural' and indicate that the influence of humans on the flow regimes of these catchments is modest. This data is available for all the CAMELS-GB catchments and data is provided for each catchment on whether it is part of the UK Benchmark Network or not.

6.8.2 Abstraction and Discharges

The abstraction data consists of monthly abstraction data from January 1999 – December 2014 that are reported by abstraction licence holders to the Environment Agency. These data are the actual abstraction returns and represent the total volume of water removed by the licence holder for each month over the time period. A mean daily abstraction rate for all English catchments is provided in CAMELS-GB for groundwater and surface water sources. The monthly returns for each abstraction

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licence in the database were averaged to provide a mean monthly abstraction from 1999 – 2014. All abstraction licences that fell within each catchment boundary (using the catchment masks outlined in section 4) were then summed for surface water and groundwater abstractions respectively and converted into mm day^{-1} using catchment area. The mean daily abstraction rate is provided alongside attributes describing the use of the abstracted water (agriculture, amenities, environmental, industrial, energy or for water supply). The discharge data consists of daily discharges into water courses from water companies and other discharge permit holders reported to the Environment Agency from 1st January 2005 – 31st December 2015. To calculate a mean daily discharge rate for each catchment, the daily discharge data for each discharge record was averaged and then all discharge records that fell within the catchment boundary were summed and then converted into mm day^{-1} using catchment area.

There are several important caveats associated with these data. Firstly, these data are only available for England. Consequently, there are many catchments where no data are available (identified by 'NaN') and only a proportion of the abstractions may have been accounted for catchments which lie on the border of England/Wales or England/Scotland. Furthermore, not all licence types/holders are required to submit records to the Environment Agency, therefore this is not the full picture of human intervention within each catchment. Secondly, the abstractions and discharges data cover different time periods. Thirdly, the topographical catchment mask was used to define which abstraction returns were included in each catchment. Groundwater abstractions that lie within the topographical catchment may not have a direct impact on the catchment streamflow and instead may impact a neighbouring catchment that shares the same aquifer. Conversely, groundwater abstractions that lie outside the catchment could have an impact on the catchment streamflow. Fourthly, there is a large inter-annual and intra annual variation in the abstraction and discharges data and their impacts will be different across the flow regime. Consequently, it is important that the mean abstraction totals are used as a guide to the degree of human intervention in each catchment rather than absolute totals of the abstraction for any given month. Finally, although 'abstractions' represent removed from surface water or groundwater sources, some of this water will be returned to catchment storages. The discharge data provided accounts just for treated water from sewage treatment works and does not provide information on other water returns that may be fed back into catchment storages. As such, the mean totals used here are a very broad guide. Other (less widely available) metrics have been applied in the UK which use modelling approaches to assess the net impact of abstractions/discharges across the whole flow regime (for example the Low Flows Enterprise methodology; see also Hannaford et al. 2013).

6.8.3 Reservoirs

Reservoir attributes are derived from an open source UK reservoir inventory (Durant and Counsell, 2018) supplemented with information from SEPA's publicly available controlled reservoirs register. The UK reservoir inventory includes reservoirs above 1,600 megalitre (ML) capacity, covering approximately 90% of the total reservoir storage in the UK. This dataset was collected from the Environment Agency through a Freedom of Information request, the UK Lakes Portal (CEH) and subsequent internet searches. It includes information on the location of the reservoir, its capacity, use and year the reservoir was built. To check the accuracy of this dataset, we cross-referenced the reservoirs in the UK reservoir inventory with reservoirs in the Global Reservoir and Dam (GRanD v1.3) database (Lehner et al., 2011). While many of the reservoirs and their capacity data was consistent for reservoirs for England and Wales, many Scottish reservoirs contained in the GRanD database were not present in the UK reservoir inventory or reported very different storage capacities. This is likely due to the estimation of storage capacities of Scottish reservoirs in the UK reservoir inventories (see Hughes et al., 2004) rather than actual storage capacities. Consequently, for reservoirs in Scotland, we used information from SEPA's publicly available controlled reservoirs register (http://map.sepa.org.uk/reservoirsfloodmap/Map.htm, last access: 11 December, 2019) including the reservoir name, location and storage capacity, and then supplemented this information with the year the reservoir was built and reservoir use by cross-referencing data from the UK reservoir inventory (users should be aware that reservoir use and the year the reservoir was built were not available for every reservoir).

For CAMELS-GB several reservoir attributes are derived for each catchment by determining the reservoirs that lie within the catchment mask from the reservoir locations and then calculating (1) the number of reservoirs in each catchment, (2) their combined capacity, (3) the fraction of that capacity that is used for hydroelectricity, navigation, drainage, water supply, flood storage and environmental purposes, and (4) the year when the first and last reservoir in the catchment was built.

6.9 Regional Variability in Catchment Characteristics

Figure 2 highlights some of the key catchment variables and in this section we discuss their regional variability (according to the regions in Figure 2a). Spatial maps of all catchment attributes can be found in the supplementary information.

There are distinct regional differences in climate across GB (Figure 2b). Precipitation is typically higher in the west and north of GB corresponding with the areas of high elevation and prevailing winds from the west that bring significant rainfall. The wettest areas of the UK are in mountainous regions with a maximum of 9.6 mm day^{-1} (3500mm year^{-1}) in the north-west. Snow fractions are generally very low across Great Britain (median snow fraction of 0.01) except for catchments in the Cairngorm mountains in north-east Scotland where the fraction of precipitation falling as snow can reach 0.17 (see supplementary information, Figure S4e). Precipitation is lowest in the south and east of GB with a minimum of 1.5mm day^{-1} in the east. In contrast, potential evapotranspiration (PET) is much less variable across GB with mean daily totals ranging from 1 to 1.5mm day^{-1}. PET is highest in the south (where temperatures are highest) and lowest in the north. Mean flow varies from 10 to 0.09mm day^{-1} and is typically higher in the north and west, reflecting the regional variability in precipitation and PET. This is also reflected in Figure 2c, where catchments in the north and west of GB tend to be wetter with higher runoff coefficients and catchments in the south and east are much drier with lower runoff coefficients. Figure 2c also shows that annual precipitation totals exceed annual PET totals; the aridity index is below 1 for all catchments reflecting the temperate and humid climate of GB. It is important to note that these estimates are dependent on the underlying data. For example, there can be significant variability in the calculation of PET, depending on the methods and assumptions used (e.g. Tanguy et al., 2018) and here we have used a PET estimate where canopy interception is not accounted for. Interception is an important component of the water cycle in GB, which experiences a large amount of low to moderate rainfall intensities (Blyth et al., 2019), thus using the CHESS PETI estimate instead would increase the aridity index above one in some locations.

There is also regional variability in baseflow index (the ratio of mean daily baseflow to daily discharge), which is typically higher in the south and east of GB and lower in the north-west. Some of these differences can be attributed to regional aquifers that have high/moderate productivity which are more prevalent in the south-east, east and north-east (see Figure 2b).

From Figure 2c, it is notable that runoff deficits significantly exceed total potential evapotranspiration for many of the CAMELS-GB catchments in the south-east – this could be due to water loss to regional aquifers, the issue of catchment areas not mapping onto the contributing area and/or due to the choice of PET used (see above). There are also seven catchments where the runoff exceeds total rainfall – this could be due to water gains from regional aquifers, catchment areas not mapping onto the contributing area, inter-basin transfers, uncertainties in the rainfall and/or under-estimation of rainfall. Many of the widely-used hydrological models and analysis techniques will not be able to reproduce catchment water balances which are outside the water and energy limitations shown in Fig 2c, unless the models or analysis techniques are explicitly adapted to consider the sources of

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uncertainty, potential unmeasured groundwater flow pathways and/or human influences that we have noted. We encourage users of the data to consider whether the assumptions of their methods are consistent with the uncertainties we have documented.

Land cover and human modifications can also impact river flows. Crops and grassland tend to be the dominant land cover for GB catchments, with crops typically the dominant land cover for catchments in the east and grassland for catchments in the west (Figure 2d). There is also a higher percentage of catchments in the east which are dominated by urban land cover. The highest proportion of reservoirs is concentrated in the more mountainous northern regions of GB, particularly in the North-East (Figure 2e).

7 Data Availability

The CAMELS-GB dataset (Coxon et al., 2020) detailed in this paper is freely available via the UK Centre for Ecology & Hydrology Environmental Information Data Centre (https://doi.org/10.5285/8344e4f3-d2ea-44f5-8afa-86d2987543a9). The data contain catchment masks, catchment time series and catchment attributes as described above. A full description of the data format is provided in the supporting documentation available on the Environmental Information Data Centre.

8 Conclusions

This study introduces the first large sample, open-source catchment dataset for Great Britain, CAMELS-GB (Catchment Attributes and MEteorology for Large-sample Studies), consisting of hydro-meteorological catchment timeseries, catchment attributes and catchment boundaries for 671 catchments. A comprehensive set of catchment attributes are quantified describing a range of catchment characteristics including topography, climate, hydrology, land cover, soils and hydrogeology. Importantly, we also derive attributes describing the level of human influence in each catchment and the first set of national discharge uncertainty estimates that quantify discharge uncertainty across the flow range.

The dataset provides new opportunities to explore how different catchment characteristics control river flow behaviour, develop common frameworks for model evaluation and benchmarking at regional-national scales and analyse hydrologic variability across the UK. To ensure the reproducibility of the dataset, many of the codes and datasets are made available to users.

While a wealth of data is provided in CAMELS-GB, there are many opportunities to expand the dataset that were outside the scope of this study. In particular, future work will concentrate on 1) expanding the dataset to include higher resolution data (such as hourly rainfall e.g. Lewis et al., 2018, and flow timeseries) and datasets for the analysis of trends (such as changes in land cover over time), 2) improving the comparability of CAMELS-GB with other CAMELS datasets by using common, global hydrometeorological and geophysical datasets to derive catchment timeseries and attributes, and 3) refining the characterisation of uncertainties in catchment attributes and forcing (particularly for rainfall data).

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Appendices

Appendix A Base flow index

The baseflow separation followed the Manual on Low-flow Estimation and Prediction of the World Meteorological Organization (2008). It relies on identifying local minima in daily streamflow series and producing a continuous baseflow hydrograph by linear interpolation between the identified local streamflow minima. The baseflow separation was performed using the R package lfstat (Koffler et al., 2016). The streamflow minima were identified using non-overlapping periods of N = 5 (block size) consecutive days and f = 0.9 as turning point parameter value.

Appendix B Land cover classes

We used the following classification to map the 21 land cover classes contained in the UK Land Cover Map 2015 to the eight land cover classes used in CAMELS-GB.

Table A1 Band ID and name from Land Cover Map (LCM) 2015 and corresponding land cover classes used in CAMELS-GB

Band	LCM2015 Band Name	CAMELS-GB Land Cover Classes
1	Broad-leaved Woodland	Deciduous woodland
2	Coniferous Woodland	Evergreen woodland
3	Arable and Horticulture	Crops
4	Improved Grassland	Grass and pasture
5	Neutral Grassland	Grass and pasture
6	Calcareous Grassland	Grass and pasture
7	Acid Grassland	Grass and pasture
8	Fen, marsh and swamp	Grass and pasture
9	Heather	Medium scale vegetation (shrubs)
10	Heather Grassland	Medium scale vegetation (shrubs)
11	Bog	Medium scale vegetation (shrubs)
12	Inland Rock	Bare soil and rocks
13	Saltwater	Not classified
14	Freshwater	Inland water
15	Supra-littoral Rock	Bare soil and rocks
16	Supra-littoral Sediment	Bare soil and rocks
17	Littoral Rock	Not classified
18	Littoral Sediment	Not classified
19	Saltmarsh	Inland water
20	Urban	Urban and suburban
21	Suburban	Urban and suburban

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Appendix C Soil pedo-transfer functions

We estimated the saturated hydraulic conductivity and porosity (also referred to as maximum water content, saturated water content, satiated water content) using two pedo-transfer functions.

The first was the widely-applied regressions based on sand and clay fractions first proposed by Cosby et al., (1984):

$$K_s = 2.54 * 10^{-0.6+0.012Sa-0.0064Cl}$$

$$\theta_s = 50.5 - 0.142Sa - 0.037Cl$$

Where K_s is saturated hydraulic conductivity in cm hour^{-1} and θ_s is porosity in percent (m^{3}m^{-3}). Predictor variables are Sand (Sa) and Clay (Cl).

The second, was the HYPRES continuous pedotransfer functions using silt and clay fractions, bulk density and organic matter content (Wösten et al., 1999; Wösten, 2000):

$$K_s = 0.04167\, e^{(7.755+0.0352Si+0.93Tp-0.967Db^2-0.000484Cl^2-0.000322Si^2+0.001Si^{-1}-0.0748Om^{-1}-0.643\ln(Si)-0.01398DbCl-0.1673DbOm+0.02986TpCl-0.03305TpSi)}$$

$$\theta_s = 0.7919 + 0.001691Cl - 0.29619Db - 0.000001491Si^2 + 0.0000821Om^2 + 0.02427Cl^{-1} + 0.01113Si^{-1} + 0.01472\ln(Si) - 0.0000733OmCl - 0.000619DbCl - 0.001183DbOm - 0.0001664TpSi$$

Where K_s is saturated hydraulic conductivity in cm hour^{-1} and θ_s is porosity (m^{3}m^{-3}). Predictor variables are Sand (Sa) and Clay (Cl). Predictor variables are Percentage Silt (Si), Percentage Clay (Cl), Percentage Organic Matter (Om), Bulk density (Db), and a binary variable for topsoil (Tp).

Appendix D Hydrogeological classes

For CAMELS-GB, we combined the BGS Hydrogeology map and superficial deposits layer. The table below provides a summary of the different classes in each dataset and how these were amalgamated to form the nine classes used in CAMELS-GB.

Table A2 Data source, class and description of the hydrogeological datasets

	Original Data		CAMELS-GB	
Data Source	**Class ID**	**Description**	**Class ID**	**Description**
British Geological Survey Hydrogeology Map (BGS, 2019)	1	Aquifers with significant intergranular flow – highly productive	1	Significant intergranular flow – high productivity
	2	Aquifers with significant intergranular flow – moderately productive	2	Significant intergranular flow – moderate productivity
	3	Aquifers with significant intergranular flow – low productivity	3	Significant intergranular flow – low productivity
	4	Aquifers in which flow is virtually all through fractures – highly productive	4	Flow through fractures – high productivity
	5	Aquifers in which flow is virtually all through fractures – moderately productive	5	Flow through fractures – moderate productivity

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	6	Aquifers in which flow is virtually all through fractures – low productivity	6	Flow through fractures – low productivity
	7	Rocks with essentially no groundwater	7	Rocks with essentially no groundwater
British Geological Survey Superficial Deposits Layer	8	Moderate productivity	2	Significant intergranular flow – moderate productivity
	9	Low productivity	3	Significant intergranular flow – low productivity
	10	Generally low productivity but some not a significant aquifer	8	Generally low productivity (intergranular flow) but some not a significant aquifer
	11	Generally not a significant aquifer but some low productivity	9	Generally not a significant aquifer but some low productivity (intergranular flow)
	12	Not a significant aquifer	7	Rocks with essentially no groundwater

Author Contribution

G Coxon initiated and led the project. G Coxon produced the CAMELS-GB dataset with the following contributions: 1) N Addor derived the climatic indices and hydrologic signatures, 2) R Lane derived the soil attributes, 3) M Lewis derived the superficial deposits geological layer and provided guidance with J Bloomfield on deriving the hydrogeological attributes, 4) E Robinson provided guidance on the meteorological datasets (CHESS-met and CHESS-PE) and mapping the land cover data, and 5) M Fry provided the streamflow data, catchment masks and all catchment attributes sourced from the National River Flow Archive. All co-authors contributed to the design of the dataset. The manuscript was prepared by G Coxon with contributions from all co-authors.

Competing Interests

The authors declare that they have no conflict of interest.

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Tables

Table 1 Summary table of catchment hydro-meteorological timeseries available in CAMELS-GB

Timeseries Class	Timeseries Name	Description	Unit	Data Source
Meteorological Timeseries (available from 1st October 1970 – 30th September 2015)	precipitation	catchment daily averaged precipitation	mm day^{-1}	CEH-GEAR (Keller et al., 2015; Tanguy et al., 2016)
	pet	catchment daily averaged potential evapotranspiration for a well-watered grass (Penman-Monteith equation)	mm day^{-1}	CHESS-PE (Robinson et al., 2017a, 2017b)
	peti	catchment daily averaged potential evapotranspiration for a well-watered grass (Penman-Monteith equation with a correction added for interception on days where rainfall has occurred)	mm day^{-1}	
	temperature	catchment daily averaged temperature	°C	CHESS-met (Robinson et al., 2017a)
	windspeed	catchment daily averaged wind speed	m s^{-1}	
	humidity	catchment daily averaged specific humidity	g kg^{-1}	
	shortwave_rad	catchment daily averaged downward short wave radiation	W m^{-2}	
	longwave_rad	catchment daily averaged longwave radiation	W m^{-2}	
Hydrological Timeseries (available from 1st October 1970 – 30th September 2015)	discharge_spec	catchment specific discharge (converted to mm day^{-1} using catchment areas described in Section 6.1)	mm day^{-1}	UK National River Flow Archive using the NRFA API*
	discharge_vol	catchment discharge	m^{3} s^{-1}	

* https://nrfaapps.ceh.ac.uk/nrfa/nrfa-api.html, data downloaded on the 27th March 2019, last access to website 11 December 2019

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Table 2. Summary table of catchment attributes available in CAMELS-GB

Attribute Class	Attribute Name	Description	Unit	Data Source
Location and Topography	gauge_id	catchment identifier (corresponds to the gauging station ID provided by the NRFA)	-	UK National River Flow Archive using the NRFA API*
	gauge_name	gauge name (river name followed by gauging station name)	-	
	gauge_lat	gauge latitude	°	
	gauge_lon	gauge longitude	°	
	gauge_easting	gauge easting	m	
	gauge_northing	gauge northing	m	
	gauge_elev	gauge elevation	m.a.s.l	
	area	catchment area	km^2	CEH's Integrated Hydrological Digital Terrain Model (Morris and Flavin, 1990)
	dpsbar	catchment mean drainage path slope	$m\ km^{-1}$	
	elev_mean	catchment mean elevation	m.a.s.l	
	elev_min	catchment minimum elevation	m.a.s.l	
	elev_10	catchment 10^{th} percentile elevation	m.a.s.l	
	elev_50	catchment median elevation	m.a.s.l	
	elev_90	catchment 90^{th} percentile elevation	m.a.s.l	
	elev_max	catchment maximum elevation	m.a.s.l	
Climatic Indices (computed for 1st Oct 1970 to 30th Sept 2015)	p_mean	mean daily precipitation	$mm\ day^{-1}$	Catchment timeseries of precipitation, potential evapotranspiration and temperature described in Section 5.1 and Table 1
	pet_mean	mean daily PET (Penman-Monteith equation without interception correction)	$mm\ day^{-1}$	
	aridity	aridity, calculated as the ratio of mean daily potential evapotranspiration to mean daily precipitation	-	
	p_seasonality	seasonality and timing of precipitation (estimated using sine curves to represent the annual temperature and precipitation cycles; positive (negative) values indicate that precipitation peaks in summer (winter) and values close to zero indicate uniform precipitation throughout the year)	-	
	frac_snow	fraction of precipitation falling as snow (for days colder than 0°C)	-	
	high_prec_freq	frequency of high precipitation days (≥ 5 times mean daily precipitation)	$days\ yr^{-1}$	
	high_prec_dur	average duration of high precipitation events (number of consecutive days ≥ 5 times mean daily precipitation)	days	
	high_prec_timing	season during which most high precipitation days (≥ 5 times mean daily precipitation) occur. If two seasons register the same number of events, a value of NaN is given.	season	
	low_prec_freq	frequency of dry days (< $1mm\ day^{-1}$)	$days\ yr^{-1}$	
	low_prec_dur	average duration of dry periods (number of consecutive days < $1mm\ day^{-1}$)	days	
	low_prec_timing	season during which most dry days (< $1mm\ day^{-1}$) occur. If two seasons	season	

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		register the same number of events, a value of NaN is given.		
Hydrologic Signatures (computed for 1st Oct 1970 to 30th Sept 2015)	q_mean	mean daily discharge	mm day^{-1}	Catchment timeseries of streamflow and precipitation described in Sections 5.2 and 5.1 respectively, and Table 1
	runoff_ratio	runoff ratio, calculated as the ratio of mean daily discharge to mean daily precipitation	-	
	stream_elas	streamflow precipitation elasticity (sensitivity of streamflow to changes in precipitation at the annual timescale, using the mean daily discharge as reference)	-	
	slope_fdc	slope of the flow duration curve (between the log-transformed 33rd and 66th streamflow percentiles)	-	
	baseflow_index	baseflow index (ratio of mean daily baseflow to daily discharge, hydrograph separation performed using the Ladson et al., 2013 digital filter)	-	
	baseflow_index_ceh	baseflow index (ratio of mean daily baseflow to daily discharge, hydrograph separation performed using the Gustard et al., 1992 method described in Appendix A)	-	
	hfd_mean	mean half-flow date (date on which the cumulative discharge since 1 October reaches half of the annual discharge)	days since 1st October	
	Q5	5% flow quantile (low flow)	mm day^{-1}	
	Q95	95% flow quantile (high flow)	mm day^{-1}	
	high_q_freq	frequency of high-flow days (> 9 times the median daily flow)	days yr^{-1}	
	high_q_dur	average duration of high flow events (number of consecutive days >9 times the median daily flow)	days	
	low_q_freq	frequency of low flow days (< 0.2 times the mean daily flow)	days yr^{-1}	
	low_q_dur	average duration of low flow events (number of consecutive days < 0.2 times the mean daily flow)	days	
	zero_q_freq	frequency of days with Q = 0	%	
Land Cover Attributes	dwood_perc	percentage cover of deciduous woodland	%	1km percentage target class, Land Cover Map 2015 (Rowland et al., 2017)
	ewood_perc	percentage cover of evergreen woodland	%	
	grass_perc	percentage cover of grass and pasture	%	
	shrub_perc	percentage cover of medium scale vegetation (shrubs)	%	
	crop_perc	percentage cover of crops	%	
	urban_perc	percentage cover of suburban and urban	%	
	inwater_perc	percentage cover of inland water	%	
	bares_perc	percentage cover of bare soil and rocks	%	
	dom_land_cover	dominant land cover (the land cover class that has the highest percentage cover in each catchment)	-	
	sand_perc	percentage sand	%	

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Soil Attributes Each soil attribute (apart from percentage sand, silt, clay and organic content) is accompanied by the 5th, 50th and 95th percentile of that attribute across the catchment and the percentage missing	silt_perc	percentage silt	%	European Soil Database Derived Data product (Hiederer, 2013a, 2013b), and the modelled depth to bedrock global product (Pelletier et al., 2016b)
	clay_perc	percentage clay	%	
	organic_perc	percentage organic content	%	
	bulkdens	bulk density	g cm^{-3}	
	tawc	total available water content	mm	
	porosity_cosby	volumetric porosity (saturated water content estimated using a pedotransfer function based on sand and clay fractions)	-	
	porosity_hypres	volumetric porosity (saturated water content estimated using a pedotransfer function based on silt, clay and organic fractions, bulk density and topsoil)	-	
	conductivity_cosby	saturated hydraulic conductivity (estimated using a pedotransfer function based on sand and clay fractions)	cm h^{-1}	
	conductivity_hypres	saturated hydraulic conductivity (estimated using a pedotransfer function based on silt, clay and organic fractions, bulk density and topsoil)	cm h^{-1}	
	root_depth	depth available for roots	m	
	soil_depth_pelletier	depth to bedrock (maximum 50m)	m	
Hydrogeology Attributes	inter_high_perc	significant intergranular flow – high productivity	%	British Geological Survey hydrogeology map (BGS, 2019) and superficial deposits map
	inter_mod_perc	significant intergranular flow – moderate productivity	%	
	inter_low_perc	significant intergranular flow – low productivity	%	
	frac_high_perc	flow through fractures – high productivity	%	
	frac_mod_perc	flow through fractures – moderate productivity	%	
	frac_low_perc	flow through fractures – low productivity	%	
	no_gw_perc	rocks with essentially no groundwater	%	
	low_nsig_perc	generally low productivity (intergranular flow) but some not significant aquifer	%	
	nsig_low_perc	generally not significant aquifer but some low productivity (intergranular flow)	%	
Hydrometry	station_type	gauging station type denoted by the following abbreviations (**B** Broad-crested weir; **C** Crump profile single-crest weir; **CB** Compound broad-crested weir; **CC** Compound Crump weir; **EM** Electromagnetic gauging station; **EW** Essex weir; **FL** Flume; **FV** Flat V triangular profile weir; **MIS** Miscellaneous; **TP** Rectangular thin-plate weir; **US** Ultrasonic gauging station; **VA** Velocity-area gauging station; **VN** Triangular (V notch) thin-plate weir). Two abbreviations may be	-	UK National River Flow Archive using the NRFA API*

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		applied to each station relating to the measurement of low or high flows.		
	flow_period_start	first date that daily flow time series provided in CAMELS-GB is available for this gauging station	-	Catchment timeseries of streamflow described in Section 5.2
	flow_period_end	end date that daily flow time series provided in CAMELS-GB are available for this gauging station	-	
	flow_perc_complete	percentage of days with flow time series available from 1st October 1970 – 31st September 2015	%	
	bankfull_flow	flow at which the river begins to overlap the banks at a gauging station (obtained from stage-discharge relationships so may be derived by extrapolation)	$m^3 s^{-1}$	UK National River Flow Archive using the NRFA API*
	structurefull_flow	flow at which the river begins to the wingwalls of a structure at a gauging station (obtained from stage-discharge relationships so may be derived by extrapolation)	$m^3 s^{-1}$	
	qXX_uncert_upper	upper bound of the discharge uncertainty interval for the XX percentile flow given as a percentage of the XX percentile flow – estimates for XX values of 5, 25, 50, 75, 95, 99 are provided	%	Derived from Coxon et al (2015)
	qXX_uncert_lower	lower bound of the discharge uncertainty interval for the XX percentile flow given as a percentage of the XX percentile flow – estimates for XX values of 5, 25, 50, 75, 95, 99 are provided	%	
	quncert_meta	metadata describing the reasons why discharge uncertainty estimates are (not) provided; **Calculated discharge uncertainties**; **No stage-discharge measurements available**; **Less than 20 stage-discharge measurements available for most recent rating**; **Discharge uncertainty estimates not provided** as the estimated uncertainty bounds were deemed to not accurately reflect the discharge uncertainty or because there was no sensible relationship between stage and discharge.	-	
Human Influence Attributes	benchmark_catch	benchmark catchment (Y indicates the catchment is part of the UK Benchmark Network, while N indicates that it is not)	Y/N	UK National River Flow Archive; Harrigan et al., (2018)
	surfacewater_abs	mean surface water abstraction	mm day^{-1}	Abstractions and discharges sourced
	groundwater_abs	mean groundwater abstraction	mm day^{-1}	
	discharges	mean discharges (daily discharges into water courses from water	mm day^{-1}	

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		companies and other discharge permit holders reported to the Environment Agency)		from the Environment Agency
	abs_agriculture_perc	percentage of total (groundwater and surface water) abstractions in catchment for agriculture	%	
	abs_amenities_perc	percentage of total (groundwater and surface water) abstractions in catchment for amenities	%	
	abs_energy_perc	percentage of total (groundwater and surface water) abstractions in catchment for energy production	%	
	abs_environmental_perc	percentage of total (groundwater and surface water) abstractions in catchment for environmental purposes	%	
	abs_industry_perc	percentage of total (groundwater and surface water) abstractions in catchment for industrial, commercial and public services	%	
	abs_watersupply_perc	percentage of total (groundwater and surface water) abstractions in catchment for water supply	%	
	num_reservoir	number of reservoirs in the catchment	-	UK Reservoir Inventory (Durant and Counsell, 2018) and SEPA's publicly available controlled reservoirs register (http://map.sepa.org.uk/reservoirsfloodmap/Map.htm, last access: 11 December, 2019)
	reservoir_cap	total storage capacity of reservoirs in the catchment in megalitres	ML	
	reservoir_he	percentage of total reservoir storage in catchment used for hydroelectricty	%	
	reservoir_nav	percentage of total reservoir storage in catchment used for navigation	%	
	reservoir_drain	percentage of total reservoir storage in catchment used for drainage	%	
	reservoir_wr	percentage of total reservoir storage in catchment used for water resources	%	
	reservoir_fs	percentage of total reservoir storage in catchment used for flood storage	%	
	reservoir_env	percentage of total reservoir storage in catchment used for environmental	%	
	reservoir_nousedata	percentage of total reservoir storage in catchment where no use data were available	%	
	reservoir_year_first	year the first reservoir in the catchment was built	-	
	reservoir_year_last	year the last reservoir in the catchment was built	-	

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Figures

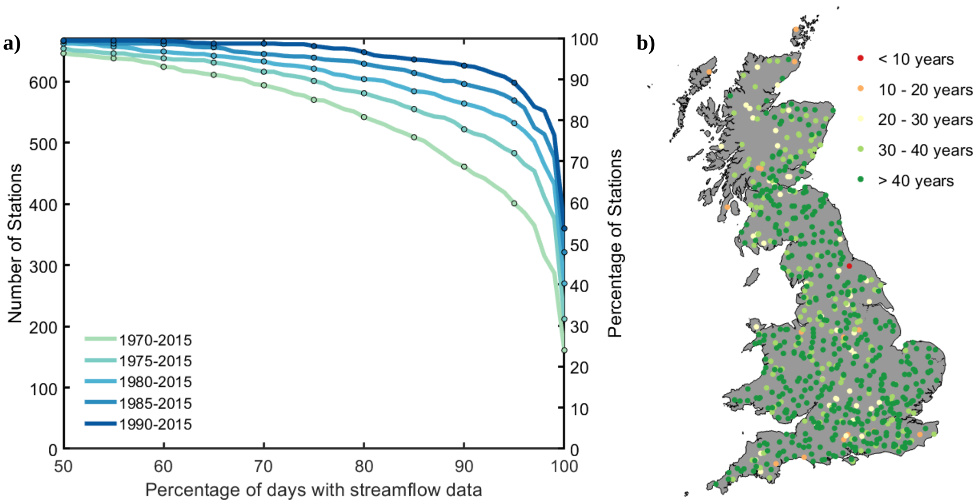

Figure 1. a) Number of stations with percentage of available streamflow data for different periods, b) Length of the flow time series for each gauge

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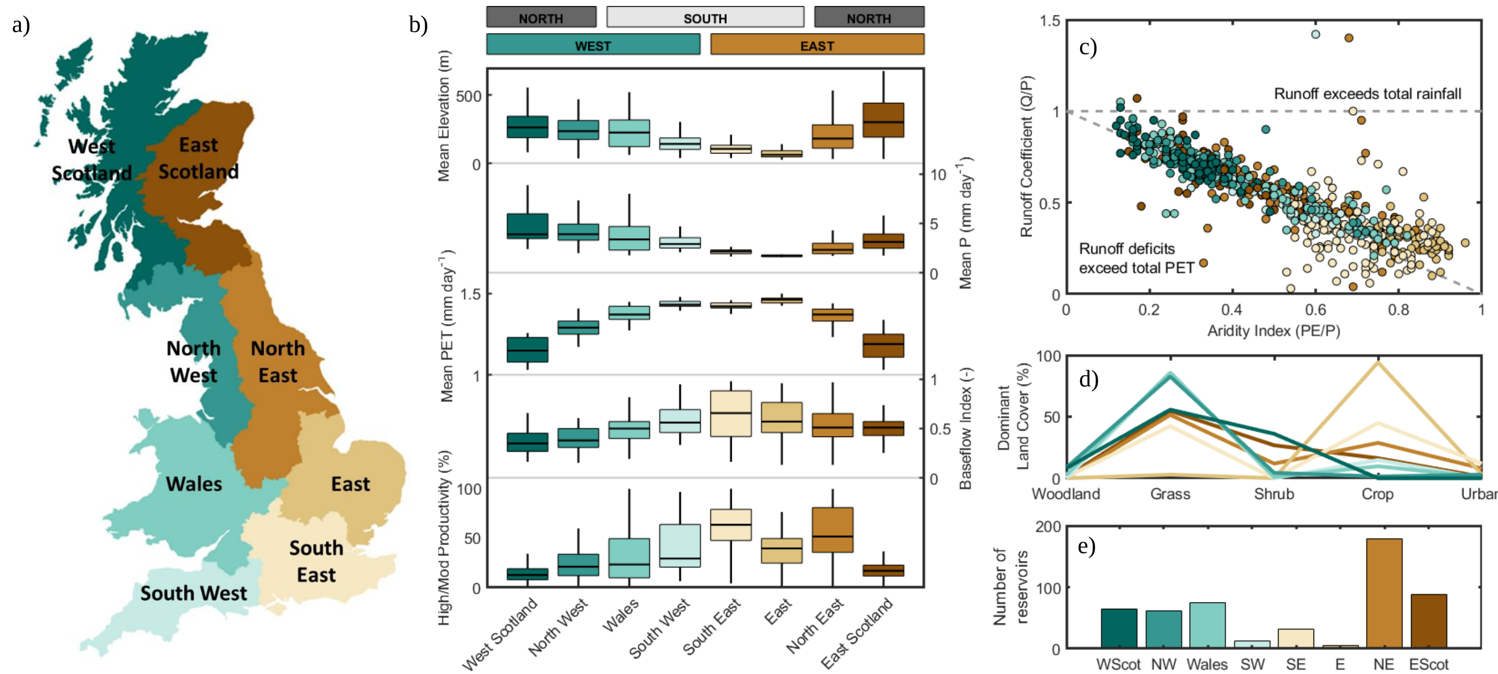

Figure 2. Key catchment variables a) Map of Great Britain river basin regions based on the UKCP09 river basins aggregated from 21 river basin districts to 8 regions, b) boxplots showing the range of key climatic, hydrology and hydrogeologic (defined by the proportion of the catchment underlain by high or moderate productivity aquifers) catchment attributes for each region, c) budyko-type curve relating runoff coefficient to aridity index, points are coloured by region and the y-axis is limited to 1.5 which means two gauges (26006 and 27038) are not shown that have a runoff coefficient of 3, d) the percentage of catchments in each region with a dominant land cover of either woodland (evergreen and deciduous), grass, shrub, crop or urban, e) the number of reservoirs in the catchments in each region