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OPEN Time series of freshwater macroinvertebrate abundances DATA DESCRIPTOR and site characteristics of European streams and rivers

Ellen A. R. Welti et al.#

Freshwater macroinvertebrates are a diverse group and play key ecological roles, including accelerating nutrient cycling, filtering water, controlling primary producers, and providing food for predators. Their differences in tolerances and short generation times manifest in rapid community responses to change. Macroinvertebrate community composition is an indicator of water quality. In Europe, efforts to improve water quality following environmental legislation, primarily starting in the 1980s, may have driven a recovery of macroinvertebrate communities. Towards understanding temporal and spatial variation of these organisms, we compiled the TREAM dataset (Time seRies of European freshwAter Macroinvertebrates), consisting of macroinvertebrate community time series from 1,816 river and stream sites (mean length of 19.2 years and 14.9 sampling years) of 22 European countries sampled between 1968 and 2020. In total, the data include > 93 million sampled individuals of 2,648 taxa from 959 genera and 212 families. These data can be used to test questions ranging from identifying drivers of the population dynamics of specific taxa to assessing the success of legislative and management restoration efforts.

Background & Summary

Freshwater ecosystems have been altered by a wide range of human impacts since the onset of the industrialization^{1,2}, which have increased in intensity in Europe since World War II³. Legislation in Europe and elsewhere (e.g., CWA 1972⁴, WFD 2000⁵) has aimed to restore freshwater and riparian habitats, especially through improvements in wastewater treatment and reductions in airborne pollutants⁶. Other stressors such as climate change, land use change, physical alterations to waterways, including straightening, deepening, and impoundments, micropollutants, and invasive species, have either increased in intensity or continue to threaten freshwater ecosystems⁷. Moreover, despite some improvements, most freshwater ecosystems still fail to meet the European Water Framework Directive's target of good ecological status⁸. Given freshwater biotas' high biodiversity, importance for ecosystem functioning and provision of ecosystem services, monitoring the state of freshwater ecosystems remains an important task for managers and scientists.

Freshwater macroinvertebrates are commonly used as indicators of the status of freshwater ecosystems because of their taxon-specific sensitivity to environmental conditions. Long-term trends of freshwater communities have typically been examined at local to national scales9-11. To study large-scale patterns in biodiversity change, many efforts have been made to compile data on the abundance and occurrence of species into harmonized datasets (e.g., BioTIME¹², Living Planet Index¹³, CESTES¹⁴). However, both freshwater biota and all invertebrates are often underrepresented in datasets of species abundance and occurrences. Currently, datasets focused on freshwater ecosystems include an earlier but smaller dataset of European macroinvertebrates¹⁵, macroinvertebrate communities in intermittent rivers¹⁶, abundance time-series data for fish¹⁷, and occurrence records for selected macroinvertebrate groups¹⁸. Freshwater insects (which constitute a large proportion of macroinvertebrates) are also represented by the InsectChange dataset¹⁹, but this dataset targeted community-level aggregated data and only the most publicly accessible time-series. In the United States, freshwater communities

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have been sampled across many sites by federal agencies, but most sites have been sampled only one to three times in the past three decades²⁰. Hence, the TREAM (Time seRies of European freshwAter Macroinvertebrates) dataset, presented here, fills an important gap in data availability.

Here, we provide a new dataset of time-series abundance data for freshwater macroinvertebrates in Europe. We used our collaborator network to identify unpublished datasets and published but inaccessible datasets collected within different studies and in different countries. In total, data were sourced from 82 data co-authors including scientists, site managers, and representatives of regulatory agencies from 22 European countries. The TREAM dataset presented here differs from previously published datasets in its inclusion of longer time series (minimum of eight sampling years), of time series reporting all sampled members of the macroinvertebrate community (e.g. not time series of only Odonata or only Ephemeroptera, Plecoptera, and Trichoptera), and its larger number of time series (1,816) within one taxonomic group. Our dataset is outlined in detail below.

We have recently used our dataset to explore how trends in freshwater macroinvertebrate communities have changed through time¹. The findings from this study indicate an overall increase in the abundance and richness of freshwater macroinvertebrates in Europe since 1968. However, the recovery has slowed since 2010. We further analysed site-level variation in trends in relation to environmental drivers, such as climate and land use¹. Our dataset can be used for population-level and community-level analyses that address a wide range of questions about spatio-temporal patterns in taxonomic or functional taxon units^{21,22}.

Methods

Data call and criteria. Data were sourced following a data call to water quality managers and freshwater ecologists in Europe from our professional networks. To be included in the TREAM dataset, data had to meet a number of criteria. First, we only included data that formed a time series (i.e., repeated sampling at the same location), with a minimum of eight sampling years, but sampling years did not have to be consecutive. Second, each time series represents whole sampled macroinvertebrate communities (i.e., all macroinvertebrates sampled were recorded). Third, we included only time series with counts or estimates of taxon abundance at the family, genus, or species level. Fourth, sampling location and method must be consistent over time within each time series, including the sampling method, the sampling effort, and the taxonomic resolution to which individual taxa were identified (see Usage notes regarding taxonomic resolution). Finally, all samples in a time series had to be collected within a period of three consecutive months of the year (e.g., June-August). Data were collected as part of 41 independent monitoring projects; project ID numbers and data provider names are listed in the file "TREAM_siteLevel.csv" within the deposited dataset²³. Projects reported data from one or more sampling locations per country (Fig. 1; Austria: 2 time series, Belgium: 82, Bulgaria: 9, Cyprus: 2, Czechia: 1, Denmark: 248, Estonia: 10, Finalnd: 10, France: 307, Germany: 151, Hungary: 87, Ireland: 16, Latvia: 3, Luxembourg: 20, Netherlands: 51, Norway: 67, Portugal: 2, Spain: 245, Sweden: 91, Switzerland: 1, United Kingdom: 406).

Data standardization. Following the data call, we standardized data by filtering subsets that did not meet our criteria listed above. For example, we excluded samples or whole time series for which sampling methods, effort, or primary level of taxonomic resolution changed during the time series (see Usage notes for a caveat regarding taxonomic resolution). When time series did not meet our inclusion criteria, including consistent sampling methods, for the minimum of eight sampling years, we excluded the whole time series. We only allowed one sampling event per year defined as sampling that occurred within a single day. For time series where multiple samples were provided within the same year, we excluded any sampling outside the most sampled period of three consecutive months. We excluded taxa that are not freshwater invertebrates, e.g., terrestrial taxa, vertebrates, and microinvertebrates that were not present in most of the time series, such as mites and copepods. Taxon names were harmonized according to freshwaterecology.info²⁴.

Data description. The resulting TREAM dataset consists of 1,816 time series from 22 European countries. Time series had a mean of 14.9 (median: 14; range: 8–32) sampling years (not necessarily continuous, spanning 8–39 years). The most commonly sampled families by abundance were Chironomidae, Gammaridae, and Hydrobiidae, and by frequency of occurrence in site sampling years were Chironomidae, Baetidae, and Elmidae. Taxonomically, samples within time series were resolved to family level or finer (762 time series mainly at species level resolution, 537 at genus/mixed level, and 517 at family level). Samples were taken from the riverbed sediments and the majority of collections used forms of kick-net sampling, with either a fixed area or time-based sampling effort. Metadata including information on sampling effort and methods in terms of sampling approach and units are provided in the Knowledge Network for Biocomplexity repository²³.

Environmental data. For each sampling location, we included stream characteristics (listed below), climate (cumulative precipitation and mean maximum temperature during the 12-month period before the sampling period), land cover (urban and agricultural land use), and dam impact score. To determine the sampling area of each site to extract these environmental data, we first assigned each sampling site to the corresponding stream segment of the Hydrography90m stream network²⁵ using the *v.net* function in GRASS GIS²⁶. The longest distance within the stream segment subcatchment was used as the distance threshold for the snapping function. We then computed the upstream catchment of each sampling site using the *r.water.outlet* function and stored the output as a raster GeoTIFF file. This raster file served as the basis for the subsequent environmental data extractions. While the point snapping and catchment delineation procedure were automated, we reviewed each resulting catchment. In cases where the site was assigned to the wrong segment, we manually corrected the stream segment assignment for a given sampling site and repeated the procedure. For stream characteristics, we extracted the topographical and topological attributes of each stream segment of the Hydrography90m dataset²⁵. Specifically, we provide the extracted area of each site's subcatchment, Strahler stream order, flow accumulation, elevation, and slope.



Fig. 1 Overview of dataset coverage including the distribution of sampling sites across 22 European countries (**a**), a histogram of the number of sampling sites per a given first year of sampling (**b**), the number of sampling years (median and interquartile range) per site within each country (**c**), and the top 15 most sampled orders and their occurrences across all surveys (**d**).

For climate data, we used the TERRA Climate monthly time series from 1967–2020 at 4-km resolution²⁷ and overlaid the monthly precipitation and temperature raster layers with the upstream catchment of each sampling site using the *r.univar* function in GRASS GIS. In addition, we extracted the local temperature layers at each sampling site using the *gdallocationinfo* function. This resulted in monthly averaged upstream aggregated precipitation (in mm) and monthly averaged local maximum temperature (in °C). For each site and year, we then calculated the average of the monthly temperatures, and the sum of the monthly precipitation values for the year preceding the average month of sampling (e.g., for a site with most sampling occurring in May, year was defined as the preceding 12 months of May until April). The use of climatic data from one year prior to the start of the first time series allows the capture of the annual period of weather prior to the first sampling event,

since population responses are typically lagged. We used Bayesian models in the program R v. 4.2²⁸ and package brms²⁹ to calculate trends of change in precipitation and temperature over time (all sampling years of a given site/time series). Code for climate trends can be found at: https://github.com/Ewelti/EuroAquaticMacroInverts/blob/main/R/Initial_Biodiversity_FuncTrait_and_climate_calcs/climateTrends_brms.R.

To collect data on land cover, we used the ESA Land Cover CCI Product³⁰ at 300-m spatial resolution to extract the percent coverage of the urban and crop land-cover categories from annual maps, available from 1992–2018. We first split each annual multi-category land cover raster layer into the single raster layers that hold only each single category. We then overlaid the urban and crop land cover category raster layers with the subcatchment raster layers and used the *r.univar* function to yield annual cover of the upstream urban and crop land cover within the catchment of each sampling site.

We used the GRanD dataset³¹ to obtain the dams across the study area domain. First, we assigned each dam to the Hydrography90m network using the *v.net* function and using the longest distance within the stream segment subcatchment as the distance threshold. Afterwards, we visually checked the spatial assignment. We then computed the distance from each sampling site to the next upstream dam using the GRASS GIS function *v.net.distance*²⁵. To create a proxy for dam impacts, we first extracted only dams within 100 km of a site that were connected to the site and upstream of the site. We then calculated a dam impact score for each site as:

$$\sum_{i=1}^{n} \frac{100 - d_i}{100}$$

where *n* represents the number of dams, *i* denotes a given dam, and *d* is the distance (km) of the dam from a site.

Data Records

Data are available from the Knowledge Network for Biocomplexity repository (https://knb.ecoinformatics.org/ view/doi:10.5063/F1NG4P4R)²³. Data are provided in three csv files listed below:

TREAM_siteLevel.csv. This file contains summary TREAM data where each row corresponds to a site (a site = location of one time series). Data include site characteristics (e.g. coordinates, number of sampling years, sampling season), data providers, and primary taxonomic resolution level, summary information on biodiversity across the time series (e.g. total taxa richness for all sampling years). The majority of available corresponding environmental data (i.e., dam impact score, mean values across sampling years of the mean temperature and precipitation in the 12 months prior to sampling, slope of temperature and precipitation over the sampled years, and percent upstream area of cropland and urban cover) and stream characteristic variables (i.e., area of subcatchment, Strahler stream order, flow accumulation, elevation, and slope) are included in this file in addition to sampling method information (e.g., sampling units, primary taxonomic resolution).

TREAM_siteYEARLevel.csv. This file contains TREAM data where each row corresponds to a site and a year of sampling. This file is the primary source for biodiversity summary metrics (e.g., taxa richness, functional redundancy, abundance of native taxa for each site and year) associated with this dataset. Environmental data that are available on an annual basis (i.e., mean temperature and precipitation in the 12 months prior to sampling, and percent upstream area of cropland and urban cover) are also provided in this file.

TREAM_alltaxa.csv. This file contains raw biodiversity data where each row corresponds to a taxon within a site and year of sampling. The abundance of each taxon per sampling unit (i.e., all sampling within one site and one time point) is included in this file (see TREAM_siteLevel.csv for sampling methods and effort information).

Technical Validation

Technical validation of the TREAM dataset was achieved through exclusion of time series data that did not match our inclusion criteria and data standardisation steps (outlined in Methods above). Any noted issues that did not adhere to the outlined standardisation within the datasets from the 41 independent projects included in this dataset were checked with data providers and corrected or removed when standardisation was not achievable (e.g., when collection methods changed over the course of the time series).

Usage Notes

Several key characteristics about these data should be noted by future data users. First, and most importantly, sampling methods and effort, and seasonality are standardised within individual time series but can vary across time series. This means that raw data are not directly comparable across the 41 independent projects included in the TREAM dataset.

Second, while we use the taxonomic backbone of freshwaterecology.info²³ as it is a common tool used by European freshwater ecologists, we are aware that it does not capture all recent changes to taxonomic names.

Third, two pairs of time series overlap in sampling locations: 1) Site ID = 100000001 (SVD) & 100000309 (Bugey_SVD) refer to the same location; 2) Site ID = 100000002 (SVG) & 100000308 (Bugey_SVG) refer to the same location. The data from these sites were collected by the Institut national de la recherche agronomique (referent: Maxence Forcellini) between 1980 and 2014, and by Électricité de France (Referent: Anthony Maire) between 2000 and 2019.

Fourth, although standardised taxonomic resolution within time series was a criterion for data inclusion, some datasets switch taxonomic resolution (e.g. from genus to species level) for a given taxon part-way through the time series. This is particularly the case for data from Denmark within the Baetidae, Brachycentridae, Chironomidae, Gammaridae, Oligochaeta and Simuliidae. We did not alter these names because they represent

the original information provided for and published in Haase *et al.*¹, and standardisation methods may vary depending on intended future use. In the time series provided, these issues could affect analyses of shifts in community composition, which could reflect a shift in identification level rather than compositional change. These issues do not affect analyses of total abundance and have little influence taxa richness or diversity, including their temporal trends, as they are typically substitutions of one unique taxon for another. As with all large datasets of ecological time series, data users should carefully check the data considering their intended use and when questions arise, contact data providers (listed in TREAM_siteLevel.csv).

Finally, since each record was assigned the Hydrography90m stream network²⁵ subcatchment, users can directly interact with the *hydrographr* R-package which facilitates subsequent network and distance analyses using the data records³².

Code availability

Code used to conduct analyses for the corresponding publication analysing these data² is available at: https://github.com/Ewelti/EuroAquaticMacroInverts. Code to calculate site and year level biodiversity metrics is available at: https://github.com/Ewelti/EuroAquaticMacroInverts/blob/main/R/Initial_Biodiversity_FuncTrait_and_climate_calcs/TaxaDiversity_calculation.R.

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Author contributions

P.H. initiated the project and put out the data call. E.A.R.W. and J.S.S. performed data cleaning. E.A.R.W. and D.E.B. designed and conducted analyses. S.D. and J.R.G. provided environmental data with assistance from G.Amatulli, J.K. and L.Shen. E.A.R.W., D.E.B., S.D. and J.R.G. wrote the first draft with assistance from P.H., and all authors contributed to additional versions, especially J.S.S. Authors who contributed macroinvertebrate data were F.A., M.Á.-C., D.G.A., G.A., I.A.J., T.A., I.A., I.B., J.B.O., C.L.B., L.B., N.B., R.B., M.C.-A., N.J.B., Z.C., T.D., E.d.E., A.D., G.D., E.D., K.A.E., J.E., T.E.E., V.E., M.J.F., M.Ferréol, M.Floury, M.Forcellini, M.A.F., R.F., N.F., J.-F.F., G.G., P.G., M.A.S.G., A.H., P.H., K.-L.H., T.C.J., R.K.J., J.I.J., L.K., A.L., P.L., L.L. M.-H.L., A.W.L., A.Maire., J.A.M.A., B.G.M., A.Millán, D.M., T.M., J.F.M., D.O., R.P., F.J.P.S., F.P., M.P., P.P., J.J.R., M.R., D.S.-F., L.Sandin, R.B.Schäfer, A.S.-K., A.Scotti, A.Skuja, S.S., M.S., R.Stubbington, H.T., V.G.T., I.T., Y.U., G.H.v., R.V., E.V., G.Várbíró, G.Velle, P.F.M.V, R.C.M.V, Y.V., and P.W.-L.

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Additional information

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