

Impacts of the global food system on terrestrial biodiversity from land use and climate change

Supplementary Methods

Details of the cropland data sources

SPAM¹ presents gridded land-use estimates for 42 individual crops at a 5-arc-minute spatial resolution, derived from satellite-based land-cover data from around the year 2010; Monfreda et al.² present similar estimates for 175 crop types, also at a 5-arc-minute spatial resolution, derived from agricultural census data and satellite-based land-cover data from around the year 2000. Where possible, we chose SPAM data over Monfreda et al.² since it is closer in time to our 2011 input-output table. However, any disjoint between the datasets should have only a minor impact on the final results, since these estimates were used only to estimate the average natural species richness across farmed areas (at a 10-km resolution), which is not likely to have changed markedly between 2000 and 2010. We aggregated SPAM's 42 crop layers to align with the EXIOBASE land-use categories of paddy rice, wheat, cereal grains nec ('not elsewhere classified'), vegetables/fruit/nuts, oil seeds, sugar cane/sugar beet and crops nec. Since SPAM does not separate fodder from non-fodder crops, we obtained spatial estimates of fodder crops from Monfreda et al.². Monfreda et al.² do not disaggregate fodder crops into the different livestock categories given by EXIOBASE so, following Marques et al.³, Monfreda et al.'s² 15 fodder crop layers were aggregated and then replicated for each EXIOBASE cropland-fodder category (fodder-cattle, fodder-pigs, fodder-poultry, fodder-meat nec, fodder-raw milk). In other words, we assumed that all 5 EXIOBASE fodder crop categories were present wherever any Monfreda et al.² fodder crop occurred. Ramankutty et al.⁴ present pasture data at 5-arc-minute spatial resolution, also derived from census data and satellite-based land-cover data from around the year 2000. Again, separating pasture into the different livestock categories given by EXIOBASE (Permanent pasture: grazing-cattle, grazing-meat animals, grazing-raw milk) was not possible so the same pasture layer was allocated to each EXIOBASE pasture category. Cells which contained less than 2.5% of each agricultural land-use type were discarded. Supplementary Table 3 shows the allocation of crops and pasture to EXIOBASE's land-use categories. The methods were implemented using the R packages raster v 3.4-5⁵ and rgdal v 1.5-23⁶.

Details of the PREDICTS model

The PREDICTS database⁷ contains 3.25 million snapshot samples of biodiversity in different land uses from 666 different studies, collected from 26,114 locations in all of the world's terrestrial biomes (see Supplementary Figure 1). The PREDICTS data represent 47,044 different species, including plants, fungi, invertebrates and vertebrates. Sites in the PREDICTS database are classified into one of six different land-use types, based on the description of the habitat given in the original source publication or following consultation with the authors of the original study: primary vegetation (natural habitat, with no record of prior destruction by human activities or extreme natural events), secondary vegetation (destroyed in the past, but now recovering toward a natural state), plantation forest (areas used to cultivate woody crops), cropland (areas used to cultivate herbaceous crops, including for use as livestock fodder), pasture (areas used directly for grazing

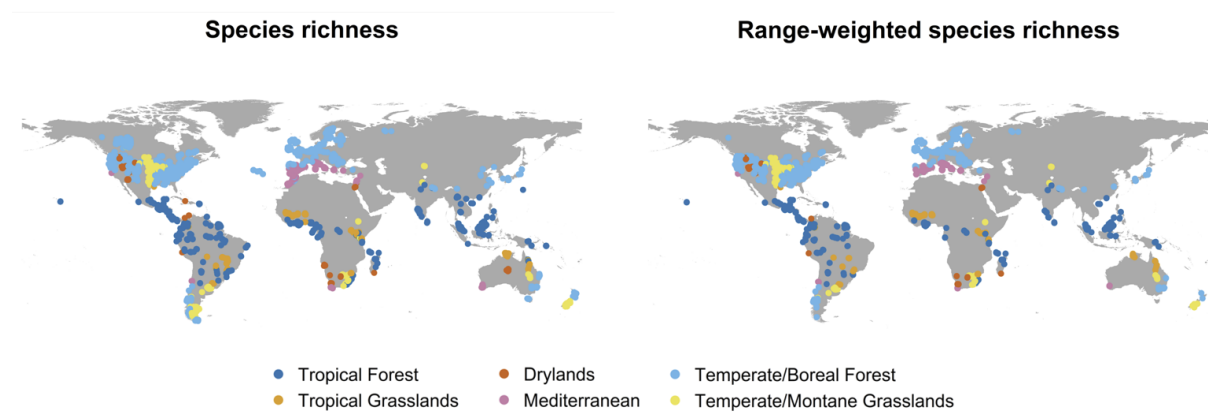
livestock, including both managed pasture and rangeland), and urban (areas of human settlement or civic amenity).

For the purposes of this study, we considered just four PREDICTS land-use classes: primary vegetation, secondary vegetation, agriculture (including both plantations and herbaceous cropland) and pasture. As in a recent study⁸, we allowed the response of biodiversity to these different land-use categories to vary among six broad biome groupings (Supplementary Table 2). Unlike the estimates of biodiversity responses to climate change, which because of data availability could be made only for vertebrates, we consider all taxonomic groups in estimating land-use responses. This broader taxonomic scope allowed us to include more data, and thus to consider how responses varied among biomes. We measured biodiversity at each site in the PREDICTS database in terms of species richness and rarity-weighted species richness. Species richness was measured, as in previous studies using the PREDICTS database e.g.⁸, as the count of the unique species sampled at a site. To account for differences in the species composition of communities, and to consider potential shifts in the representation of narrow- versus wide-ranged species within communities, previous studies have used a measure of abundance-weighted community-average ($\log_{10}$ -transformed) species range size as an additional measure of biodiversity (e.g.⁸). Species' range sizes were estimated as the number of 110-km grid cells with a record for that species in the Global Biodiversity Information Facility database e.g.⁸. Using GBIF data, rather than the expert-derived range maps provided for vertebrates by the IUCN and others (see main text), allowed us to obtain range-size estimates for non-vertebrate species. Here, we modified this measure to calculate rarity-weighted species richness. Specifically, we calculated species weights based on the inverse of estimated range size, and calculated rarity-weighted species richness as the sum of weights of all species sampled at a given site. Within each study in the PREDICTS database, these weights were rescaled linearly, such that the highest weight (i.e., for the species with the smallest range) had a value of 1. We performed this rescaling because range sizes vary enormously among regions and species, and thus among PREDICTS studies, which prevents model convergence. Given the hierarchical structure of the models (see below), the rescaling should not affect estimates of the impact of land use on relative rarity-weighted species richness.

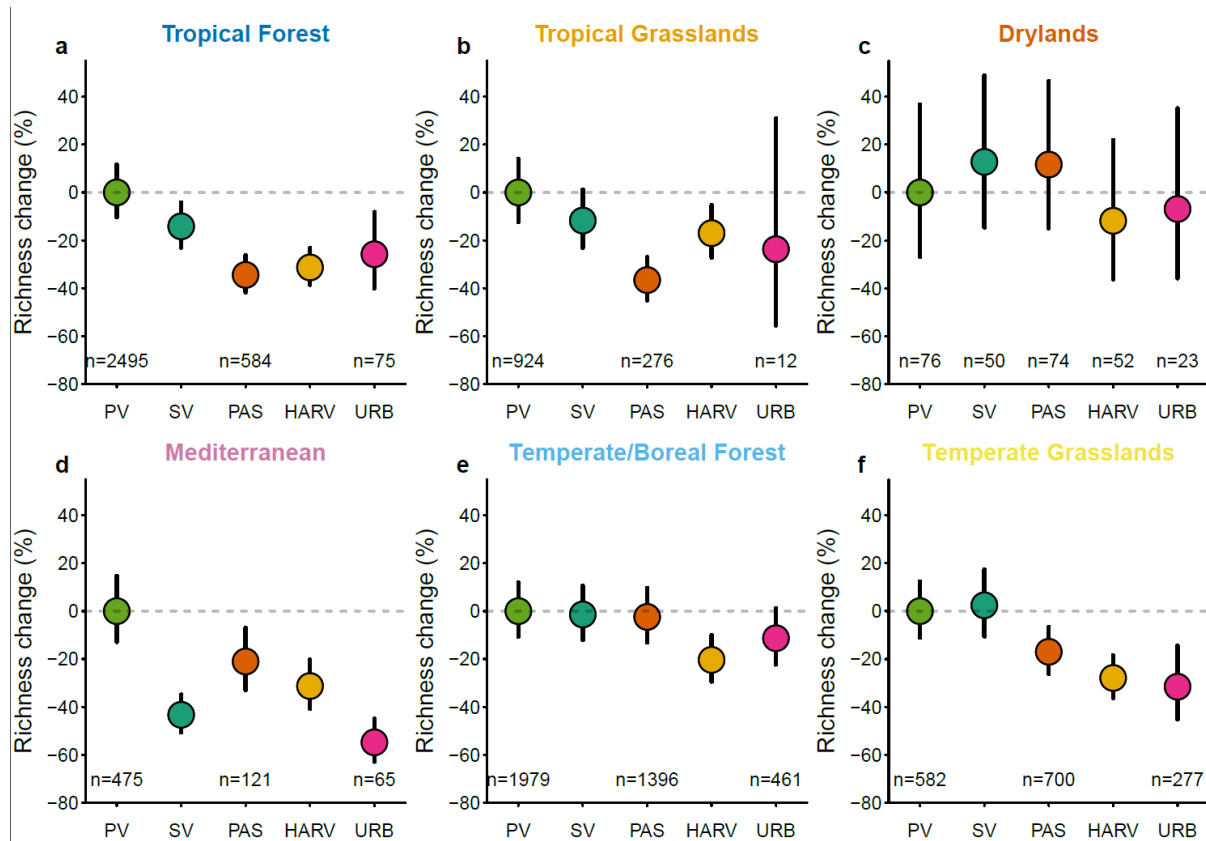
We fitted the response of species richness to land use using a generalized linear mixed-effects model with a Poisson distribution of errors (Supplementary Figure 2, Supplementary Tables 4, 5). Since rarity-weighted species richness had many non-integer values, and many values less than one, a Poisson model was not appropriate. Instead, we $\log_e$ -transformed rarity-weighted species richness values, adding 1% of the mean (0.014) to all values to deal with zeroes, and then fitted responses to land use using a linear mixed-effects model. Fixed effects for both models were land use, biome, and their interaction. Random effects were study identity (to account for variation in biodiversity among studies caused by differences in sampling methods and effort, the taxonomic groups sampled, and broad geographic differences), and for the species richness model also site nested within study (to account for the over-dispersion present)⁹.

Projected changes in the distributions of species' (and thus changes in local species richness across terrestrial areas) as a result of climate change were derived from¹⁰. The response of species' distributions to climate was captured using five different species distribution modelling algorithms (BIOCLIM, DOMAIN, Maxent, Random Forests and Generalised Linear Models). These models each related species' observed distributions according to the IUCN Red List¹¹ and Birdlife International¹², to four climatic variables shown in previous studies to be strong correlates of animal distributions: minimum temperature of the coldest month, total annual precipitation, growing degree days and water balance, derived from the Worldclim Version 1.4 database¹³, which captures average climatic conditions for the period 1961-1990. BIOCLIM fits relationships between distribution records and climatic variables using a bounding-box approach in niche space, DOMAIN by comparing the climatic similarity between observed occurrence points and potentially inhabitable areas, random forests using a machine-learning approach to identify climatic patterns in species' occurrence records, while generalized linear models and Maxent use classical parametric statistics or a maximum-entropy approach, respectively, to fit linear and quadratic relationships between species' occurrences and the climatic variables. The resulting distribution models were projected onto the Worldclim Version 1.4 estimates of future climatic conditions under the RCP scenarios, averaged across the years 2061-2080, and averaged across outputs from an ensemble of different Global Climate Models¹³. Species were assumed to be limited in their ability to disperse in response to climate change, such that mammals and birds could move 3 km year⁻¹, and reptiles and amphibians 0.5 km year⁻¹ from conditions currently estimated to be suitable for the species¹⁰. Species were further considered unable to move outside the combinations of biogeographic realm and biome estimated to be occupied in the present day according to the original range maps from the IUCN and Birdlife International¹⁰. All calculations (both distribution models and their projections) were performed on a grid with 10-km spatial resolution¹⁰. Newbold (2018)¹⁰ estimated a single metric of biodiversity, local species richness change. We also calculated rarity-weighted species richness change. To do so, each species was given a weight equal to the inverse of the number of 10-km grid cells estimated to be climatically suitable in the baseline period (1961-1990). We then calculated rarity-weighted species richness for both the baseline and future periods by summing these weights for all species for which each grid cell was estimated to be climatically suitable, and reachable given dispersal constraints. To calculate climate sensitivity, we divided projected change in species richness or rarity-weighted species richness between the baseline period and 2061-2080 by the projected change in temperature over the same time period (according to the same climate models used to drive the biodiversity projections). We focused here on projections of biodiversity and temperature under the RCP 8.5 climate scenario. This gave estimates of the sensitivity of biodiversity to climate change for every 10-km grid cell on the land surface of the world.

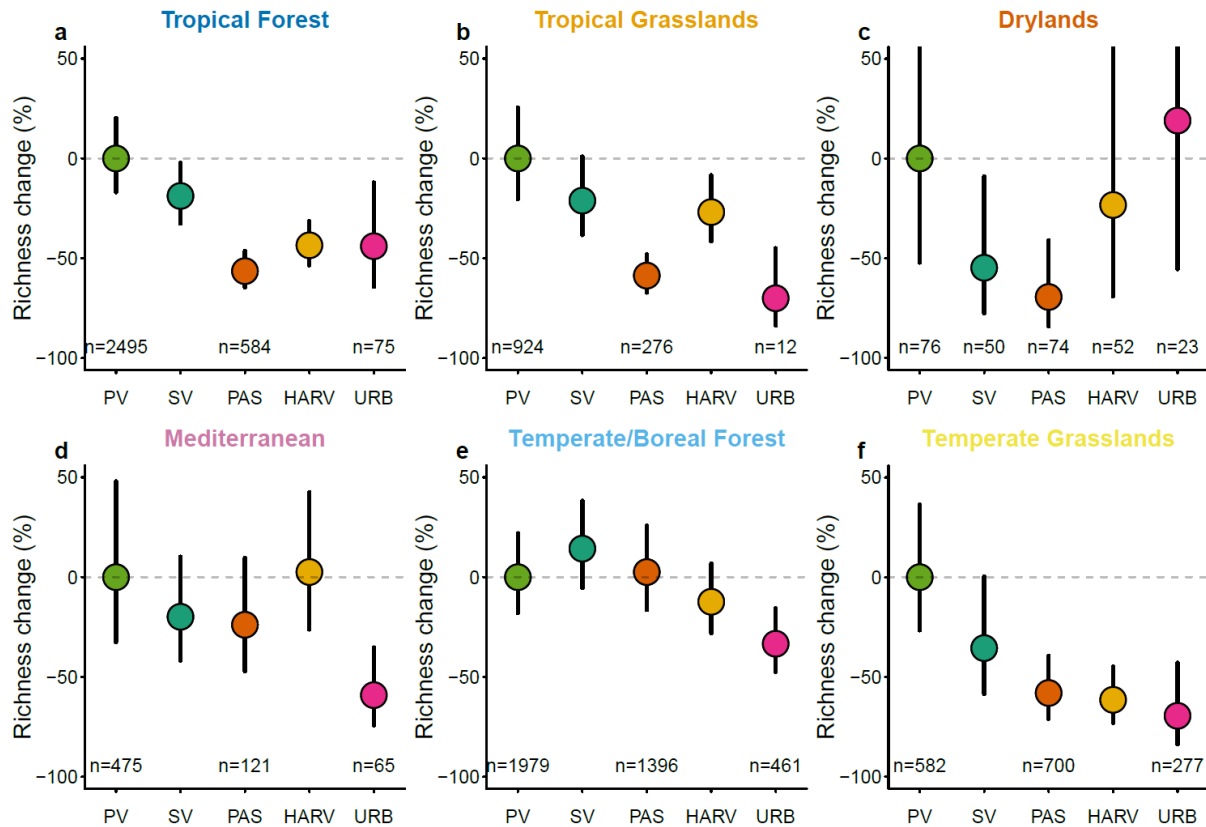
Supplementary Figures



Supplementary Figure 1. Locations of the studies in the PREDICTS database. Locations are colour-coded by biome. The map was produced using world country polygons from the World Bank Data Catalog (<https://datacatalog.worldbank.org/search/dataset/0038272/World-Bank-Official-Boundaries>) published under a Creative Commons Attribution 4.0 license.

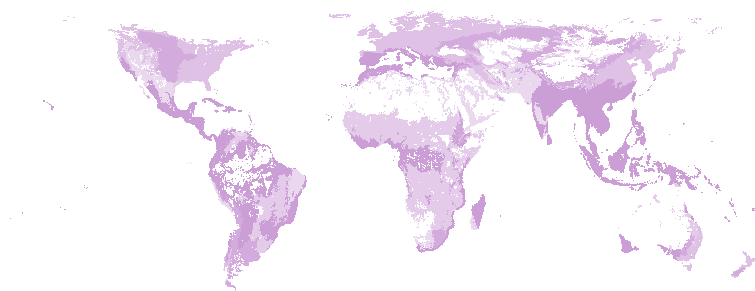


Supplementary Figure 2. Differences in total sampled species richness among land use types across different biomes. a) Tropical forest; b) Tropical grasslands; c) Drylands; d) Mediterranean; e) Temperate/Boreal forest; and f) Temperate/Montane grasslands. Plots show the percentage change in species richness compared to primary vegetation (PV), in secondary vegetation (SV), pasture (PAS), areas of harvested agriculture (woody plantations and herbaceous croplands) (HARV) and urban areas (URB). Error bars show 95% confidence intervals. Sample sizes at the bottom of each panel refer to the number of sites in each combination of land use and biome. In total 21986 sites from 637 studies were used. The final model plotted here had an $R^2_{\text{conditional}}$ of 0.61 and an R^2_{marginal} of 0.024.

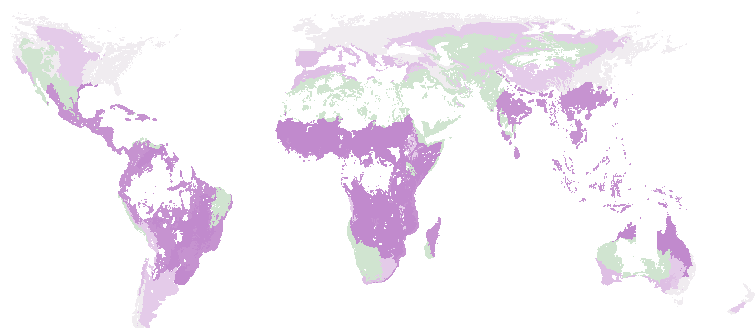


Supplementary Figure 3. Differences in total sampled rarity-weighted species richness among land use types across different biomes. a) Tropical forest; b) Tropical grasslands; c) Drylands; d)Mediterranean ; e) Temperate/Boreal forest; and f) Temperate/Montane grasslands. Plots show the percentage change in species richness compared to primary vegetation (PV), in secondary vegetation (SV), pasture (PAS), areas of harvested agriculture (woody plantations and herbaceous croplands) (HARV) and urban areas (URB). Error bars show 95% confidence intervals. Sample sizes at the bottom of each panel refer to the number of sites in each combination of land use and biome. In total 15198 sites from 403 studies were used. The final model plotted here had an $R^2_{\text{conditional}}$ of 0.63 and an R^2_{marginal} of 0.043

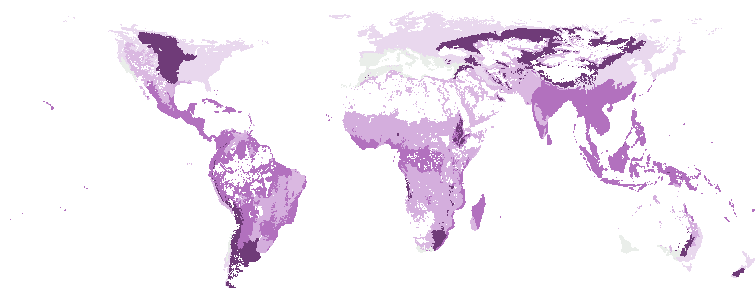
a



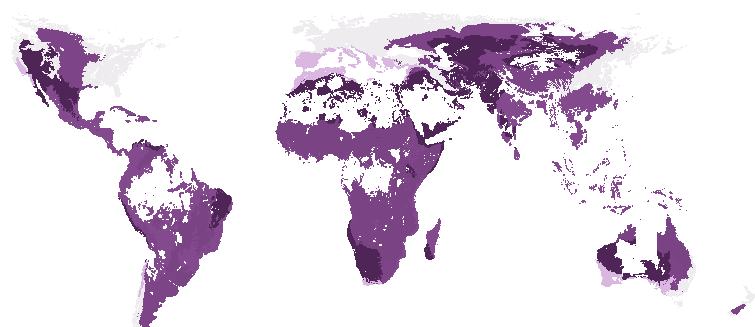
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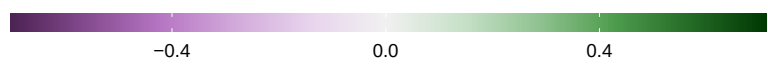
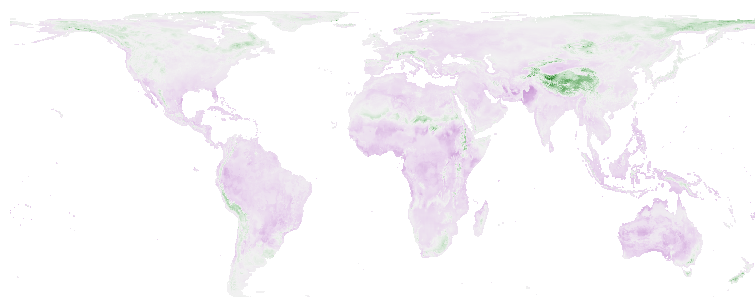
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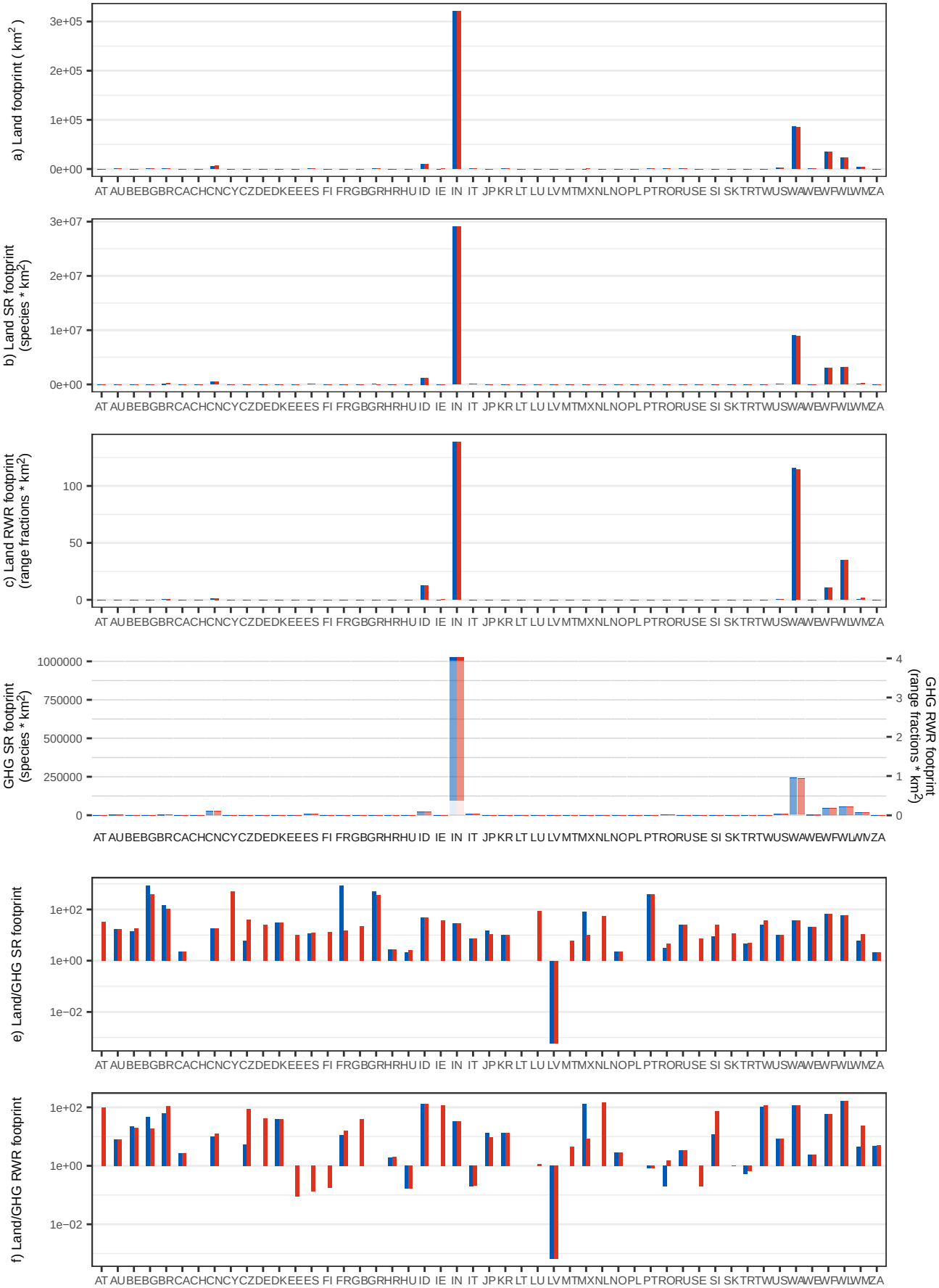
e



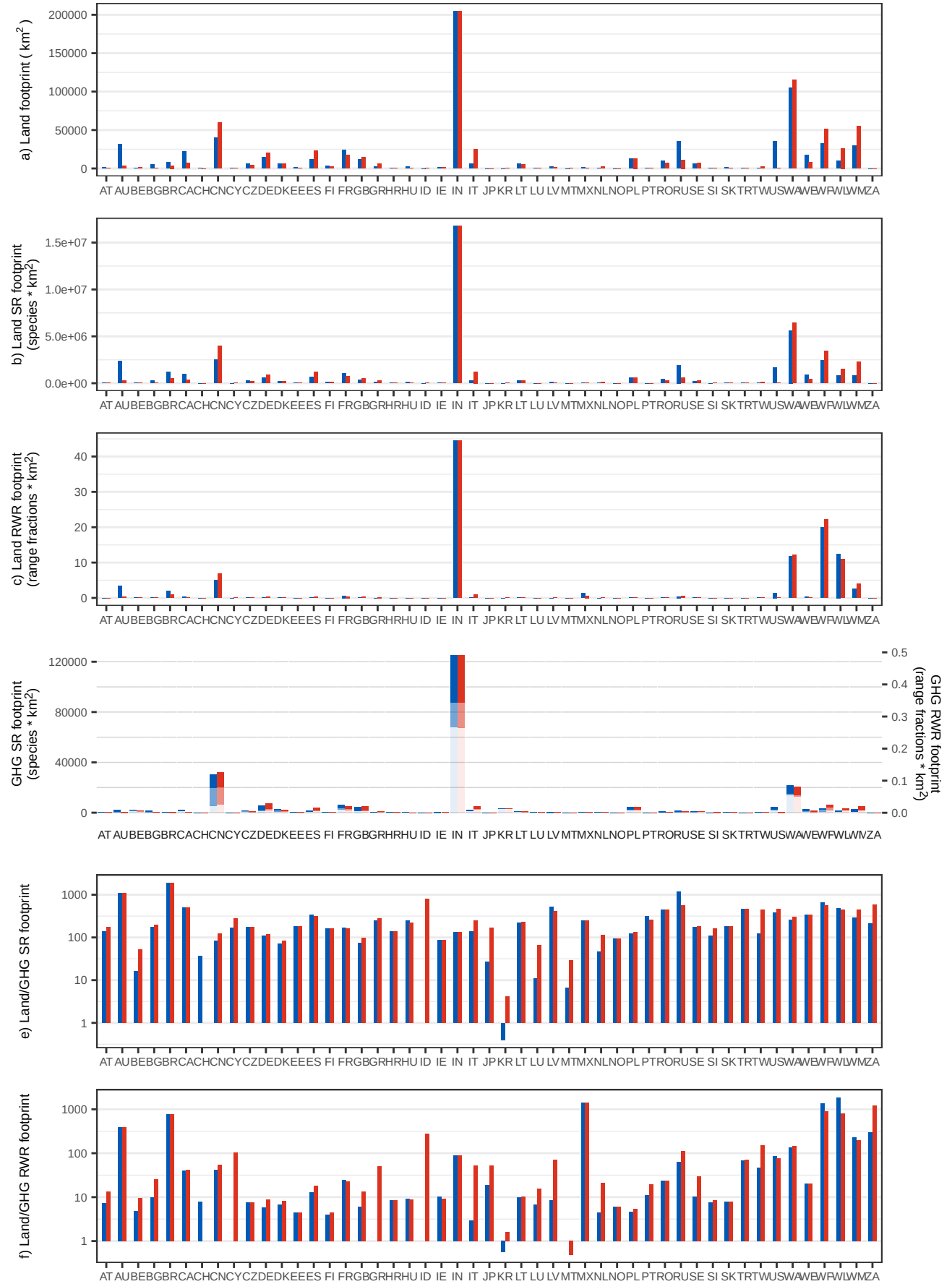
Supplementary Figure 4. Maps showing projections of land-driven and GHG-driven biodiversity change. Proportional species richness (a, b, e) and range weighted richness change (c, d, e) that would occur within a cell under crop use (a, c), pasture use (b, d), or 1° of climate warming (e). The distributions of cells containing cropland (a, c) are taken from SPAM¹ (cropland with exception of fodder crops) and EarthStat² (fodder crops). The distributions of cells containing pasture (b,c) are taken from Ramankutty, et al. ⁴. The figure was produced using the R¹⁸ packages 'ggplot2'¹⁹ and 'ggthemes'²⁰.

Supplementary Figure 1

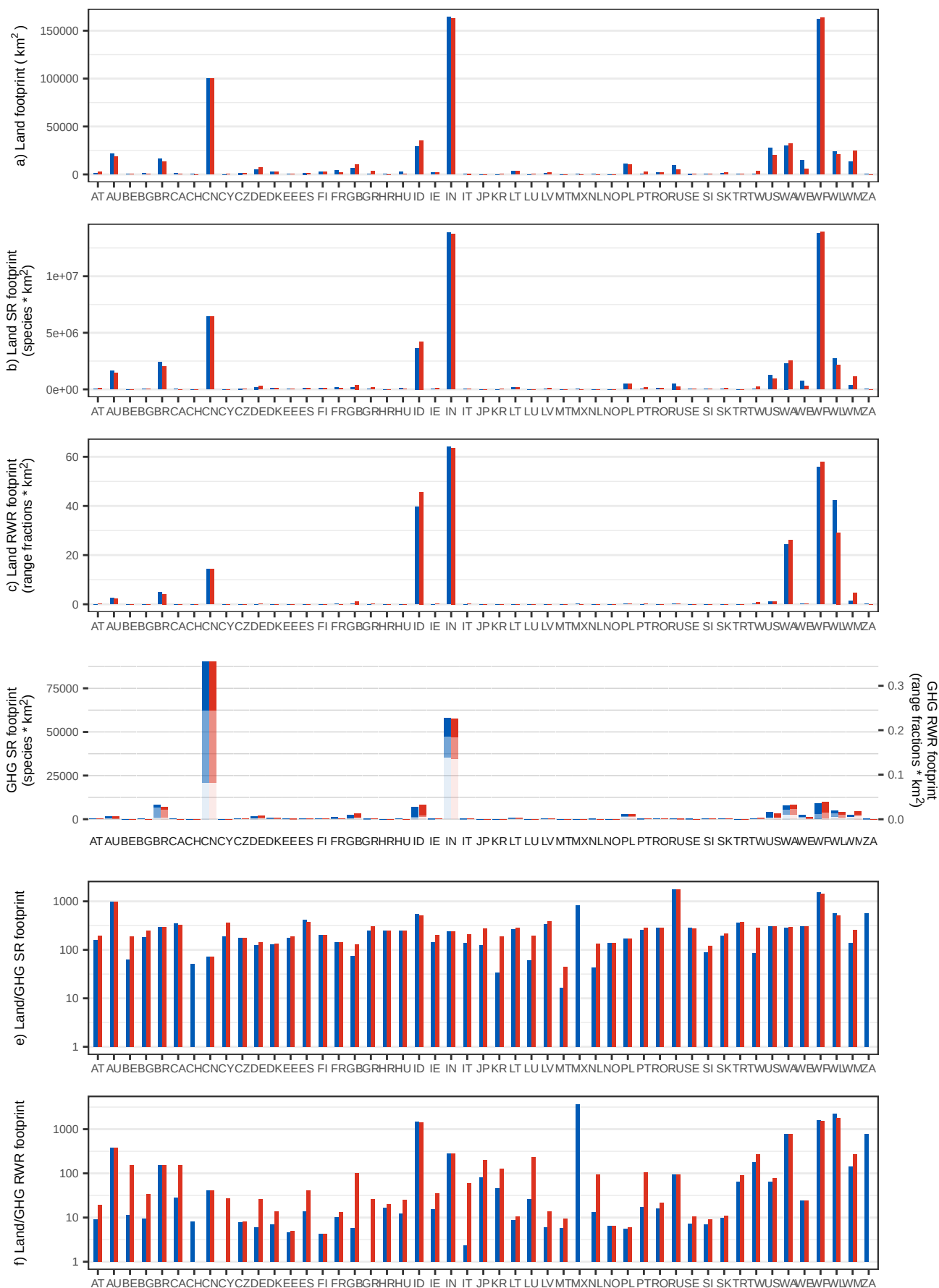
Paddy rice: production and consumption-based footprints



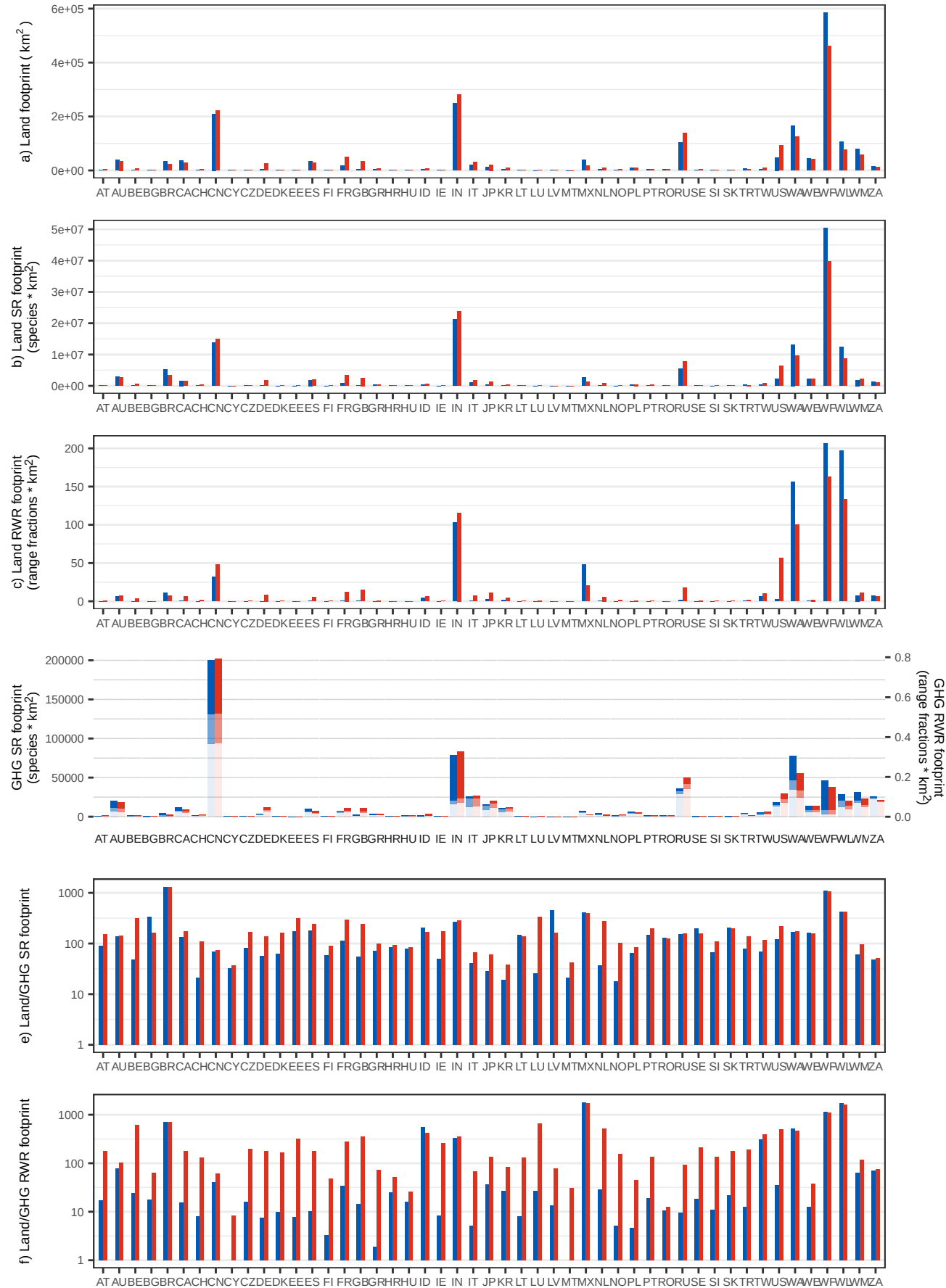
Supplementary Figure 1
Wheat: production and consumption-based footprints



Cereal grains nec: production and consumption-based footprints

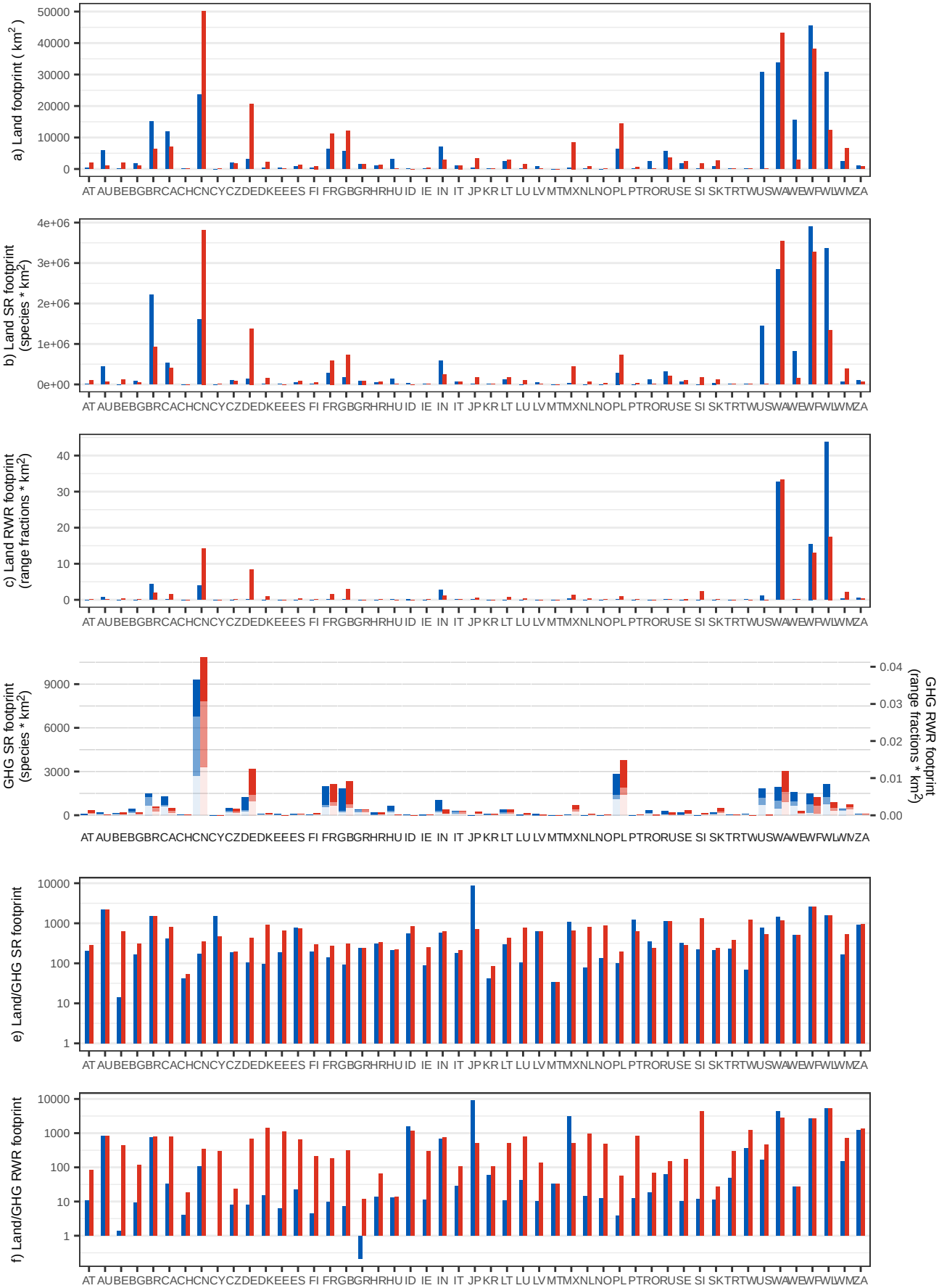


Supplementary Figure 1
Veg/Fruit/Nuts: production and consumption-based footprints

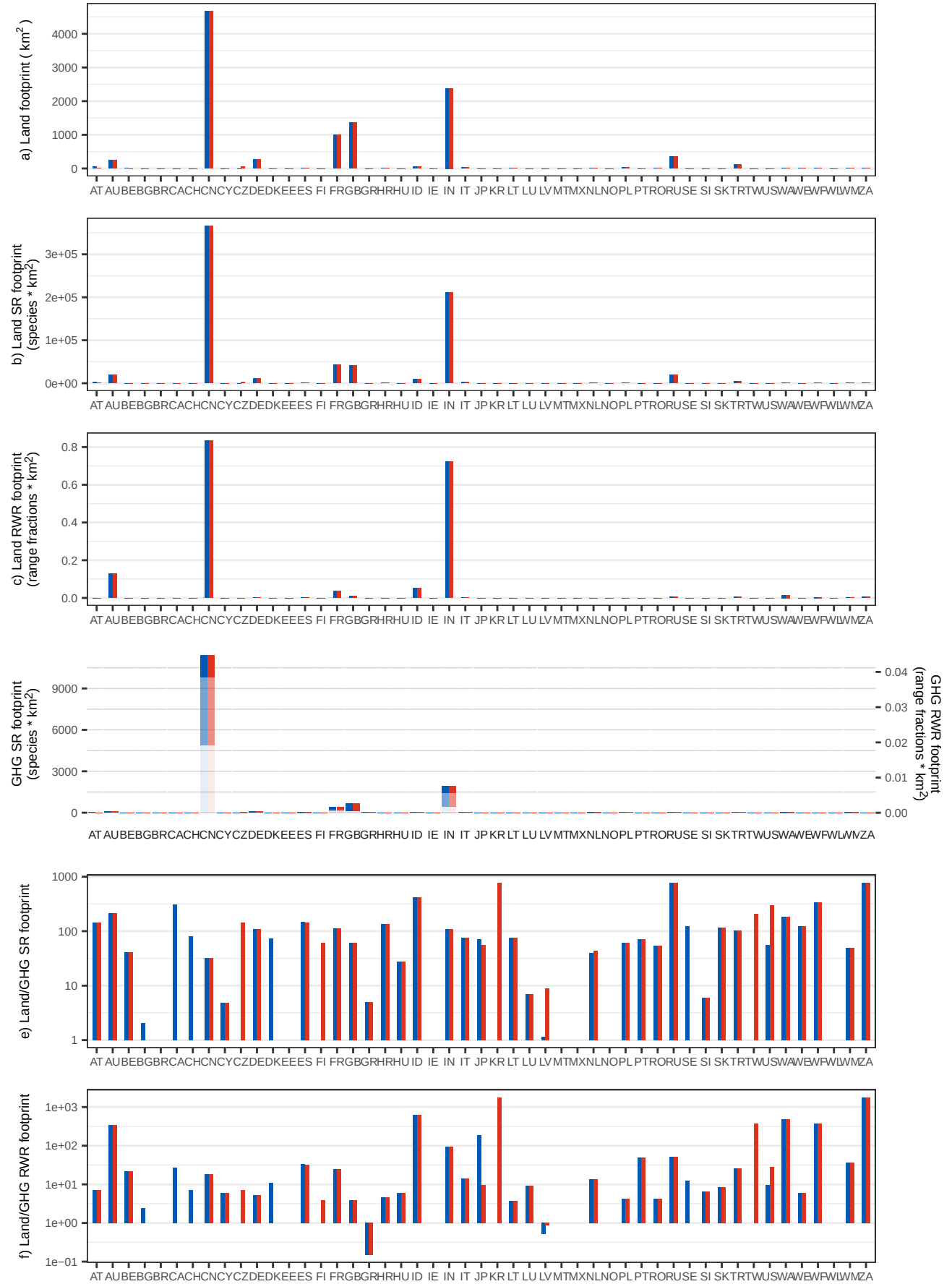


Supplementary Figure 1

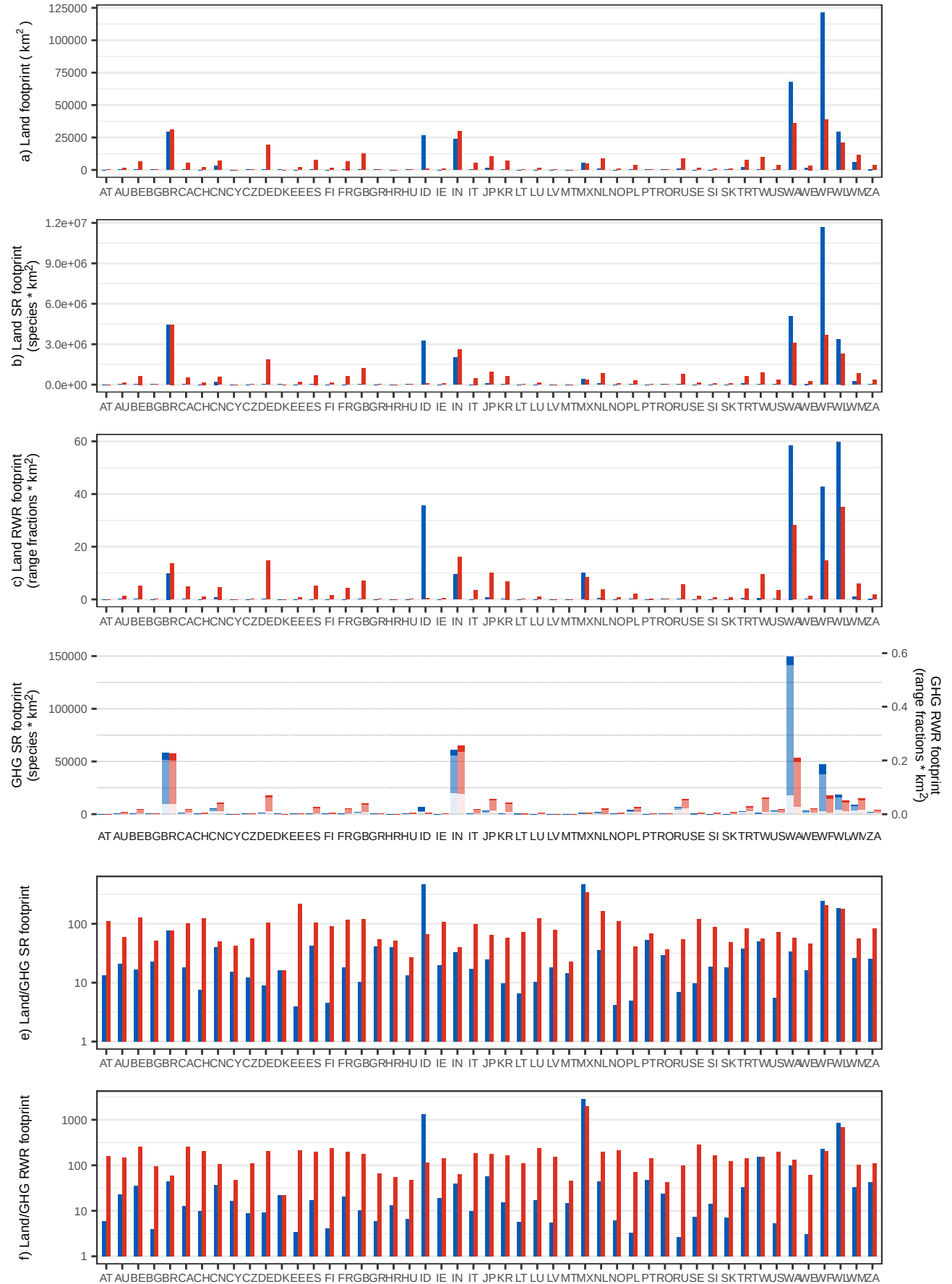
Oil seeds: production and consumption-based footprints



Supplementary Figure 1
Sugar beet/cane: production and consumption-based footprints

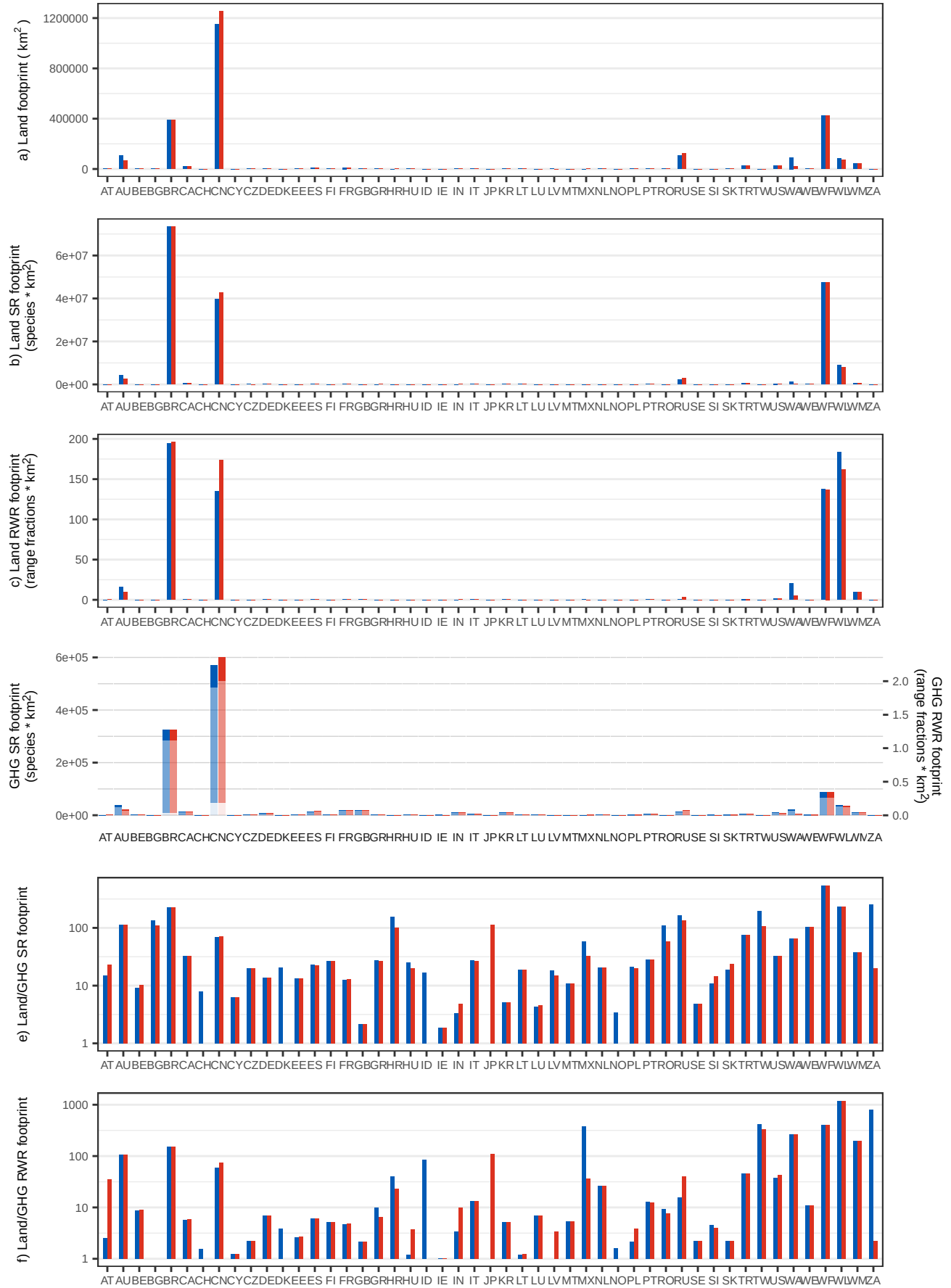


Supplementary Figure 1
Crops nec: production and consumption-based footprints

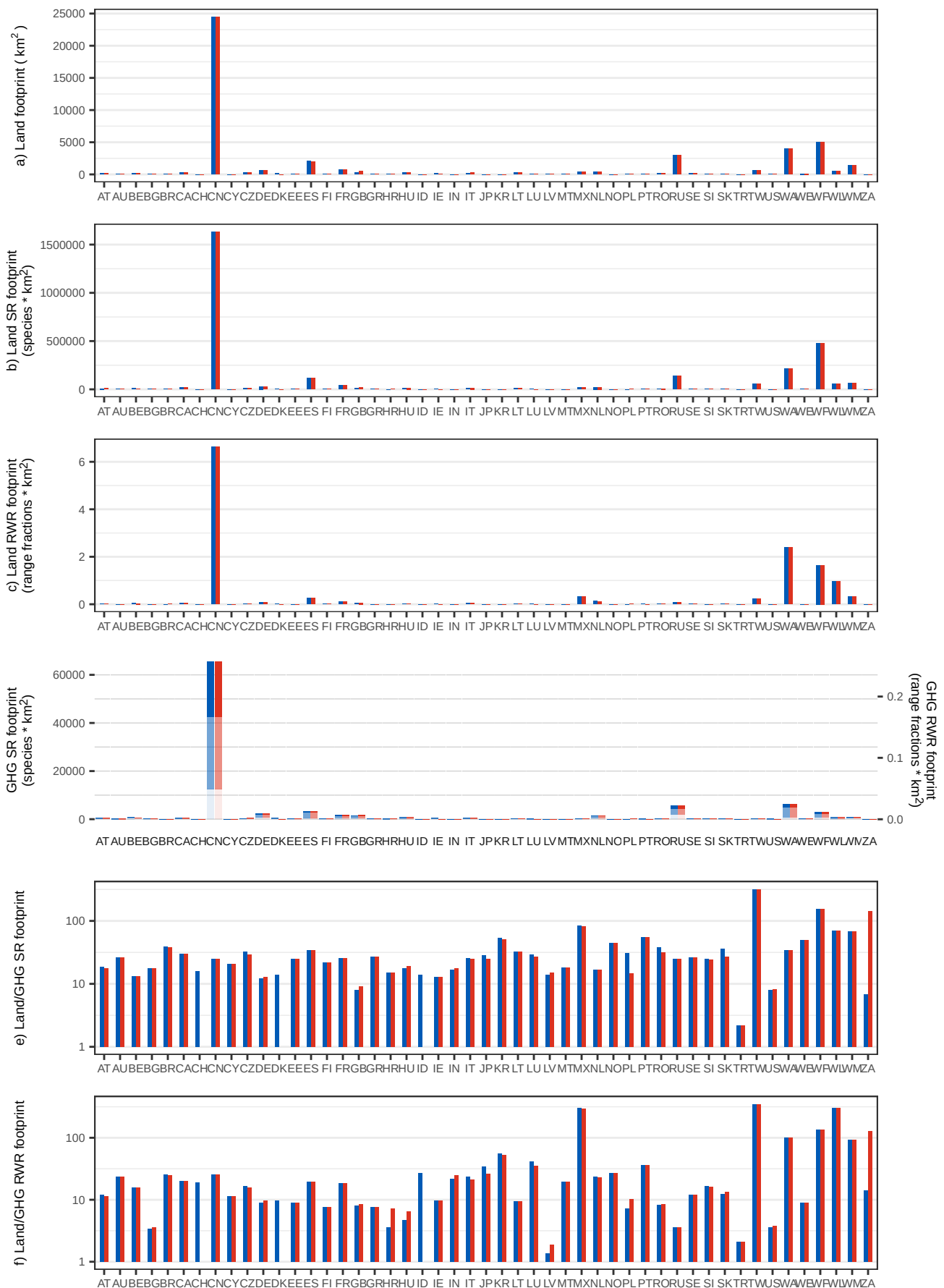


Supplementary Figure 1

Cattle: production and consumption-based footprints

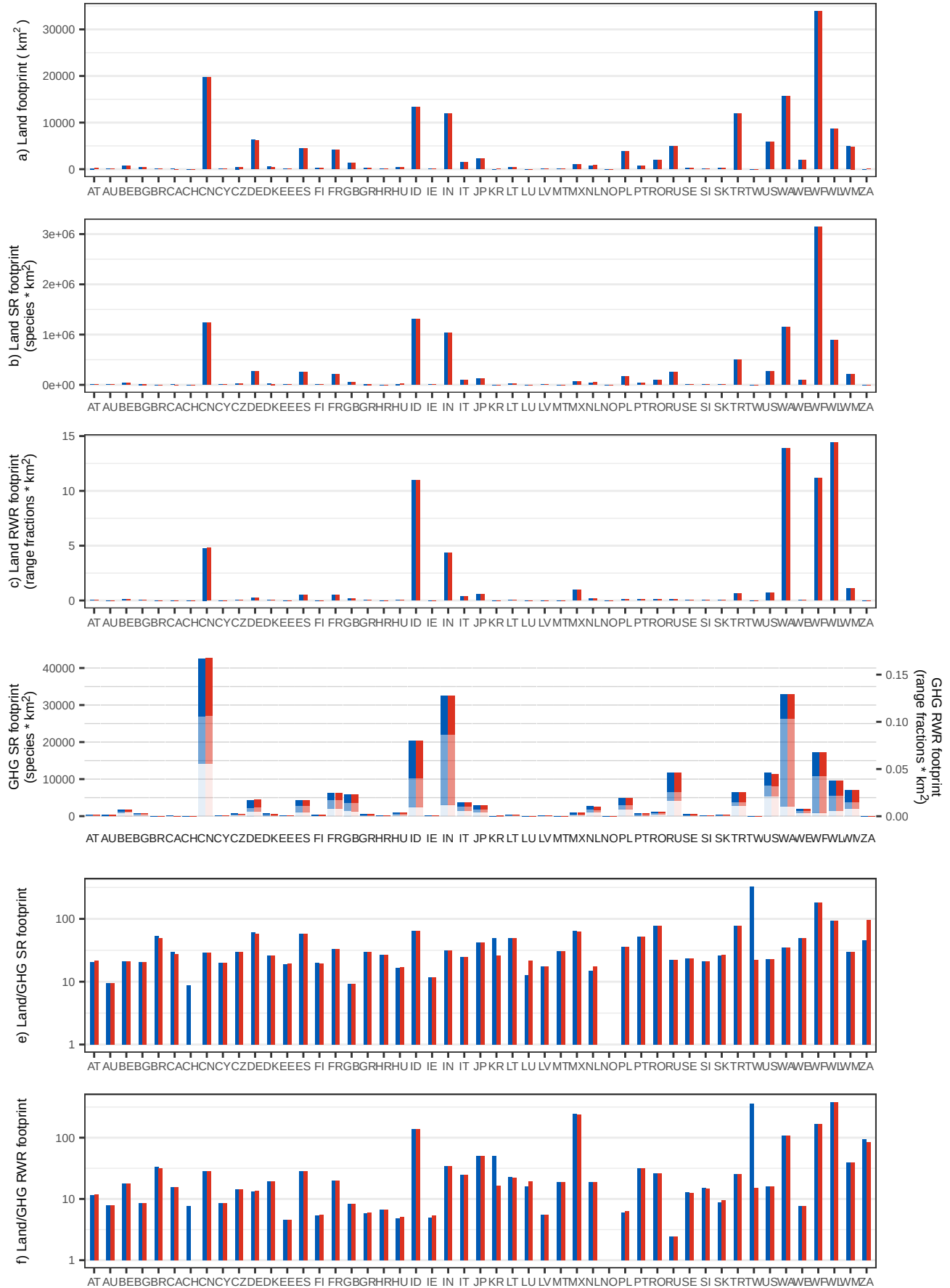


Pigs: production and consumption-based footprints

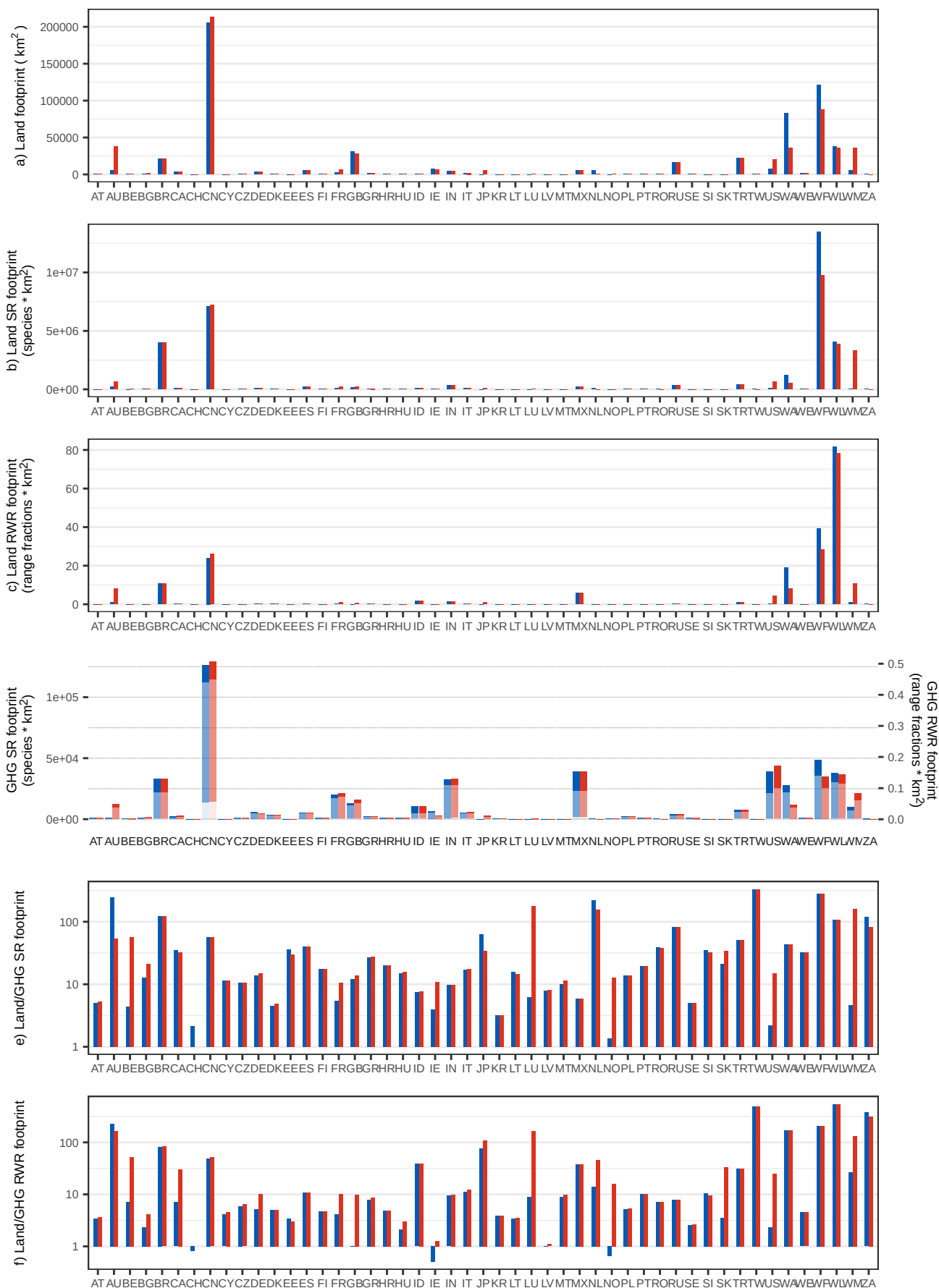


Supplementary Figure 1

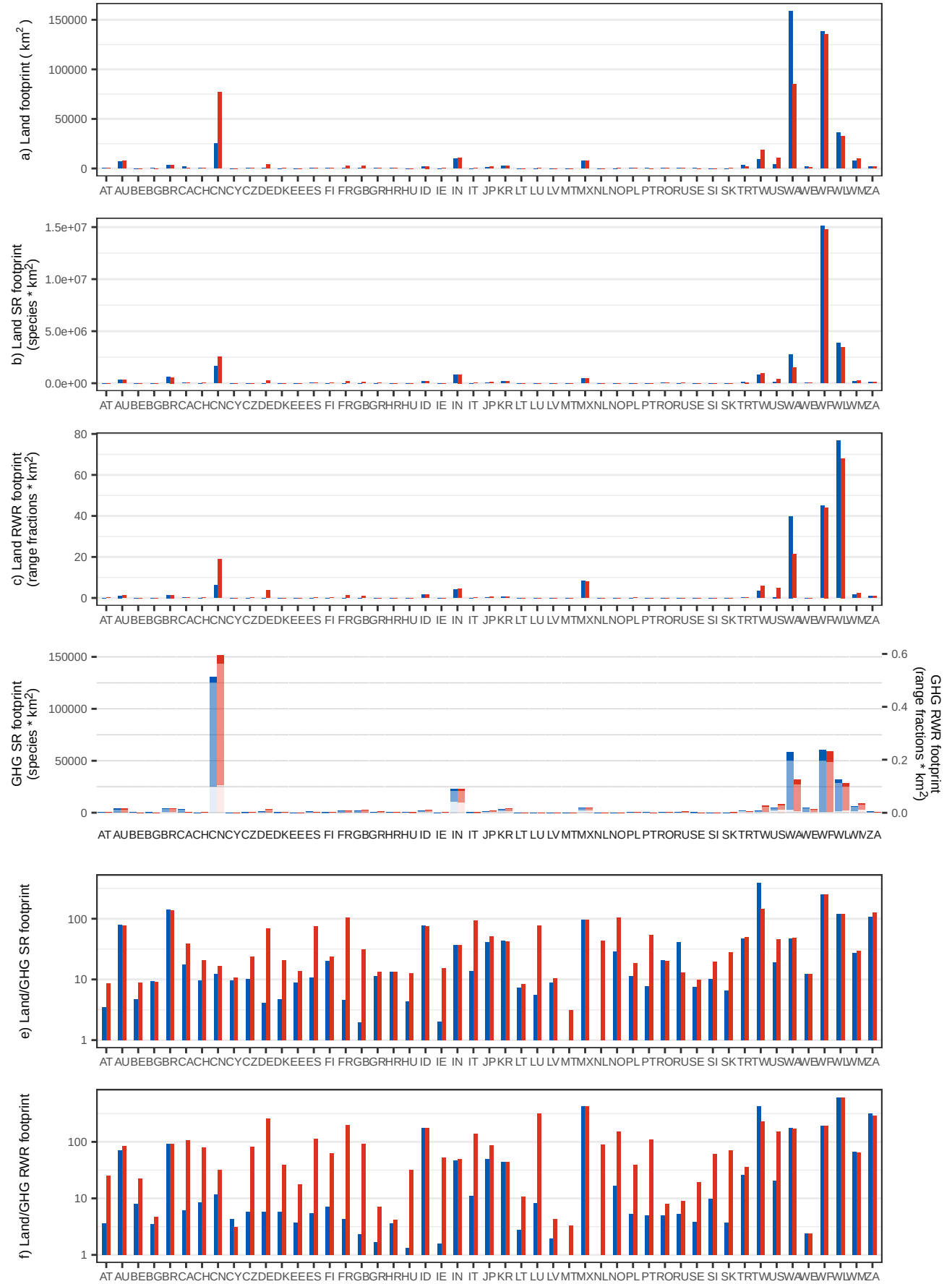
Poultry: production and consumption-based footprints



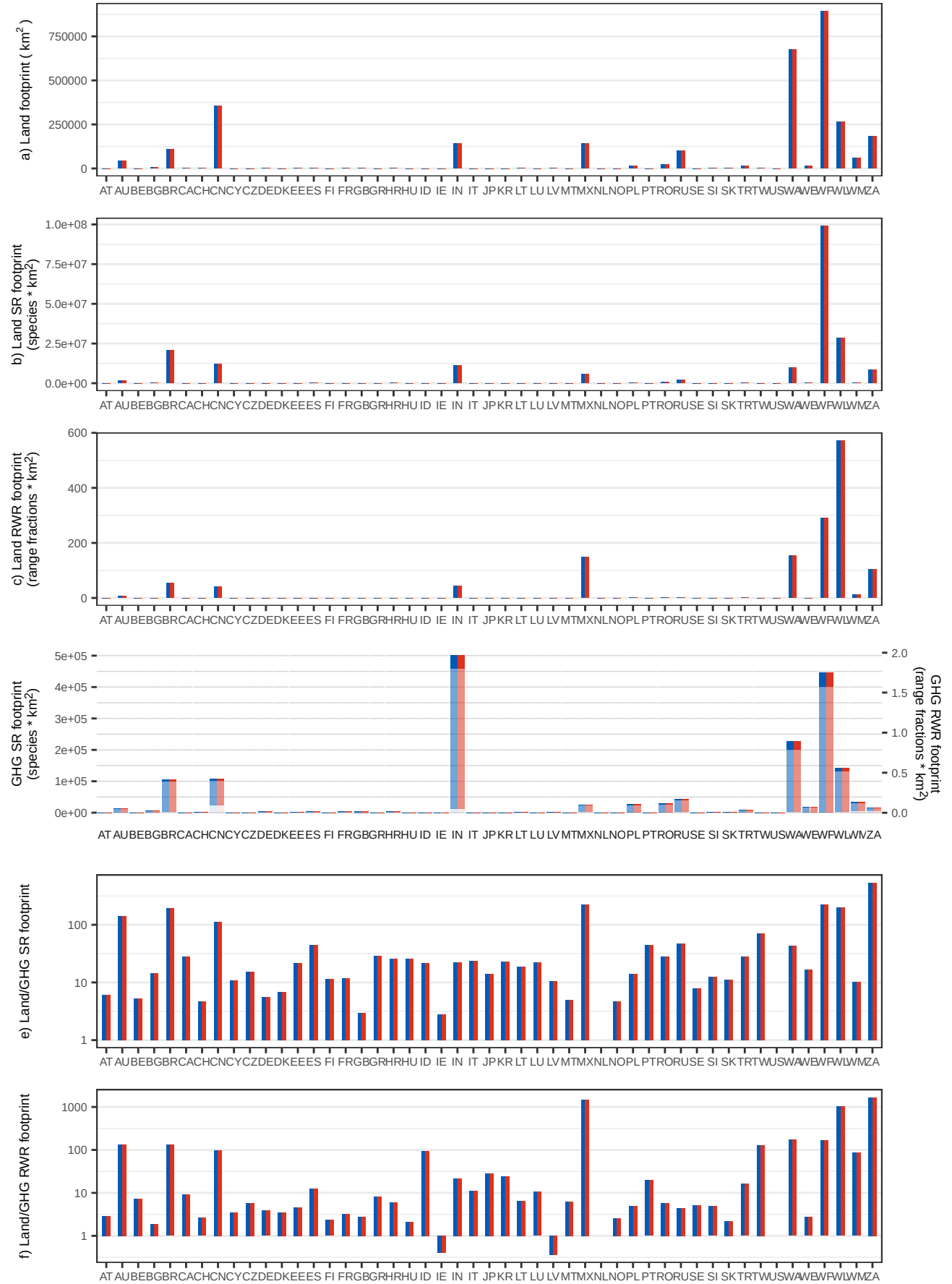
Meat animals nec: production and consumption-based footprints



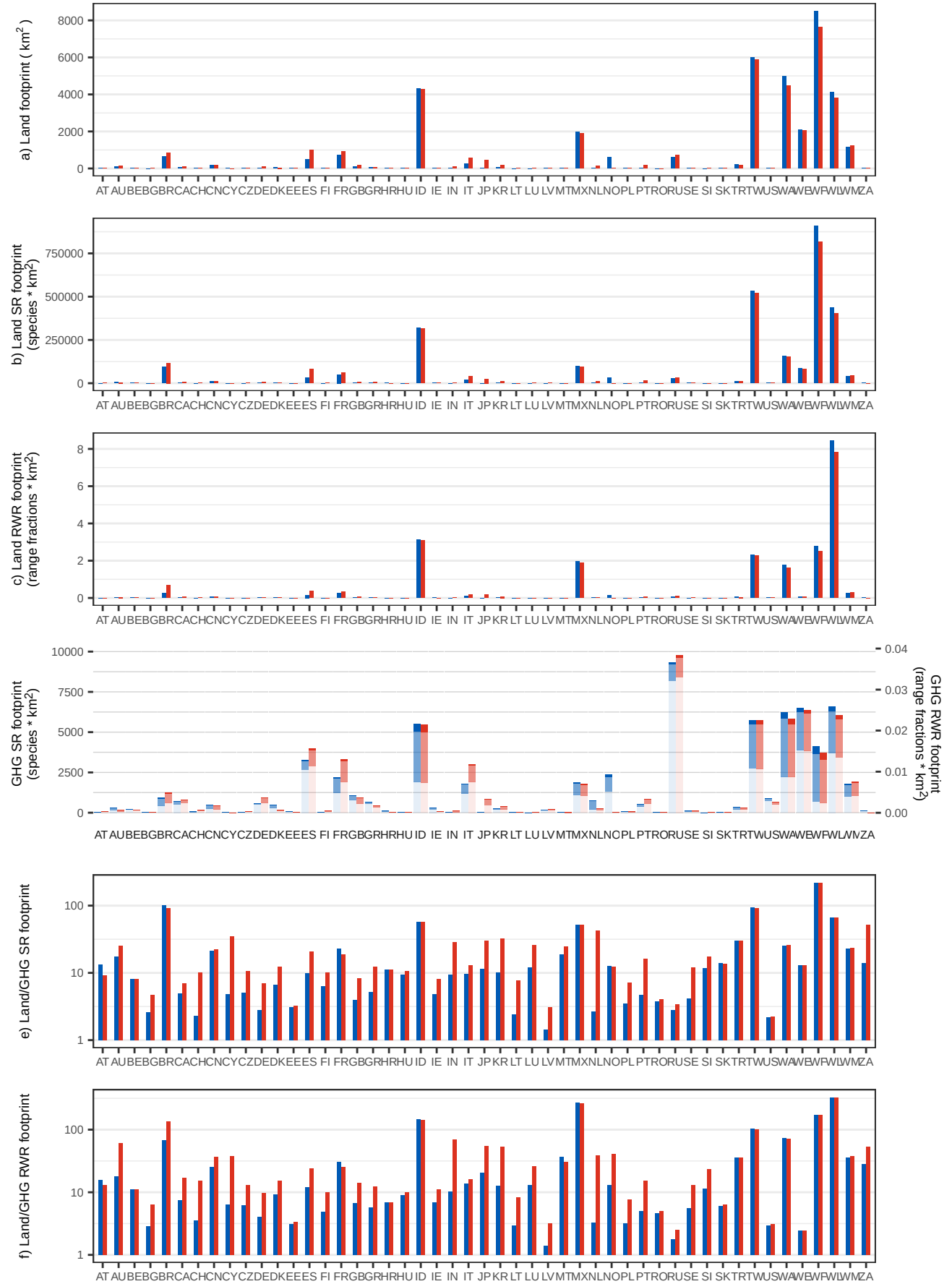
Supplementary Figure 1
Animal products nec: production and consumption-based footprints



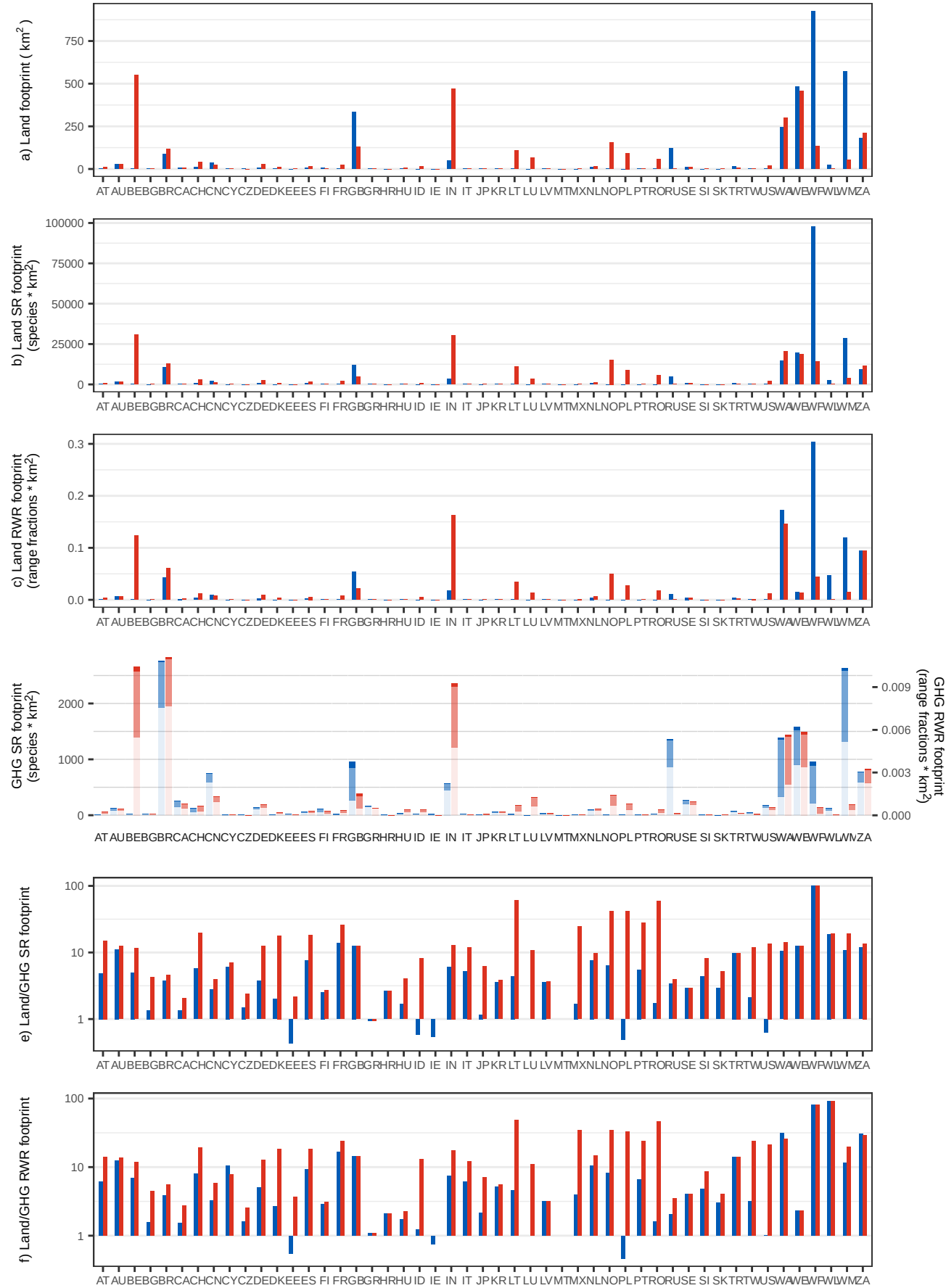
Supplementary Figure 1
Raw milk: production and consumption-based footprints



Supplementary Figure 1
Fish and other fishing products: production and consumption-based footprints

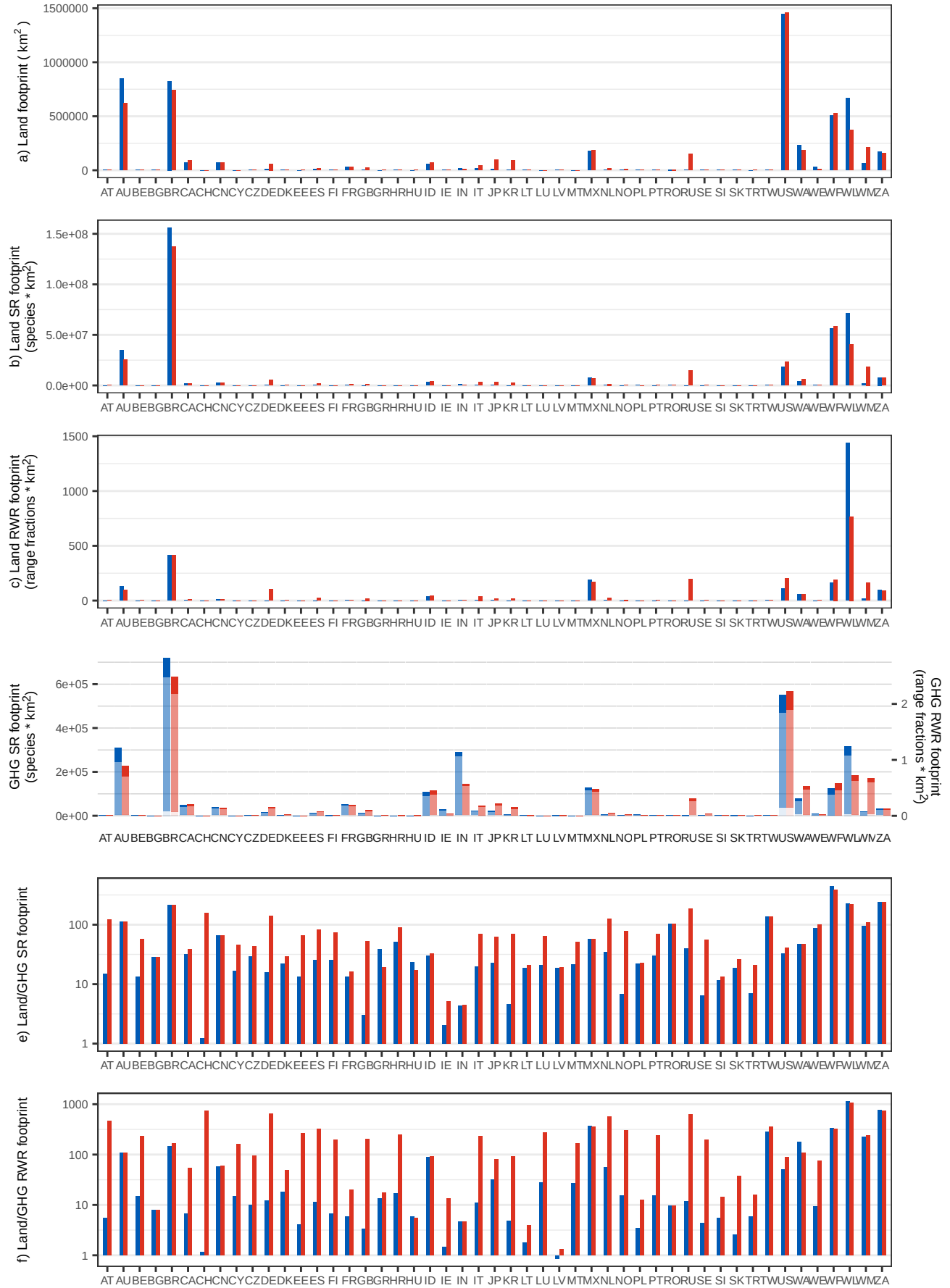


Supplementary Figure 1
Chemical and fertiliser minerals: production and consumption-based footprints

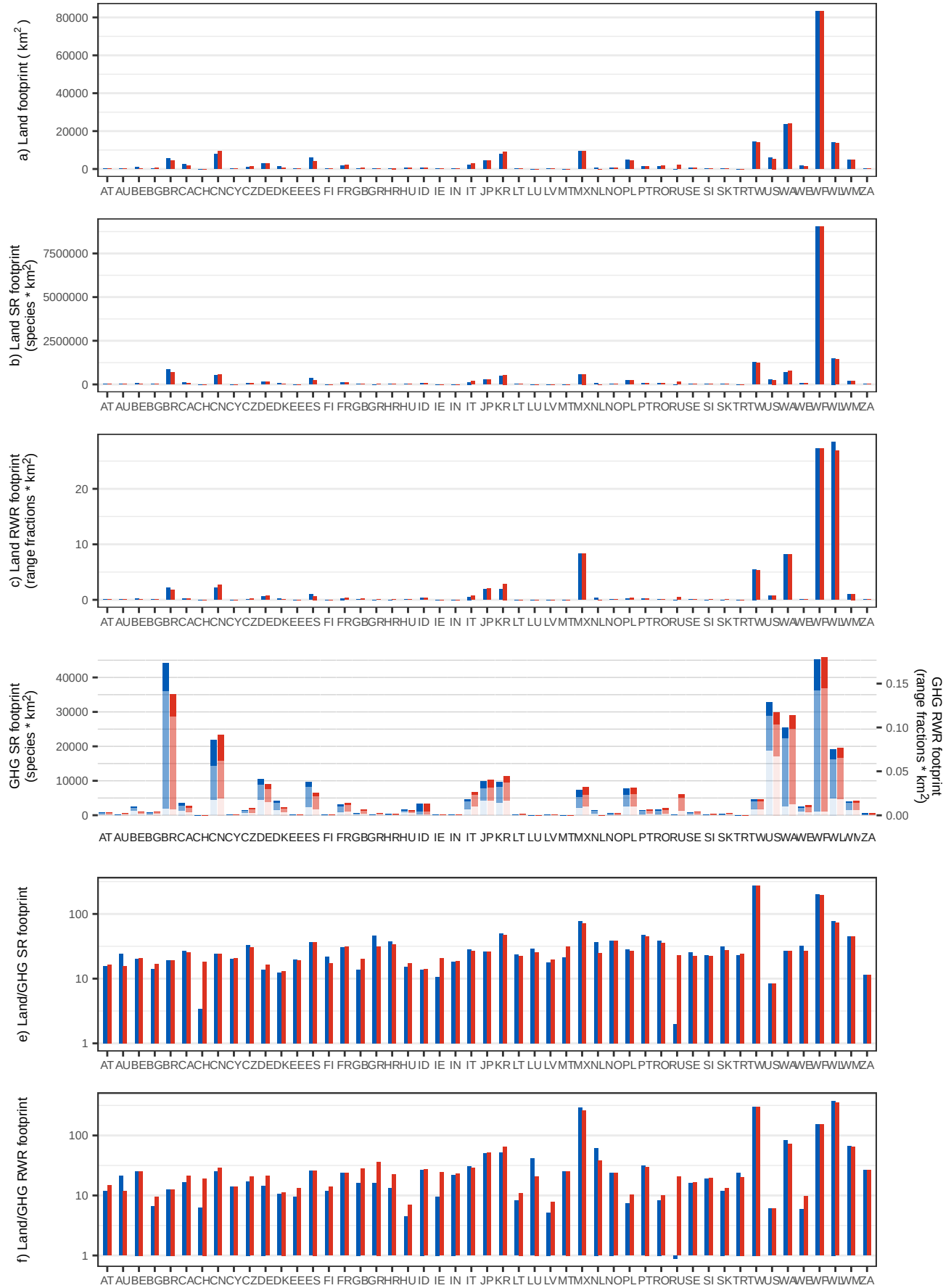


Supplementary Figure 1

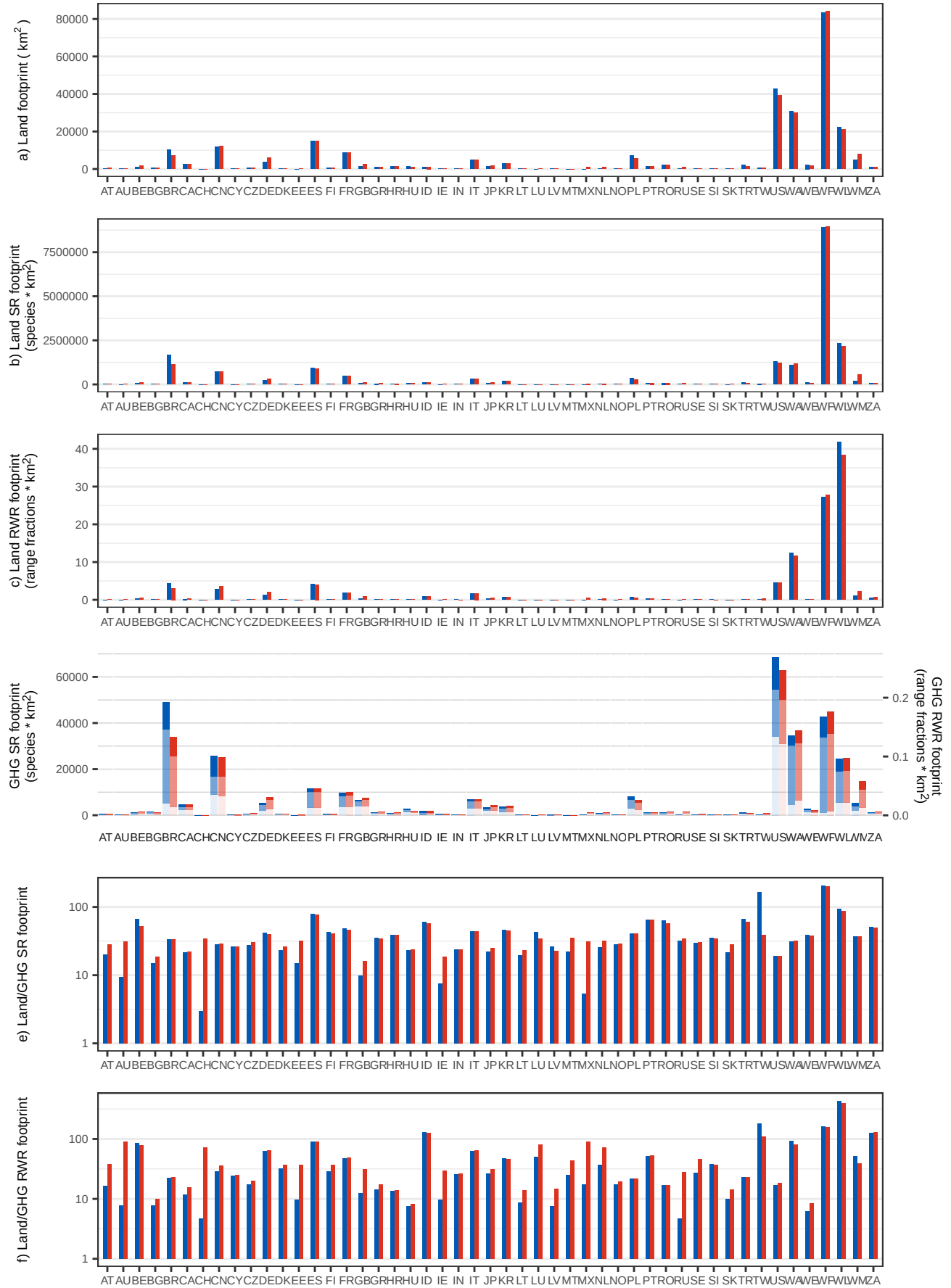
Products of meat cattle: production and consumption-based footprints



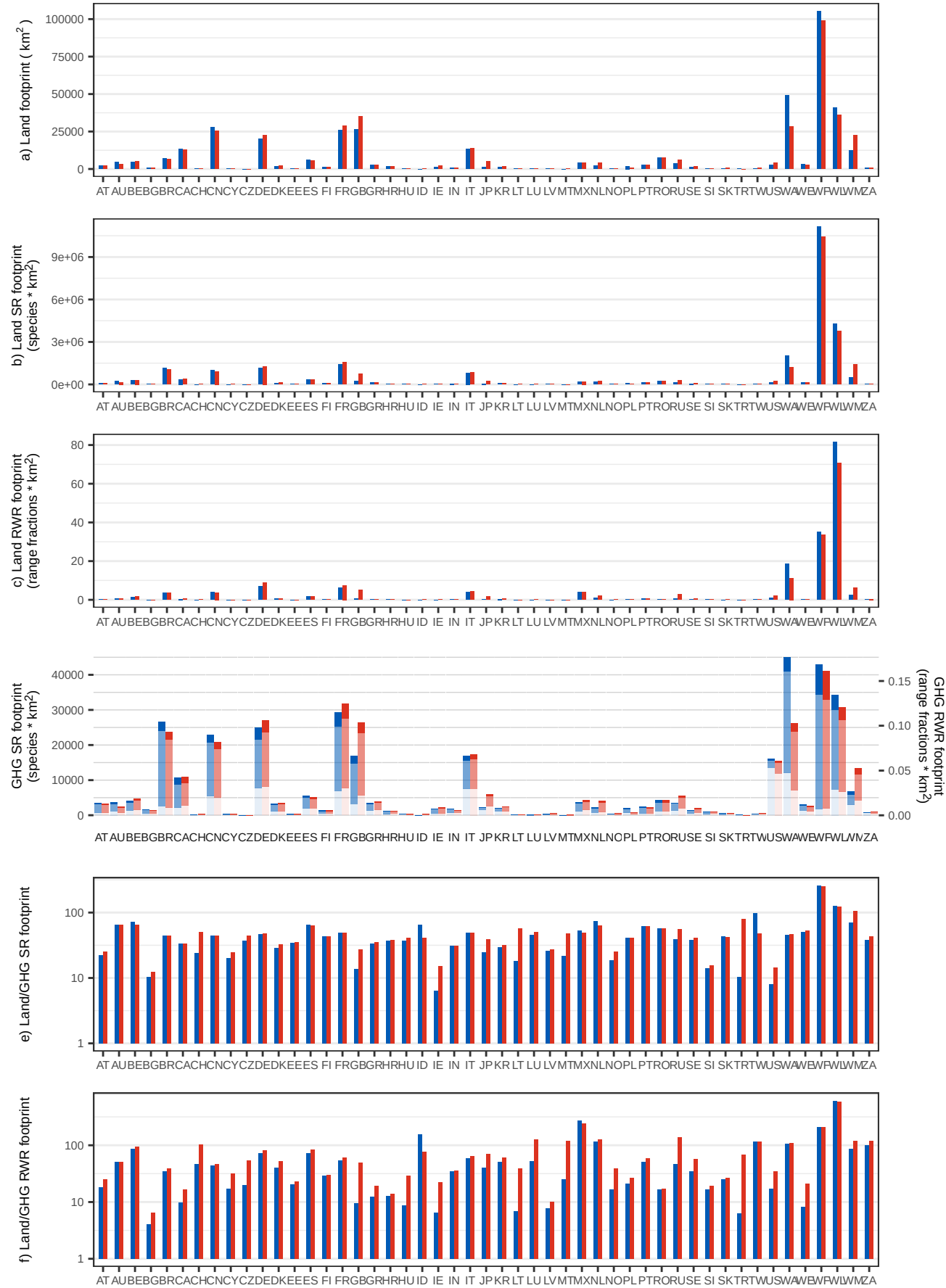
Supplementary Figure 1
Products of meat pigs: production and consumption-based footprints



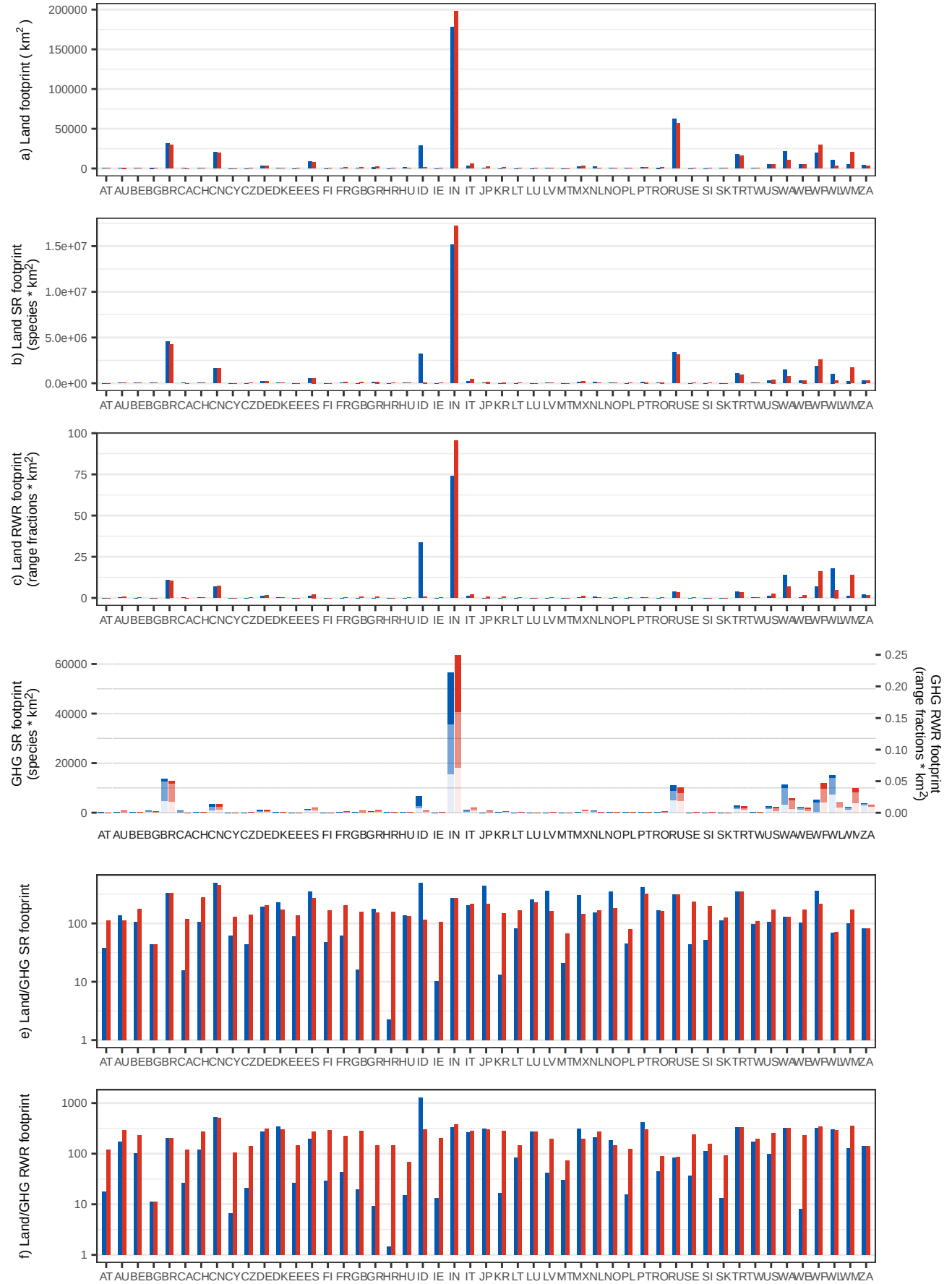
Supplementary Figure 1
Products of meat poultry: production and consumption-based footprints



Supplementary Figure 1
Meat products nec: production and consumption-based footprints

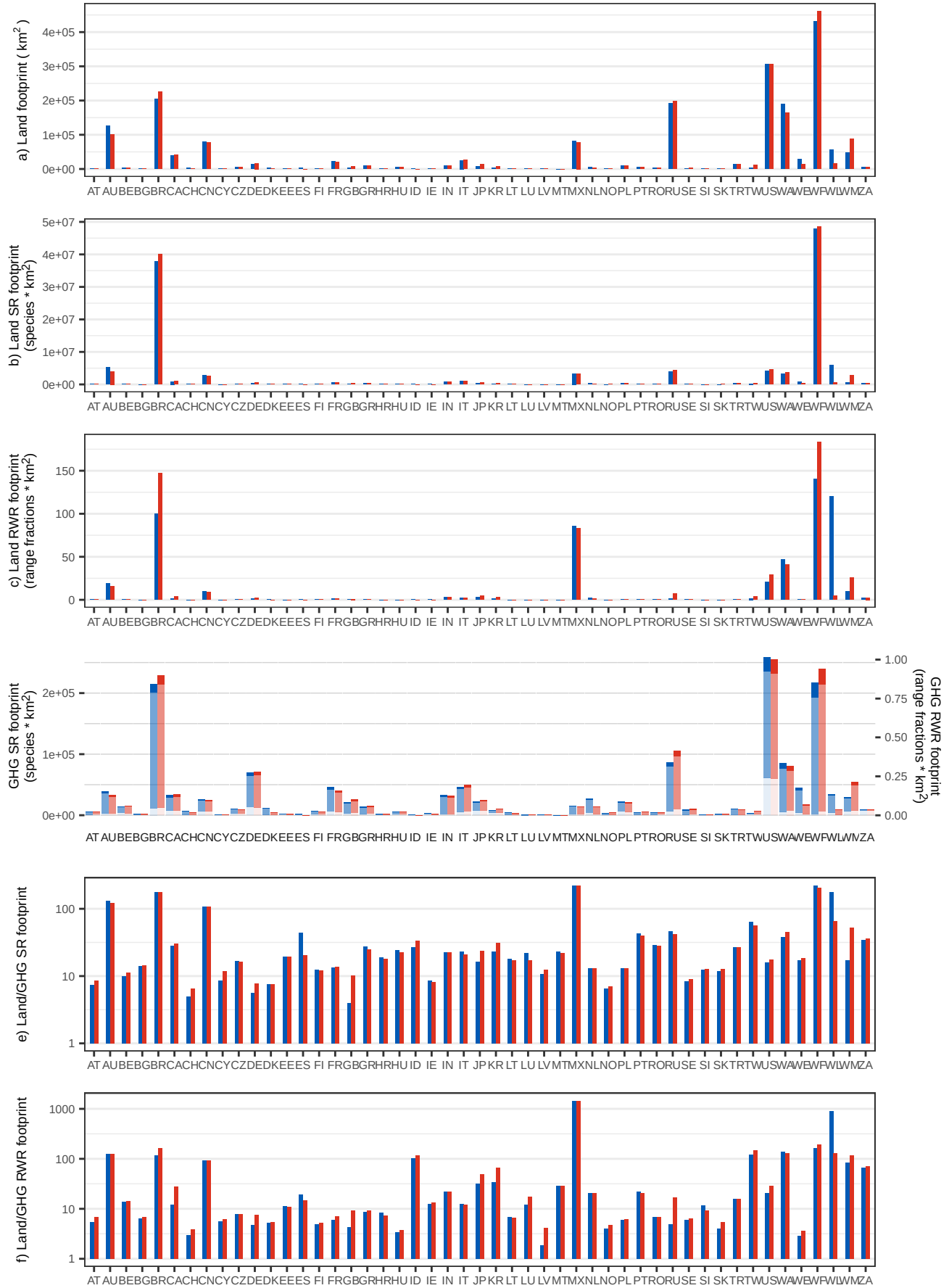


Supplementary Figure 1
Products of vegetable oils and fats: production and consumption-based footprints



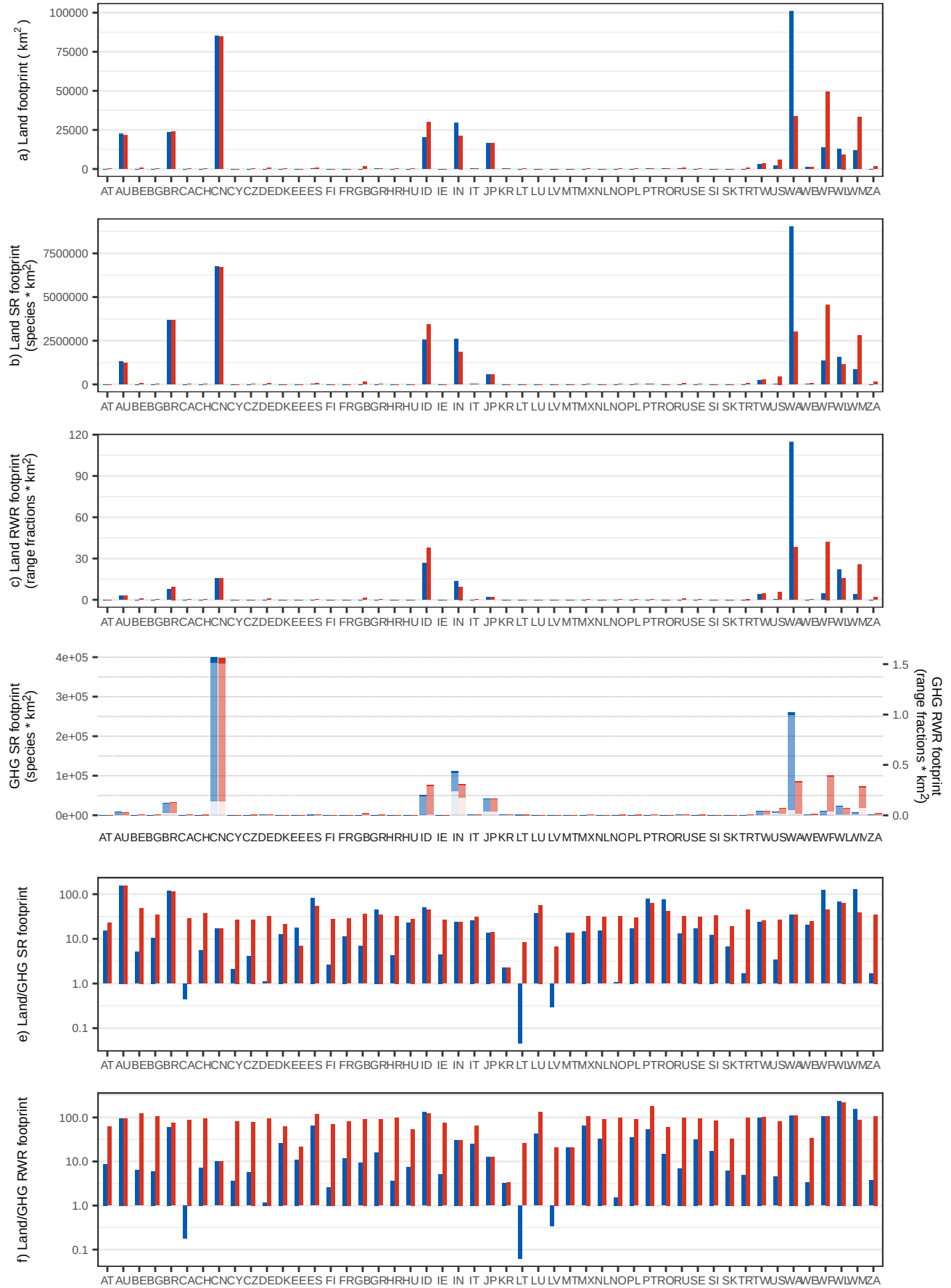
Supplementary Figure 1

Dairy products: production and consumption-based footprints



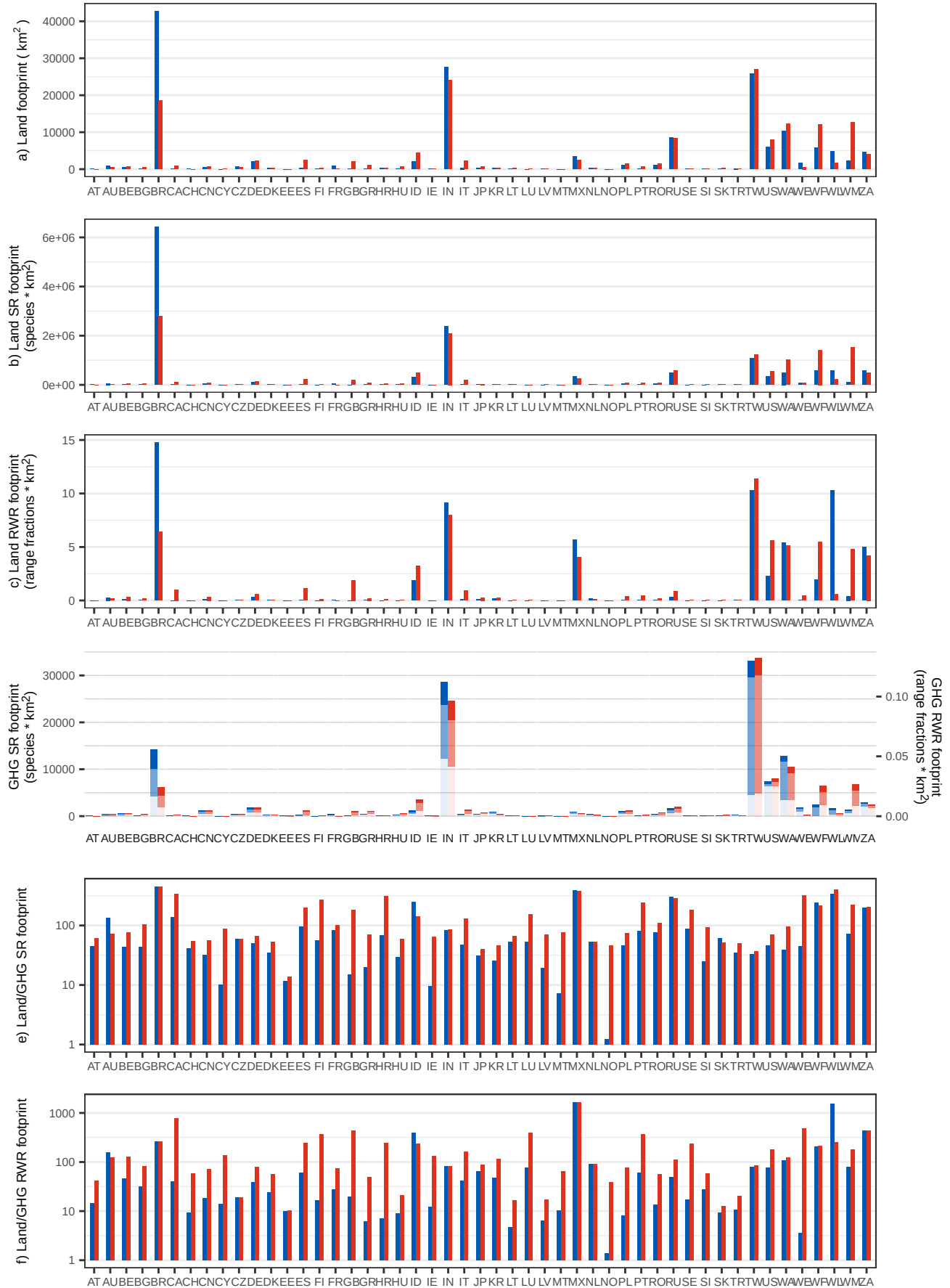
Supplementary Figure 1

Processed rice: production and consumption-based footprints



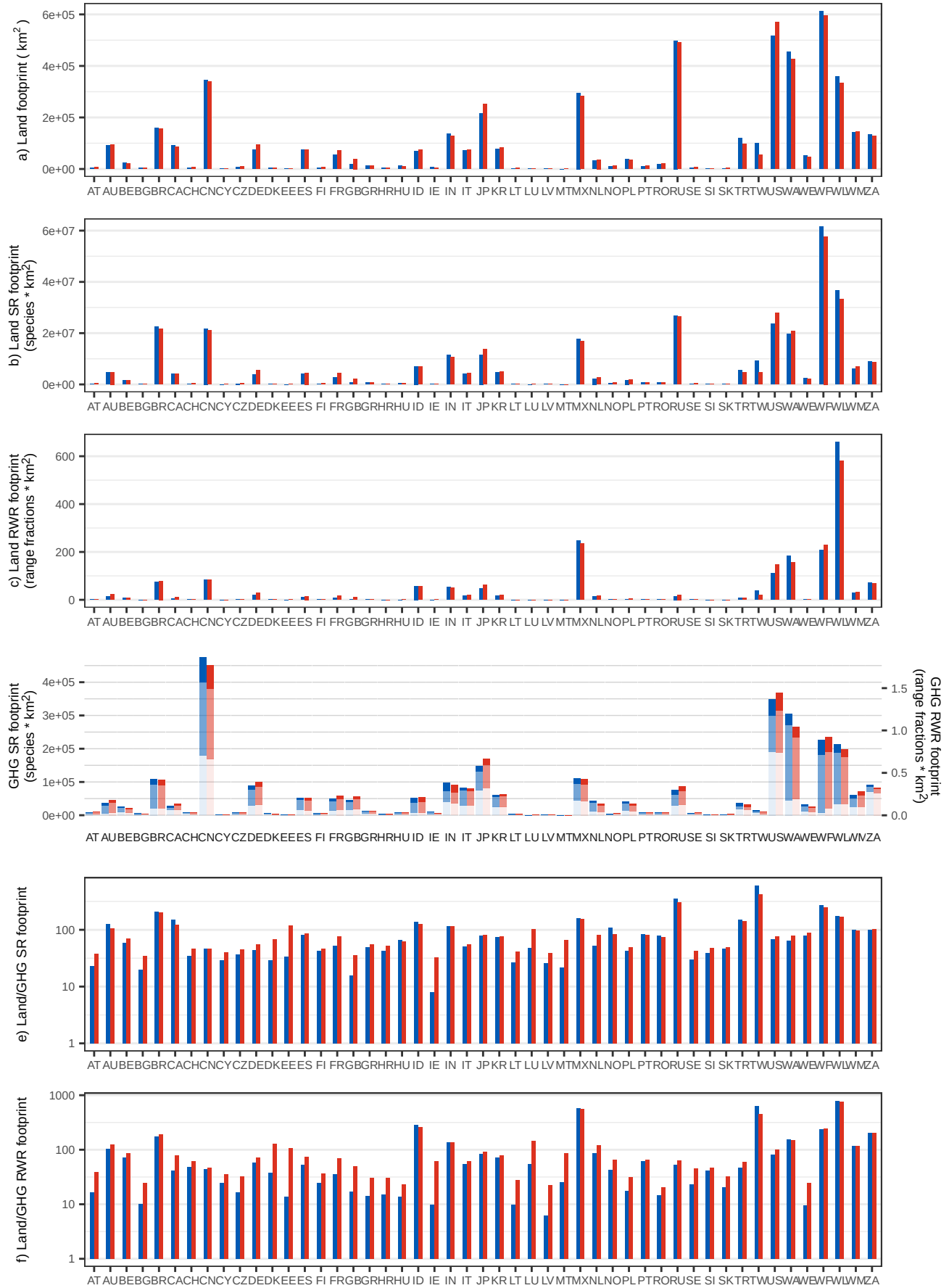
Supplementary Figure 1

Processed sugar: production and consumption-based footprints

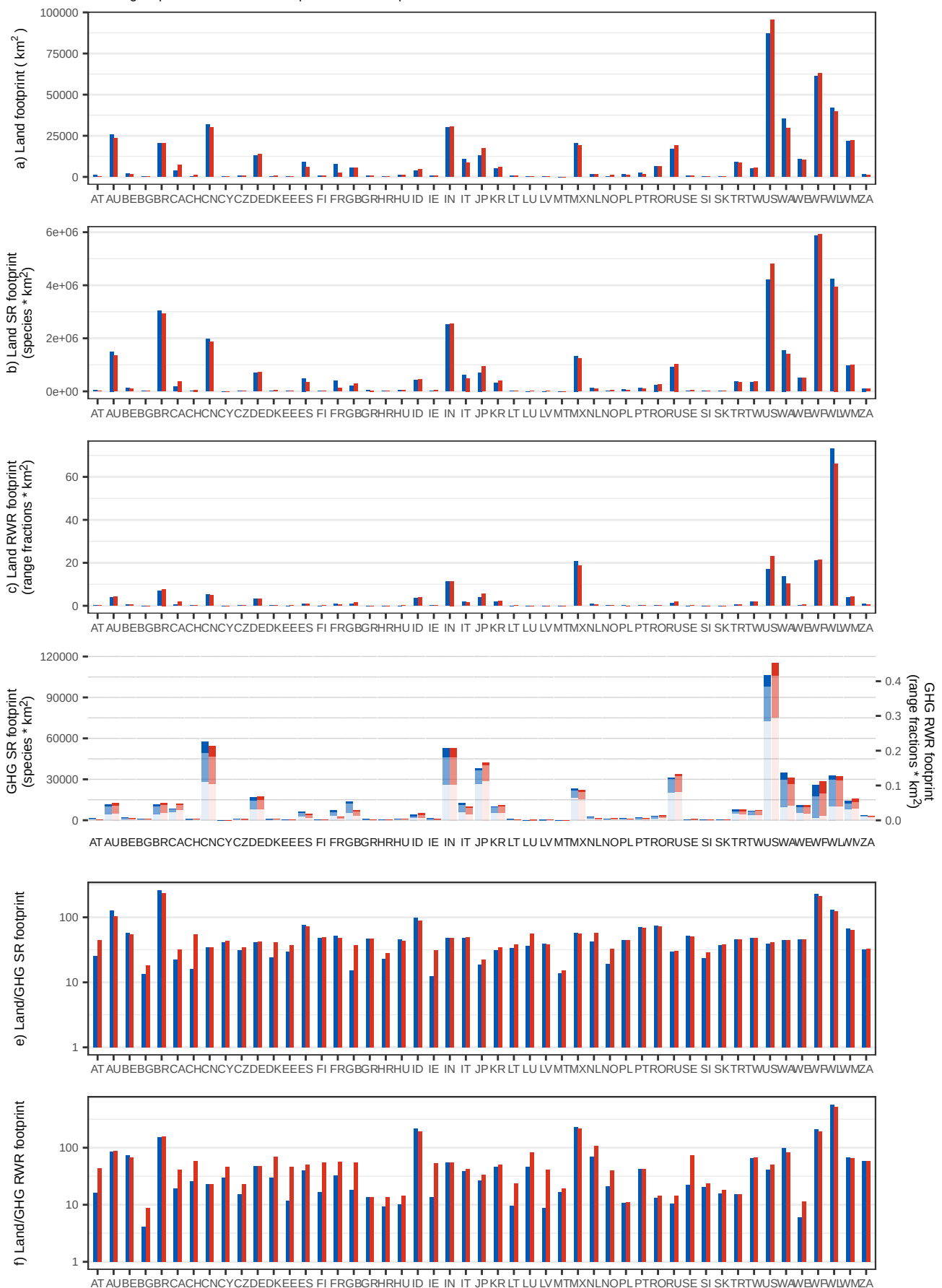


Supplementary Figure 1

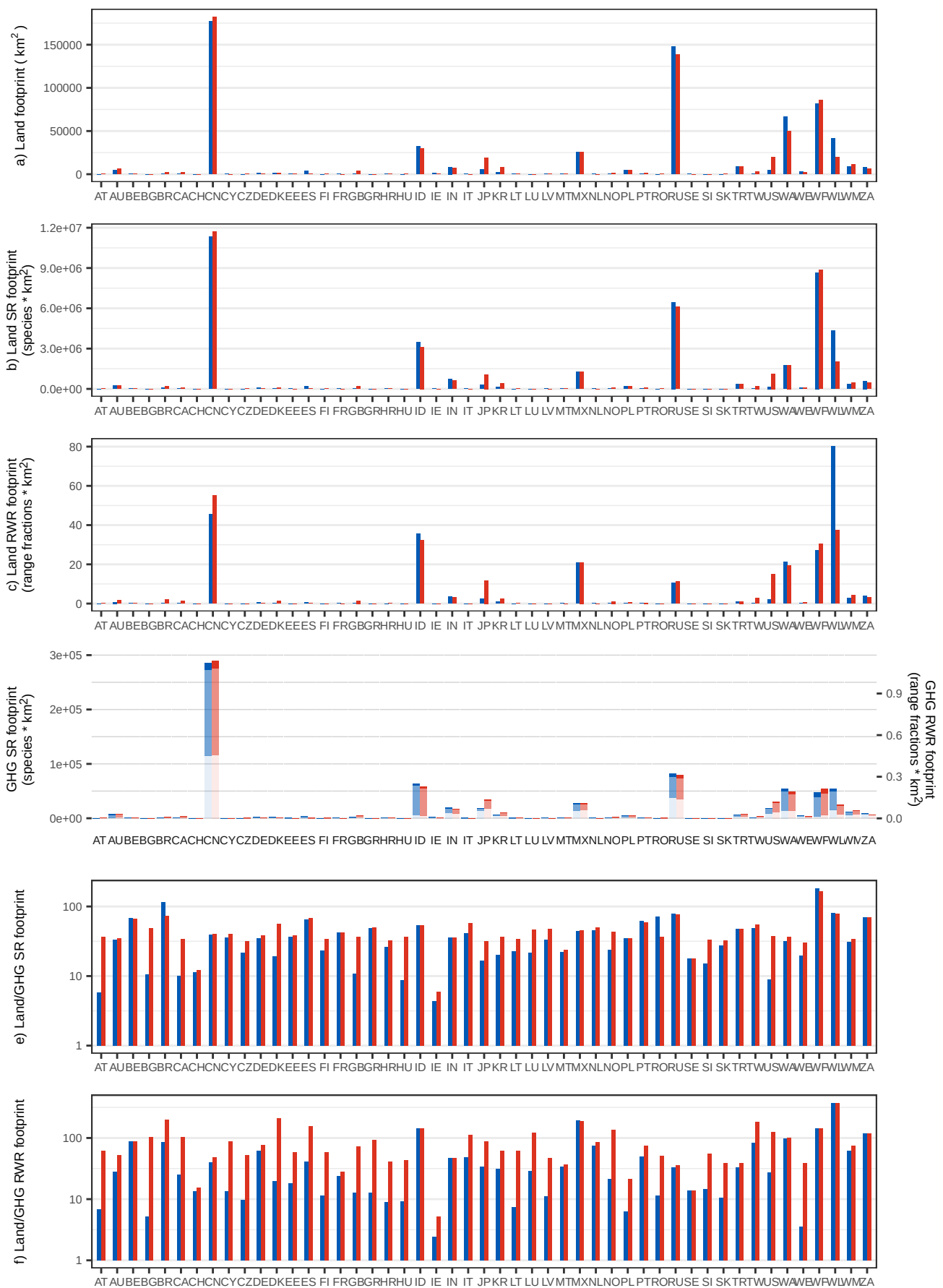
Food products nec: production and consumption-based footprints



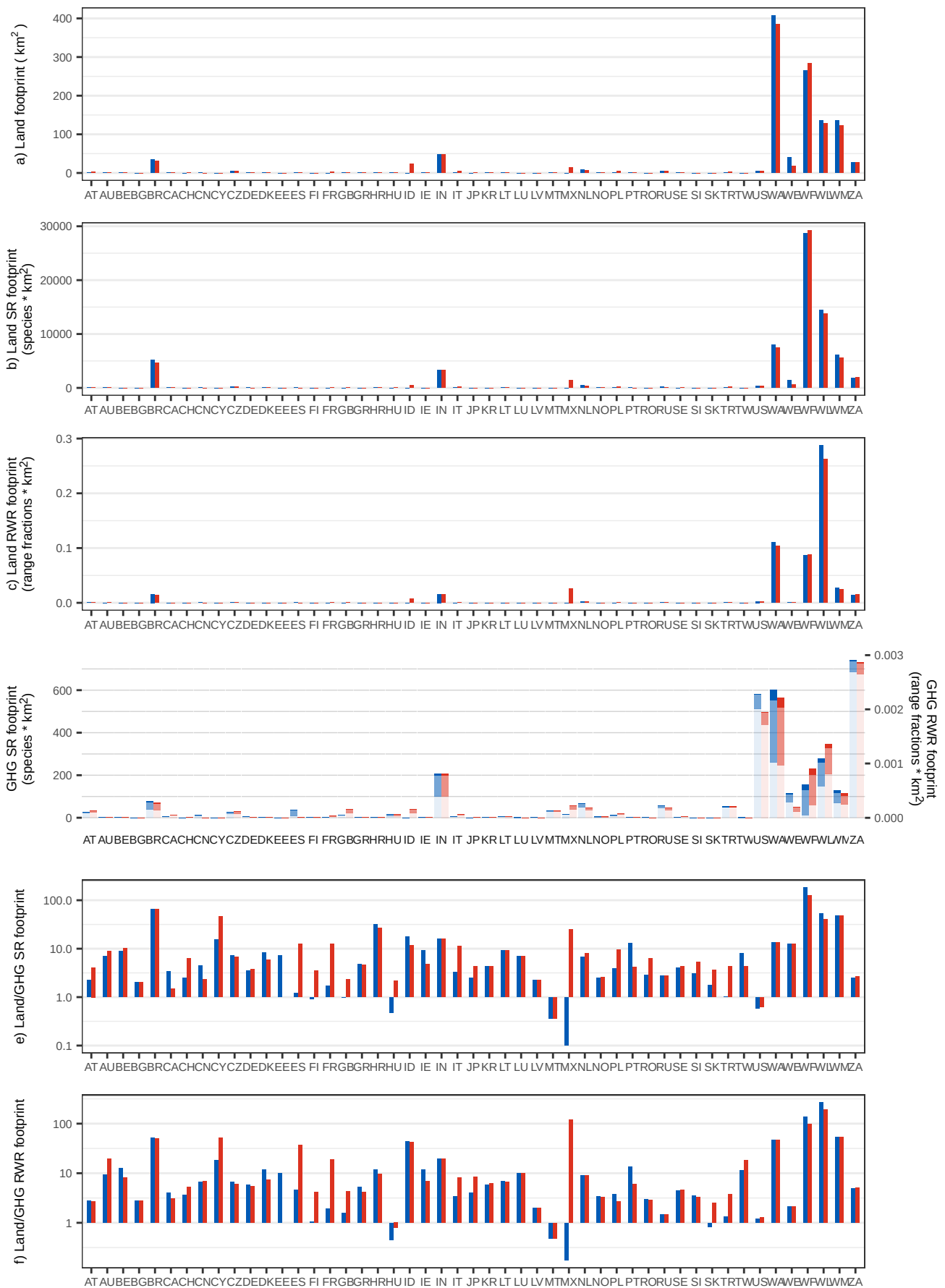
Beverages: production and consumption-based footprints



Fish products: production and consumption-based footprints

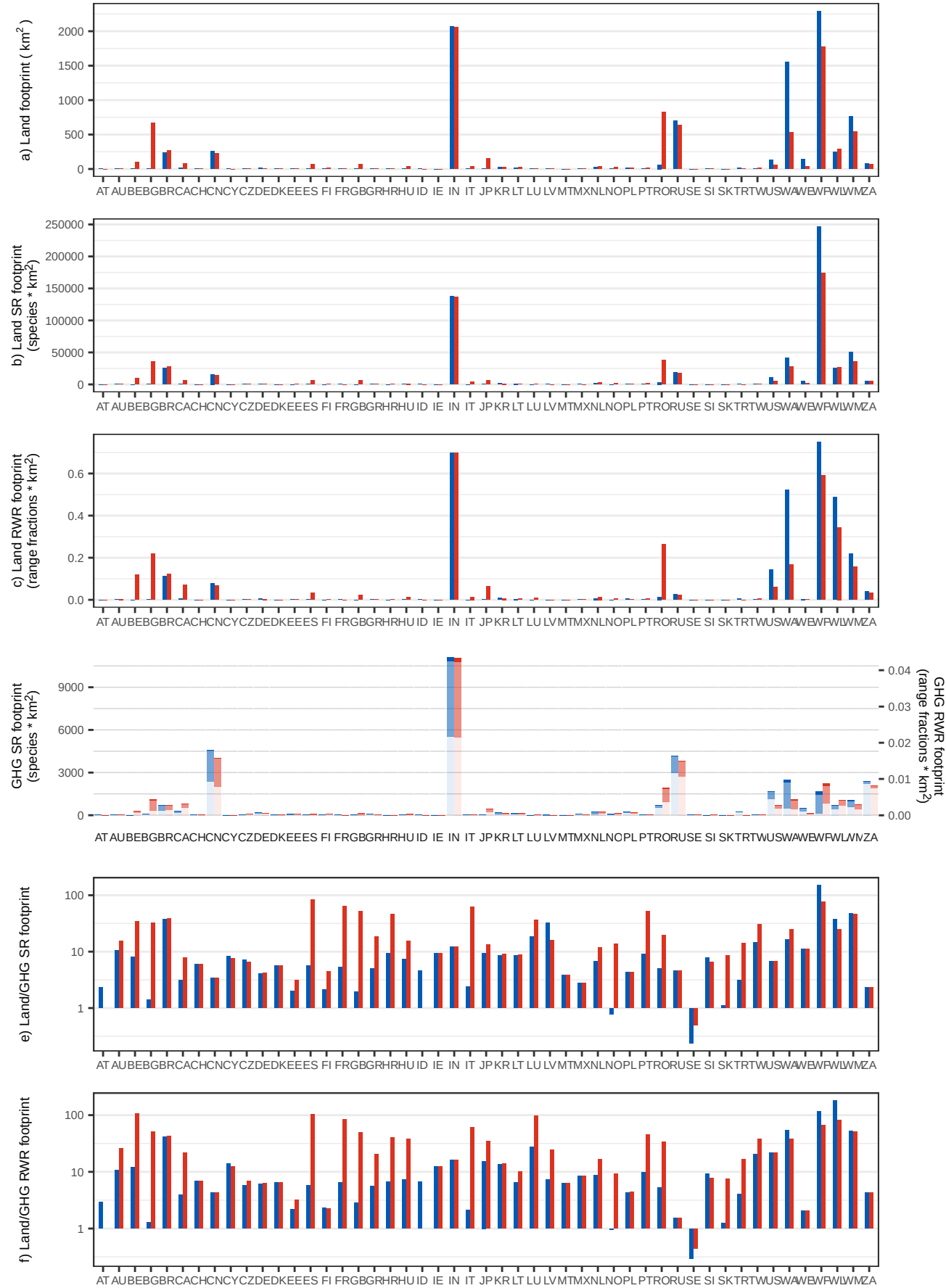


N-fertiliser: production and consumption-based footprints

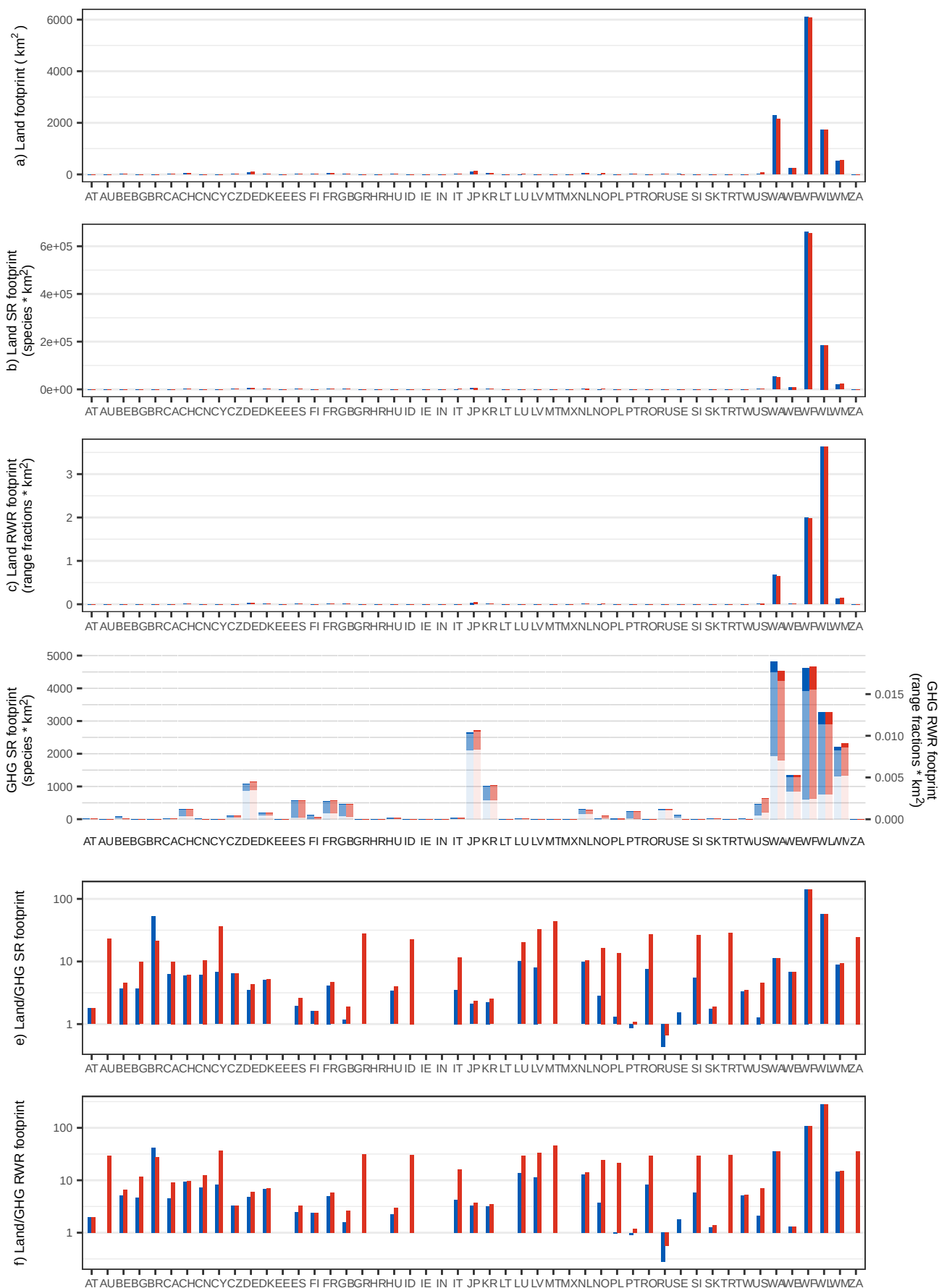


Supplementary Figure 1

P- and other fertiliser: production and consumption-based footprints

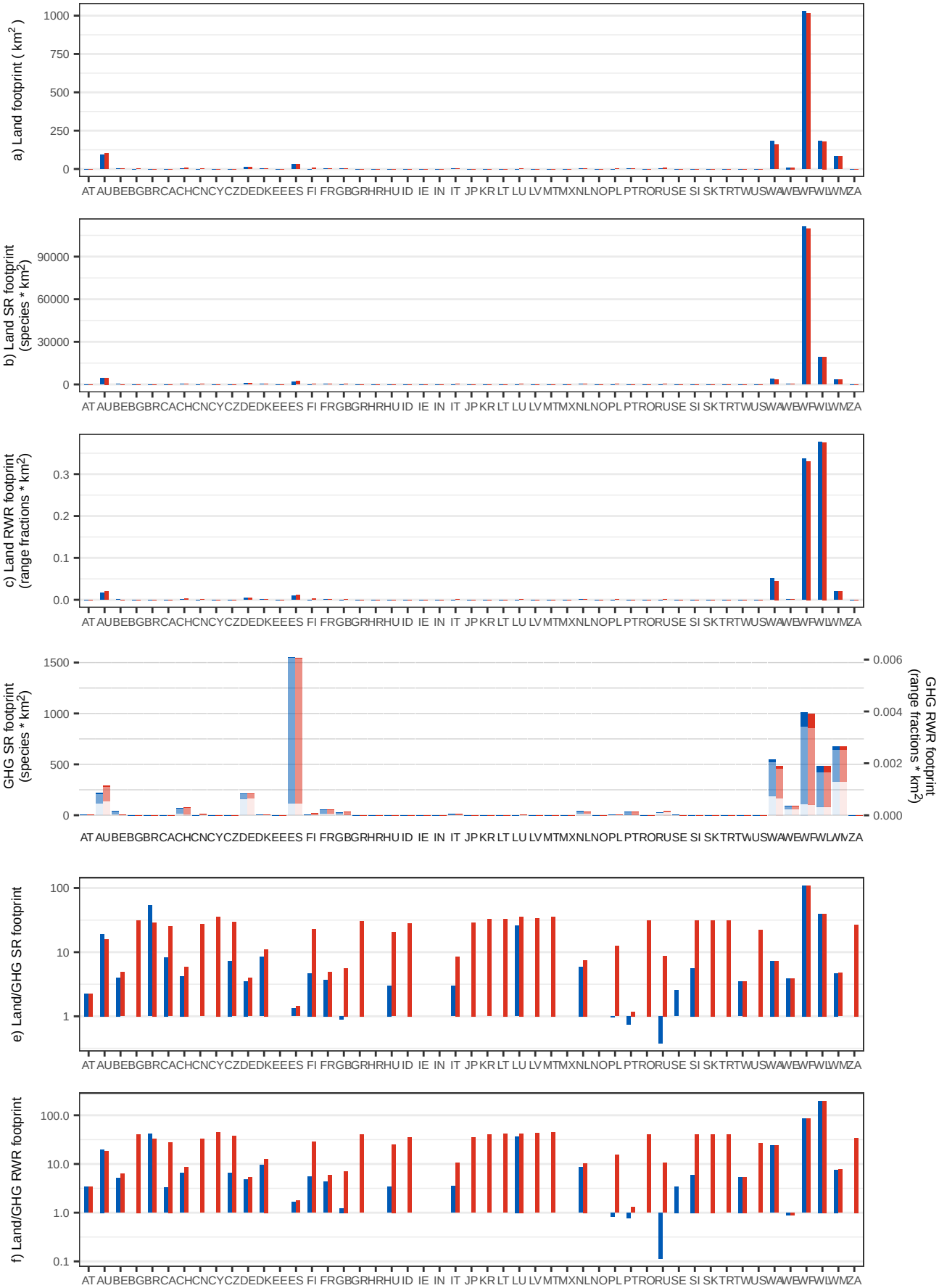


Food waste: incineration: production and consumption-based footprints



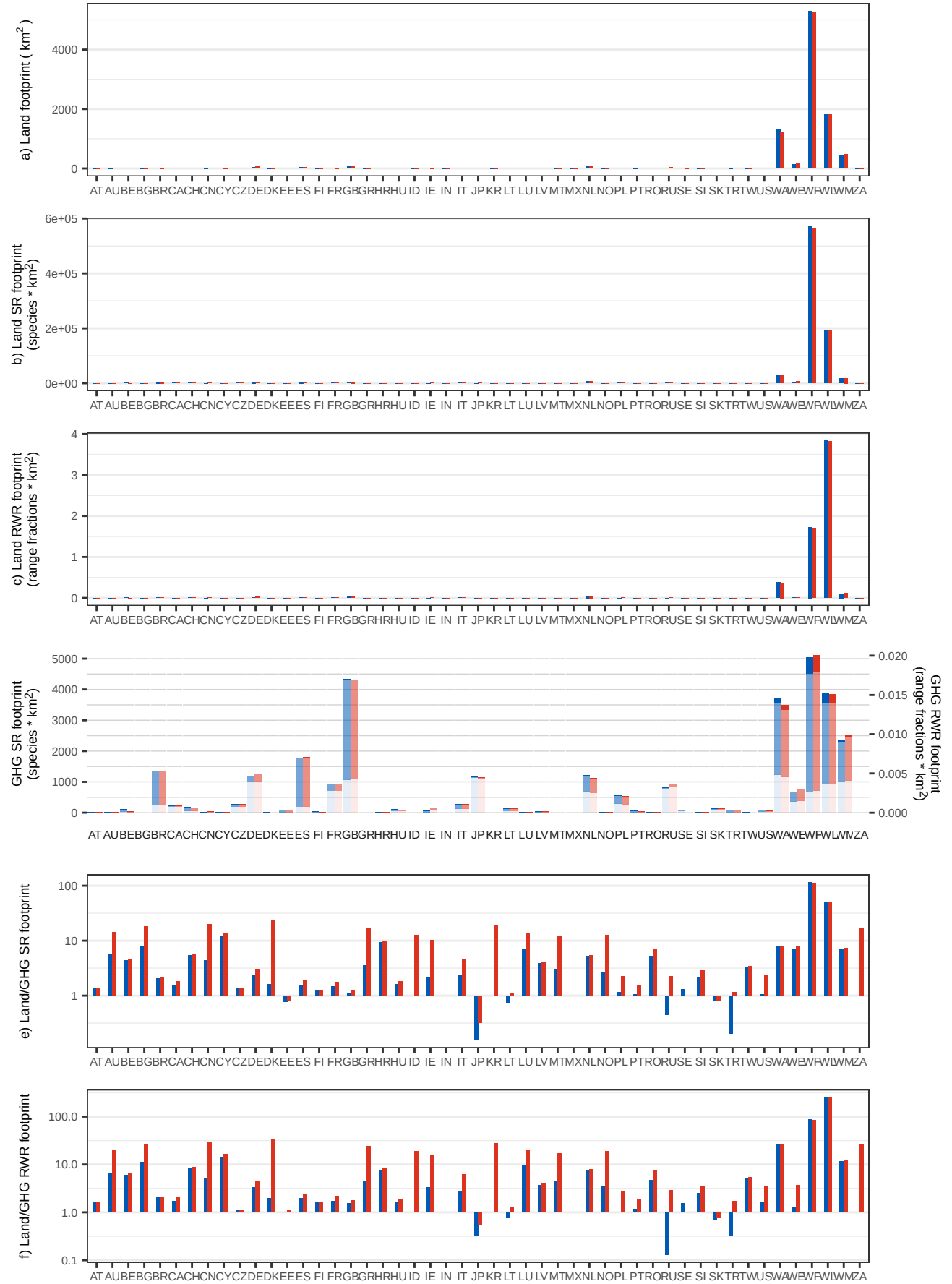
Supplementary Figure 1

Food waste: biogasification and land application: production and consumption-based footprints

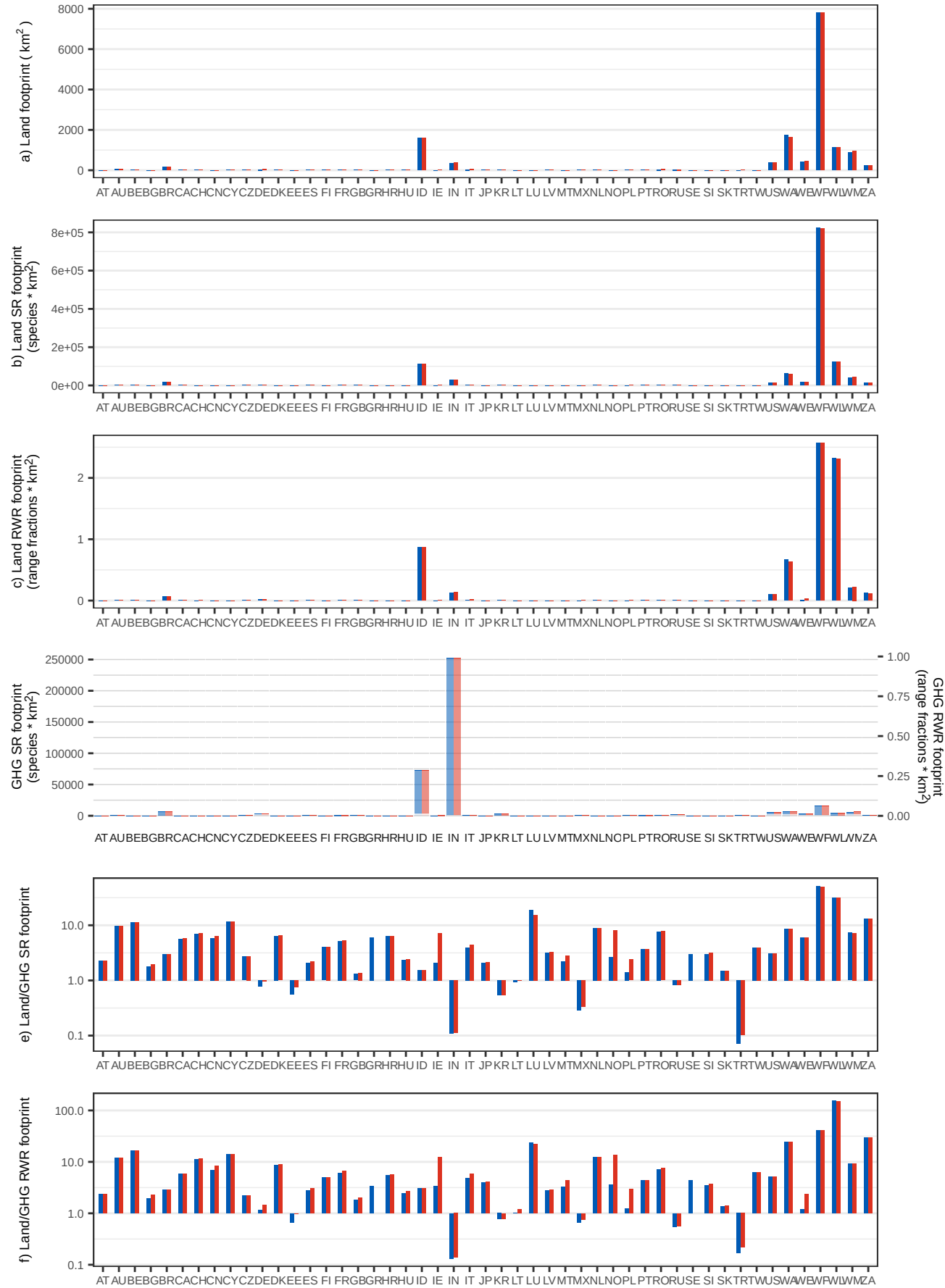


Supplementary Figure 1

Food waste: composting and land application: production and consumption-based footprints

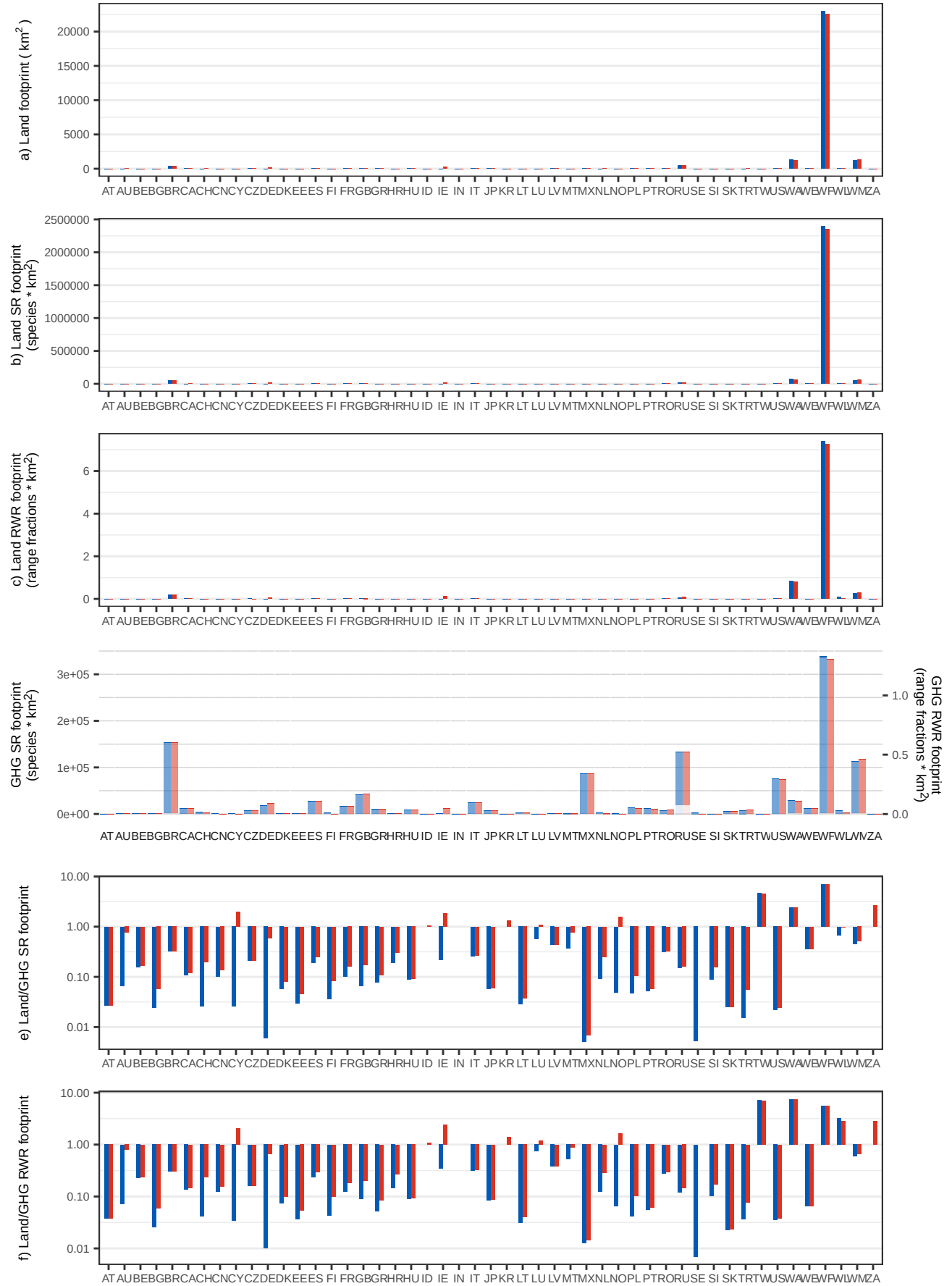


Supplementary Figure 1
Food waste: waste water treatment: production and consumption-based footprints

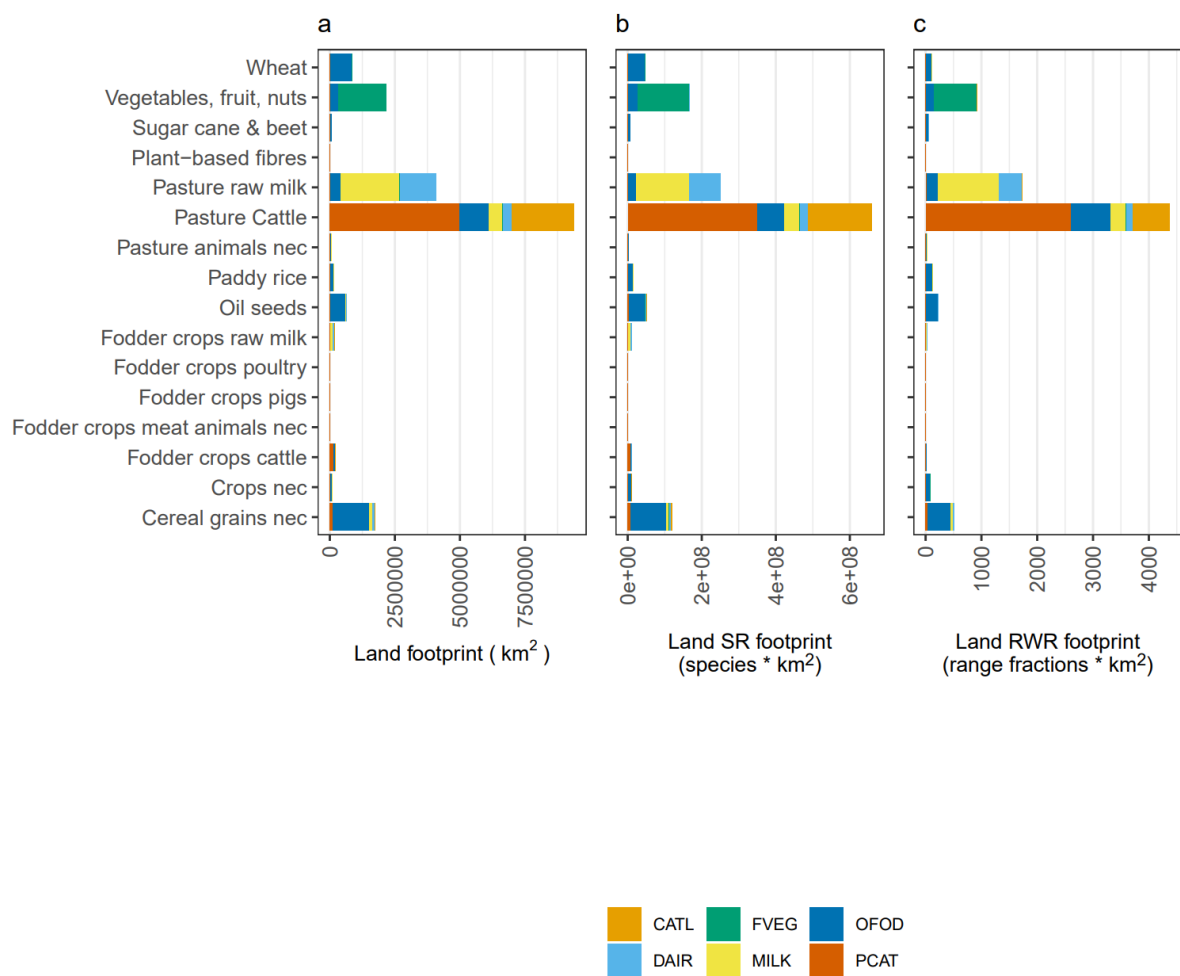


Supplementary Figure 1

Food waste: landfill: production and consumption-based footprints



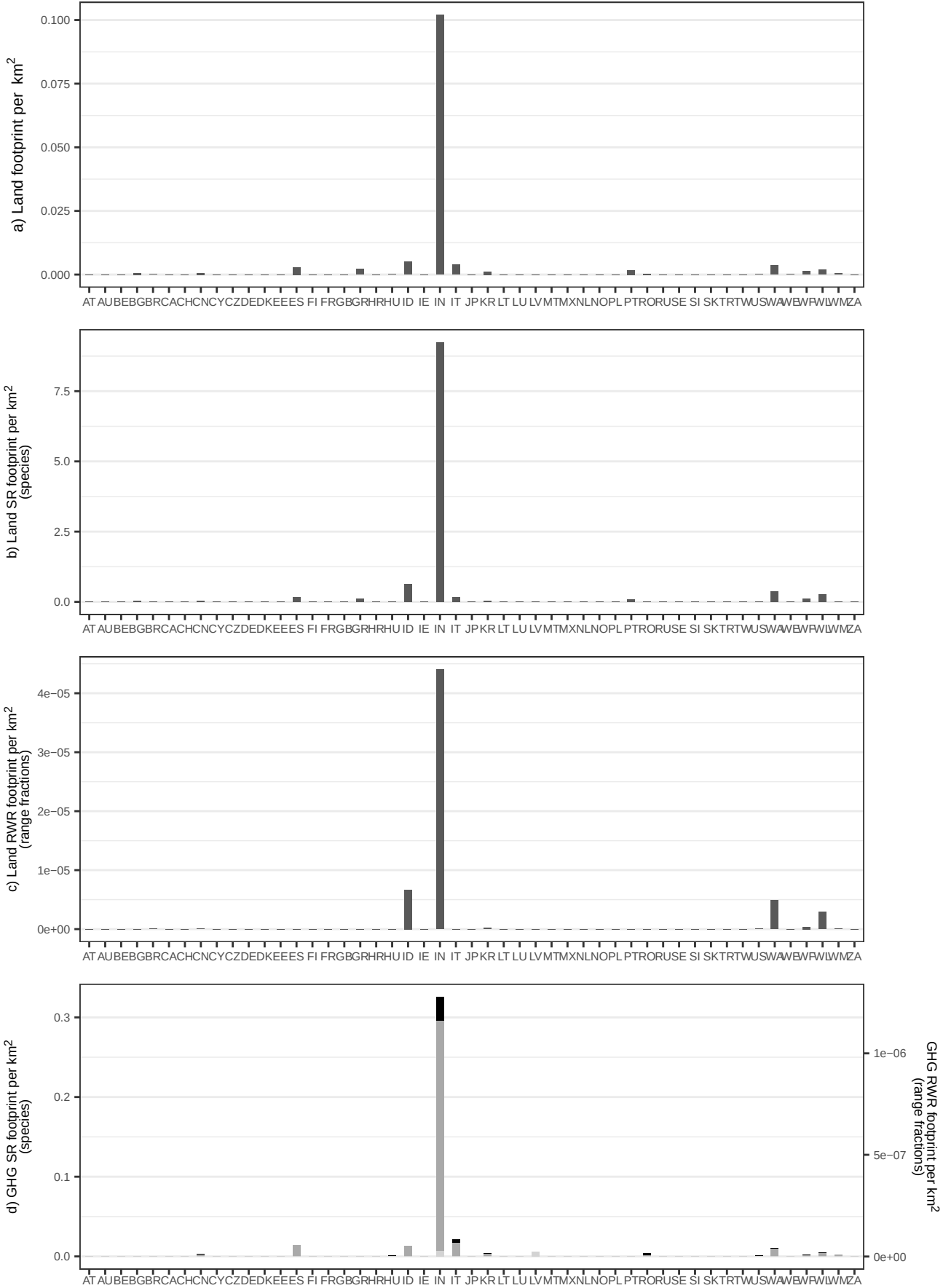
Supplementary Figure 5. Regions' production and consumption footprints of food-related products. The production (blue) and consumption (red) footprints of each individual food-related product in terms of a) land area, b) land-driven species richness loss, c) land-driven rarity-weighted richness loss, d) GHG-driven biodiversity loss split by emissions type: carbon dioxide (dark blue/red), methane (mid blue/red), nitrous oxide (light blue/red) (right-hand axis – species richness; left hand axis – rarity-weighted richness), e) the ratio of land-driven species richness loss to GHG-driven species richness loss and f) the ratio of land-driven rarity-weighted richness loss to GHG-driven rarity-weighted richness loss. Results are for the year 2011. AT = Austria, AU = Australia, BE = Belgium, BG = Bulgaria, BR = Brazil, CA = Canada, CH = Switzerland, CN = China, CY = Cyprus, CZ = Czech Republic, DE = Germany, DK = Denmark, EE = Estonia, ES = Spain, FI = Finland, GB = United Kingdom, HR = Croatia, HU = Hungary, ID = Indonesia, IE = Ireland, IN = India, IT = Italy, JP = Japan, KR = South Korea, LT = Lithuania, LV = Latvia, MT = Malta, MX = Mexico, NL = Netherlands, NO = Norway, PL = Poland, PT = Portugal, RO = Romania, RU = Russia, SE = Sweden, SI = Slovenia, SK = Slovakia, TR = Turkey, TW = Taiwan, US = United States, WA = RoW Asia & Pacific, WE = RoW Europe, WF = RoW Africa, WL = RoW CS America, WM = RoW Middle East, ZA = South Africa. Source data are provided in Supplementary Data 7.



Supplementary Figure 6. The contribution of land use types to the production footprints of selected food products. The distribution of production-based footprints of a) land-use, b) land-driven species richness and c) land-driven rarity weighted richness across the EXIOBASE land-use categories for 6 high-footprint food sectors (CATL = cattle, DAIR = dairy products, FVEG = fruit/vegetables/nuts, MILK = raw milk, OFOD = food products nec, PCAT = products of cattle). Results are for the year 2011. Source data are provided as a source data file.

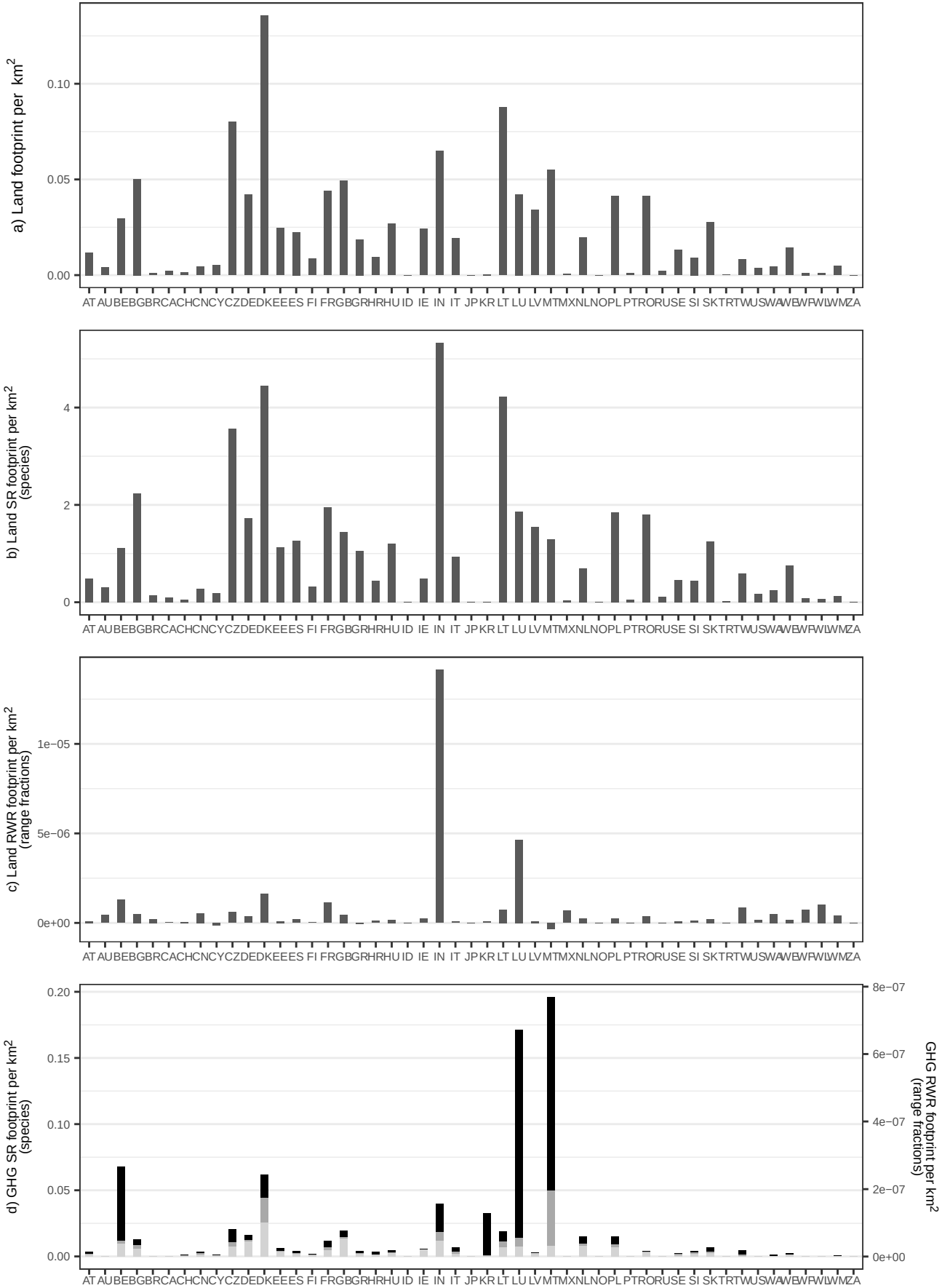
Supplementary Figure 7

Paddy rice: production-based footprints per square km



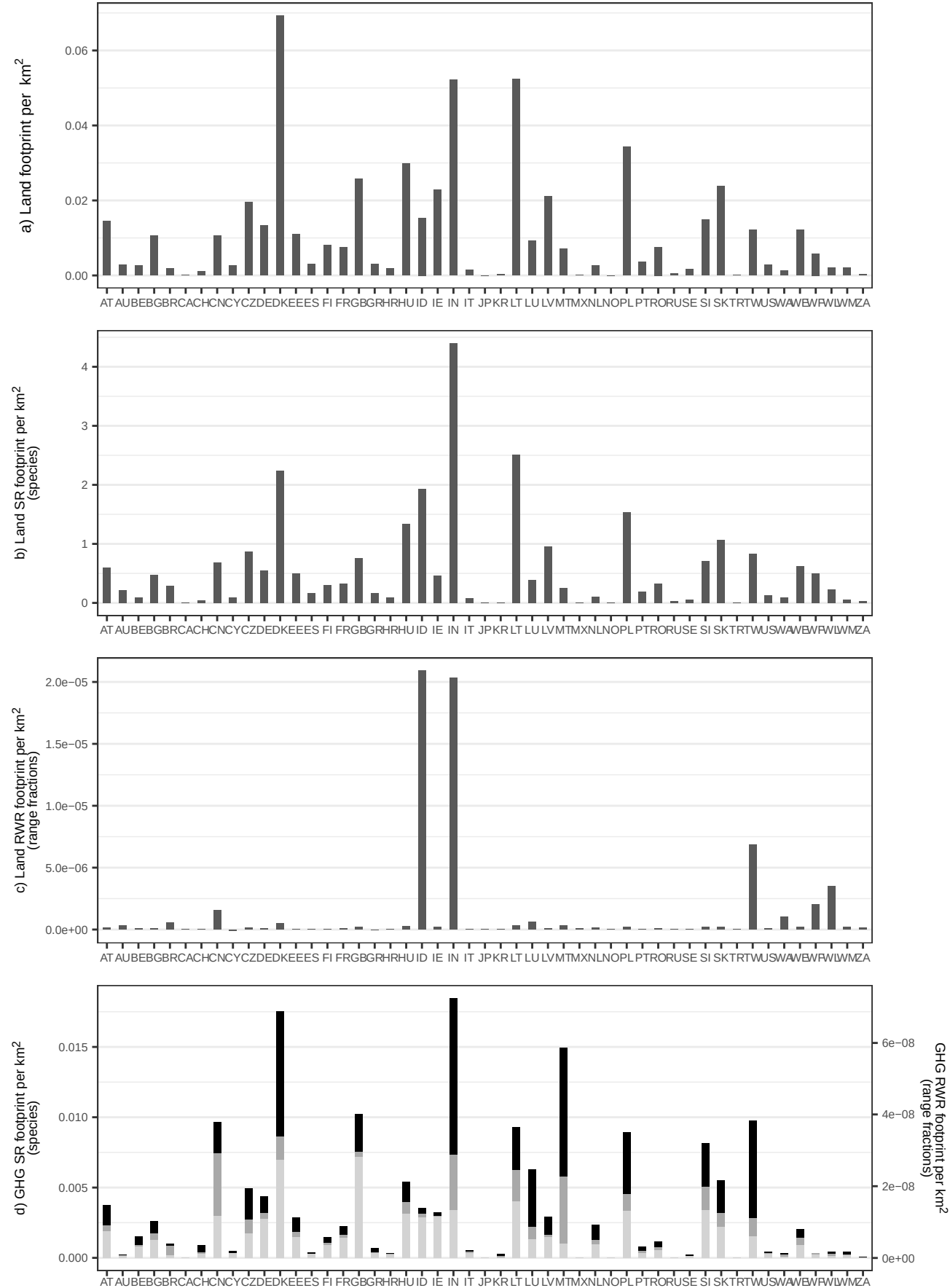
Supplementary Figure 7

Wheat: production-based footprints per square km



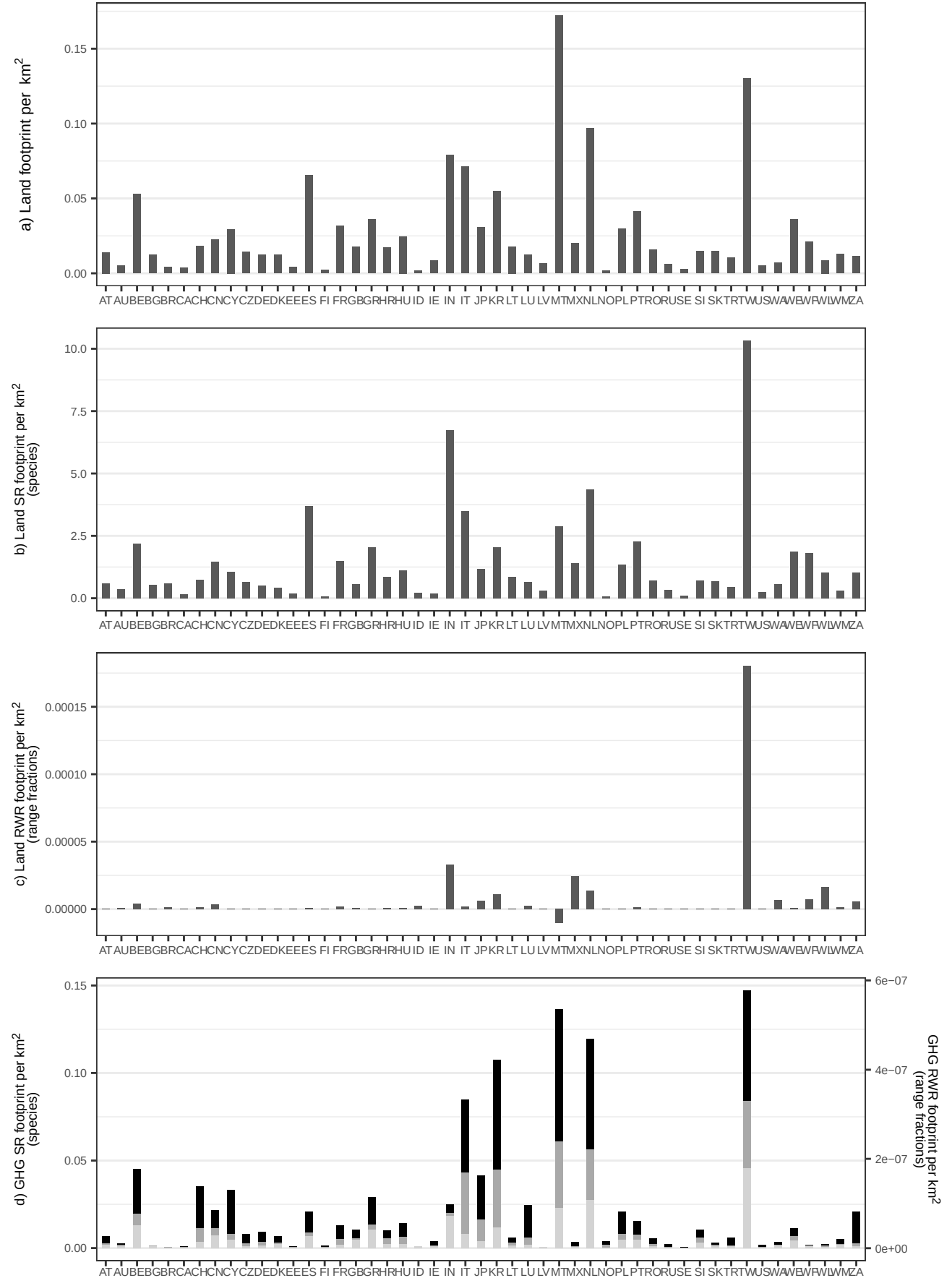
Supplementary Figure 7

Cereal grains nec: production-based footprints per square km



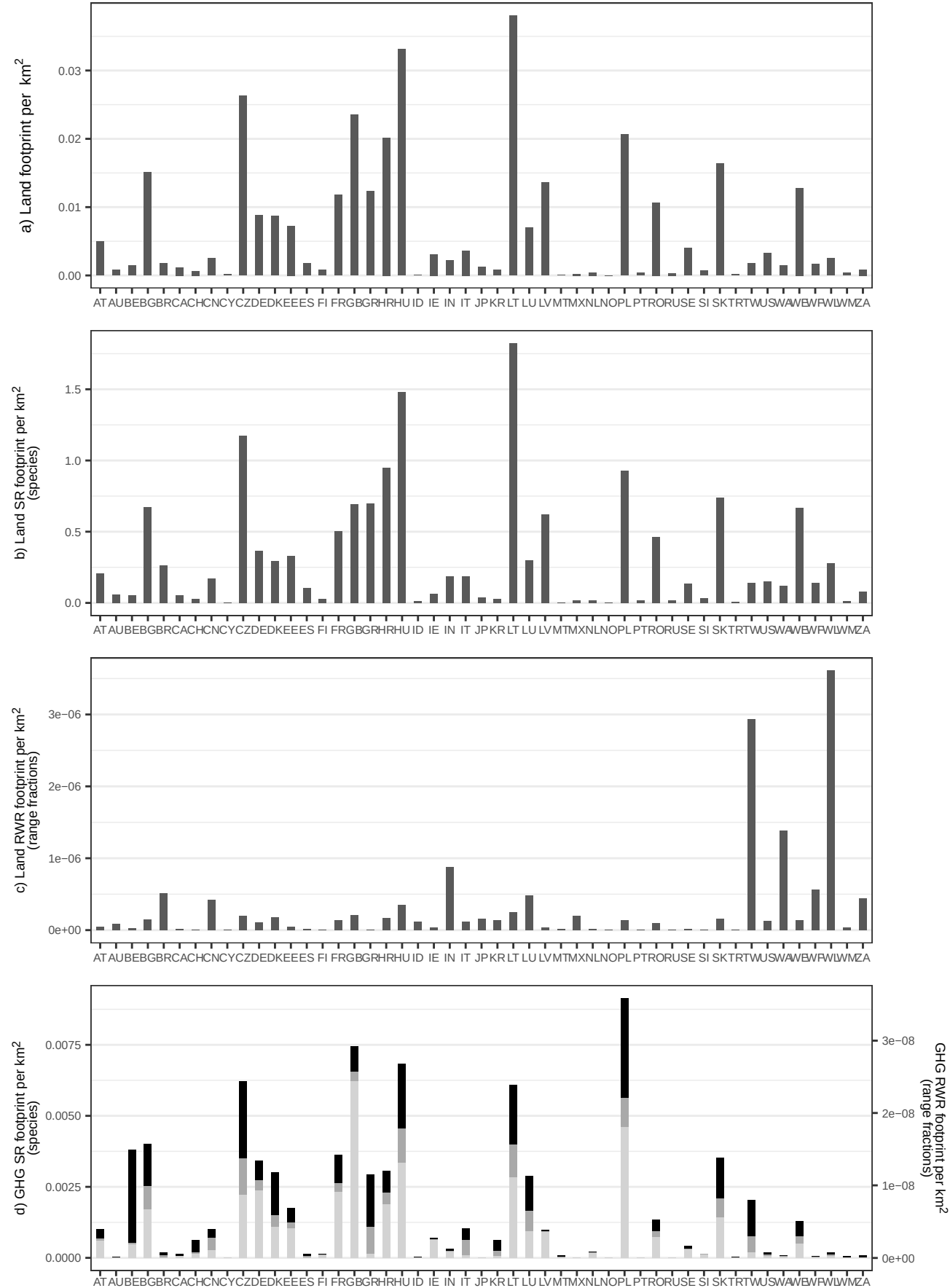
Supplementary Figure 7

Veg/Fruit/Nuts: production-based footprints per square km

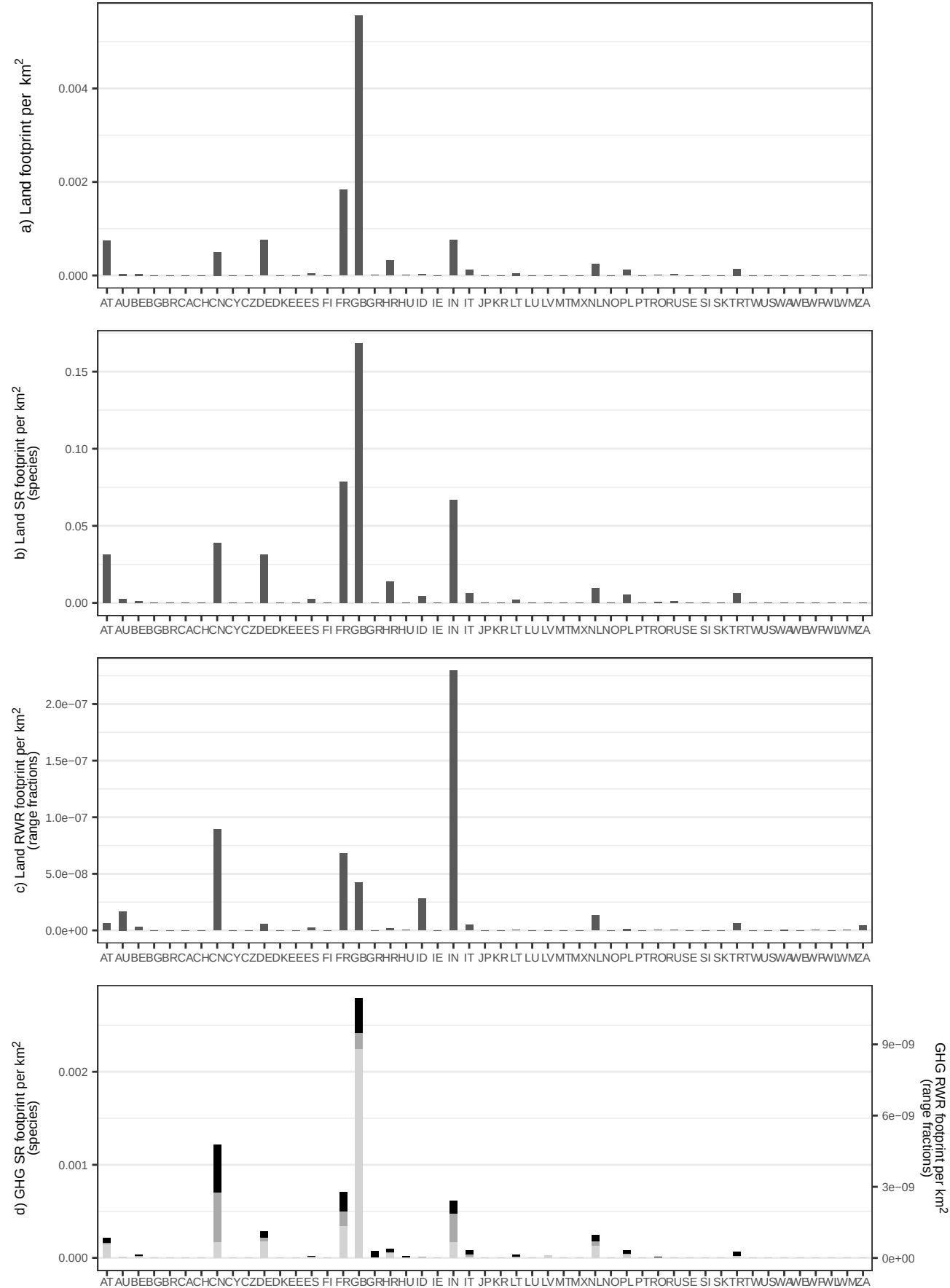


Supplementary Figure 7

Oil seeds: production-based footprints per square km

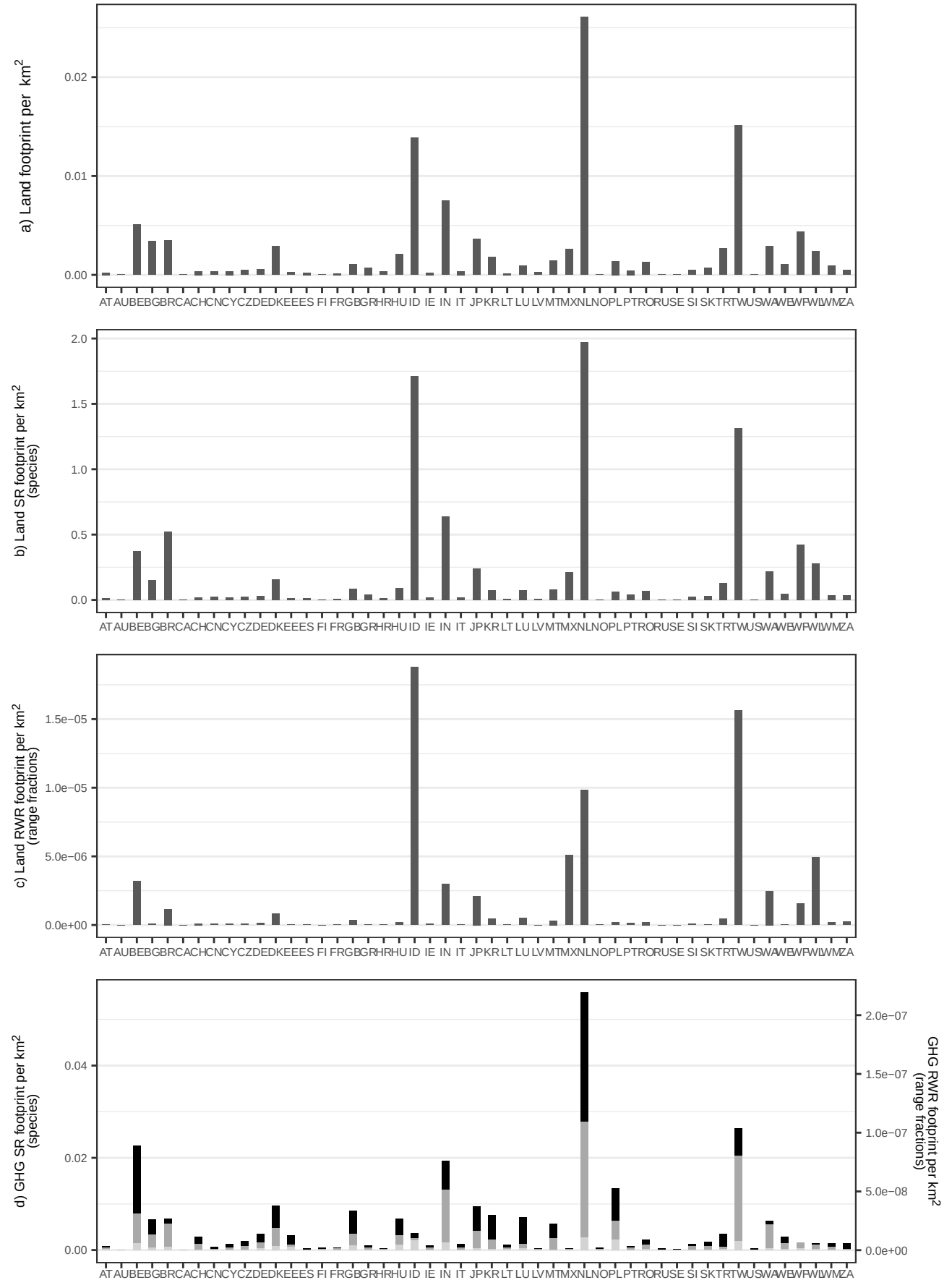


Supplementary Figure 7
Sugar beet/cane: production-based footprints per square km



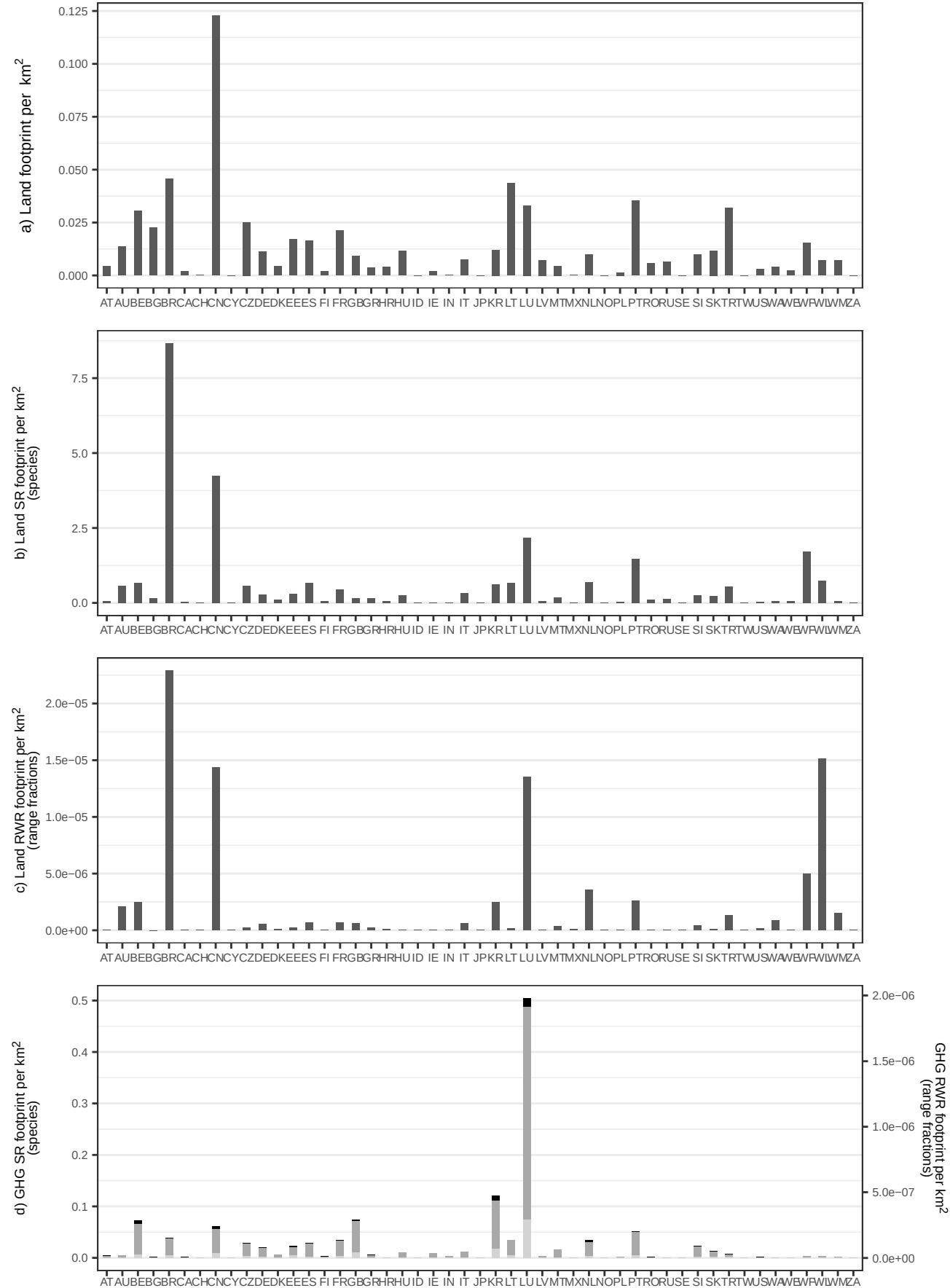
Supplementary Figure 7

Crops nec: production-based footprints per square km



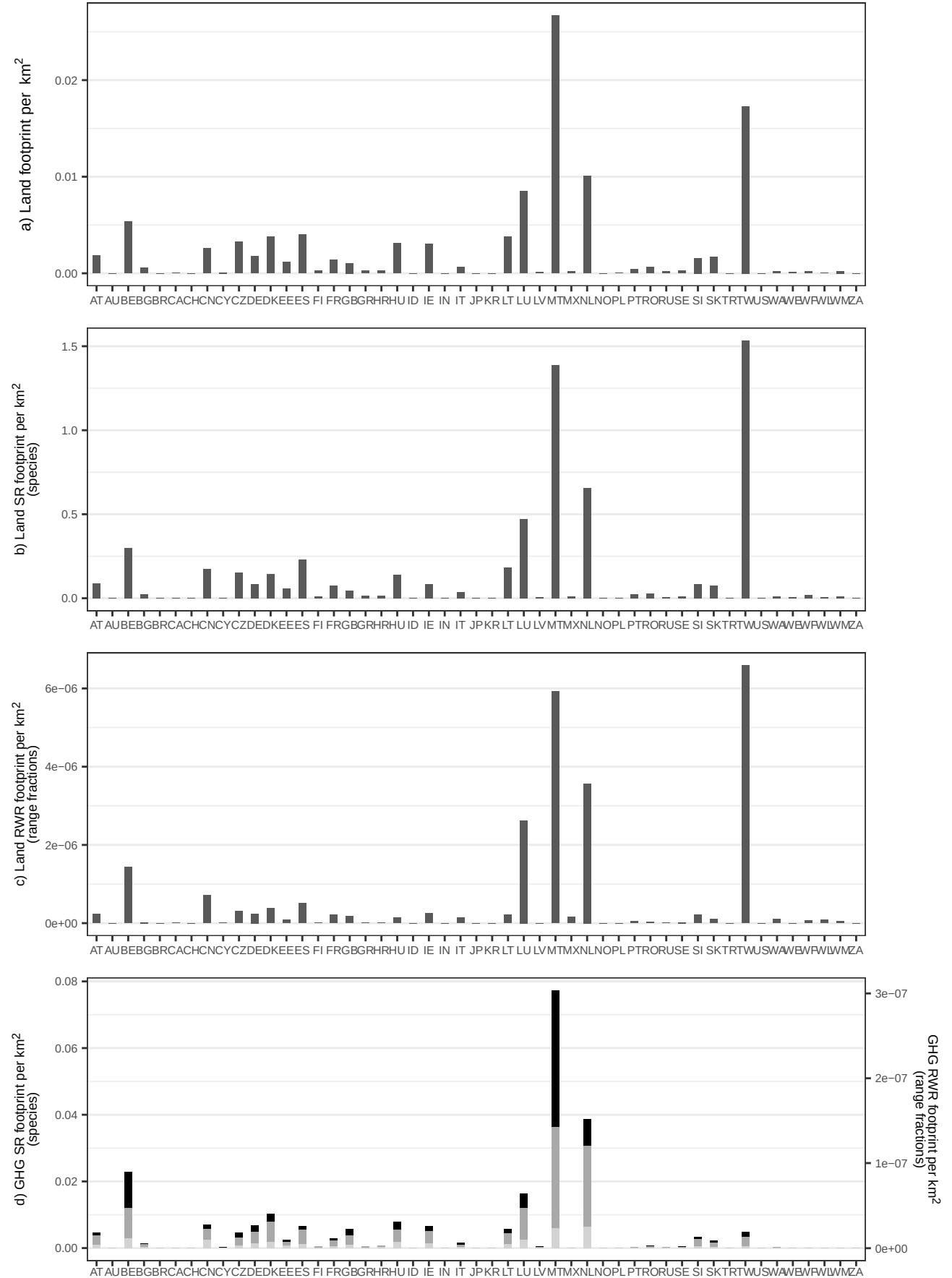
Supplementary Figure 7

Cattle: production-based footprints per square km



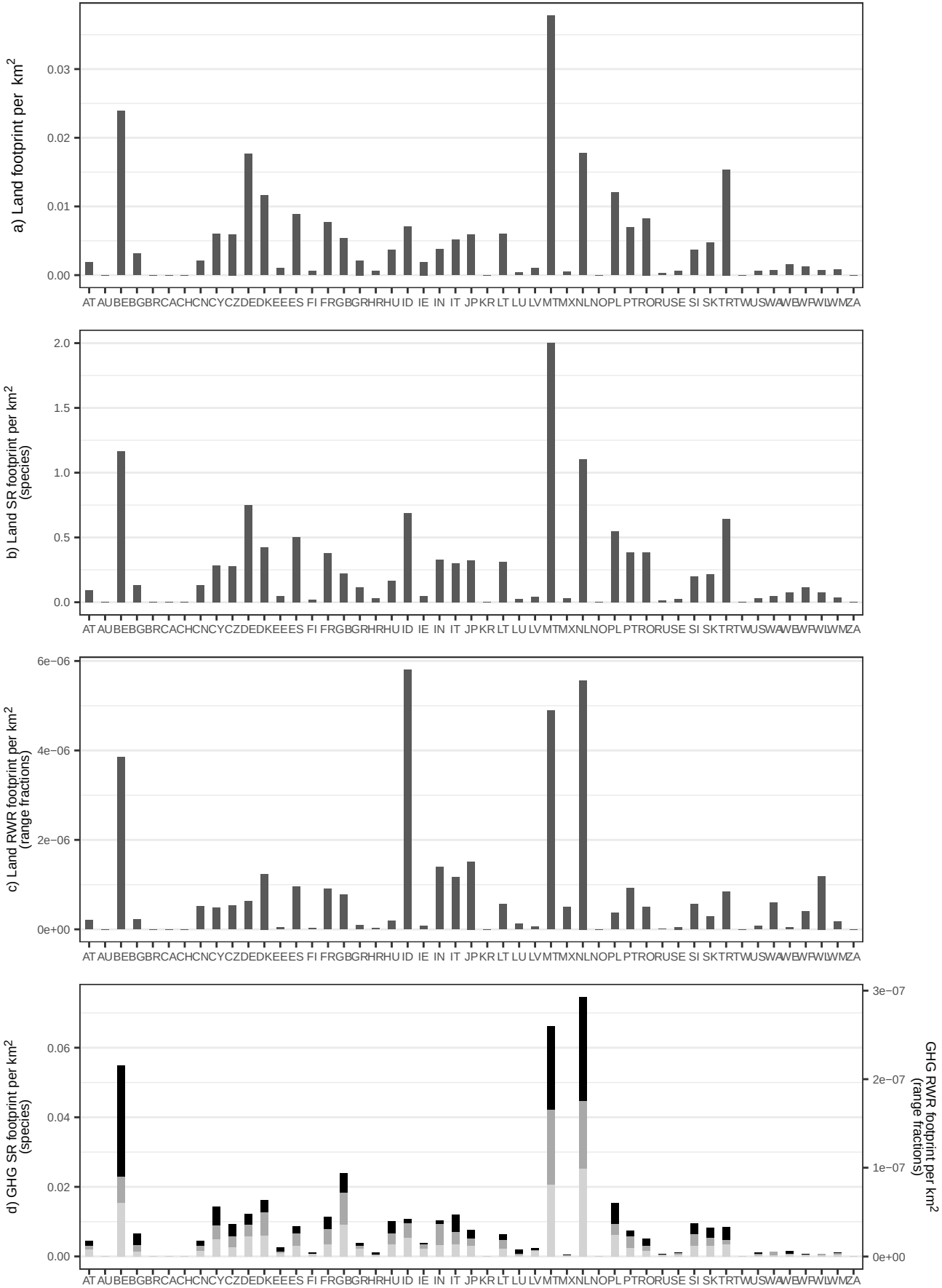
Supplementary Figure 7

Pigs: production-based footprints per square km



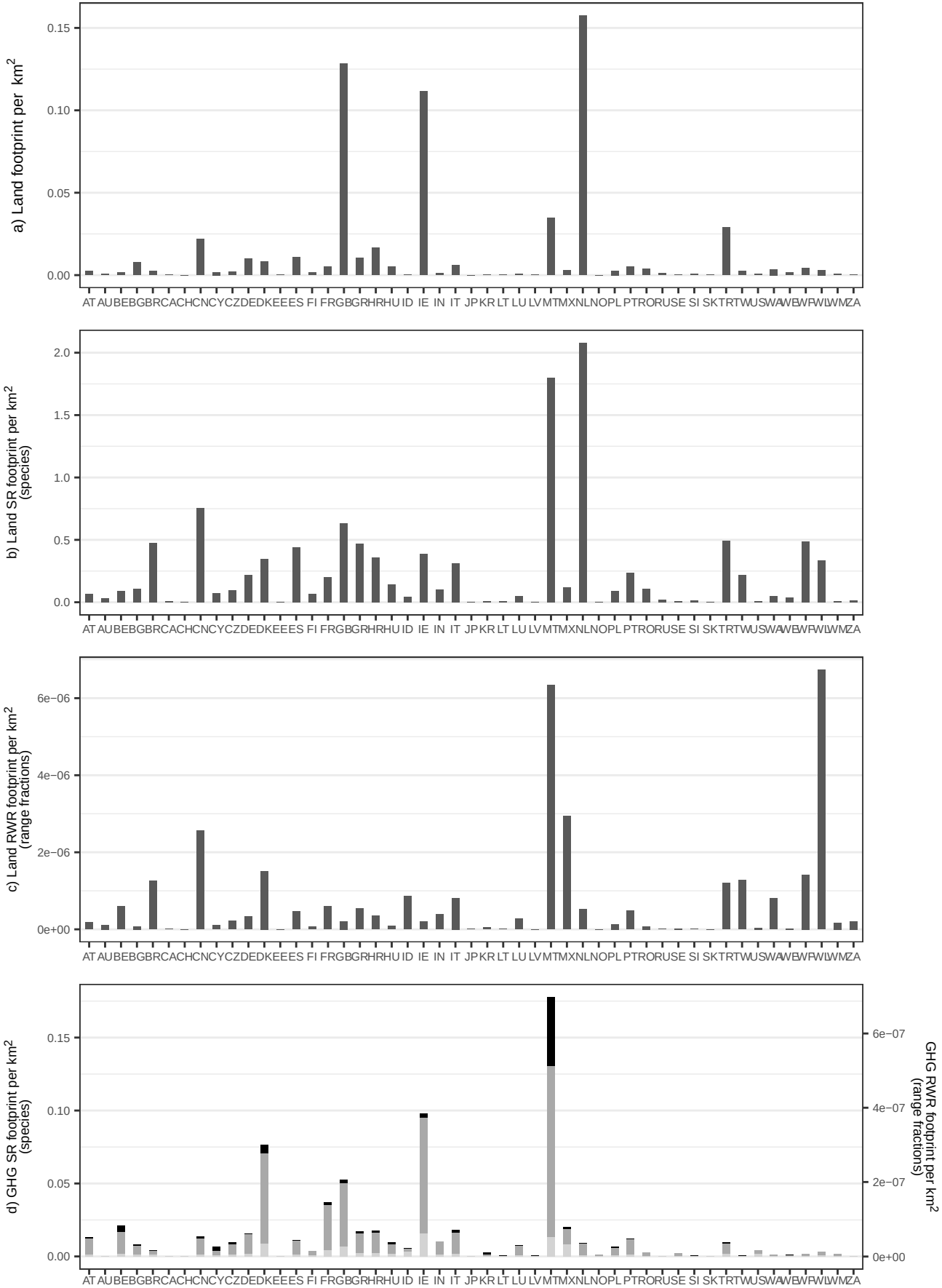
Supplementary Figure 7

Poultry: production-based footprints per square km



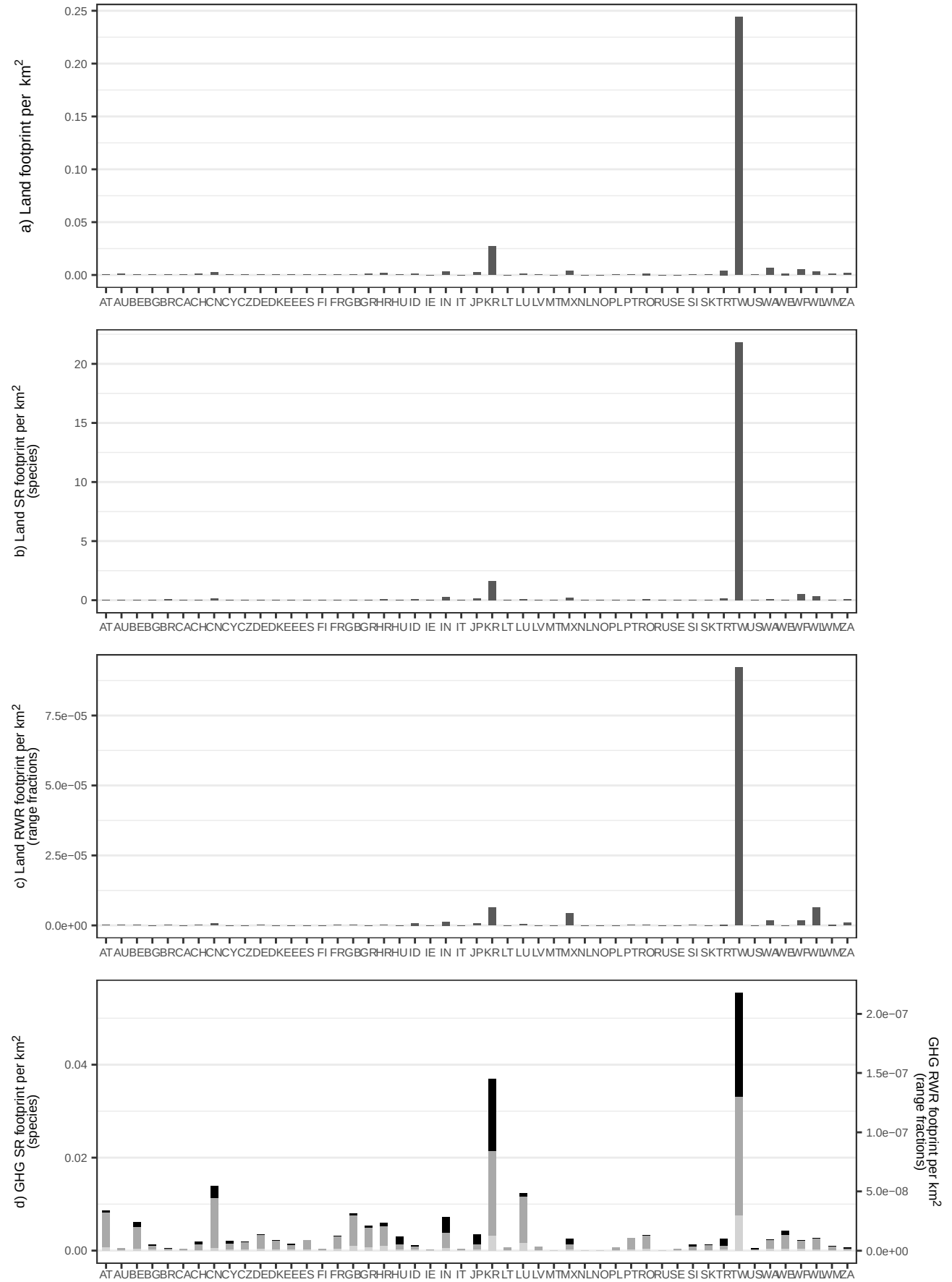
Supplementary Figure 7

Meat animals nec: production-based footprints per square km



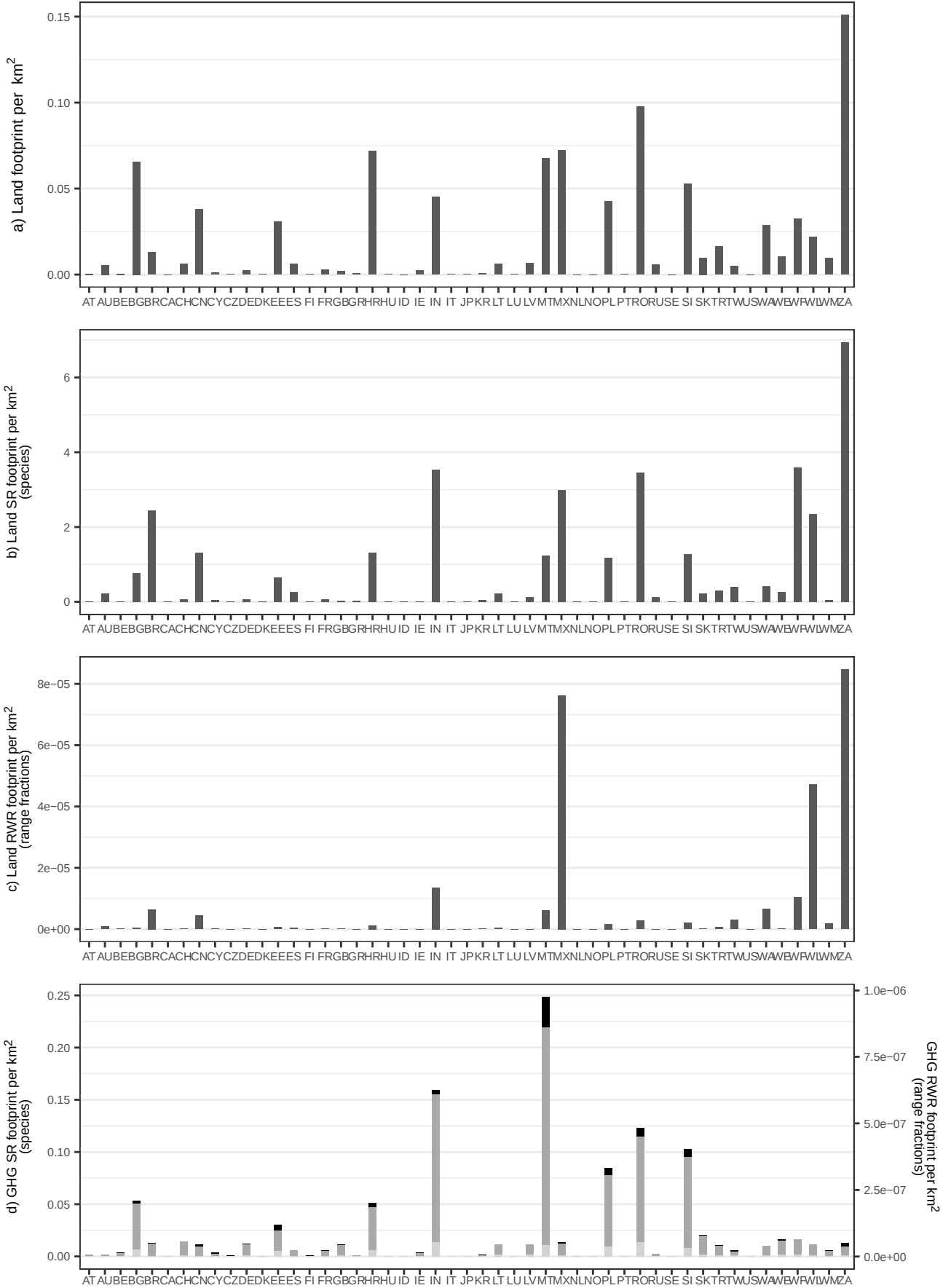
Supplementary Figure 7

Animal products nec: production-based footprints per square km



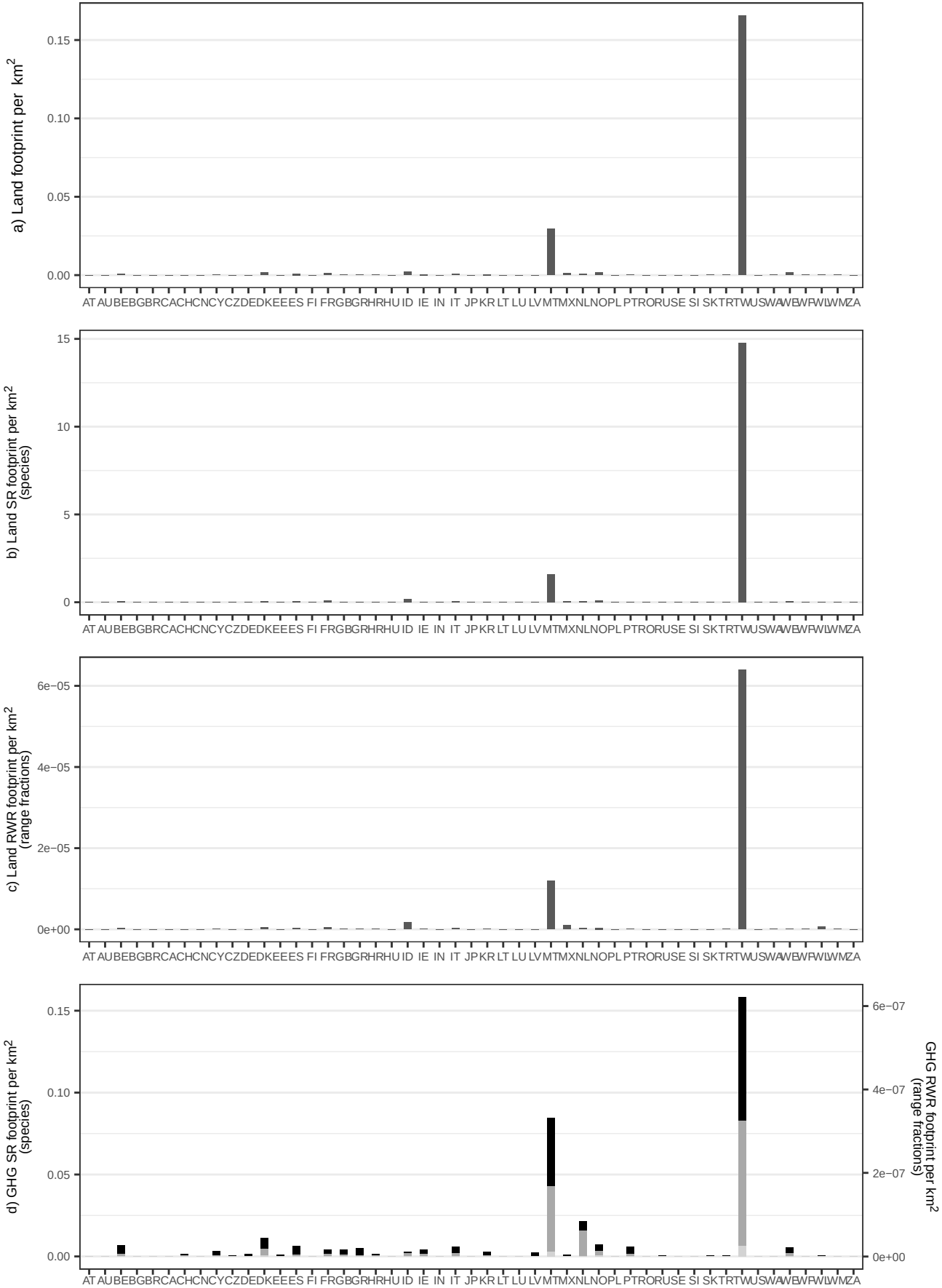
Supplementary Figure 7

Raw milk: production-based footprints per square km

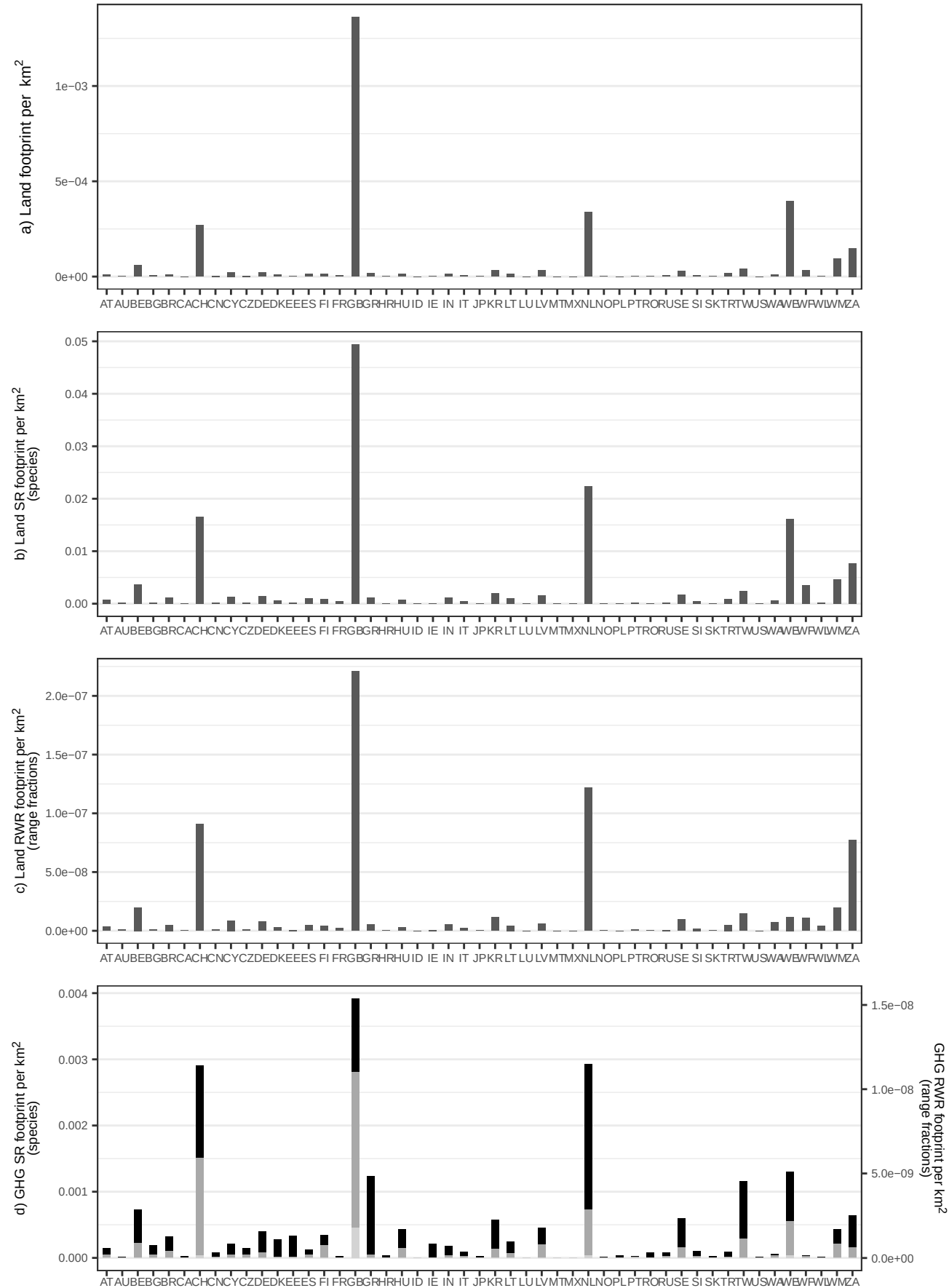


Supplementary Figure 7

Fish and other fishing products: production-based footprints per square km

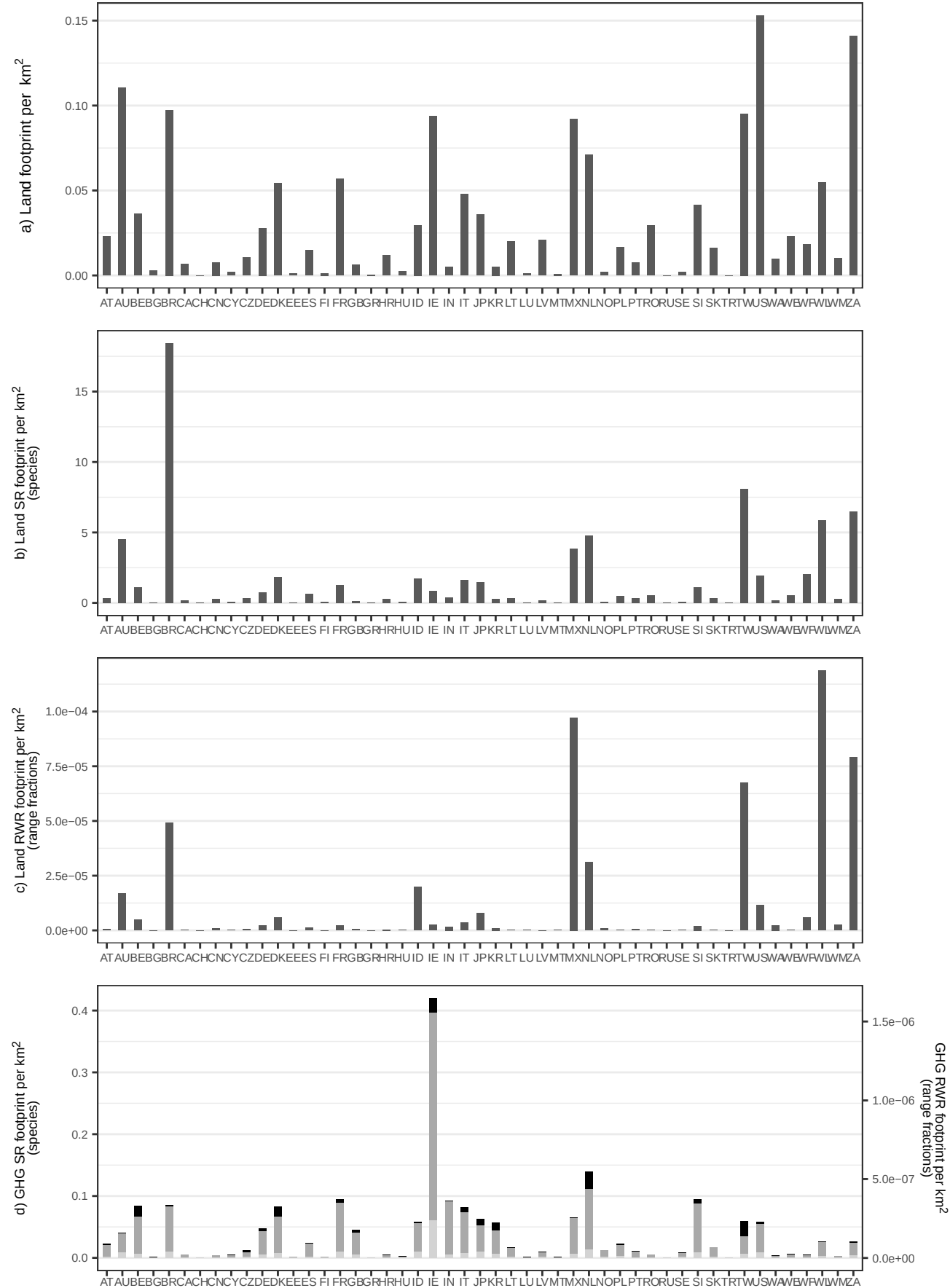


Supplementary Figure 7
Chemical and fertiliser minerals: production-based footprints per square km



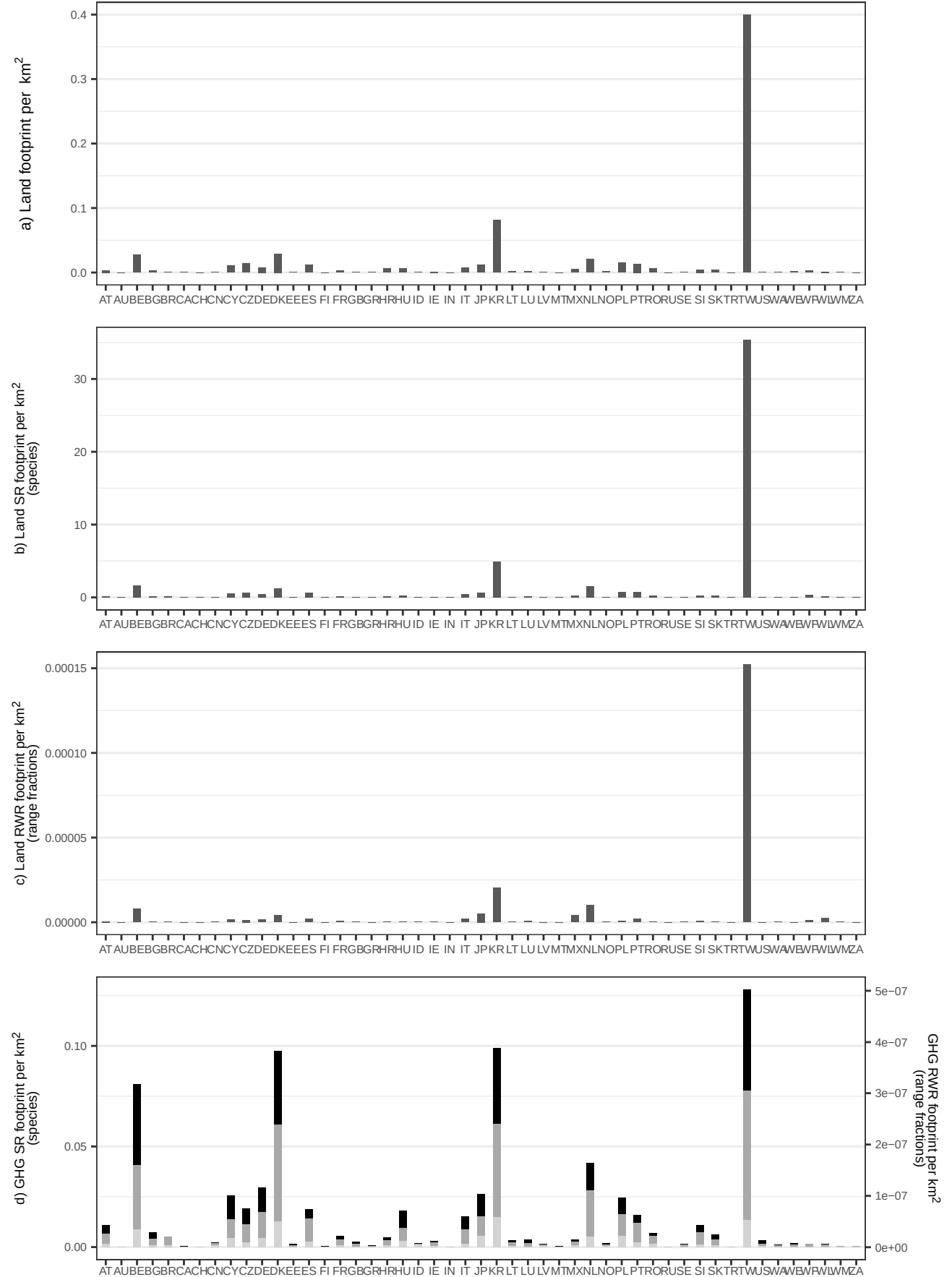
Supplementary Figure 7

Products of meat cattle: production-based footprints per square km



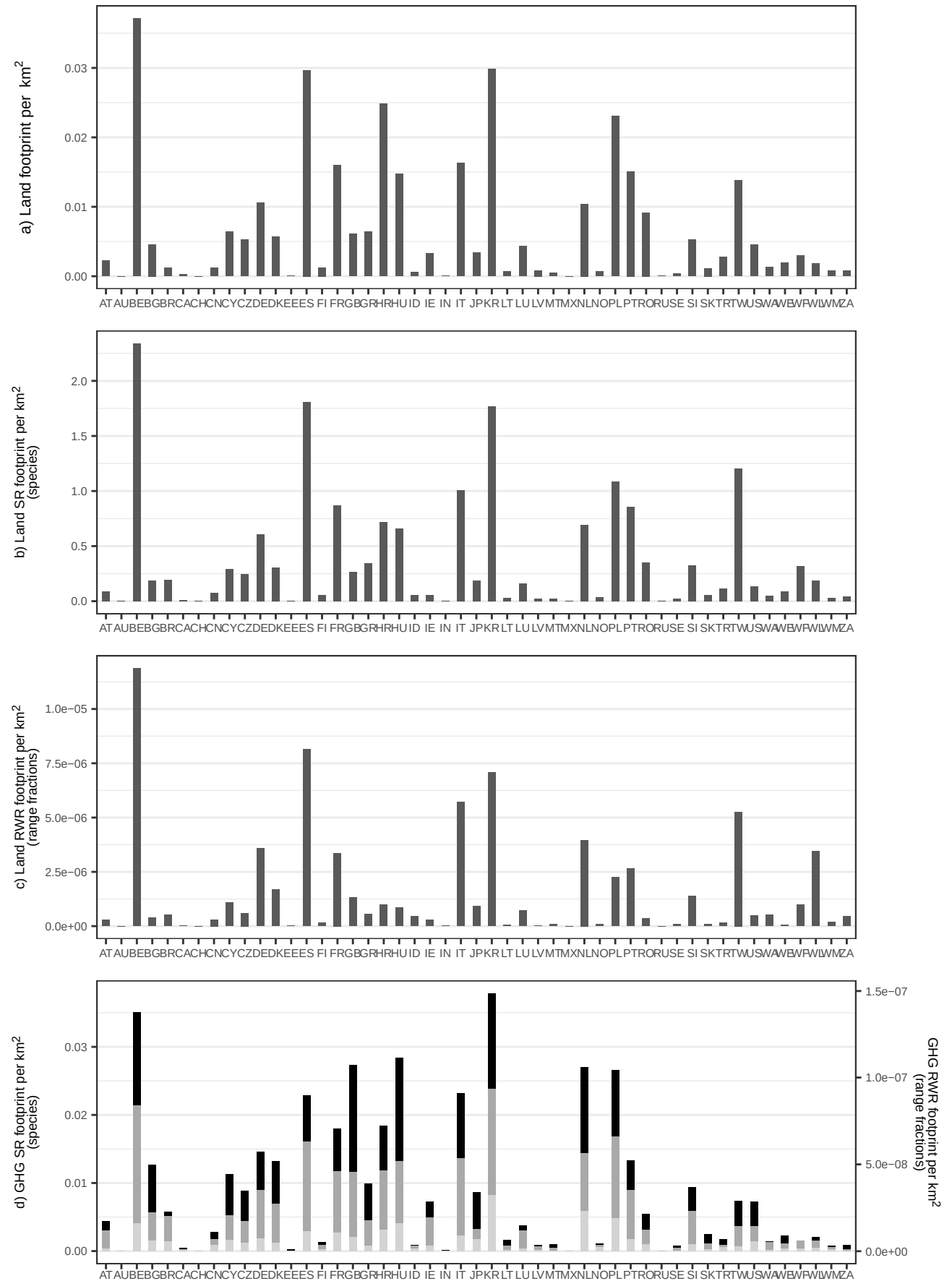
Supplementary Figure 7

Products of meat pigs: production-based footprints per square km



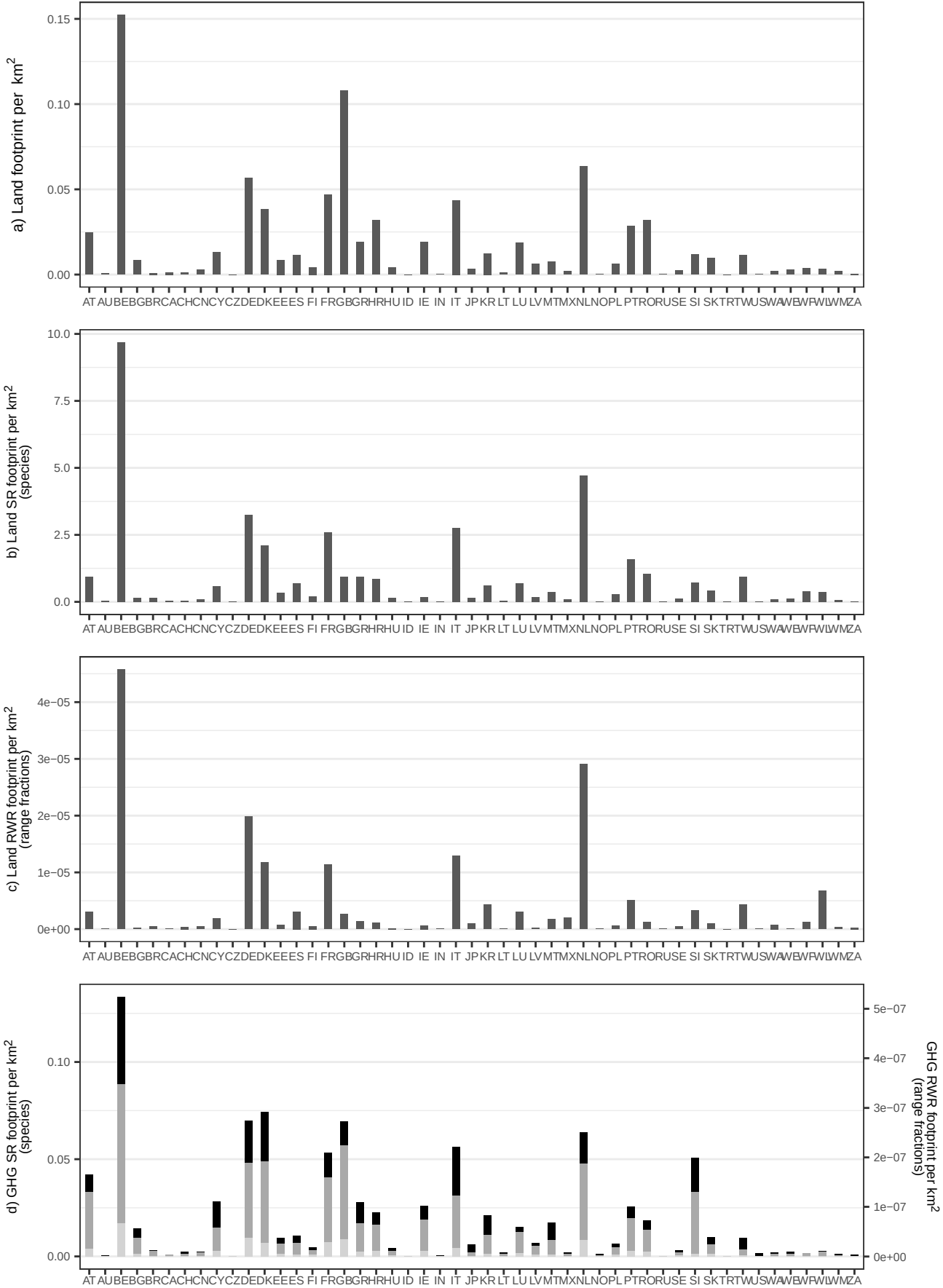
Supplementary Figure 7

Products of meat poultry: production-based footprints per square km



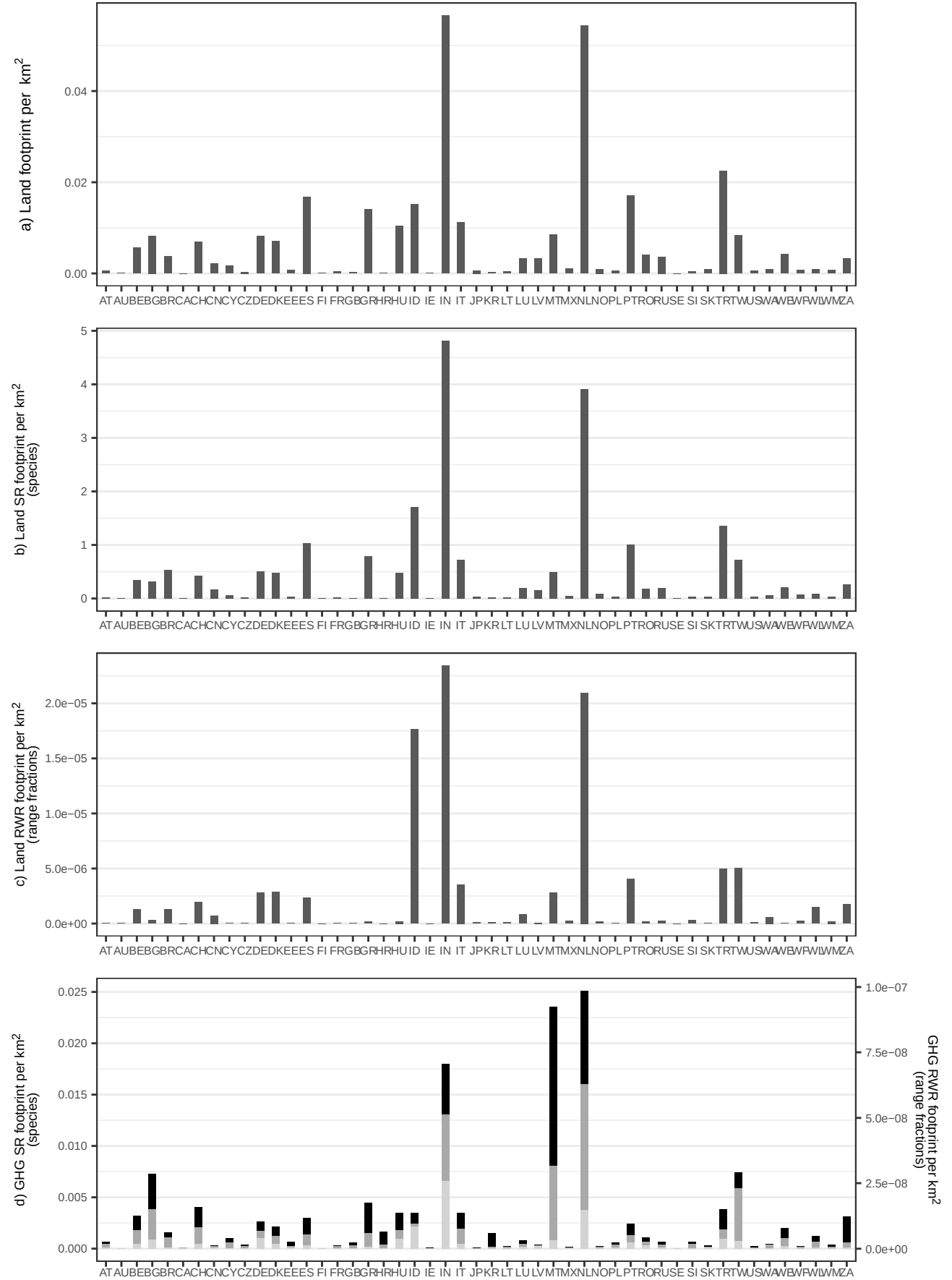
Supplementary Figure 7

Meat products nec: production-based footprints per square km



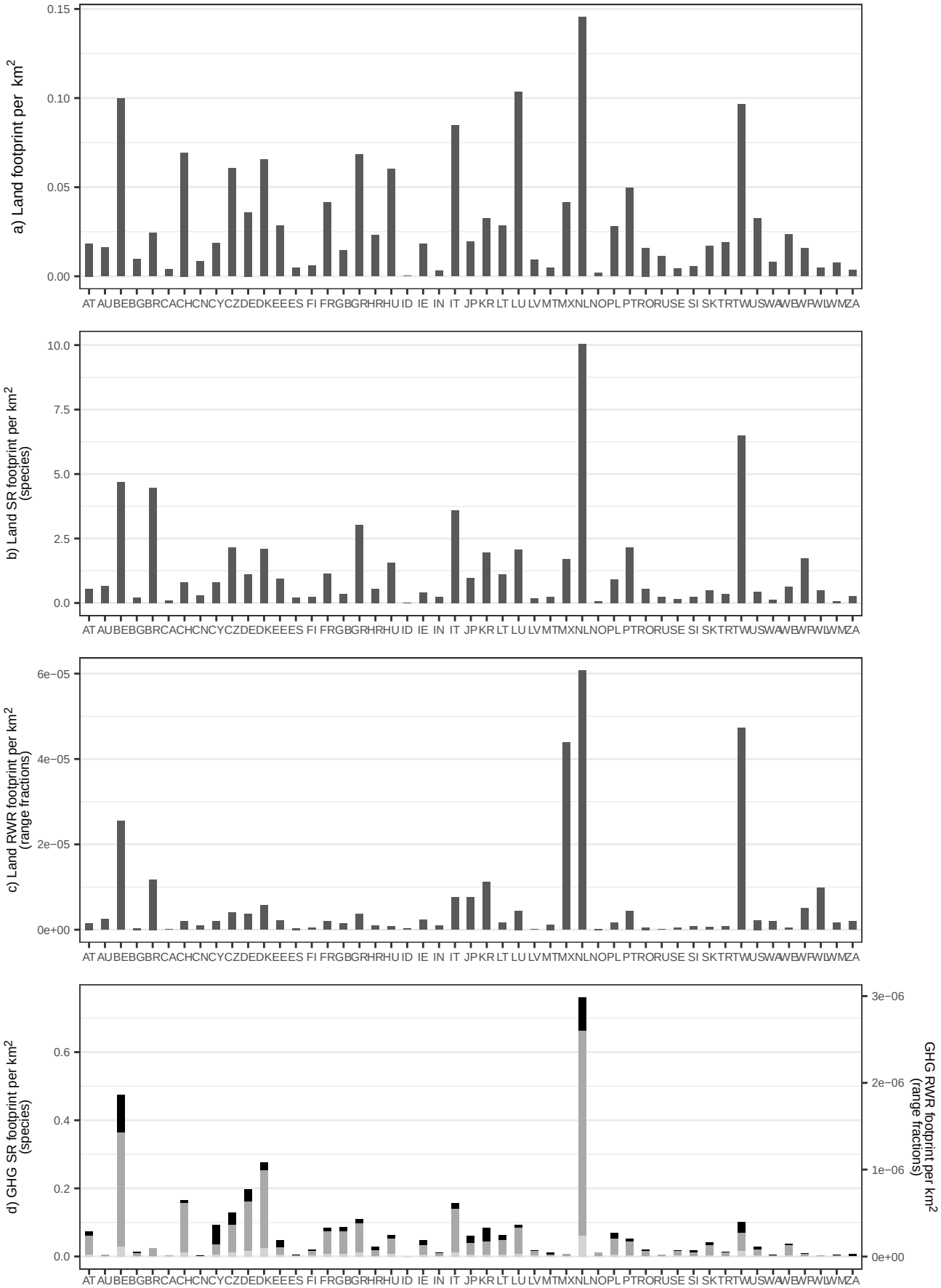
Supplementary Figure 7

Products of vegetable oils and fats: production-based footprints per square km



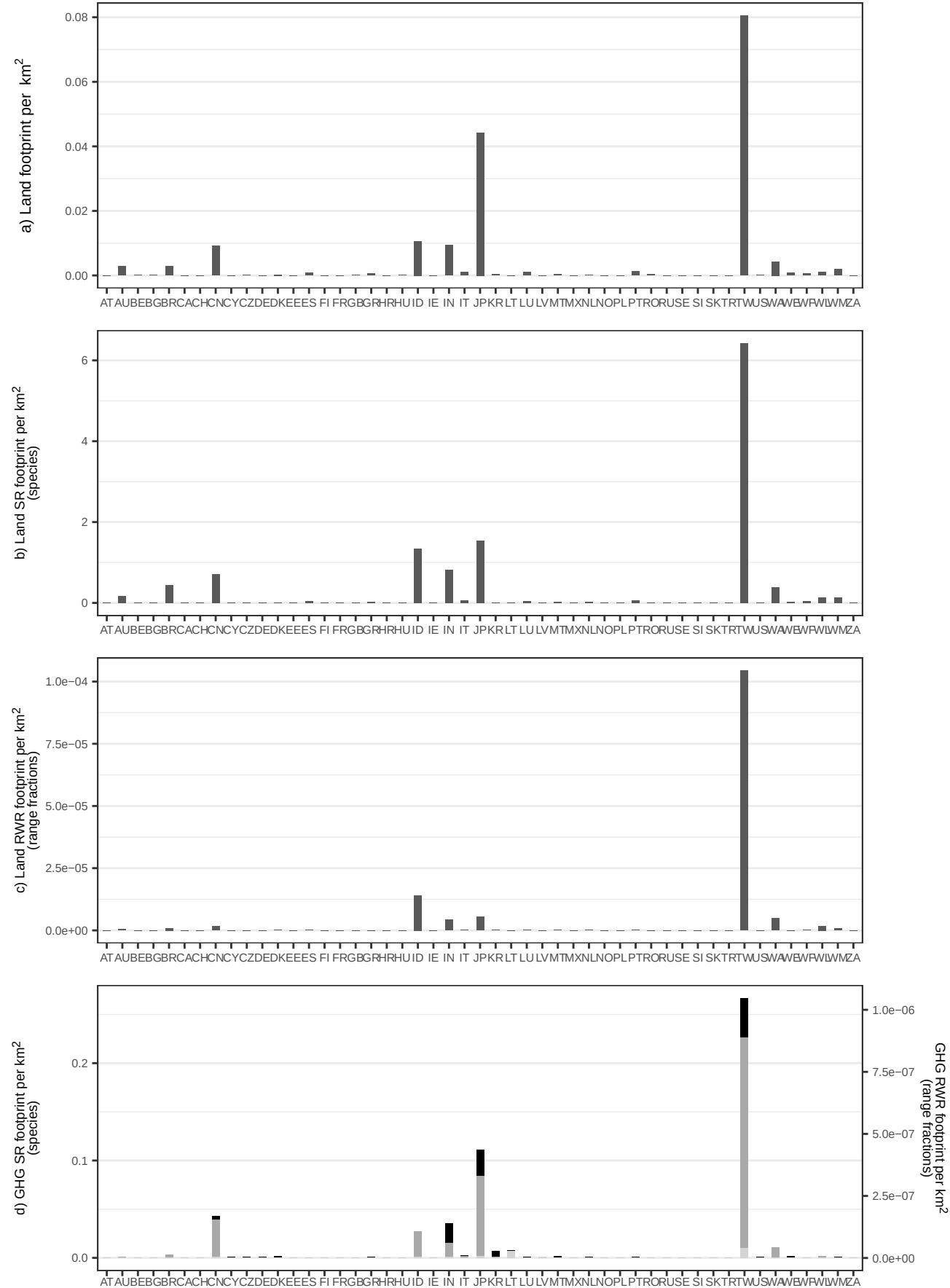
Supplementary Figure 7

Dairy products: production-based footprints per square km



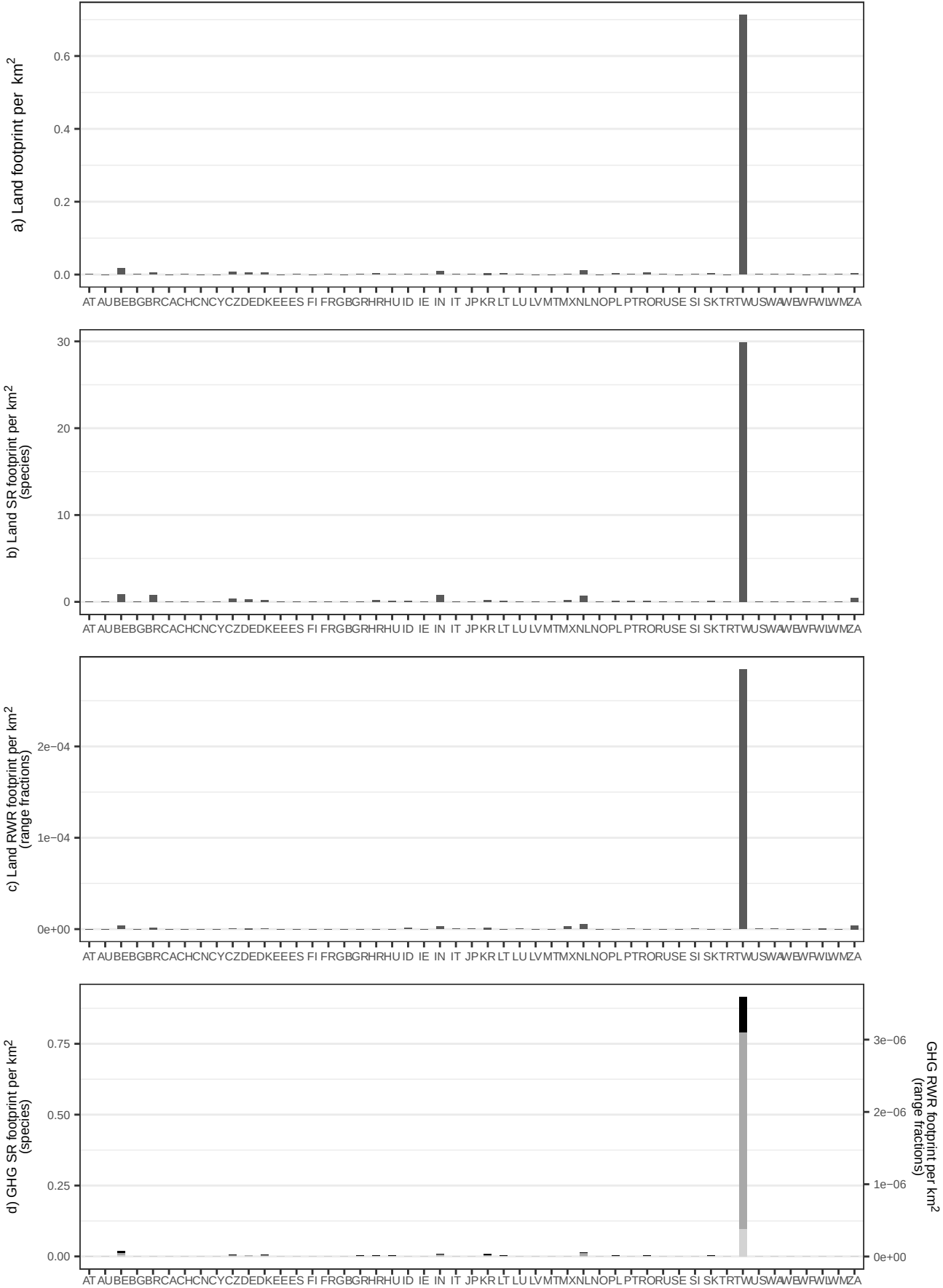
Supplementary Figure 7

Processed rice: production-based footprints per square km



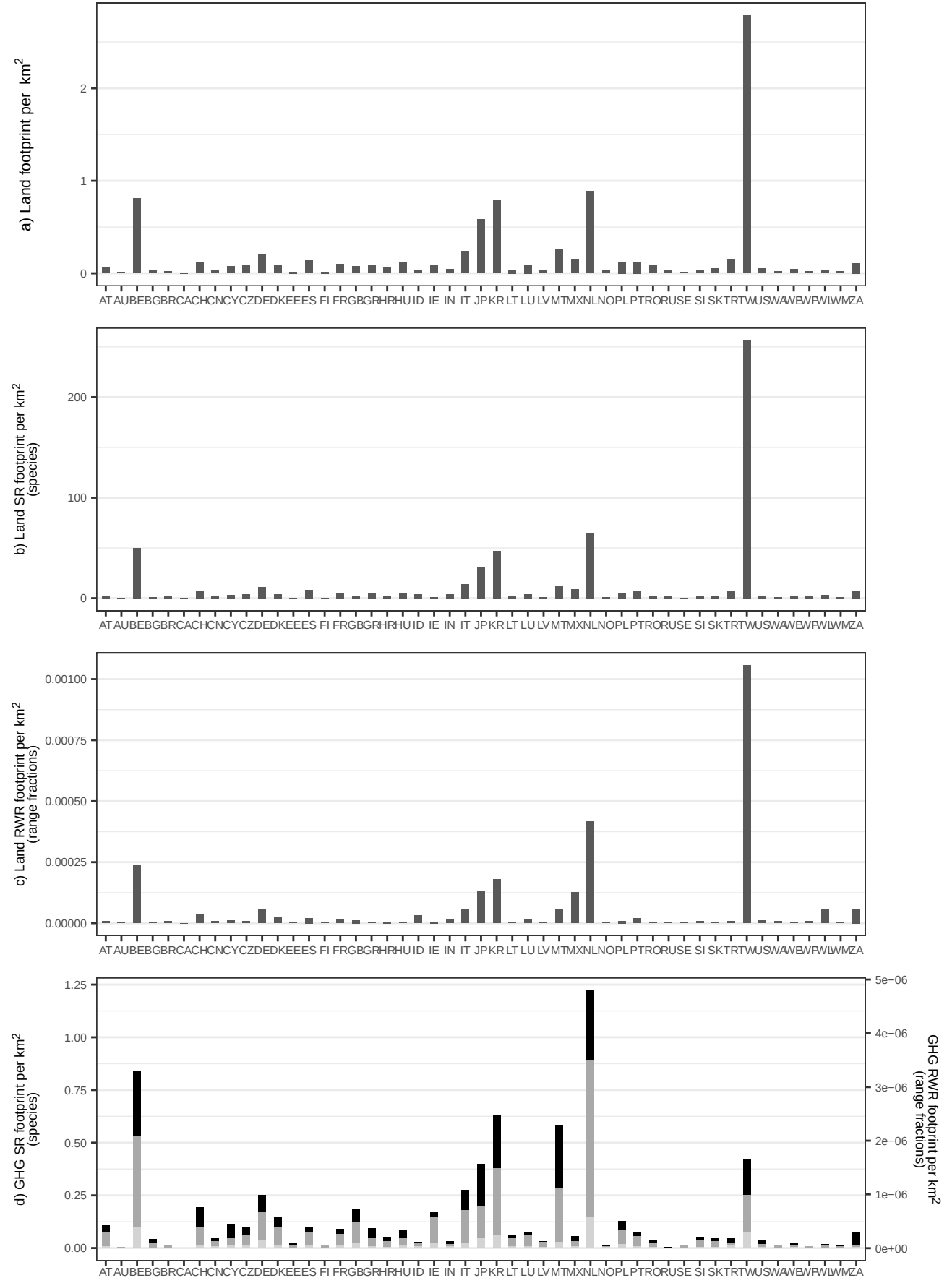
Supplementary Figure 7

Processed sugar: production-based footprints per square km

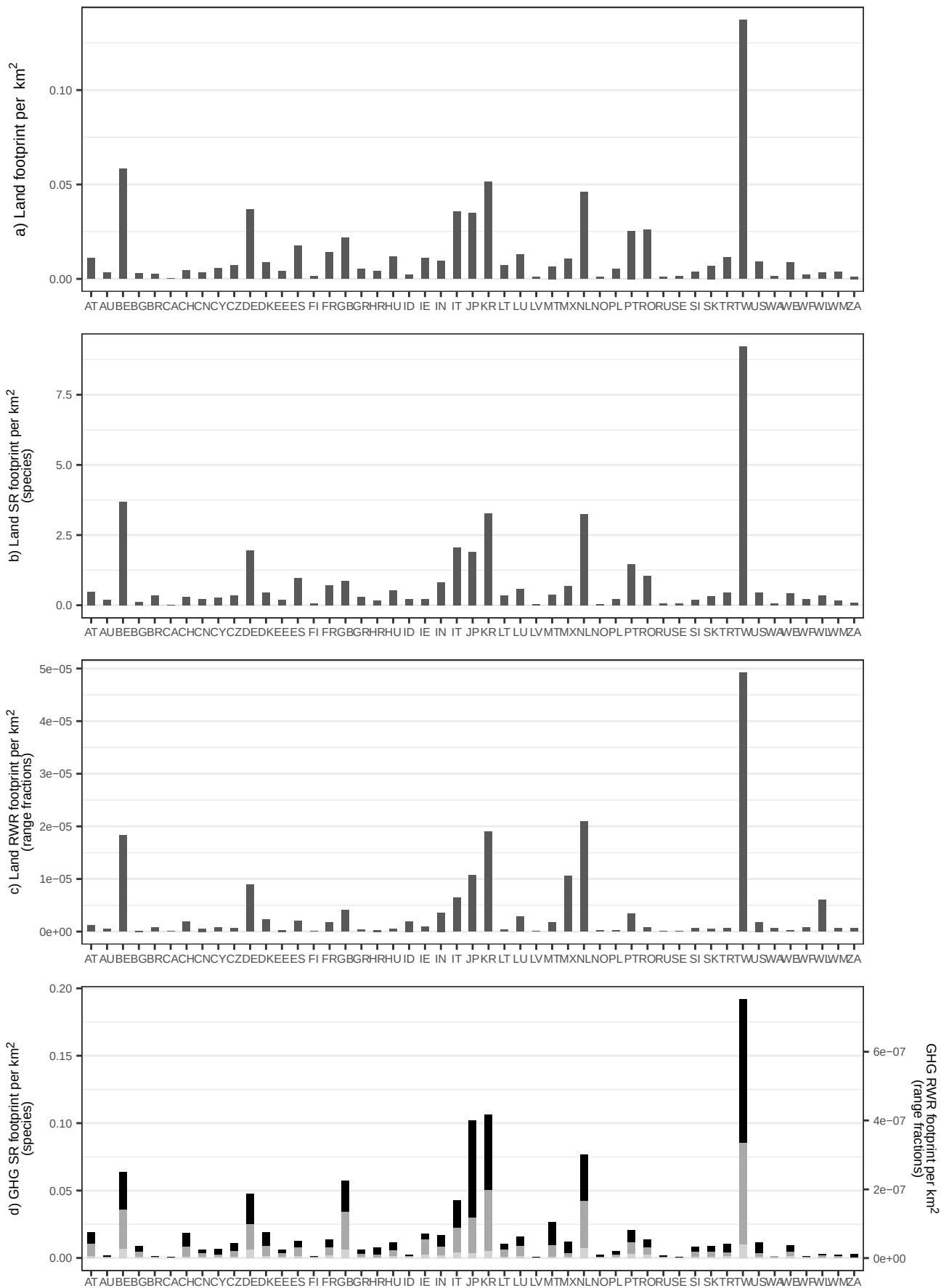


Supplementary Figure 7

Food products nec: production-based footprints per square km

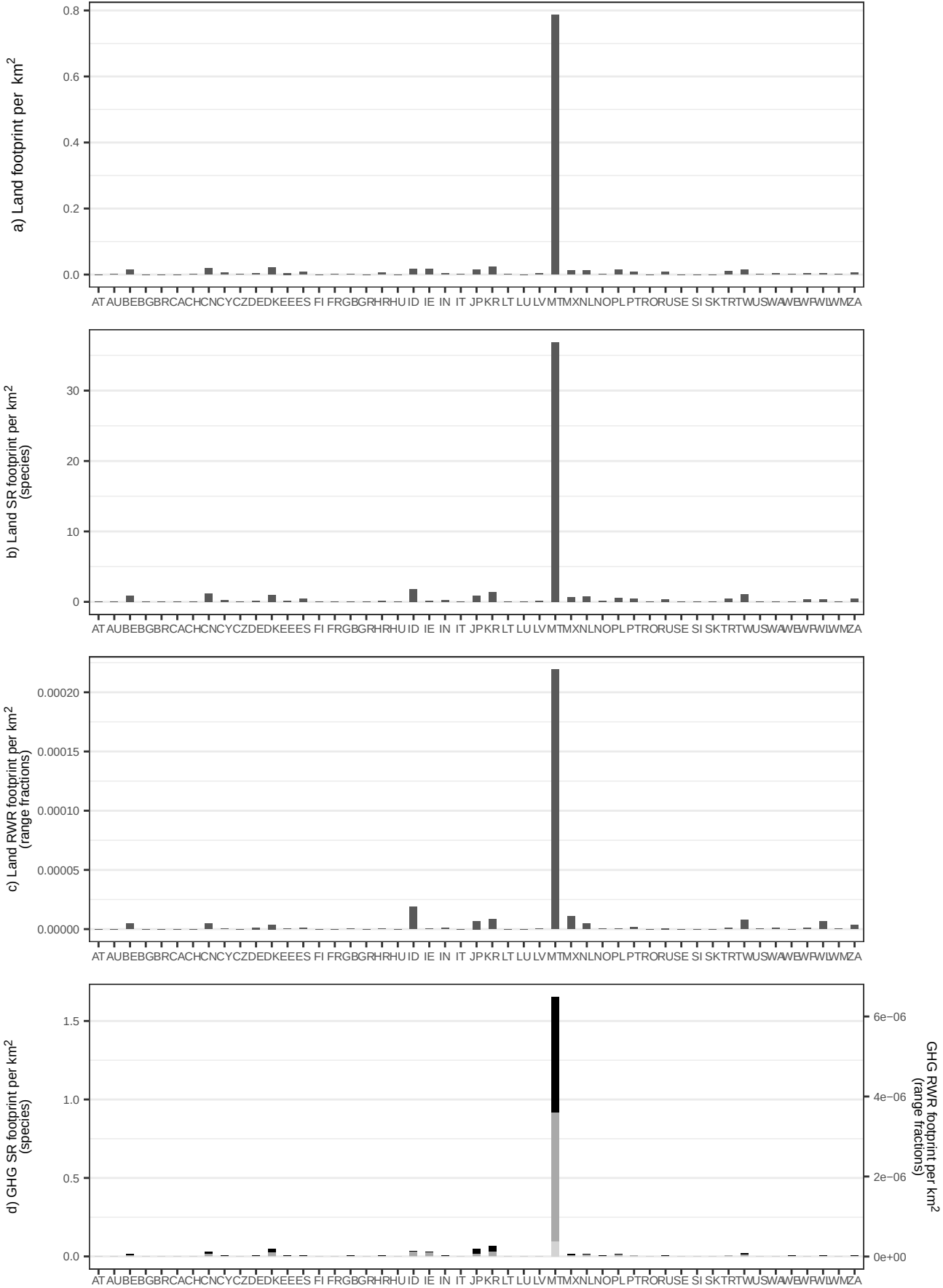


Beverages: production-based footprints per square km



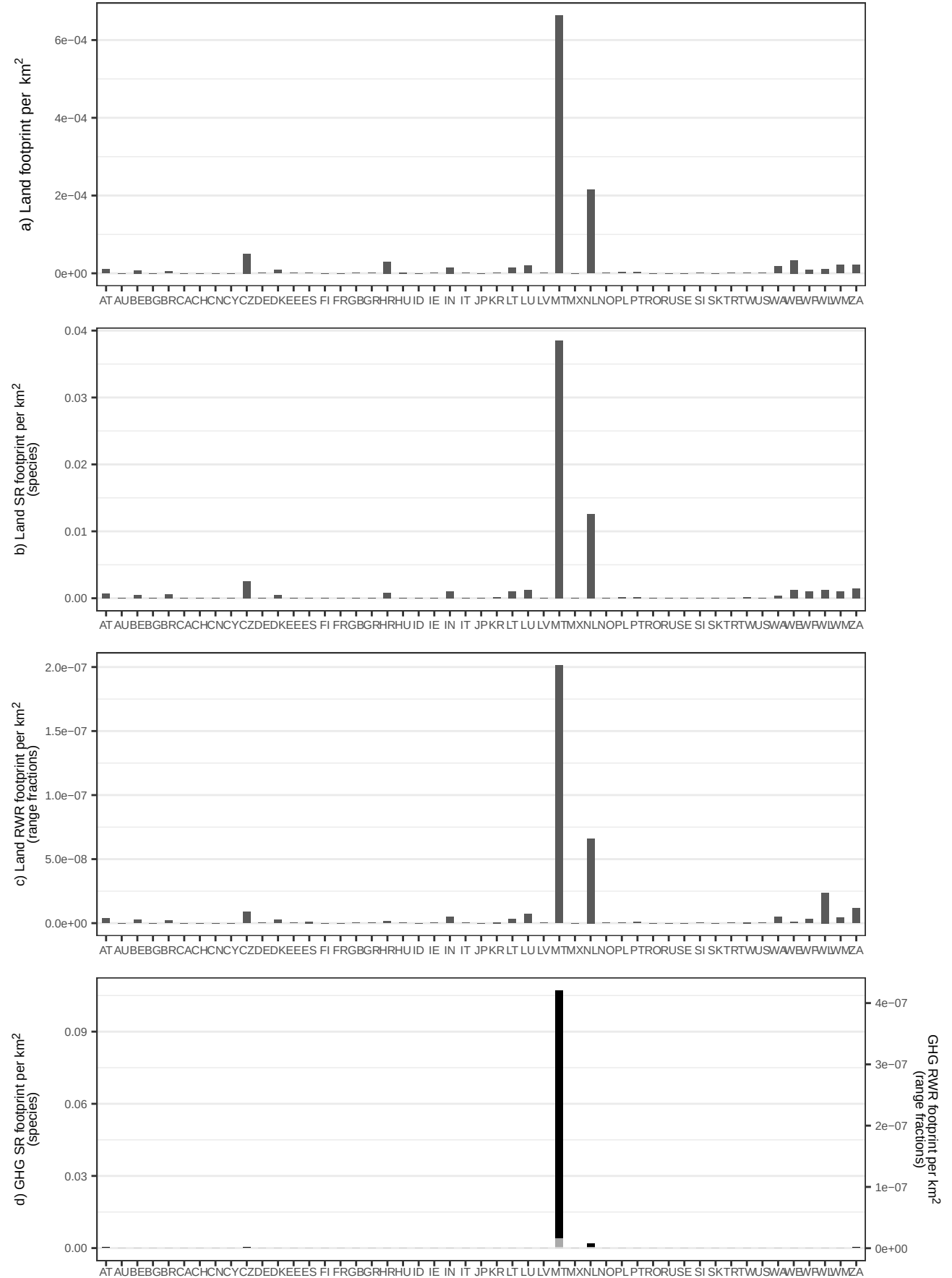
Supplementary Figure 7

Fish products: production-based footprints per square km



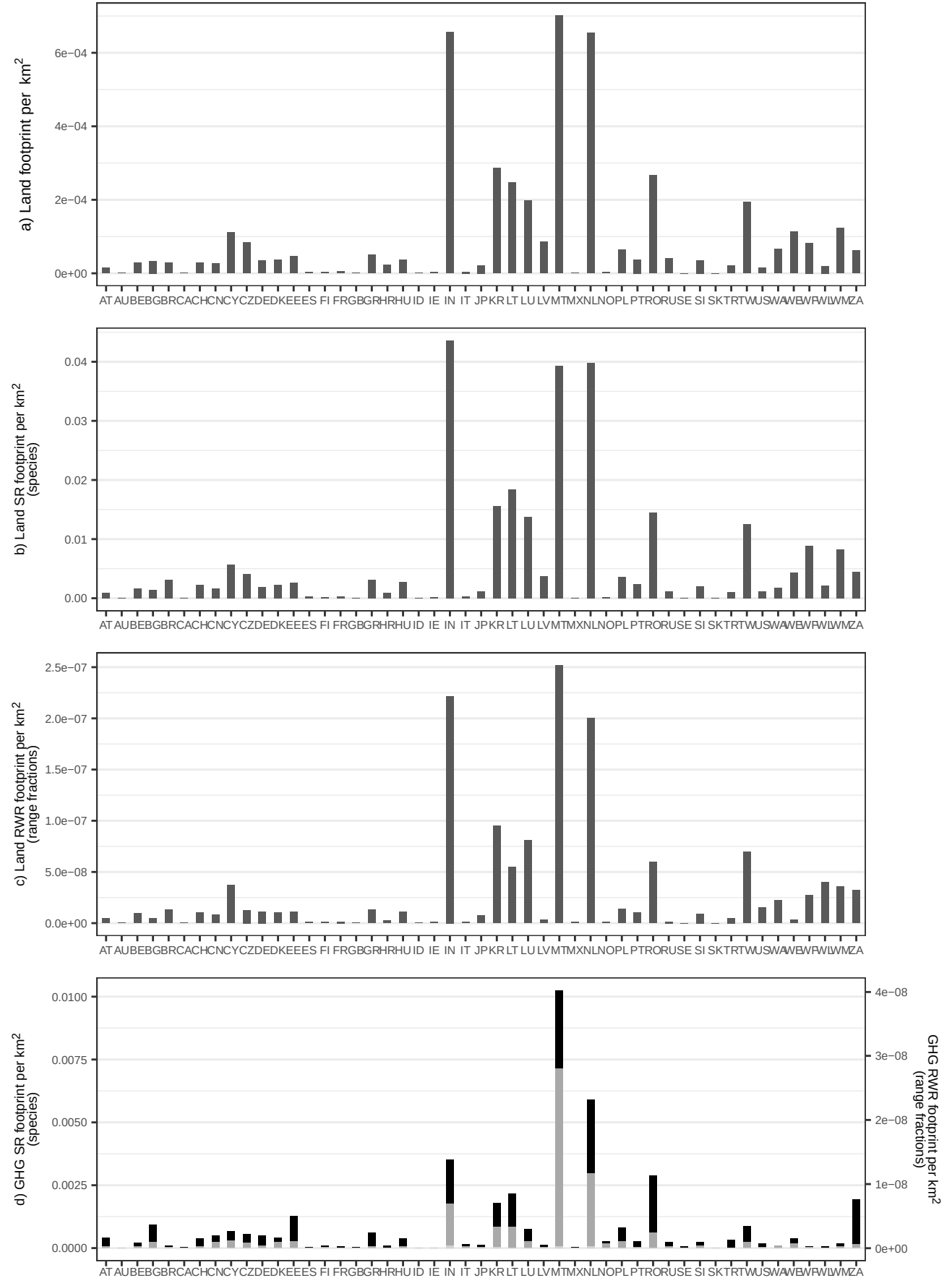
Supplementary Figure 7

N-fertiliser: production-based footprints per square km



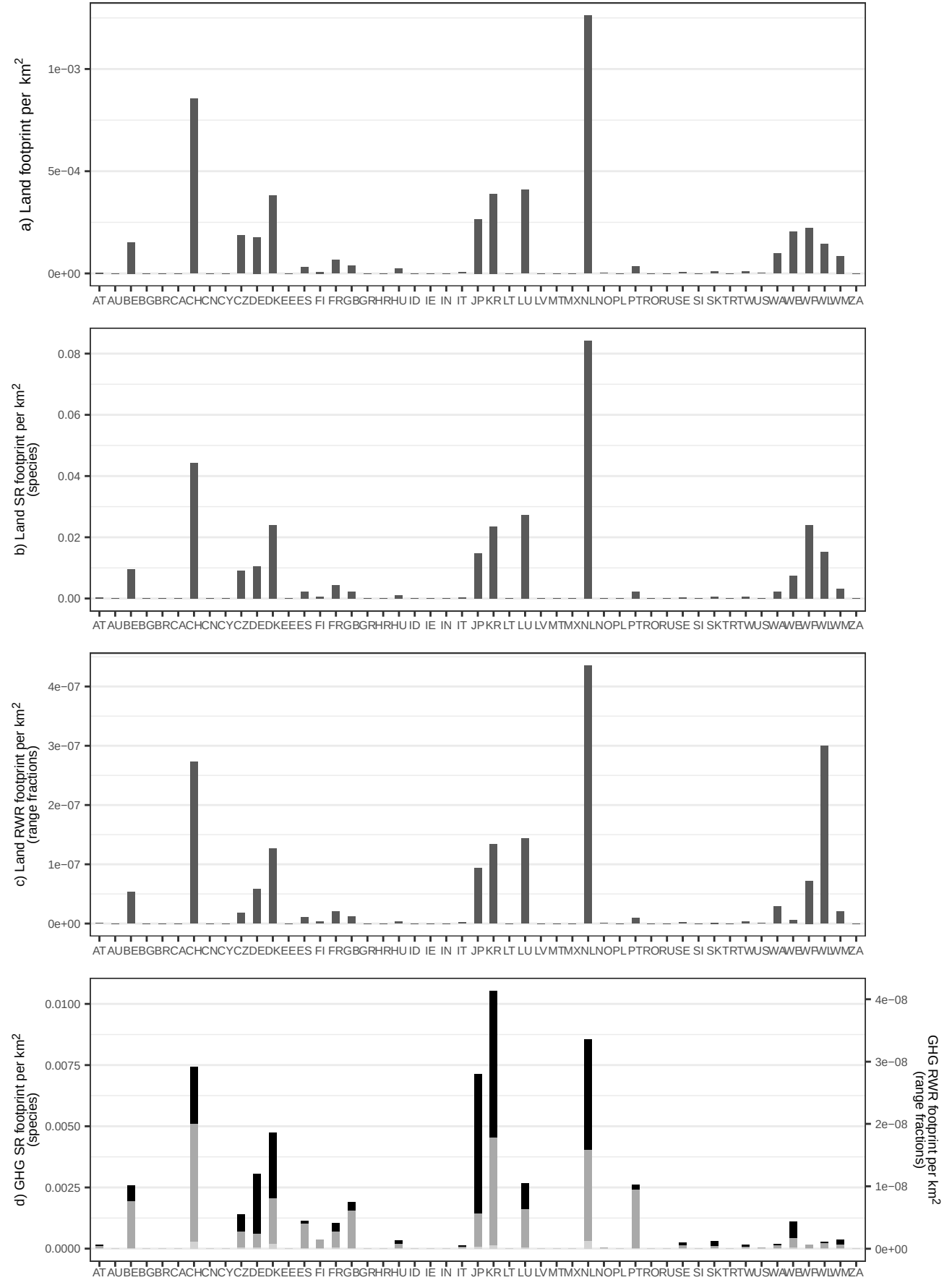
Supplementary Figure 7

P- and other fertiliser: production-based footprints per square km



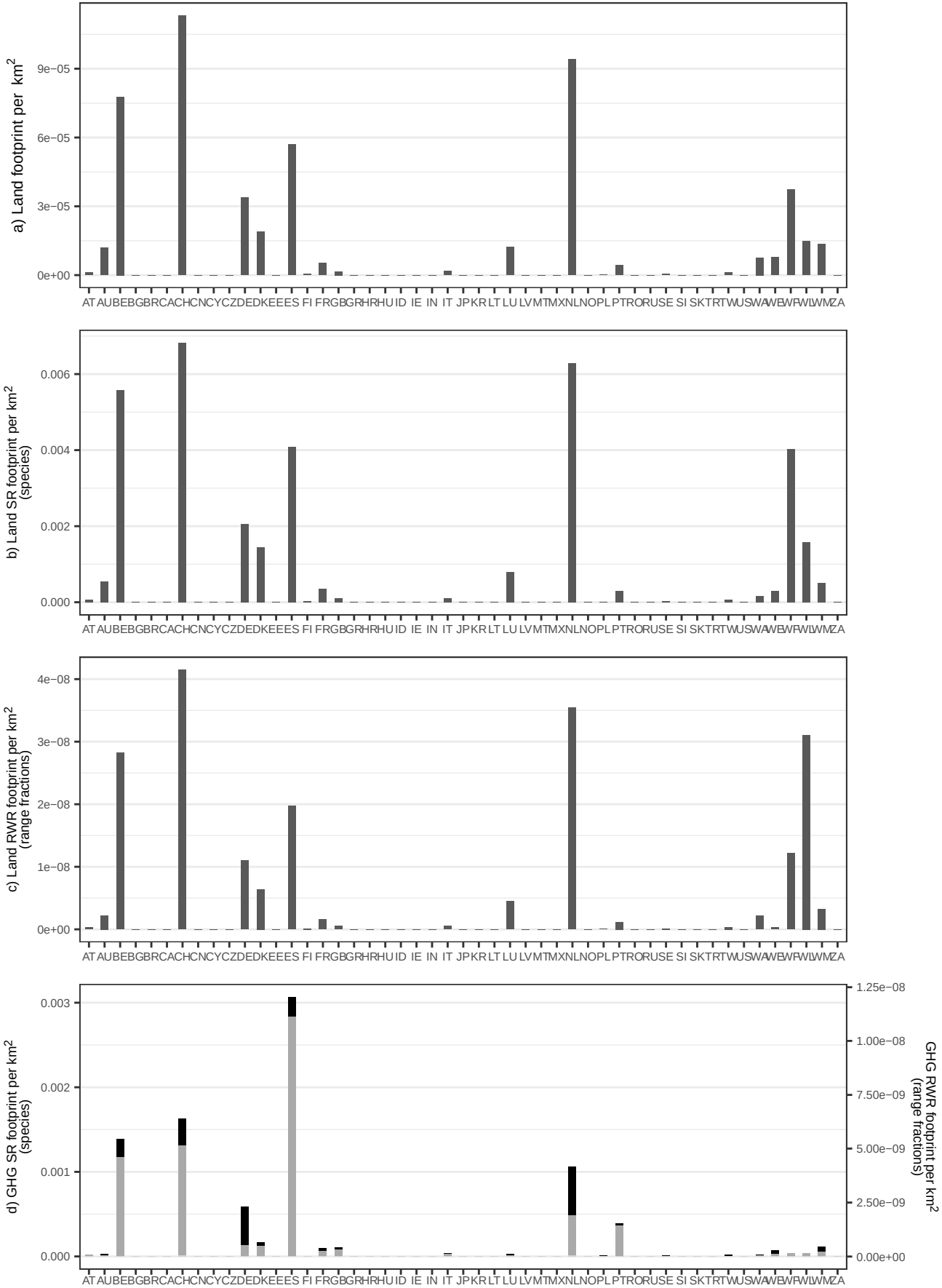
Supplementary Figure 7

Food waste: incineration: production-based footprints per square km



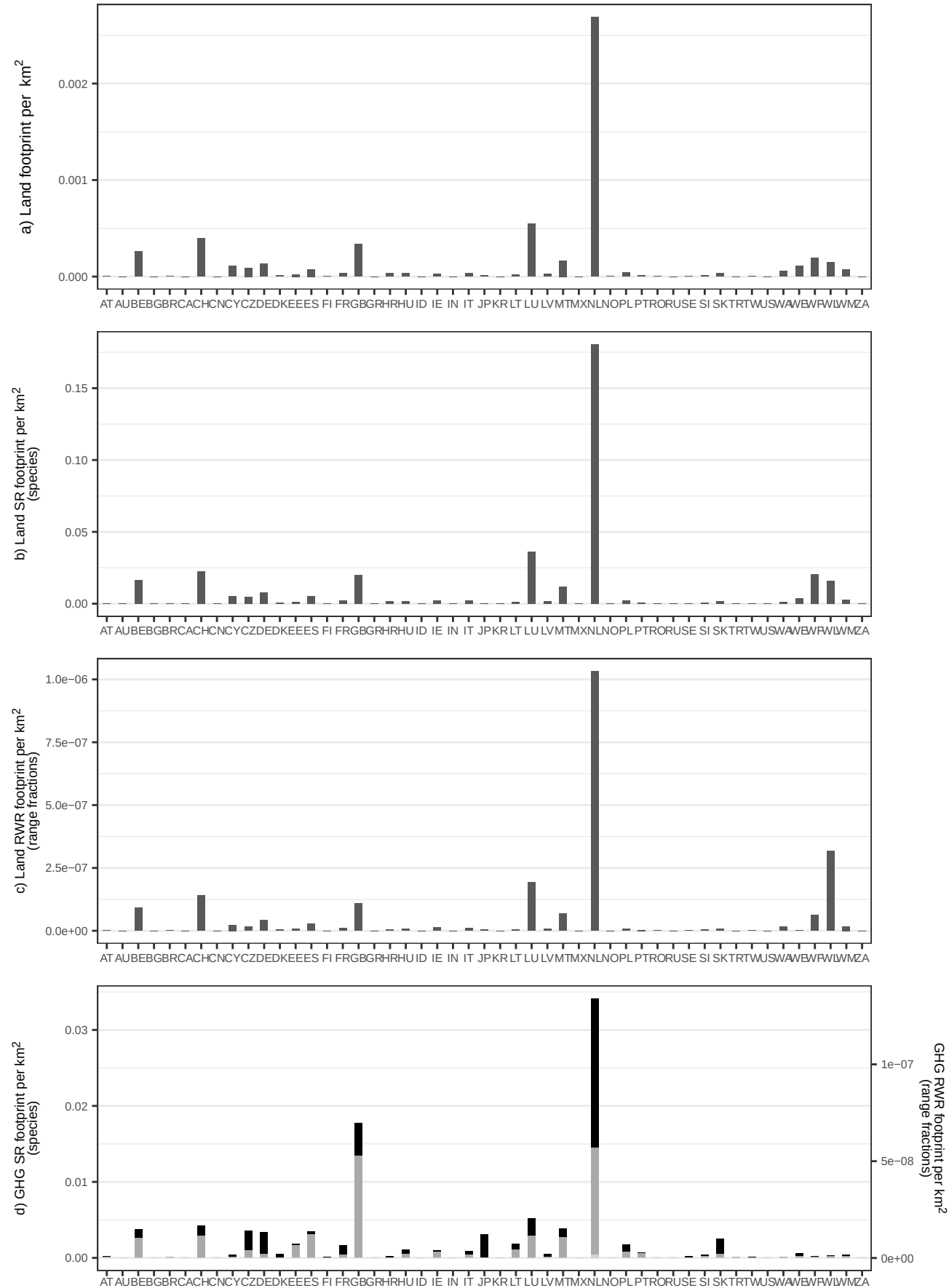
Supplementary Figure 7

Food waste: biogasification and land application: production-based footprints per square km



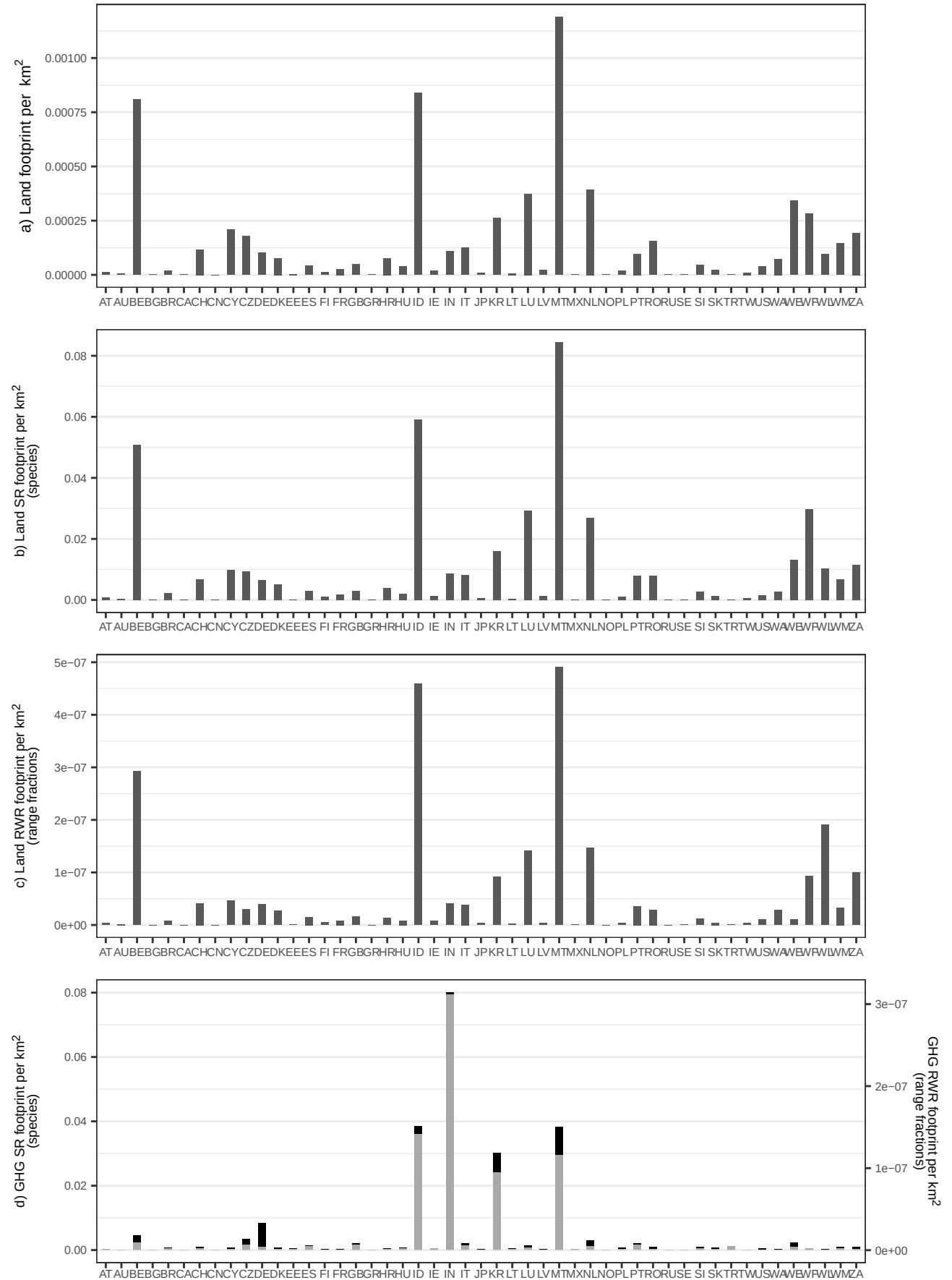
Supplementary Figure 7

Food waste: composting and land application: production-based footprints per square km



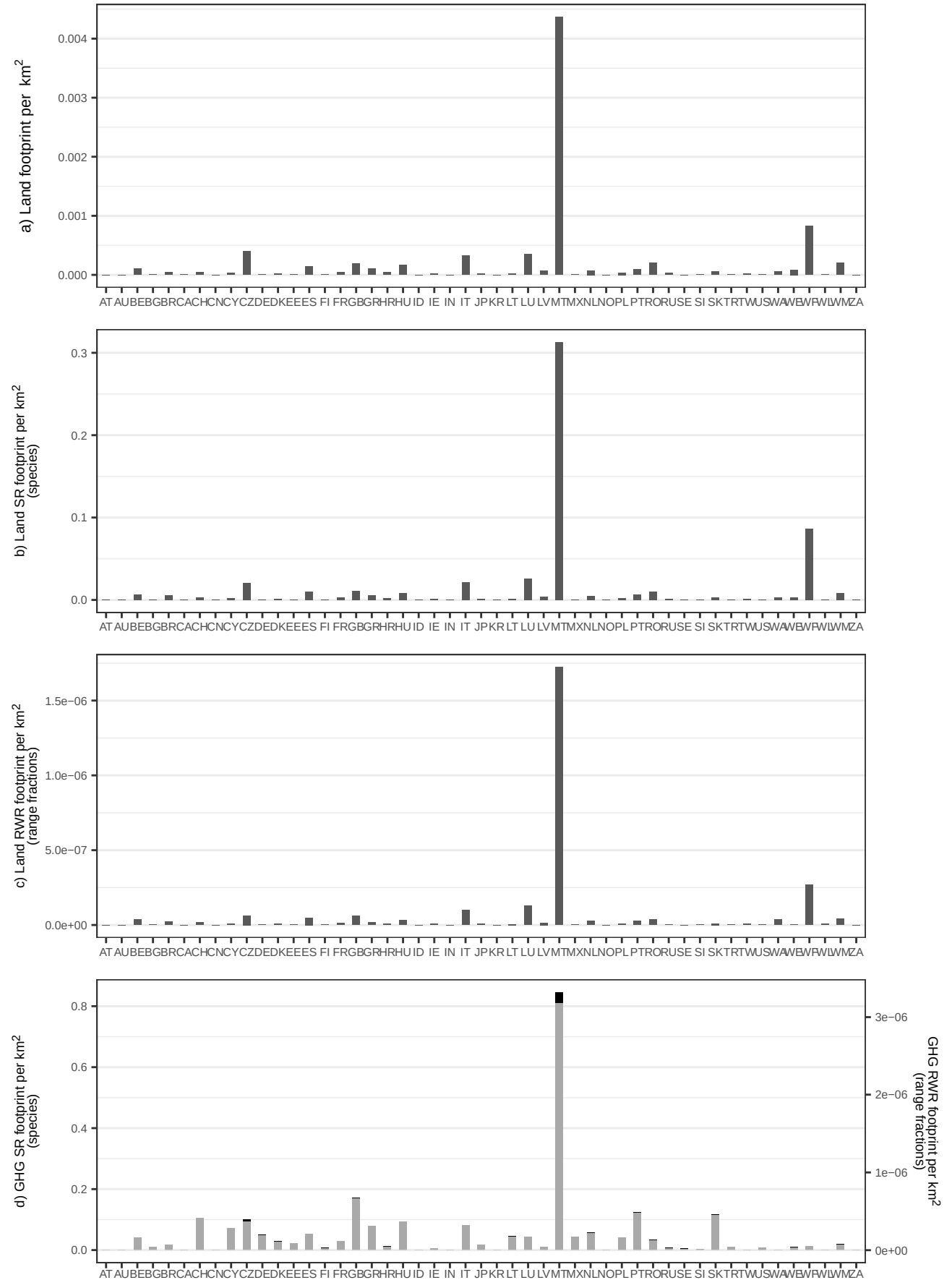
Supplementary Figure 7

Food waste: waste water treatment: production-based footprints per square km



Supplementary Figure 7

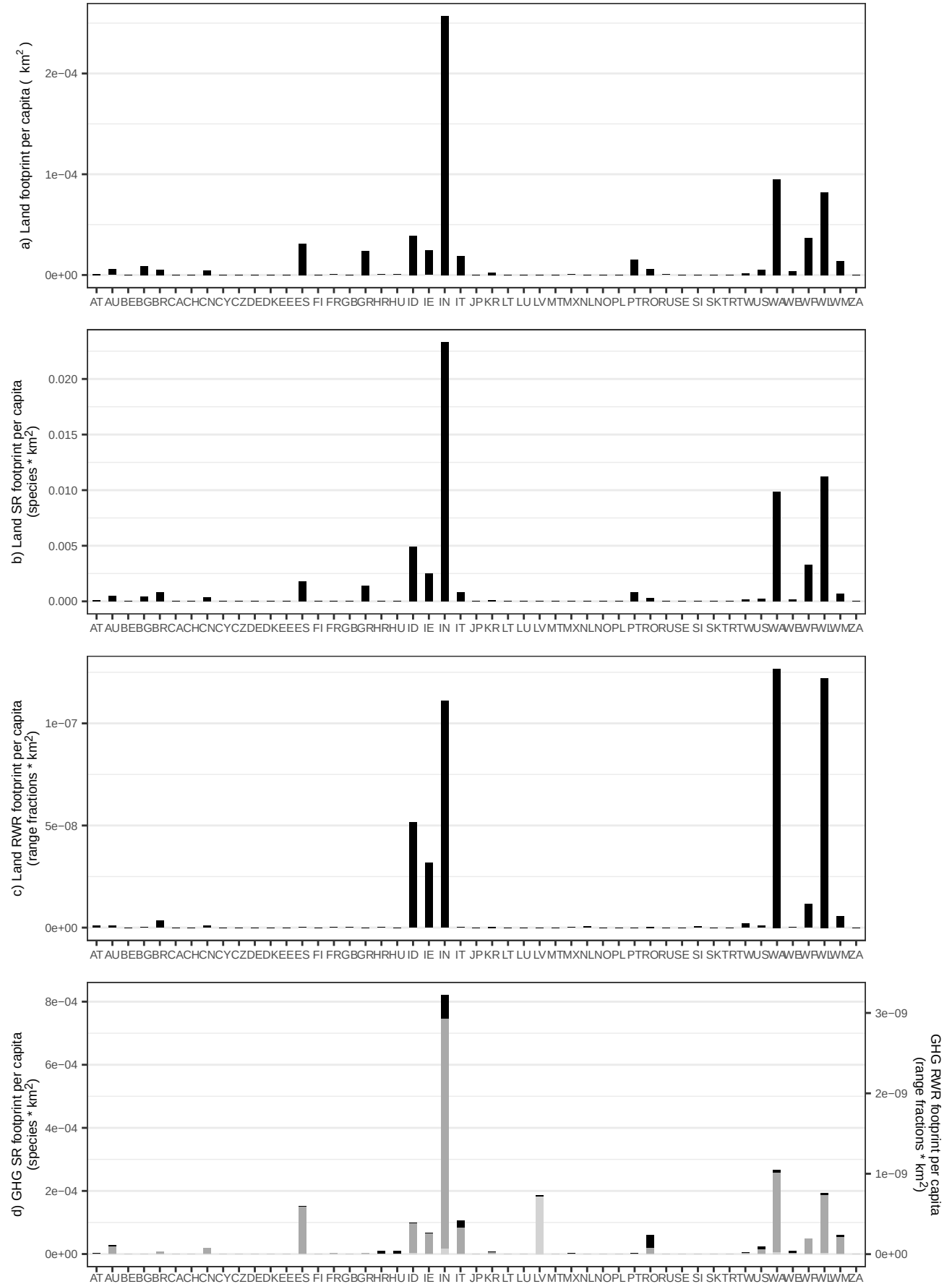
Food waste: landfill: production-based footprints per square km



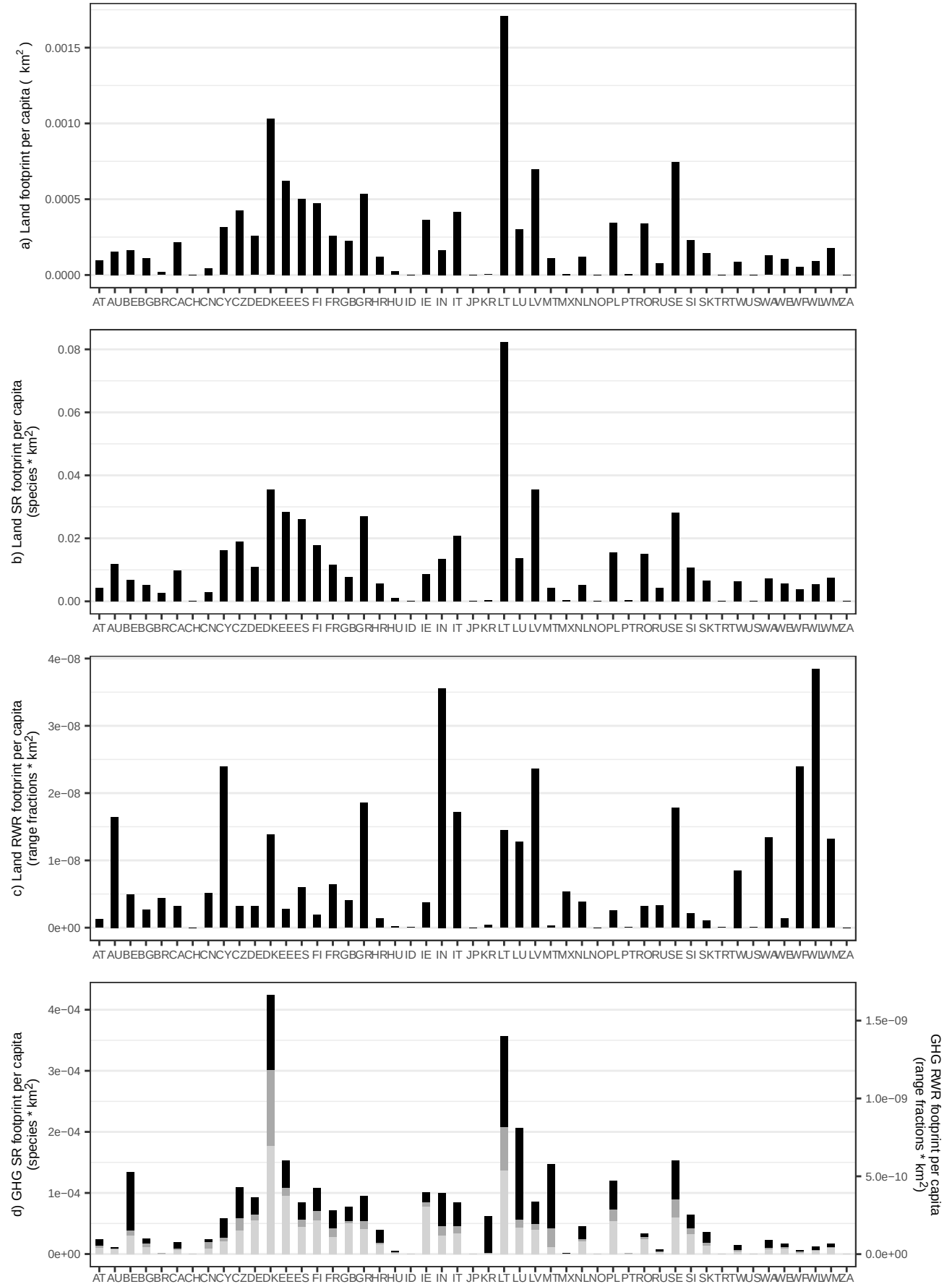
Supplementary Figure 7. Regions' production per km² footprints of food-related products. Individual products' production footprints per km² for each EXIOBASE region for a) land use, b) land-driven species richness loss, c) land-driven rarity-weighted richness loss and d) GHG-driven biodiversity loss split by emissions type: carbon dioxide (black), methane (dark grey), nitrous oxide (light grey) (right-hand axis – species richness; left-hand axis – rarity-weighted richness). Results are for the year 2011. AT = Austria, AU = Australia, BE = Belgium, BG = Bulgaria, BR = Brazil, CA = Canada, CH = Switzerland, CN = China, CY = Cyprus, CZ = Czech Republic, DE = Germany, DK = Denmark, EE = Estonia, ES = Spain, FI = Finland, GB = United Kingdom, HR = Croatia, HU = Hungary, ID = Indonesia, IE = Ireland, IN = India, IT = Italy, JP = Japan, KR = South Korea, LT = Lithuania, LV = Latvia, MT = Malta, MX = Mexico, NL = Netherlands, NO = Norway, PL = Poland, PT = Portugal, RO = Romania, RU = Russia, SE = Sweden, SI = Slovenia, SK = Slovakia, TR = Turkey, TW = Taiwan, US = United States, WA = RoW Asia & Pacific, WE = RoW Europe, WF = RoW Africa, WL = RoW CS America, WM = RoW Middle East, ZA = South Africa. Source data are provided in Supplementary Data 9.

Supplementary Figure 8.

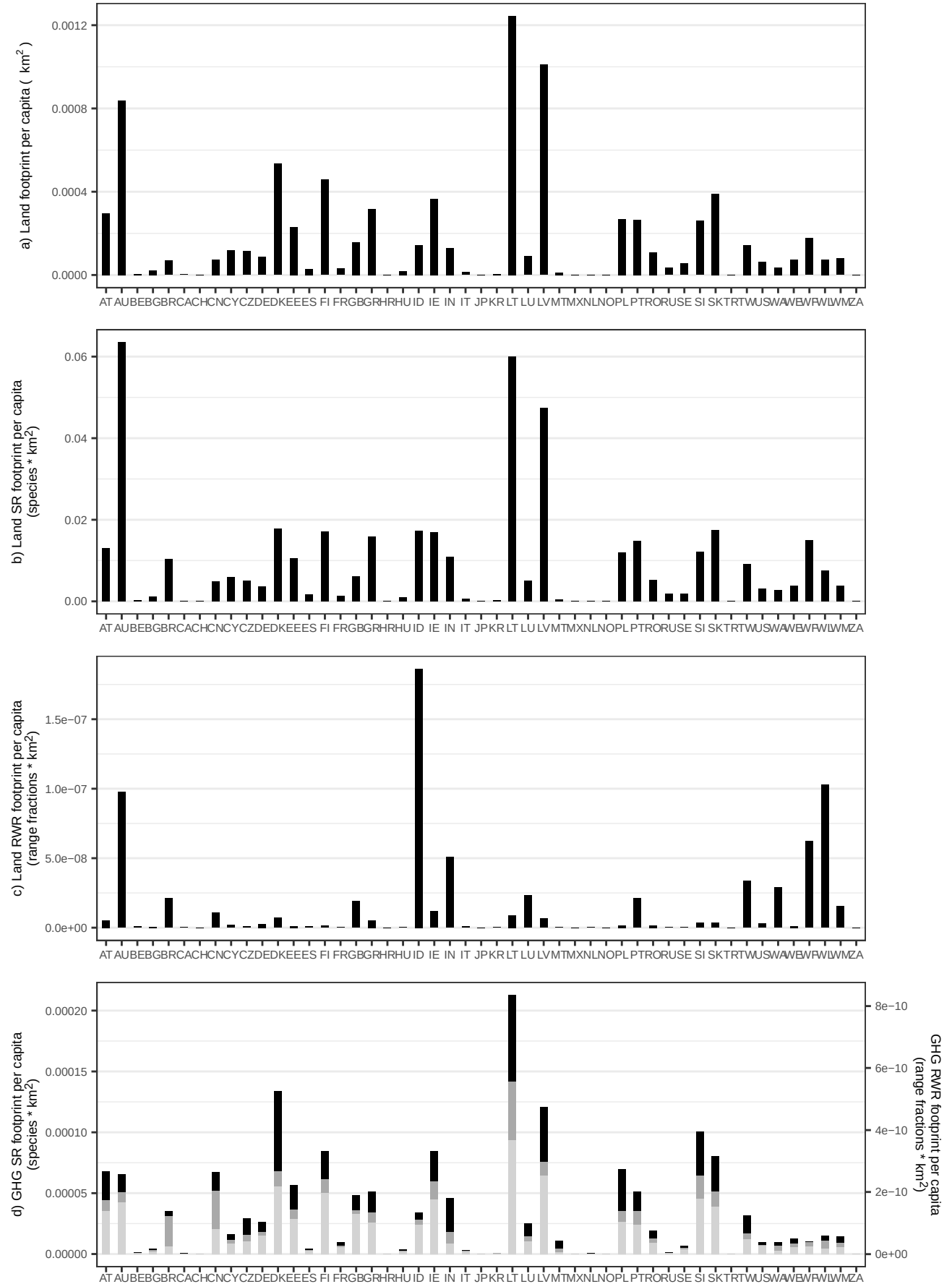
Paddy rice: consumption-based footprints per capita



Supplementary Figure 8.
Wheat: consumption-based footprints per capita

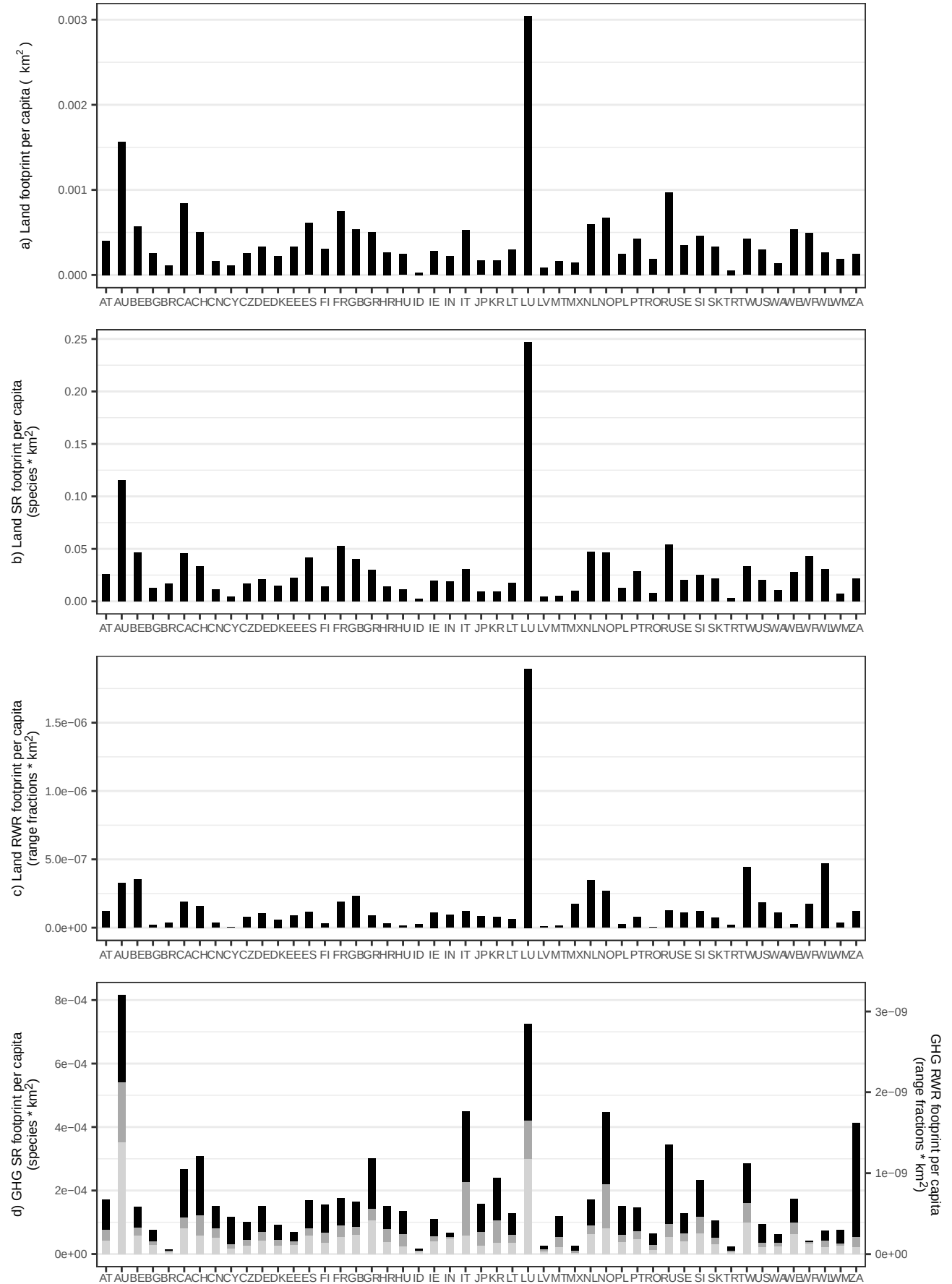


Supplementary Figure 8.
Cereal grains nec: consumption-based footprints per capita



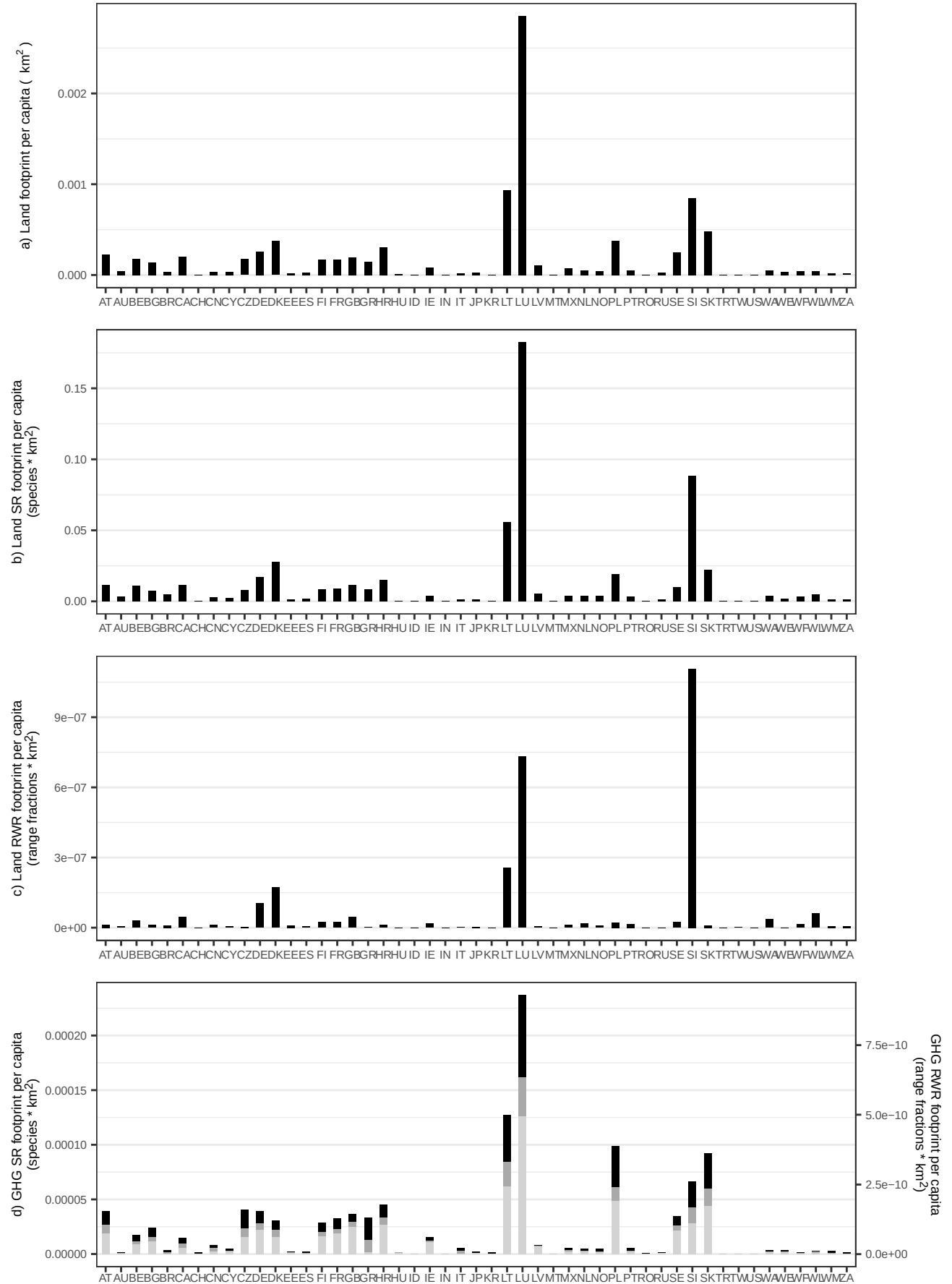
Supplementary Figure 8.

Veg/Fruit/Nuts: consumption-based footprints per capita

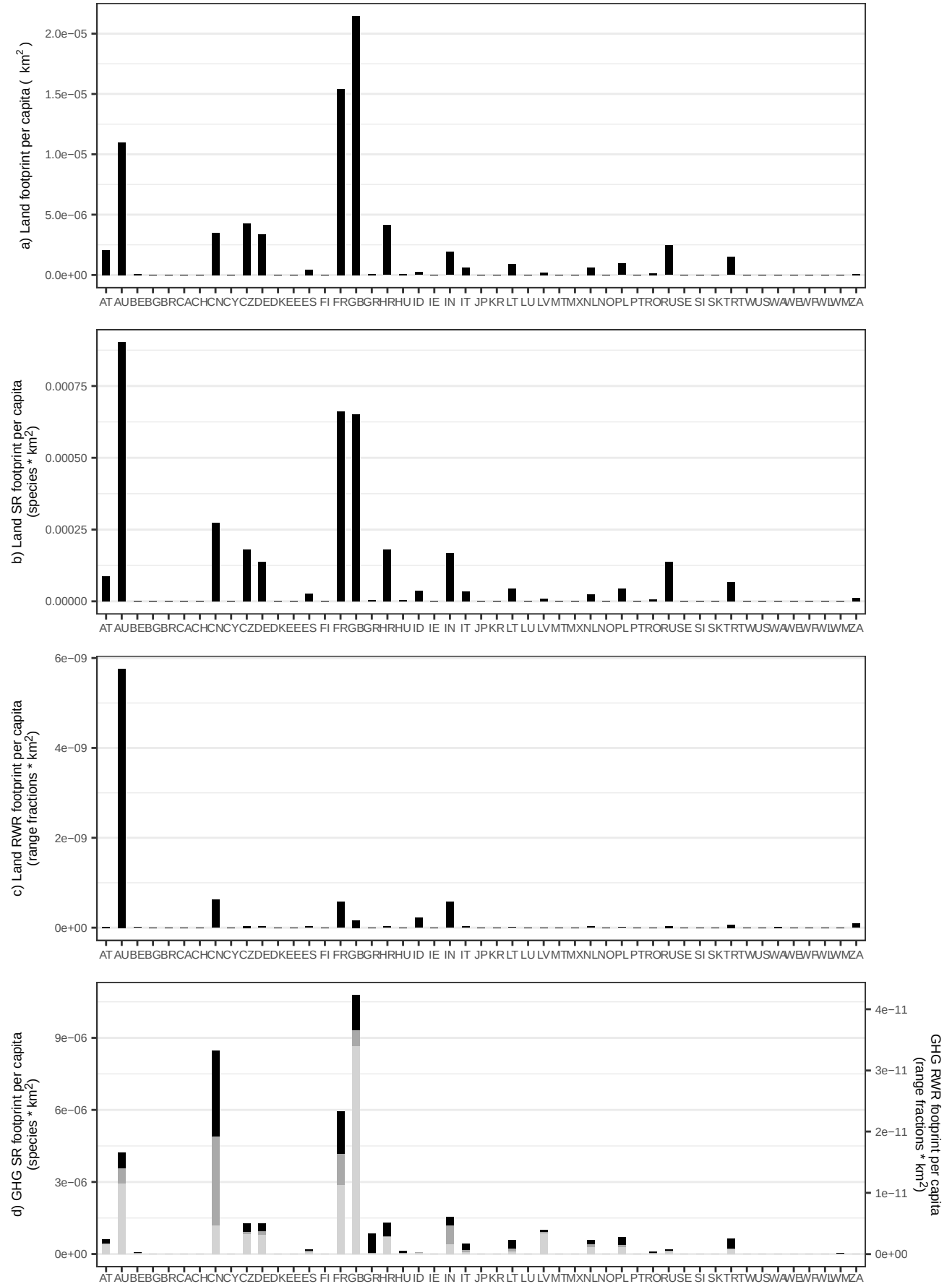


Supplementary Figure 8.

Oil seeds: consumption-based footprints per capita

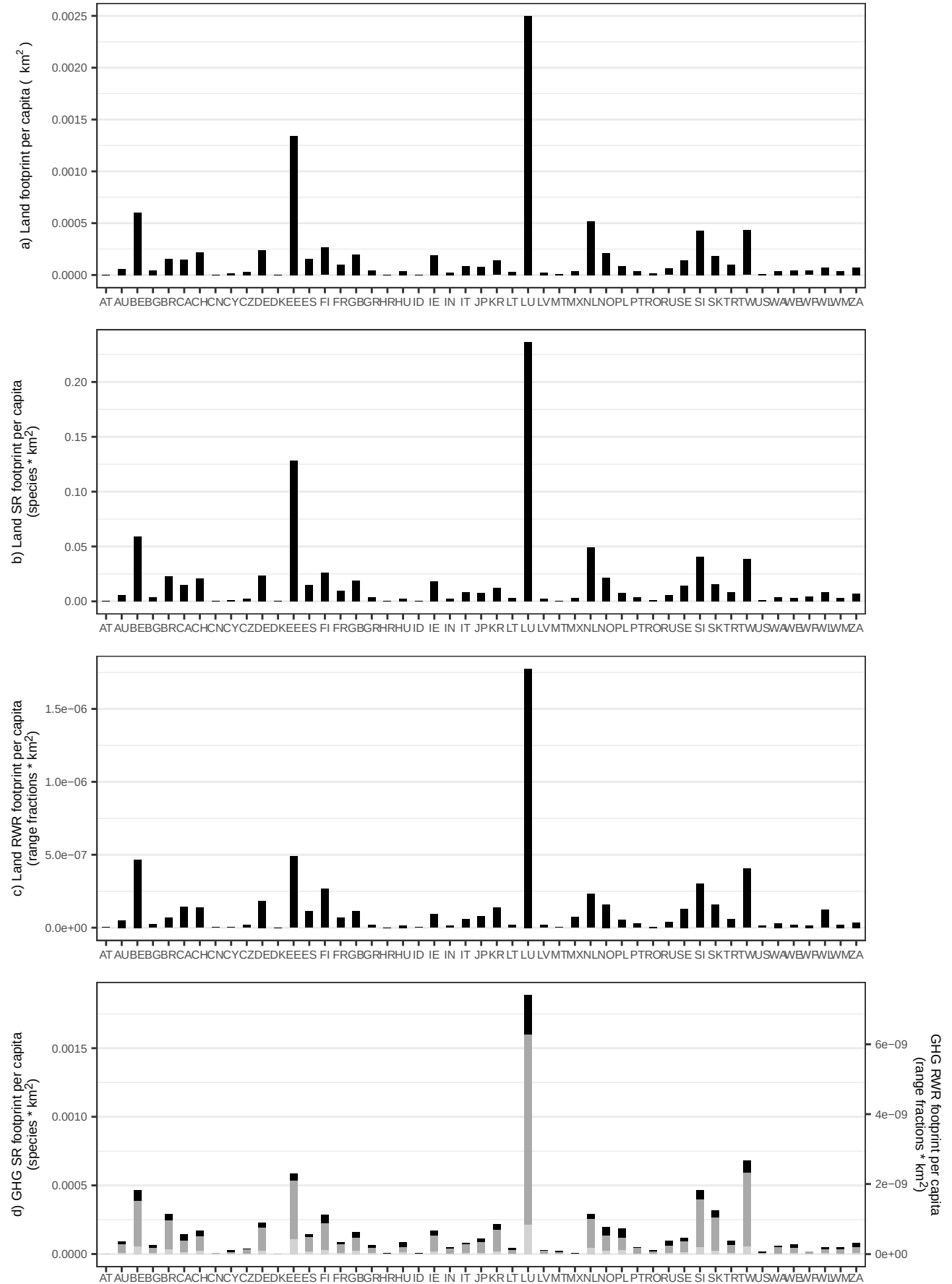


Supplementary Figure 8.
Sugar beet/cane: consumption-based footprints per capita

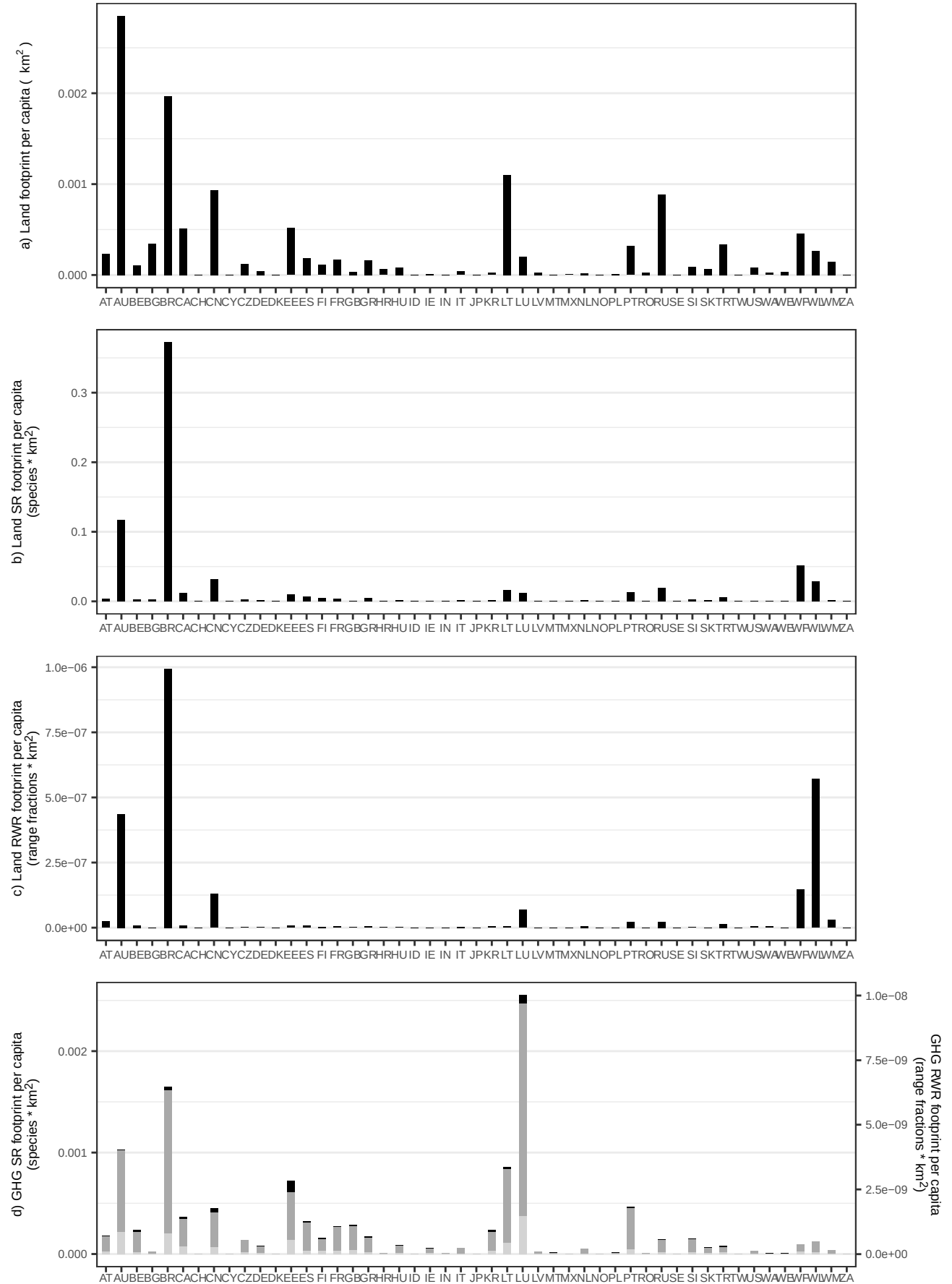


Supplementary Figure 8.

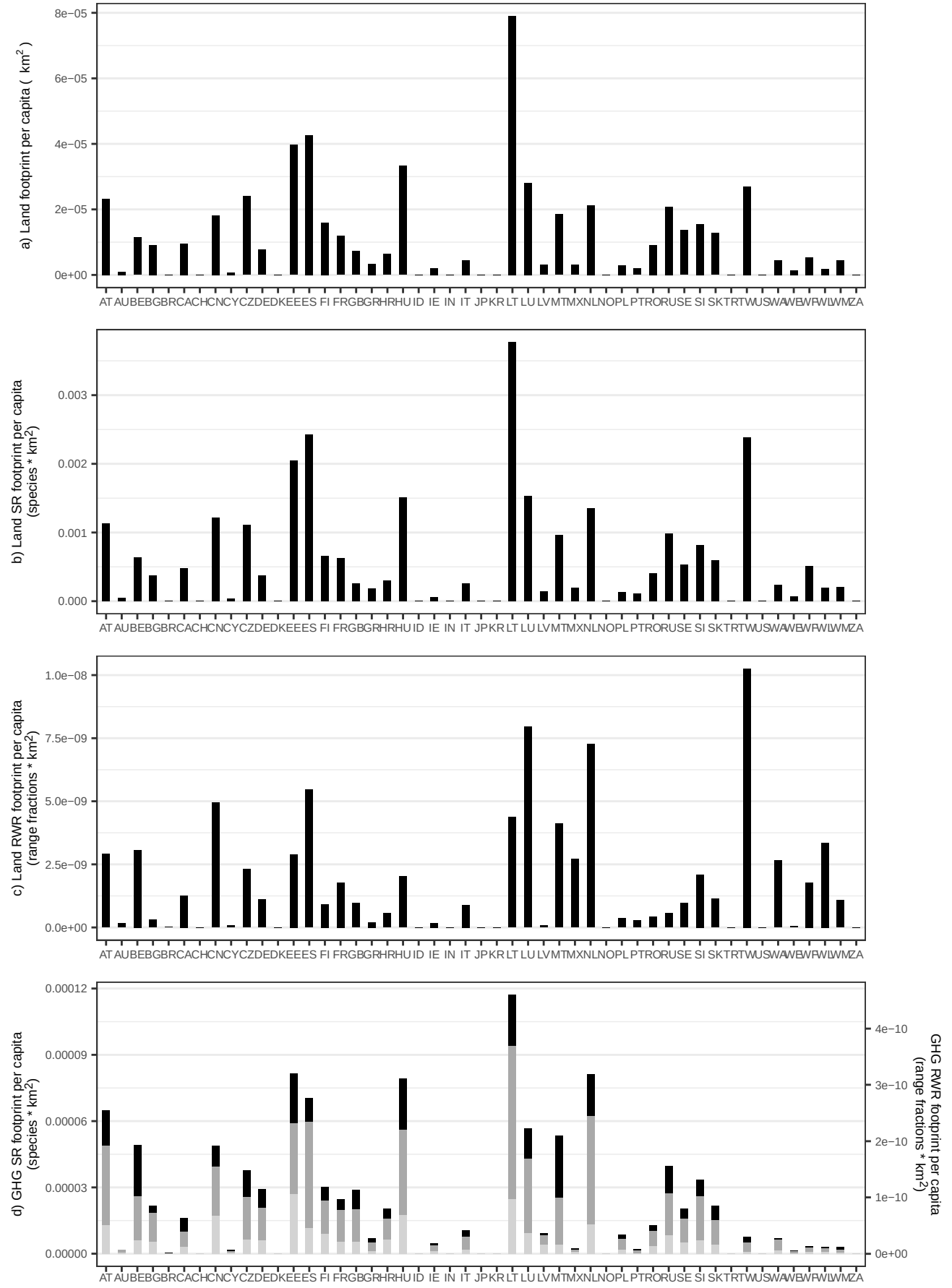
Crops nec: consumption-based footprints per capita



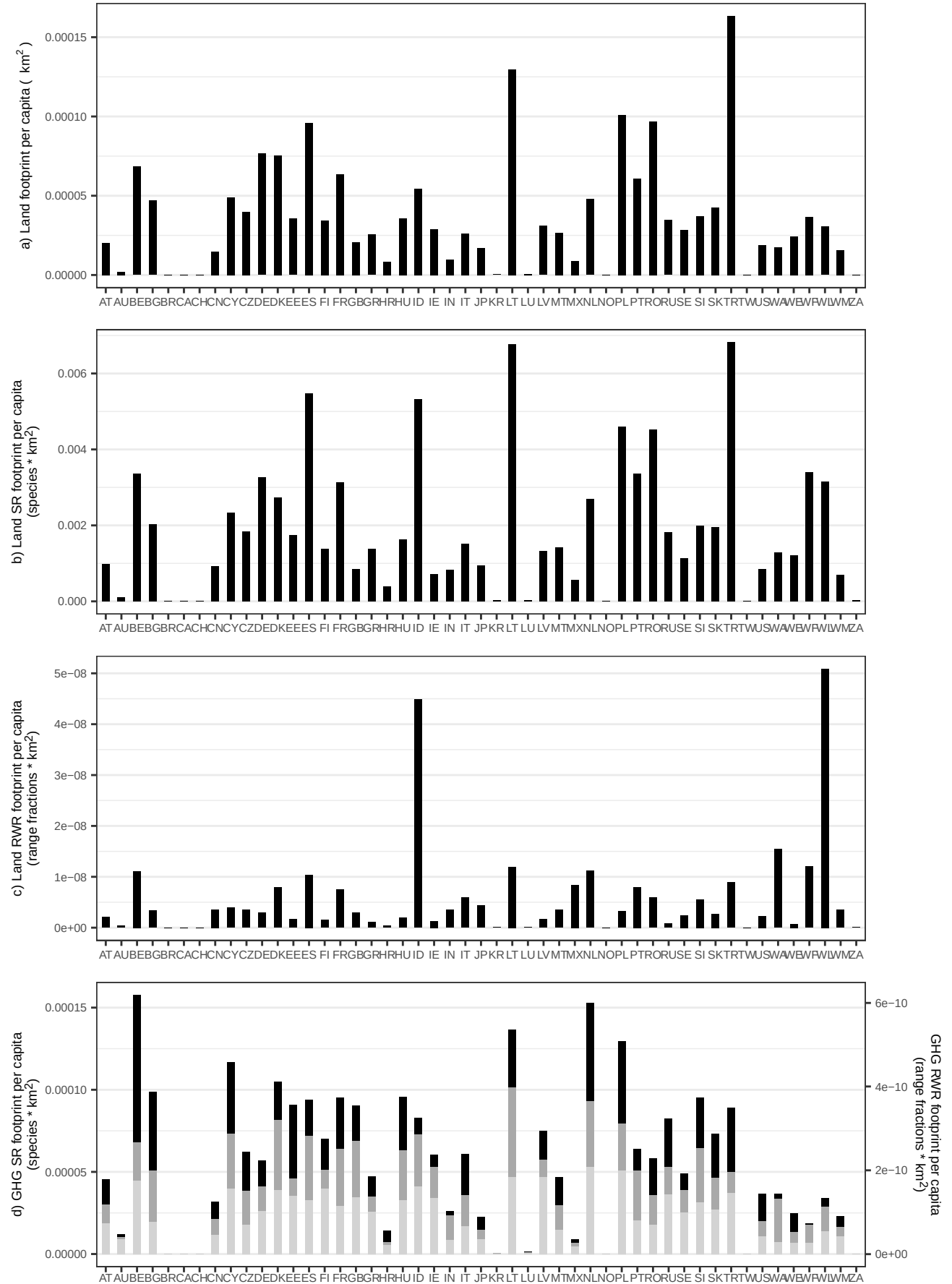
Supplementary Figure 8.
Cattle: consumption-based footprints per capita



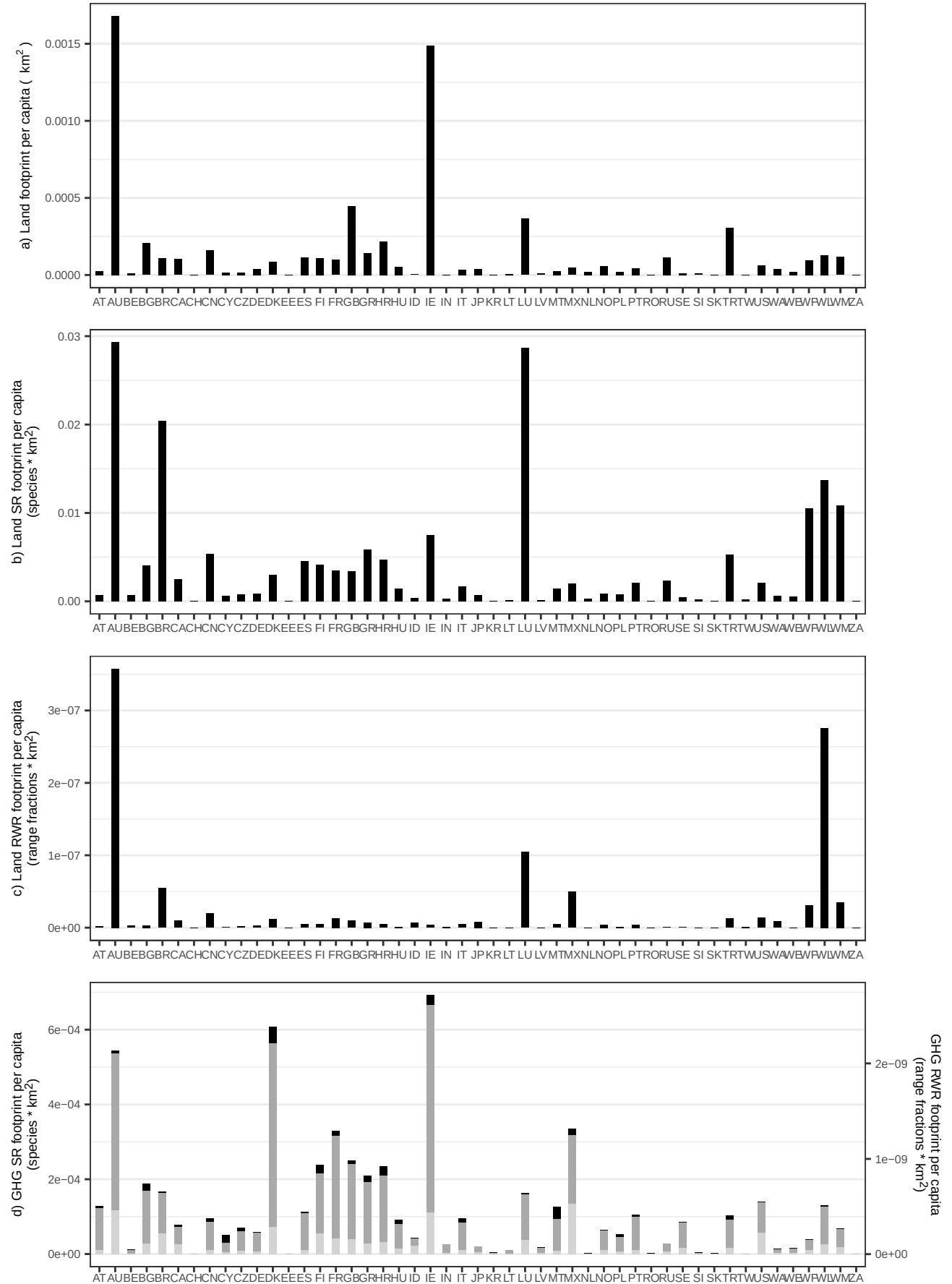
Supplementary Figure 8.
Pigs: consumption-based footprints per capita



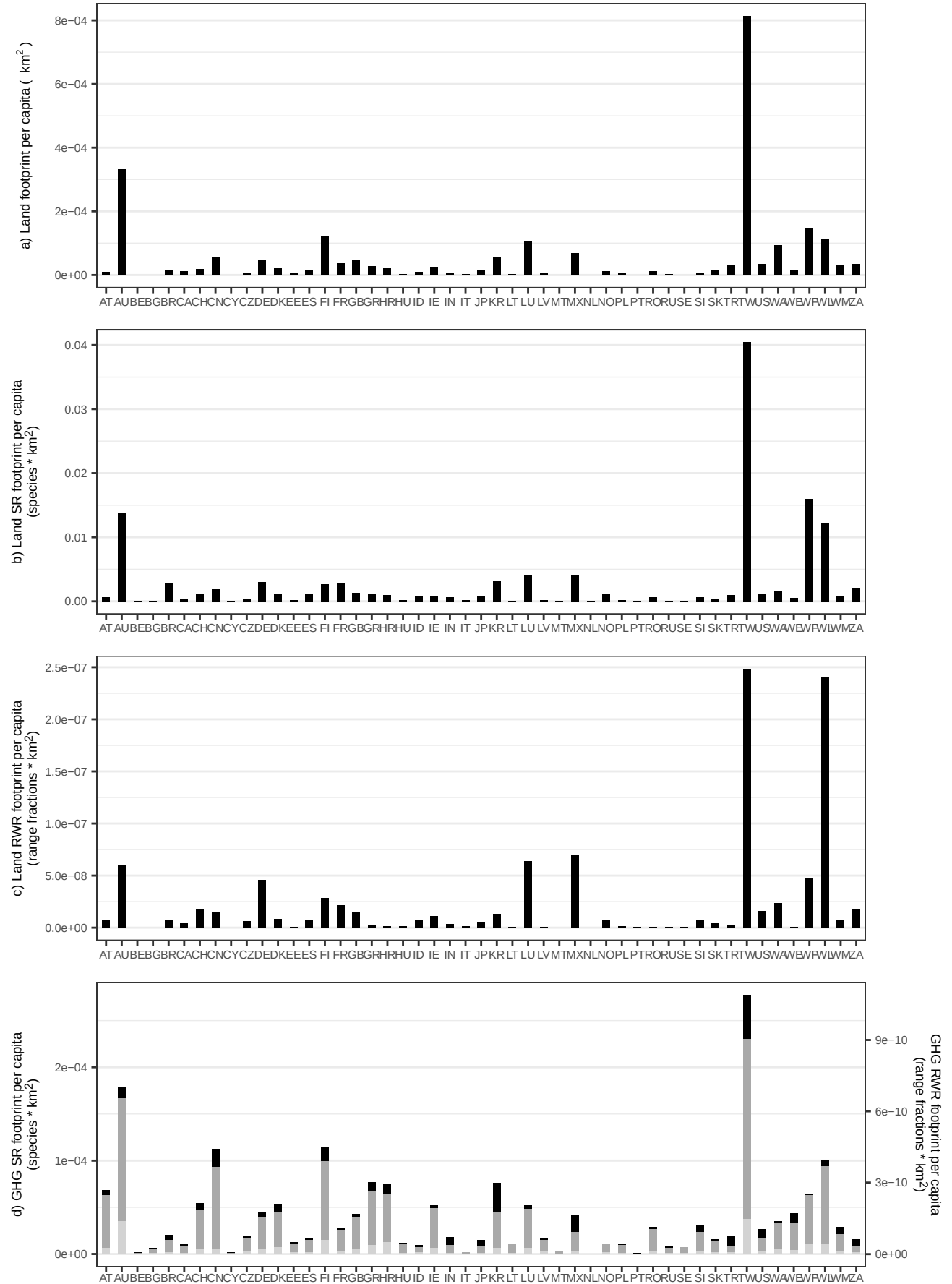
Supplementary Figure 8.
Poultry: consumption-based footprints per capita



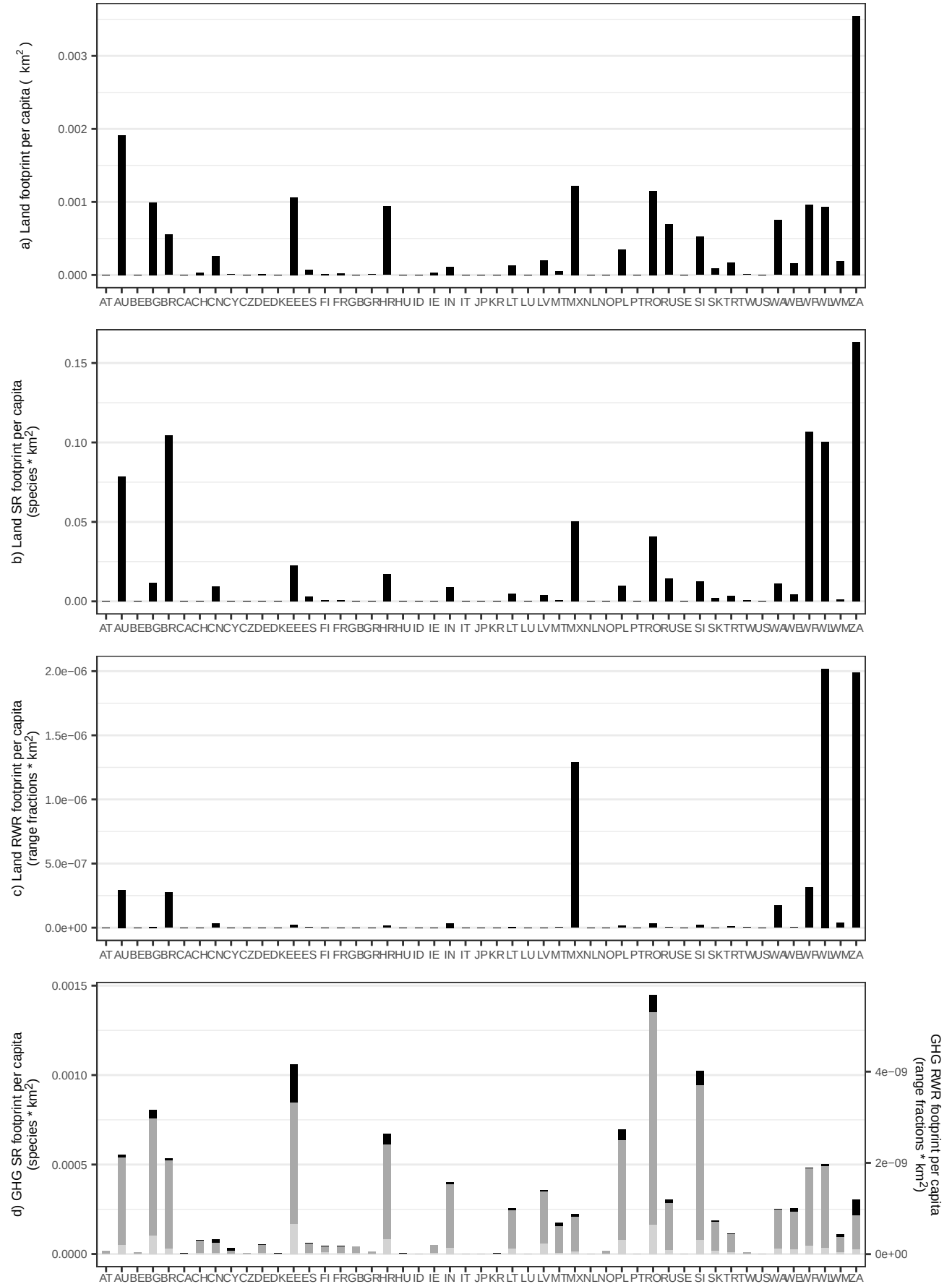
Supplementary Figure 8.
Meat animals nec: consumption-based footprints per capita



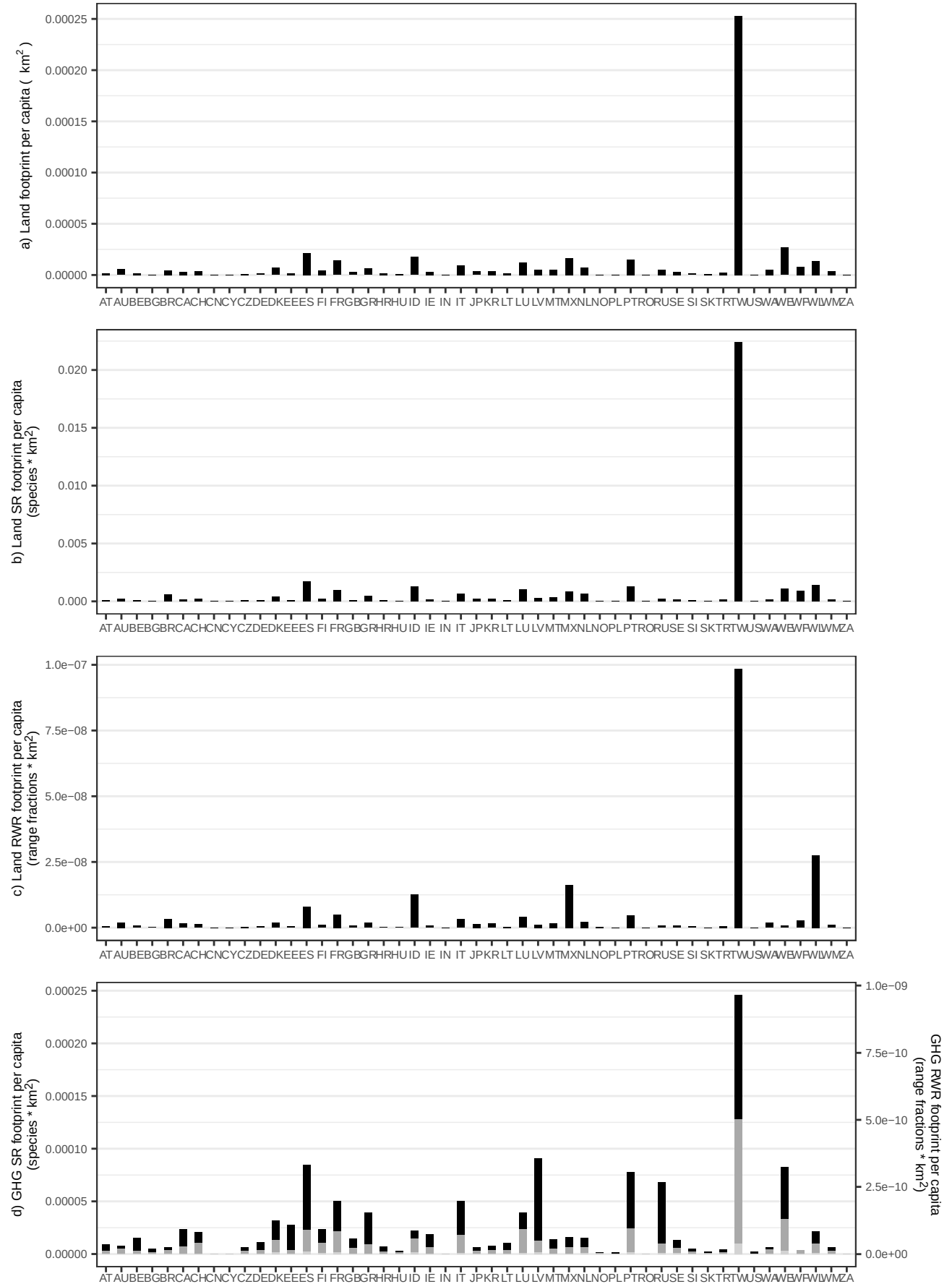
Supplementary Figure 8.
Animal products nec: consumption-based footprints per capita



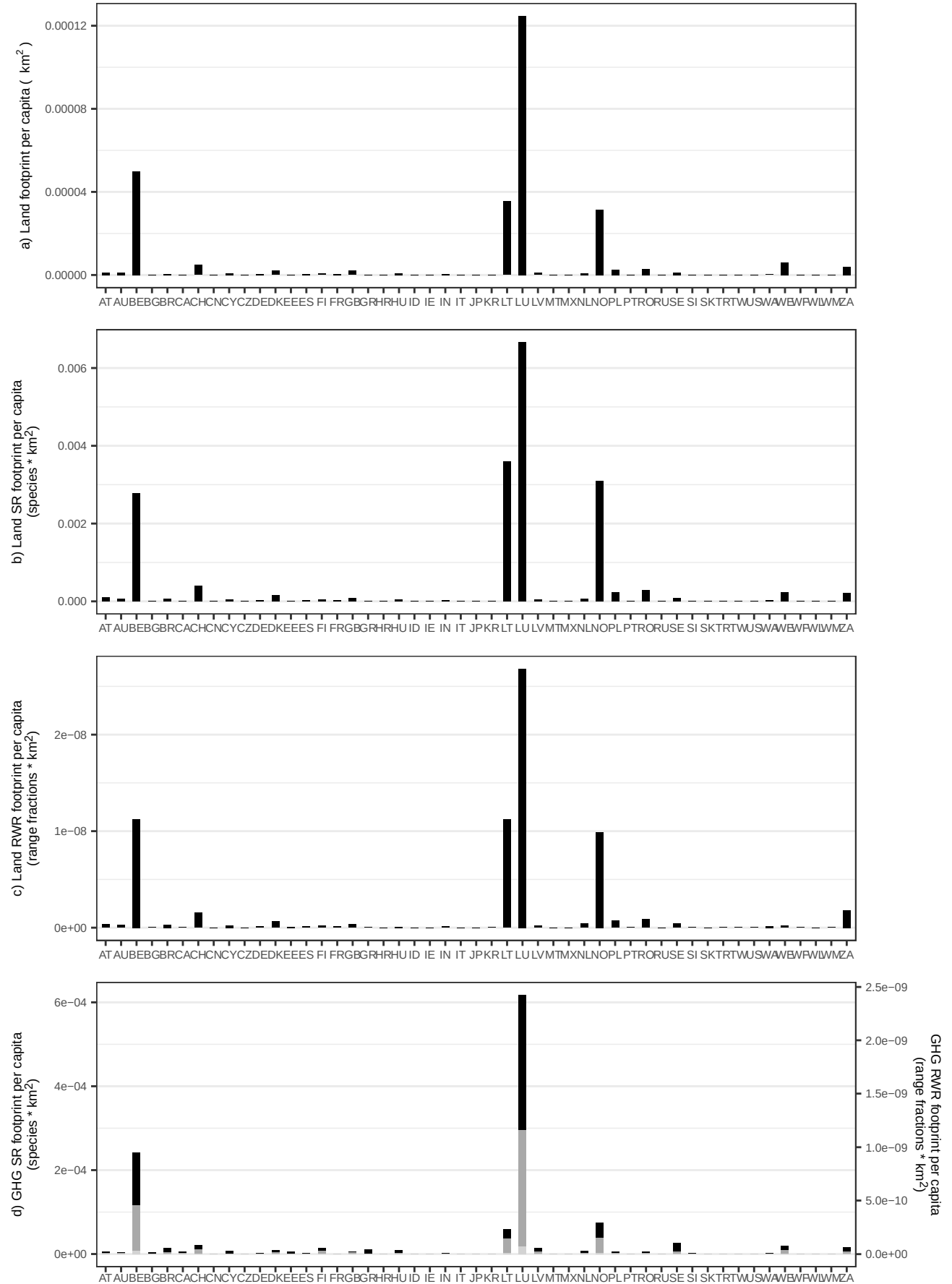
Supplementary Figure 8.
Raw milk: consumption-based footprints per capita



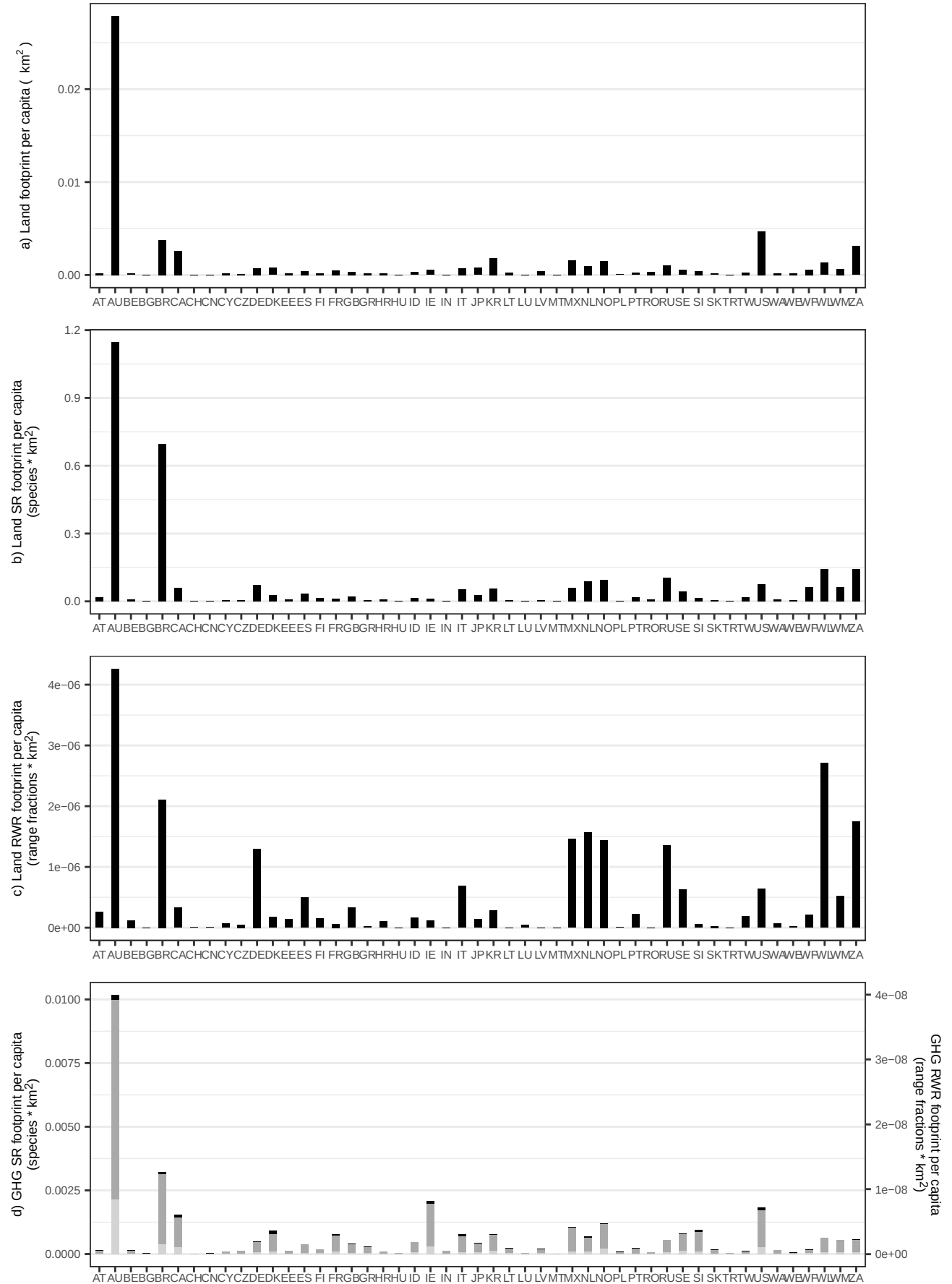
Supplementary Figure 8.
Fish and other fishing products: consumption-based footprints per capita



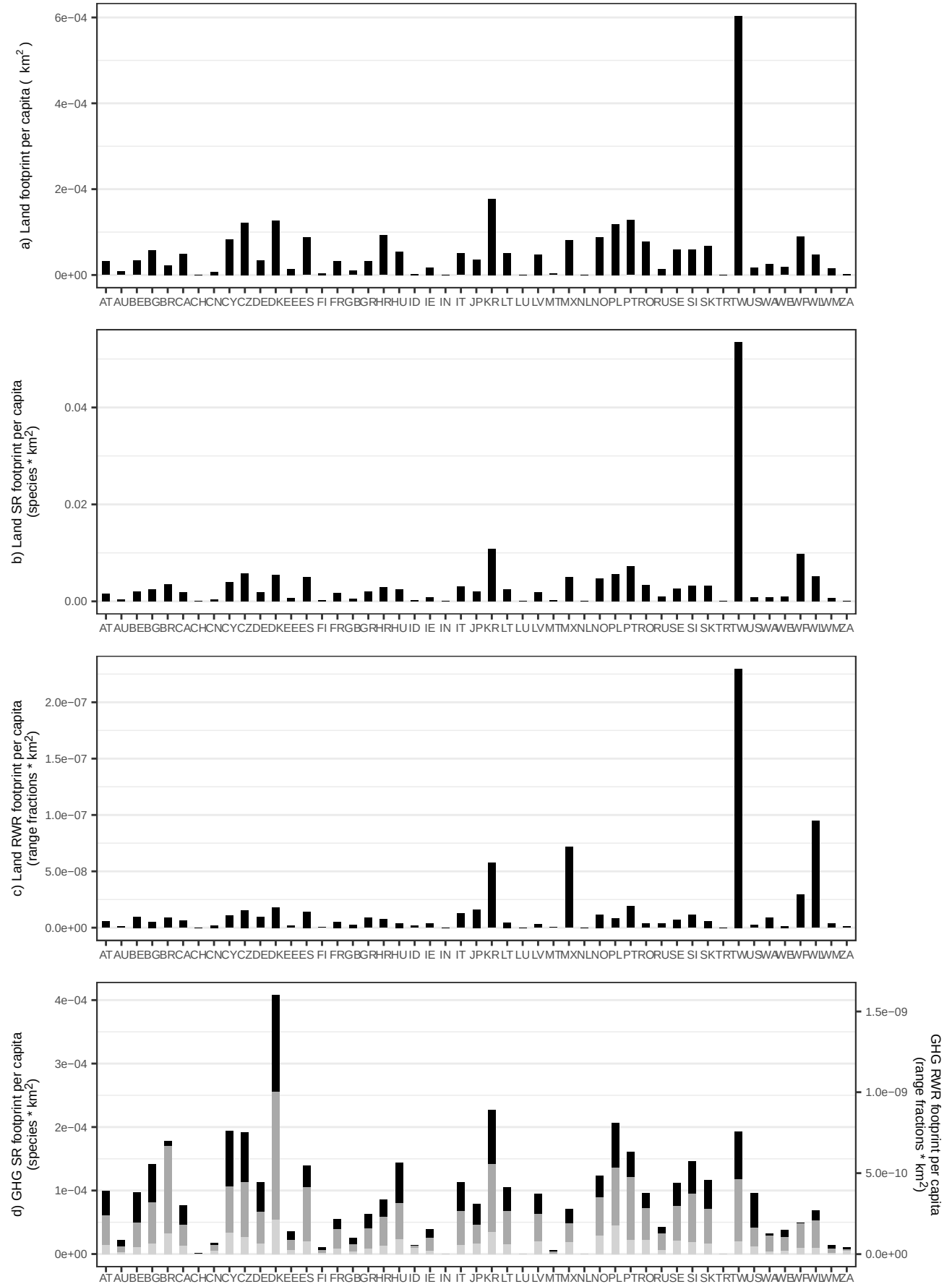
Supplementary Figure 8.
Chemical and fertiliser minerals: consumption-based footprints per capita



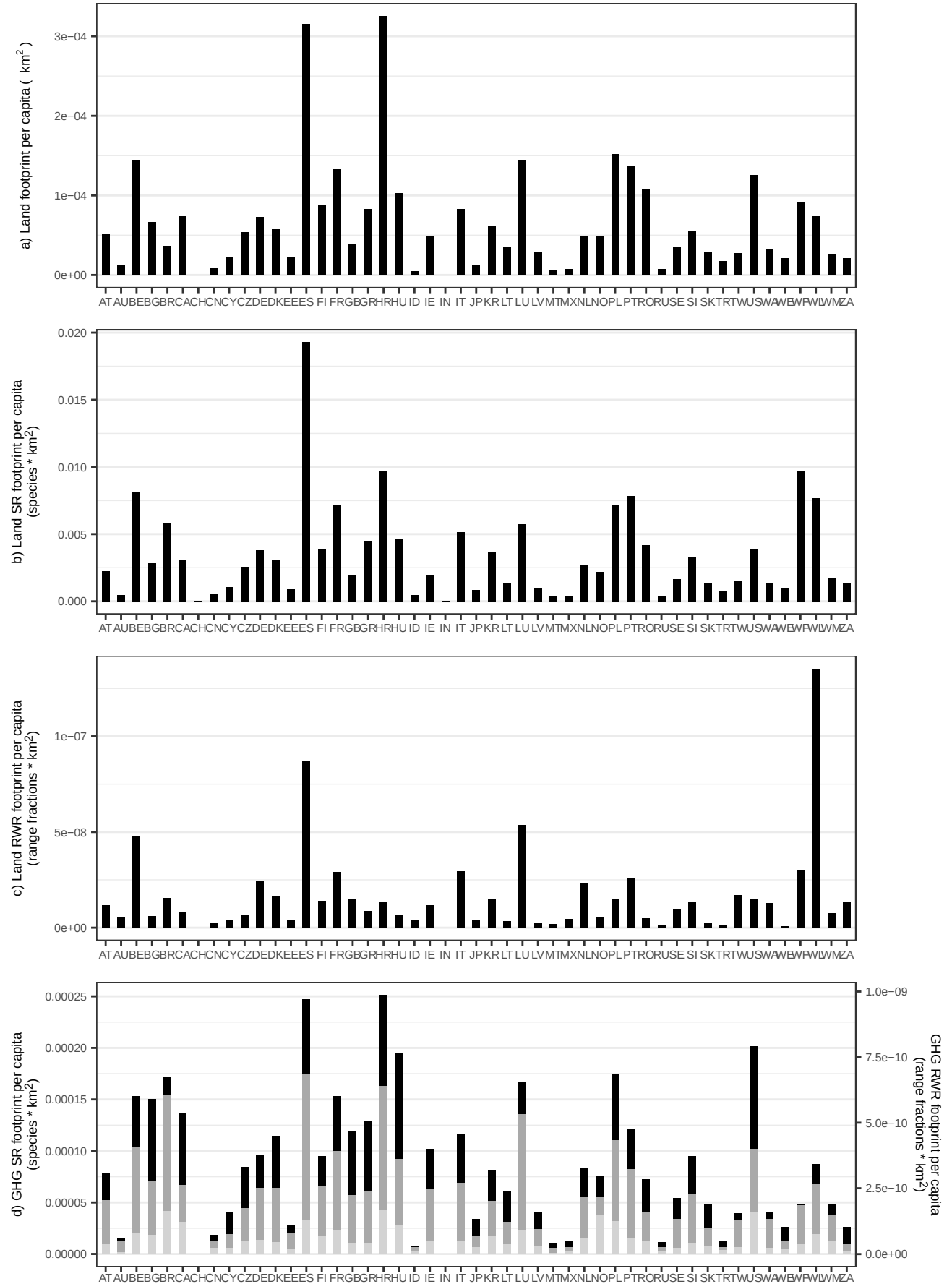
Supplementary Figure 8.
Products of meat cattle: consumption-based footprints per capita



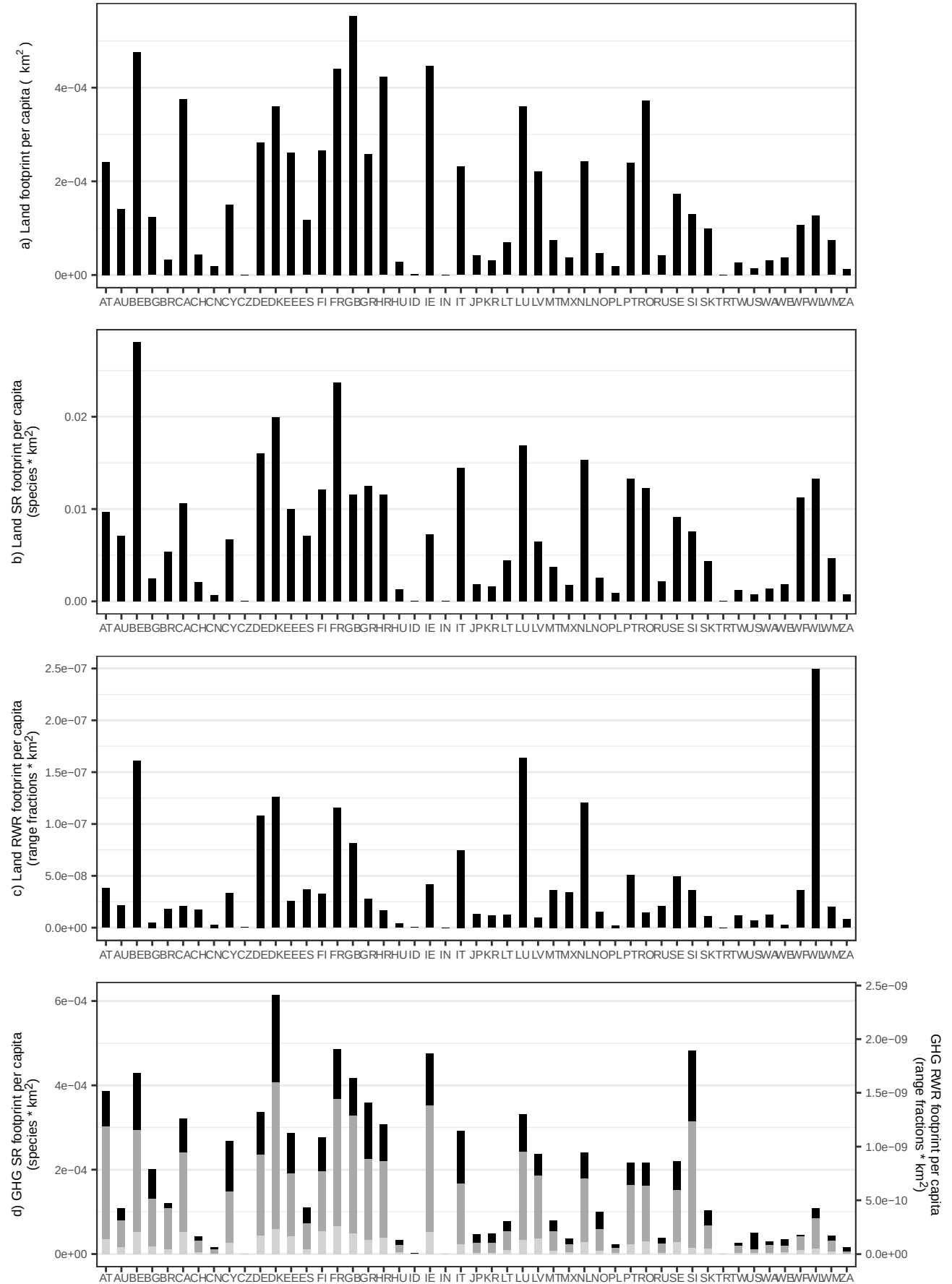
Supplementary Figure 8.
Products of meat pigs: consumption-based footprints per capita



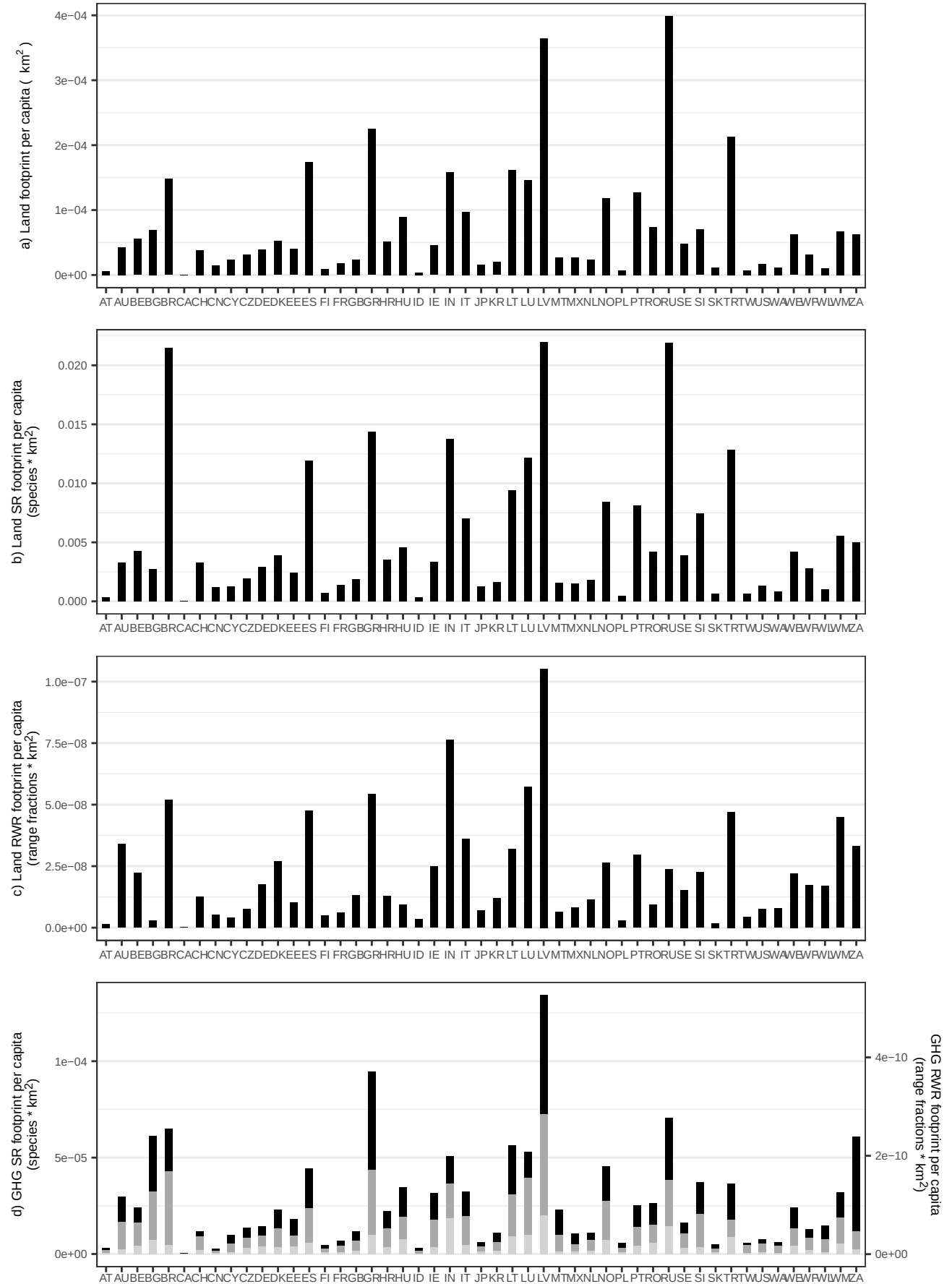
Supplementary Figure 8.
Products of meat poultry: consumption-based footprints per capita



Supplementary Figure 8.
Meat products nec: consumption-based footprints per capita

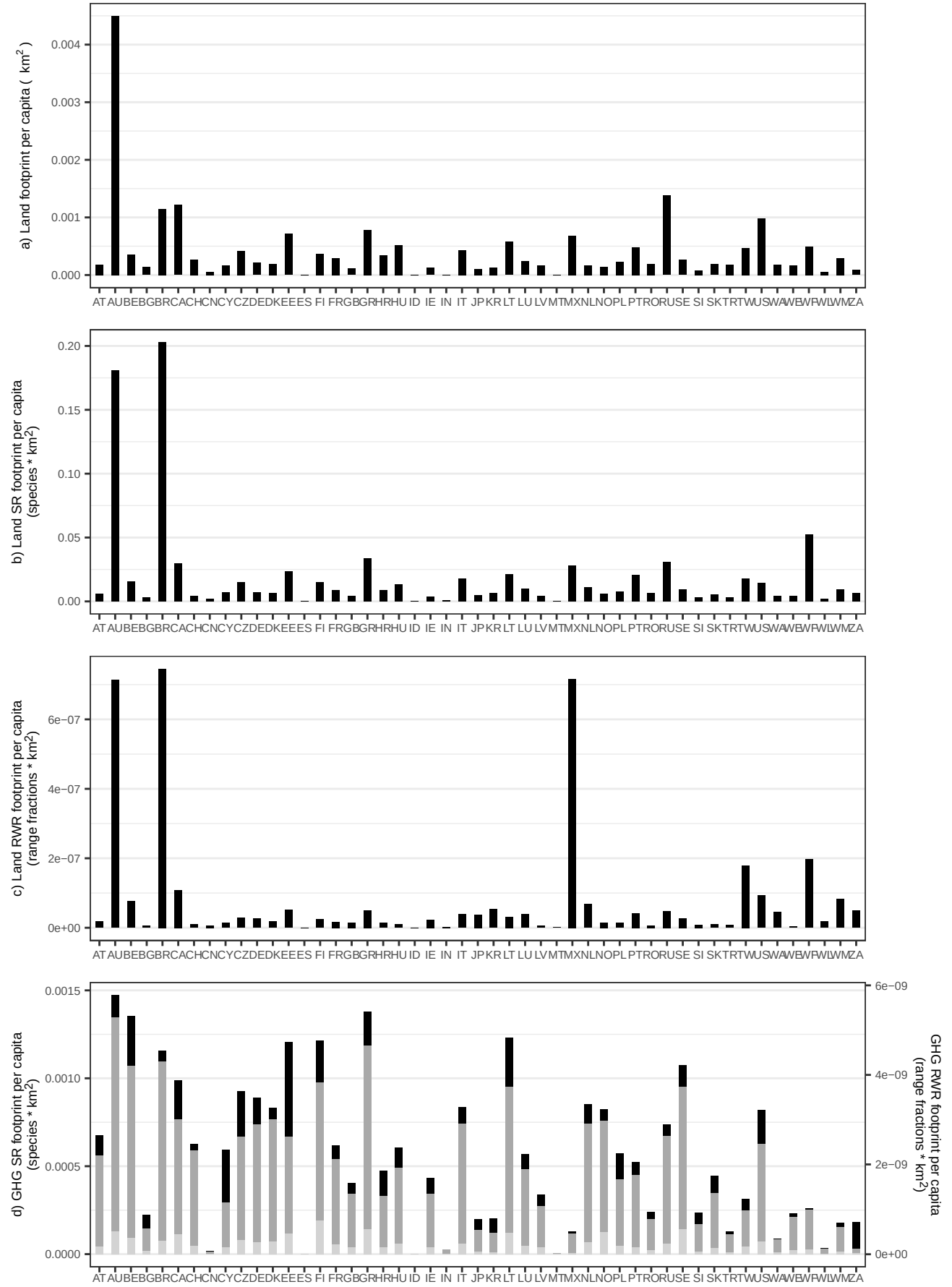


Supplementary Figure 8.
Products of vegetable oils and fats: consumption-based footprints per capita

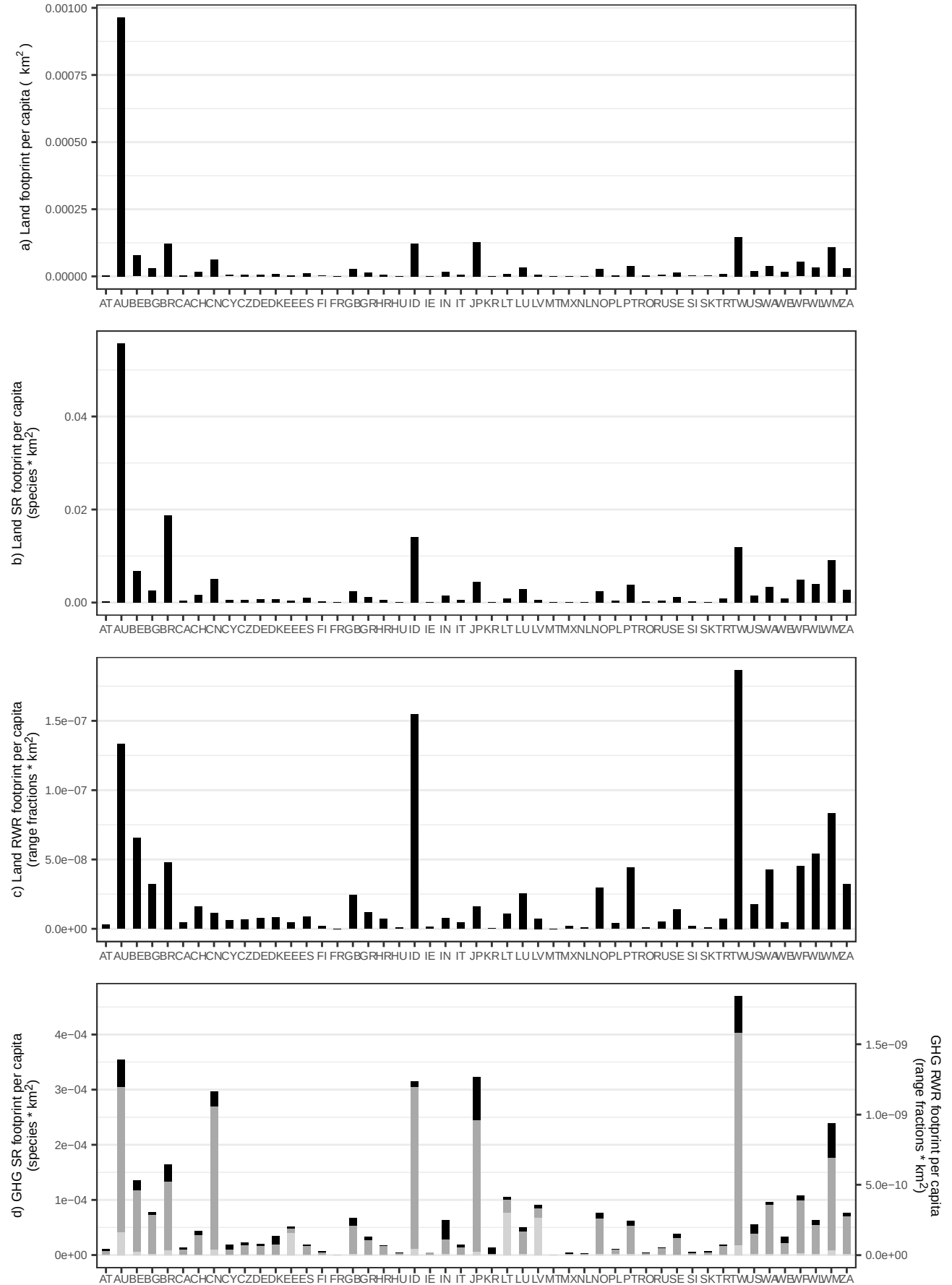


Supplementary Figure 8.

Dairy products: consumption-based footprints per capita

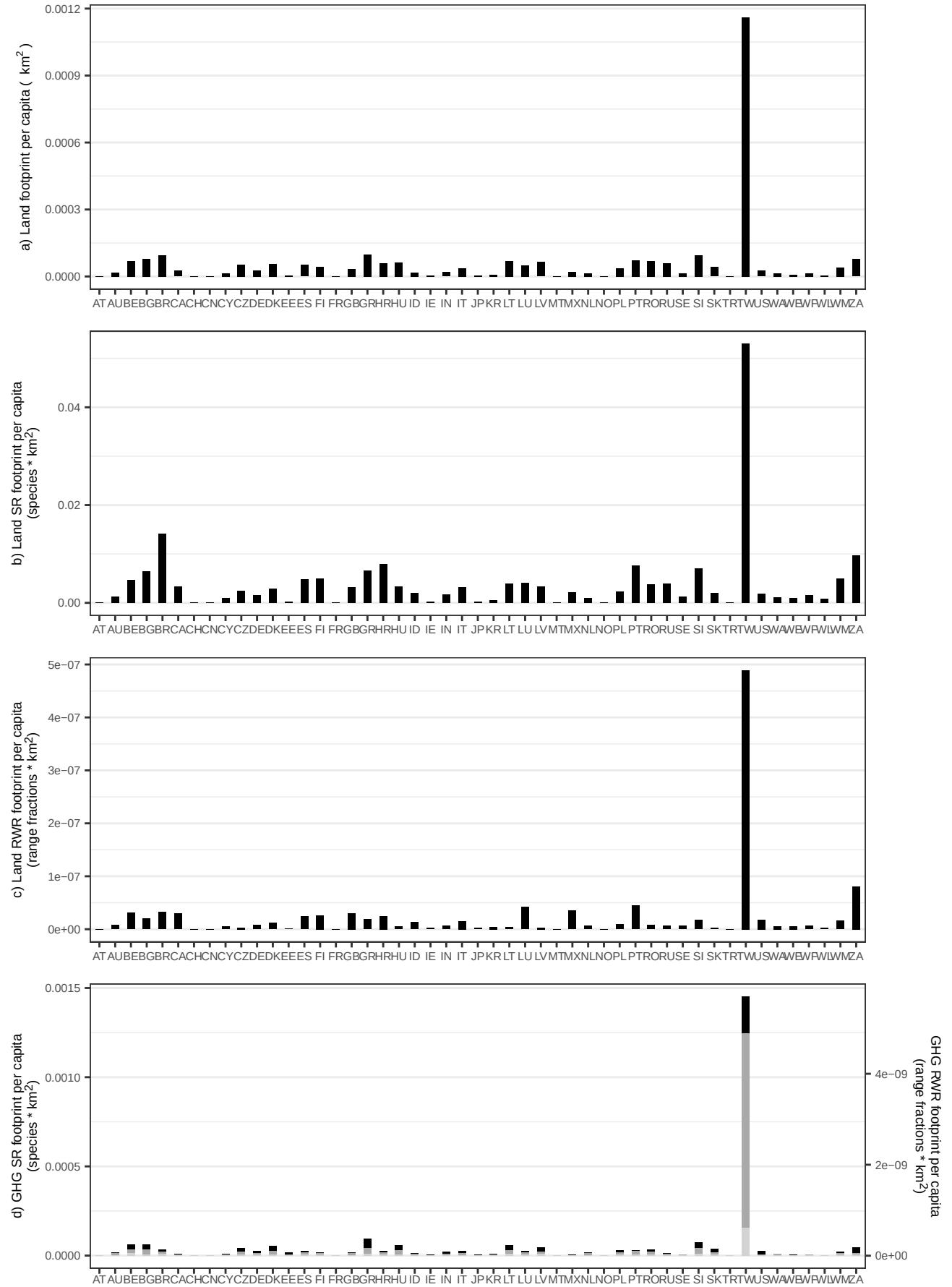


Supplementary Figure 8.
Processed rice: consumption-based footprints per capita

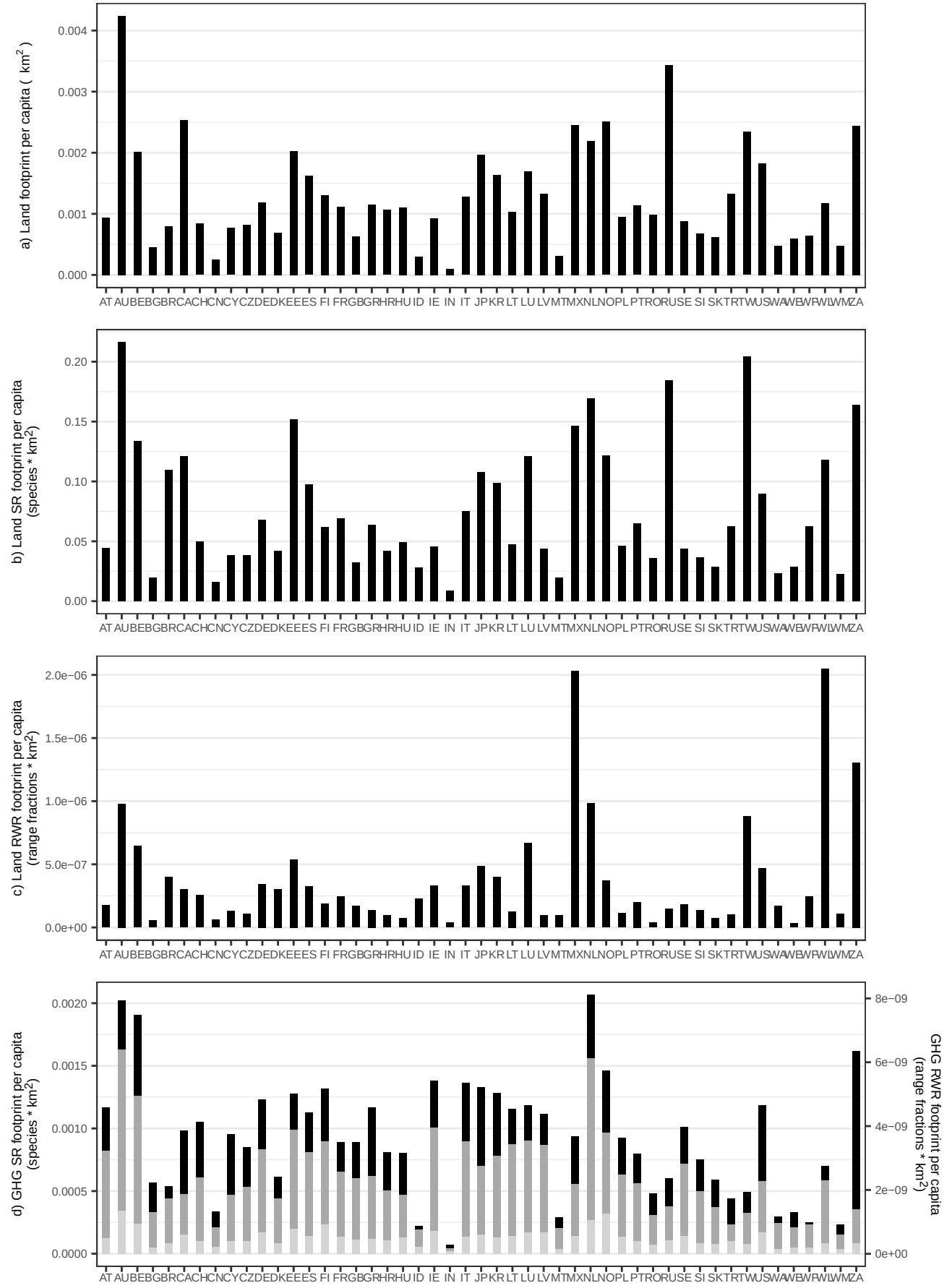


Supplementary Figure 8.

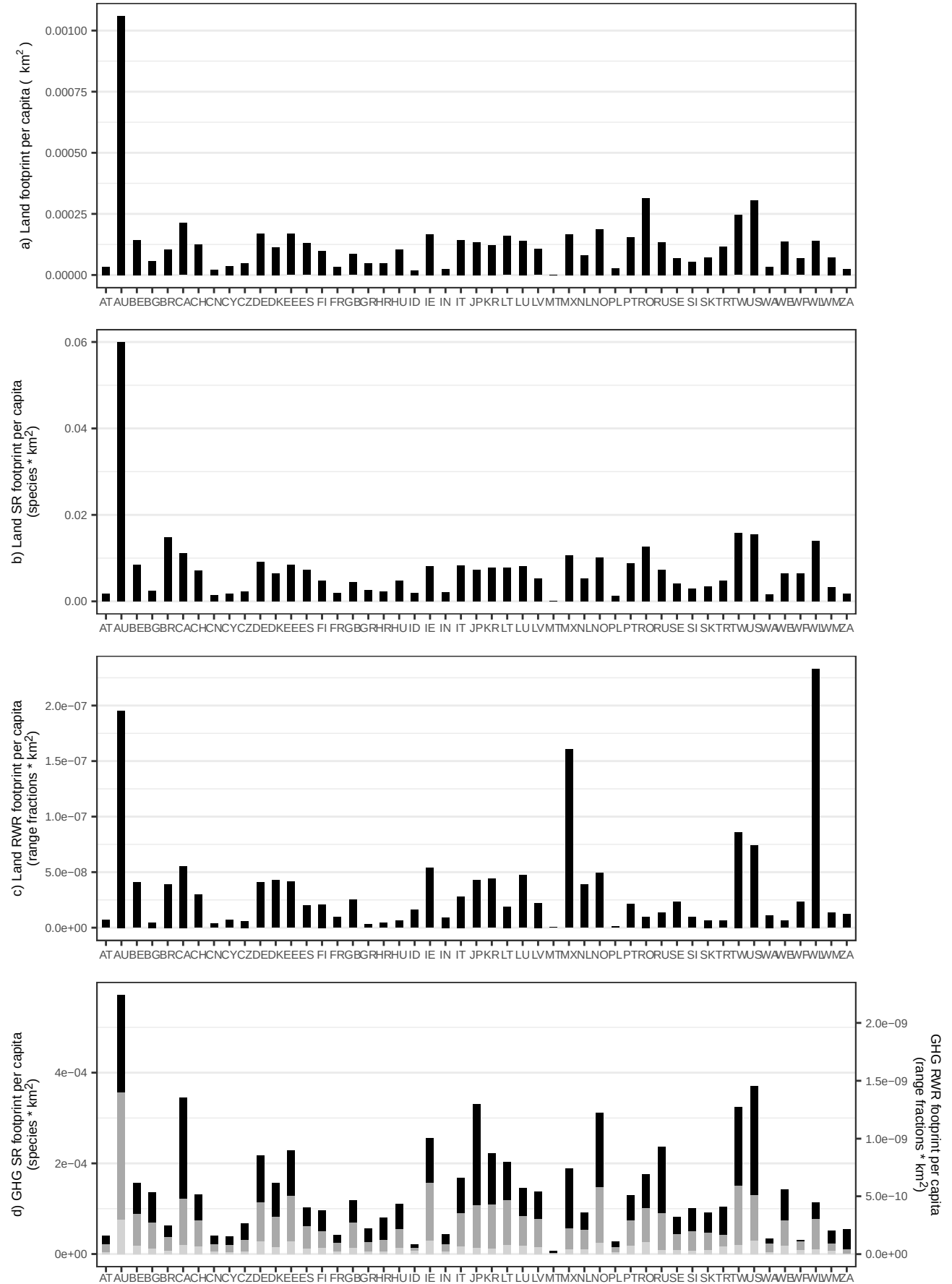
Processed sugar: consumption-based footprints per capita



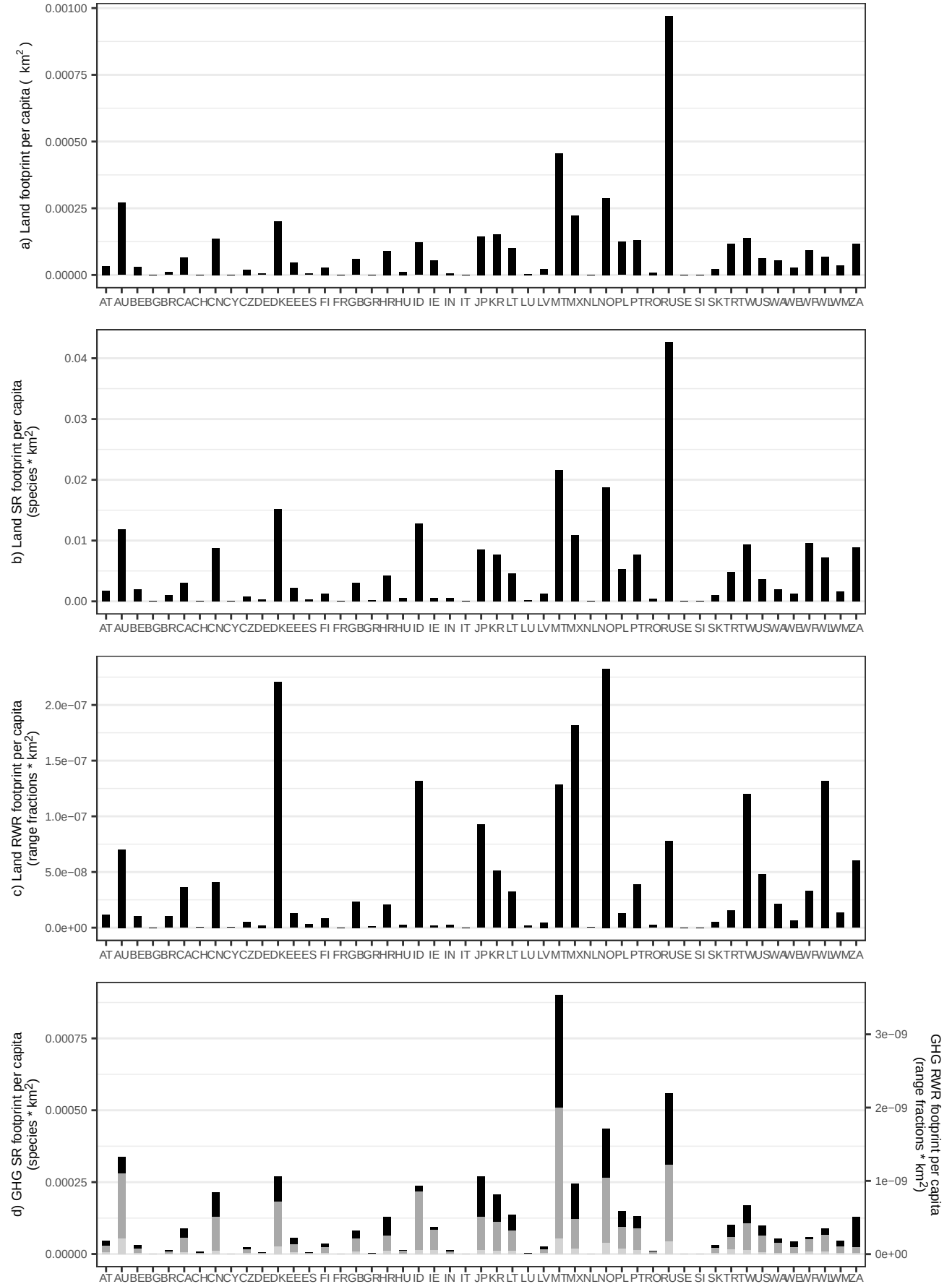
Supplementary Figure 8.
Food products nec: consumption-based footprints per capita



Supplementary Figure 8.
Beverages: consumption-based footprints per capita

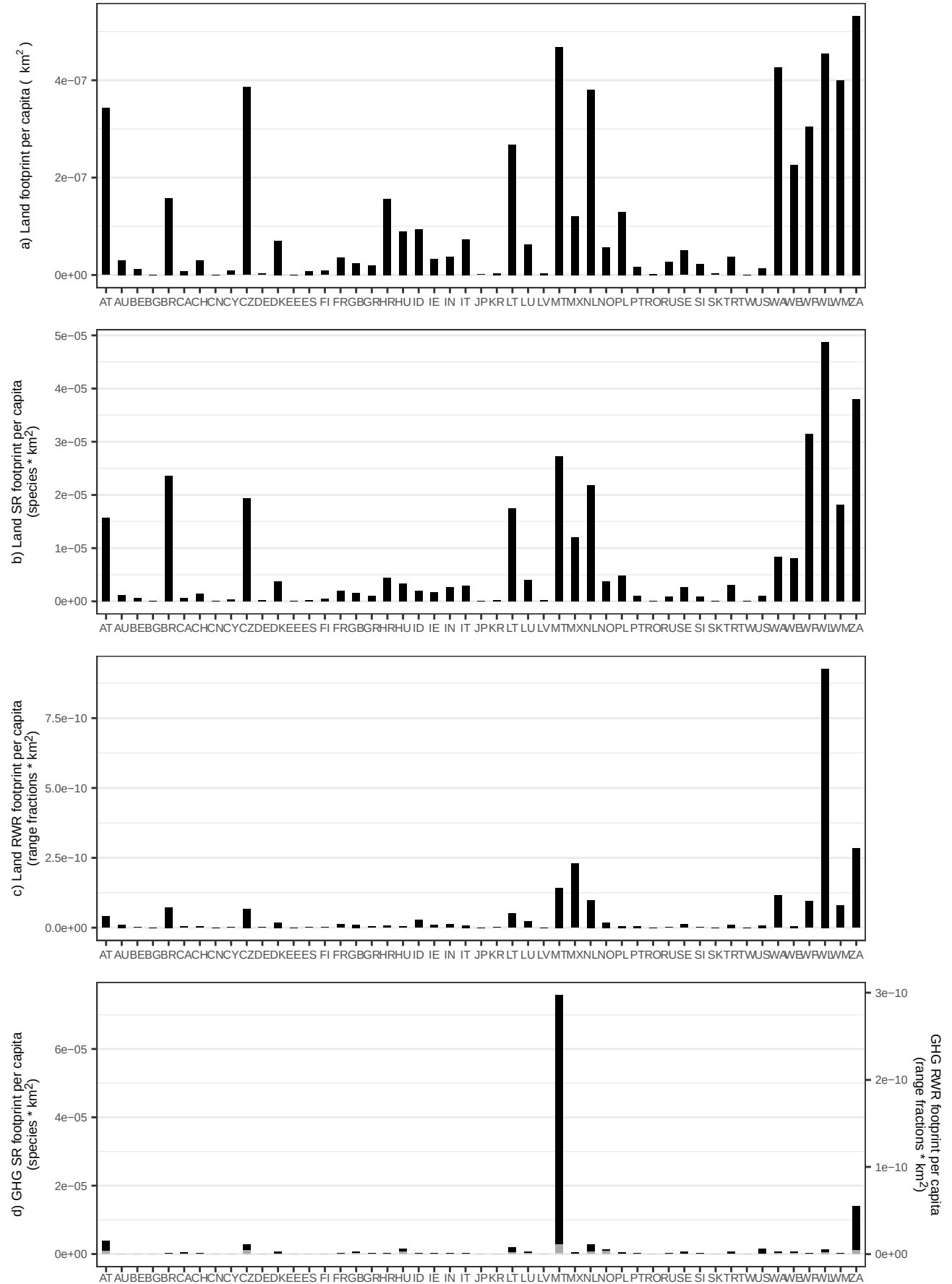


Supplementary Figure 8.
Fish products: consumption-based footprints per capita

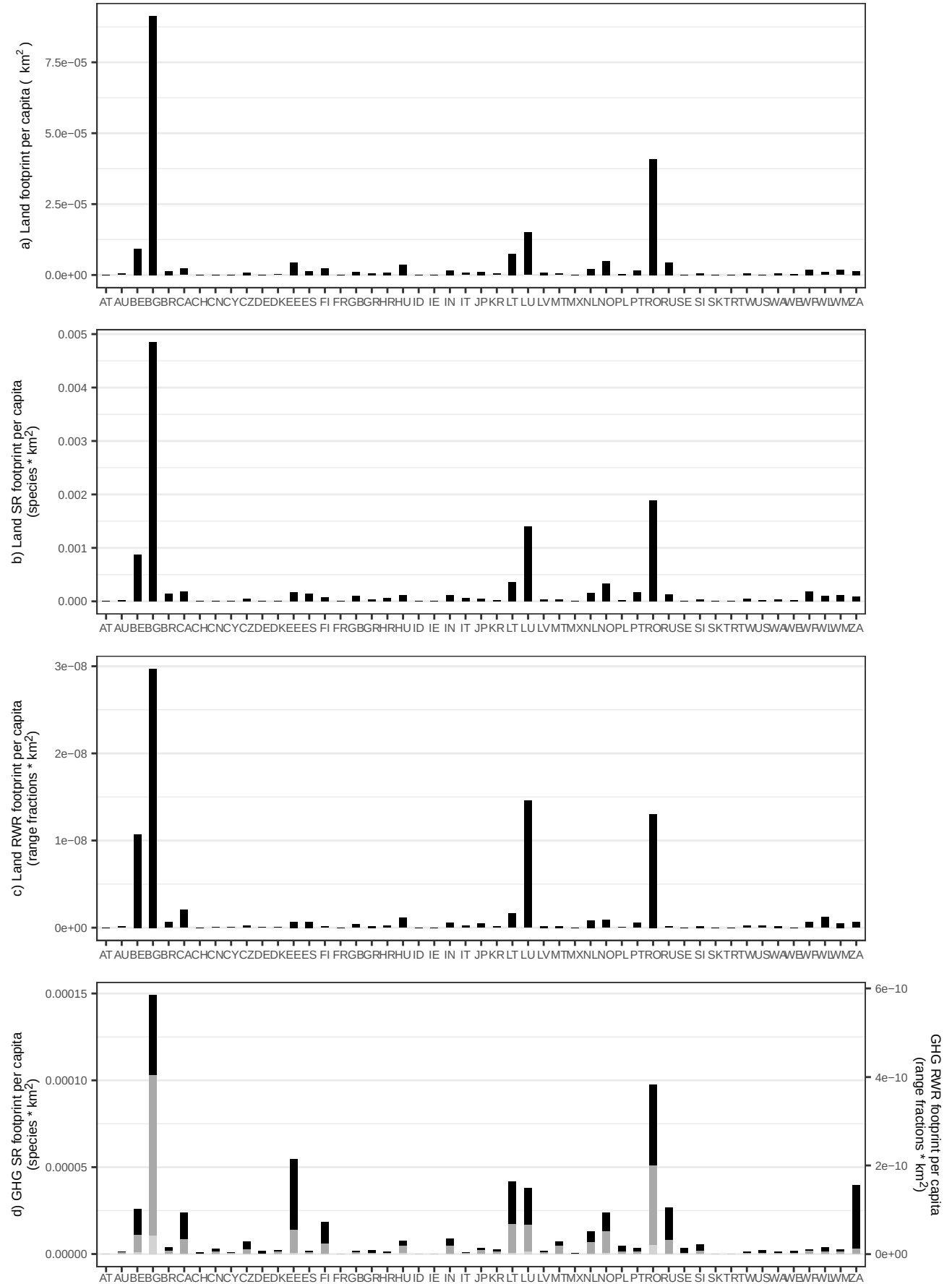


Supplementary Figure 8.

N-fertiliser: consumption-based footprints per capita

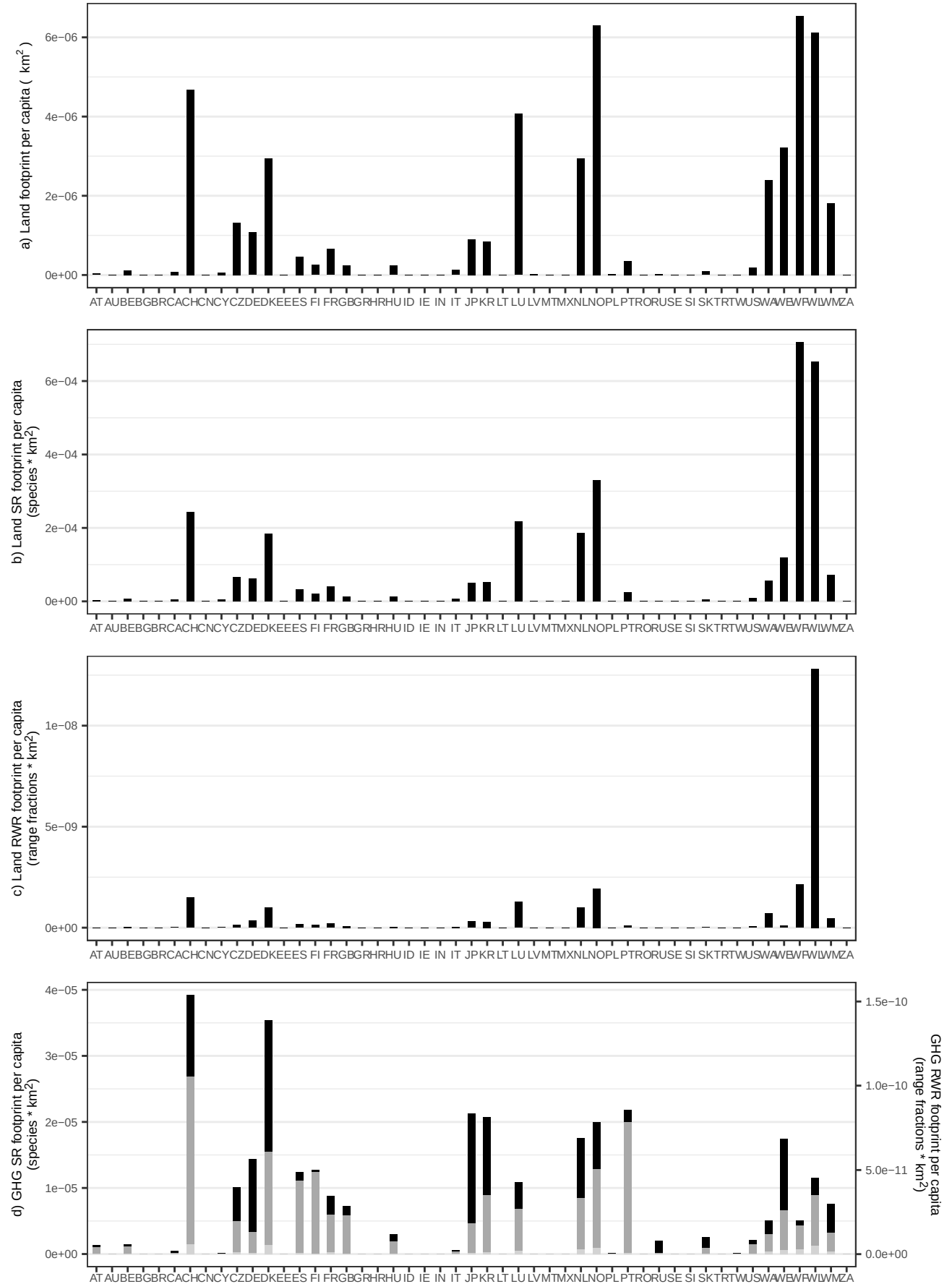


Supplementary Figure 8.
P- and other fertiliser: consumption-based footprints per capita



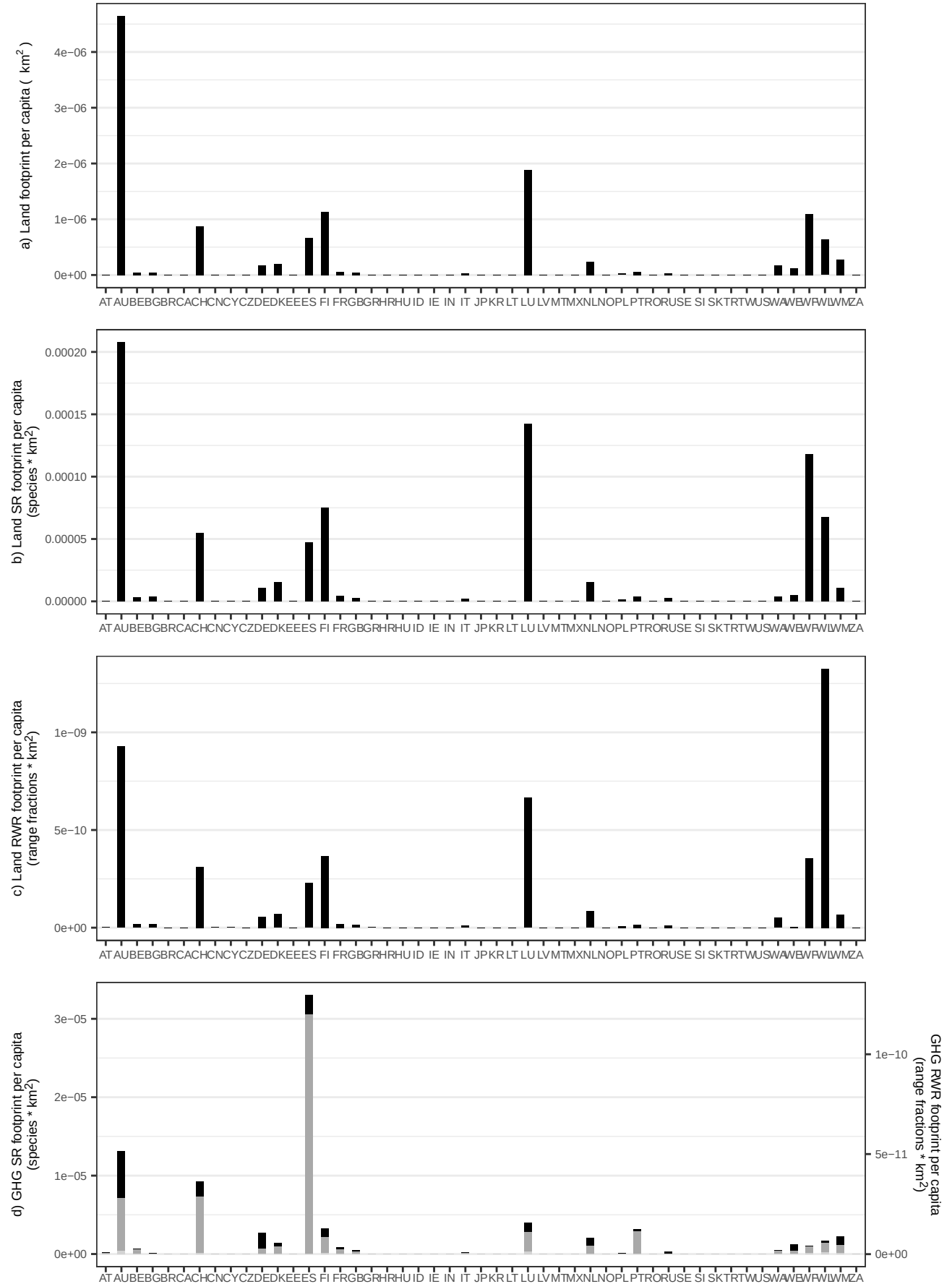
Supplementary Figure 8.

Food waste: incineration: consumption-based footprints per capita



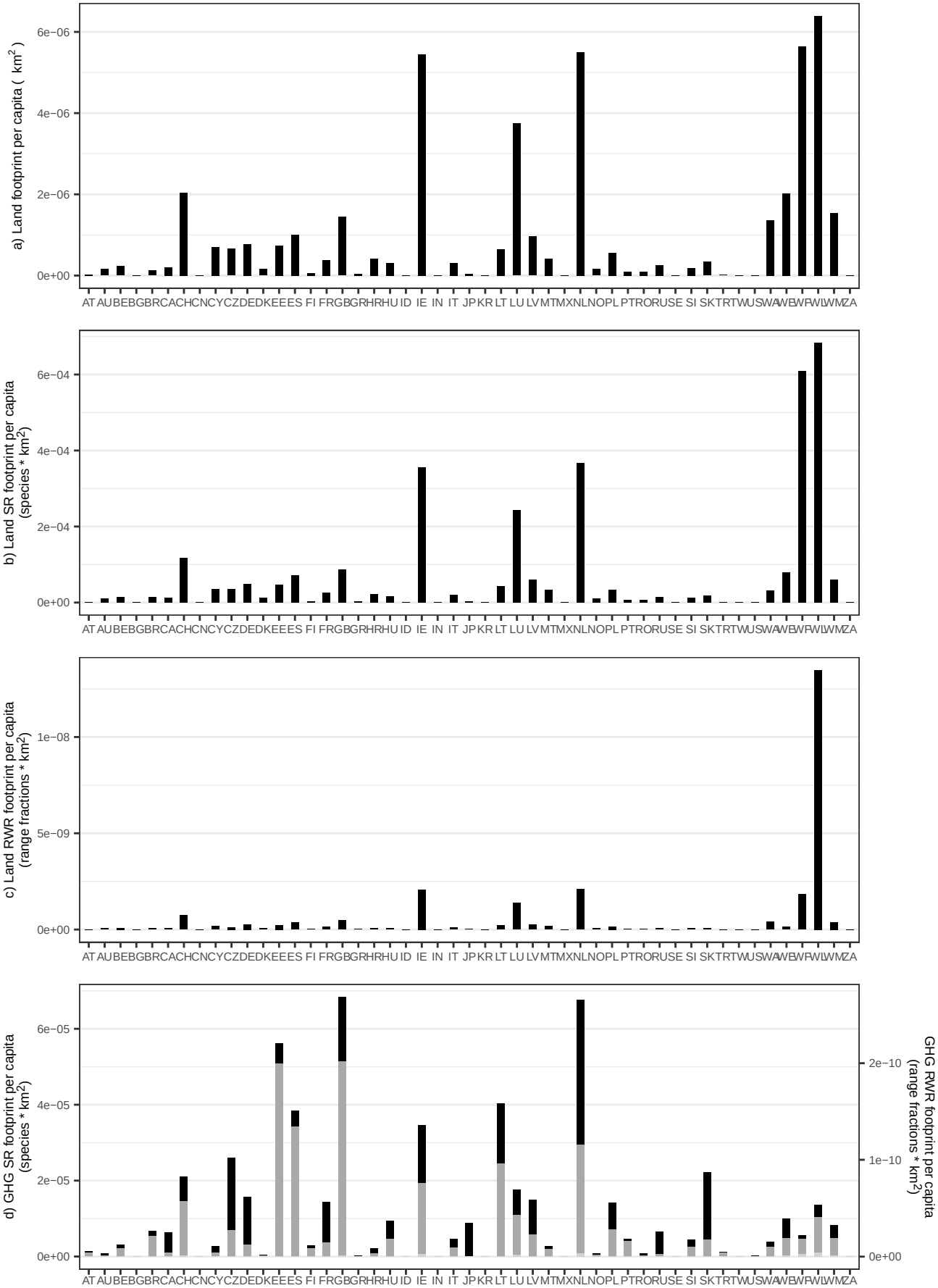
Supplementary Figure 8.

Food waste: biogasification and land application: consumption-based footprints per capita

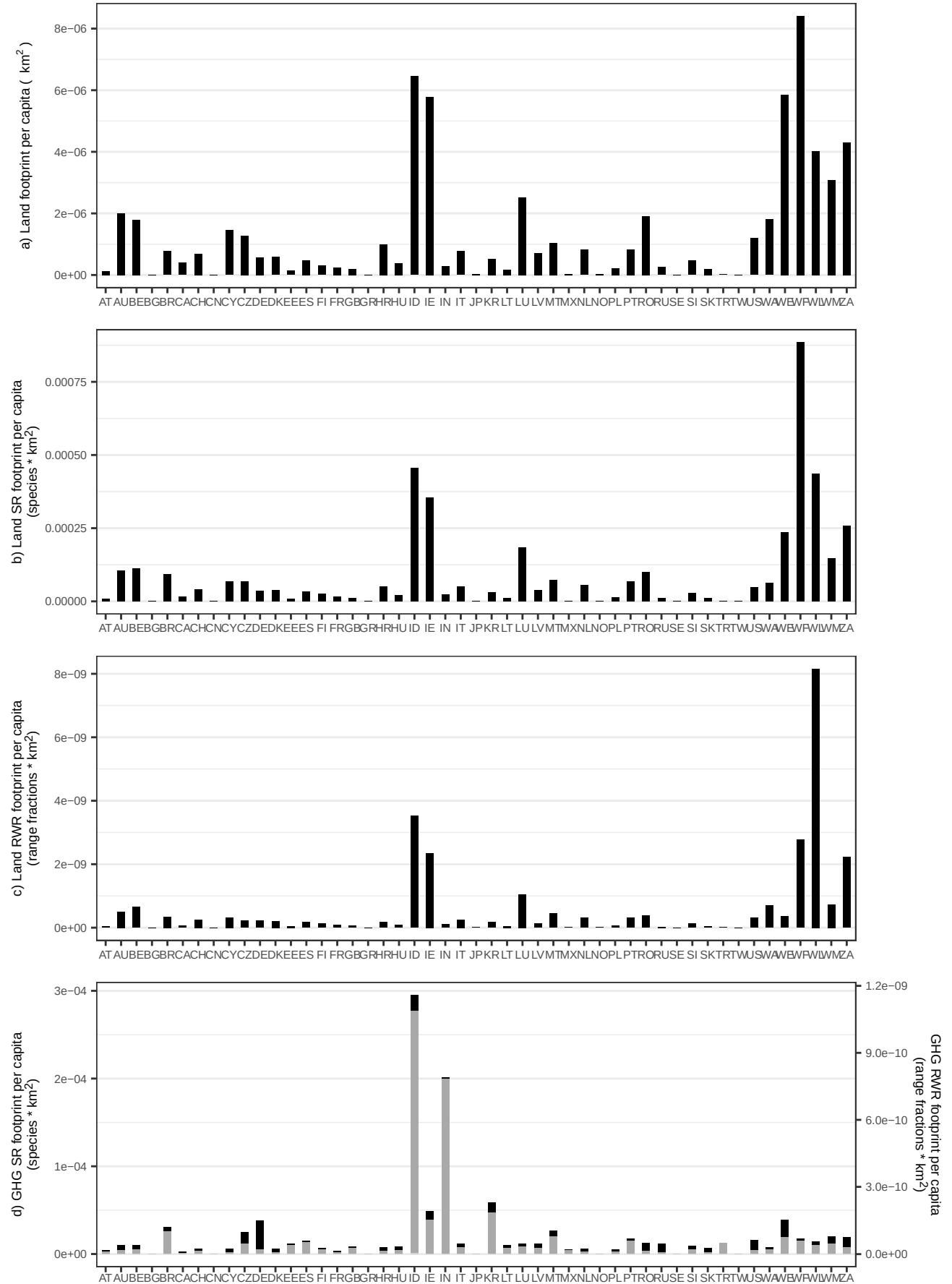


Supplementary Figure 8.

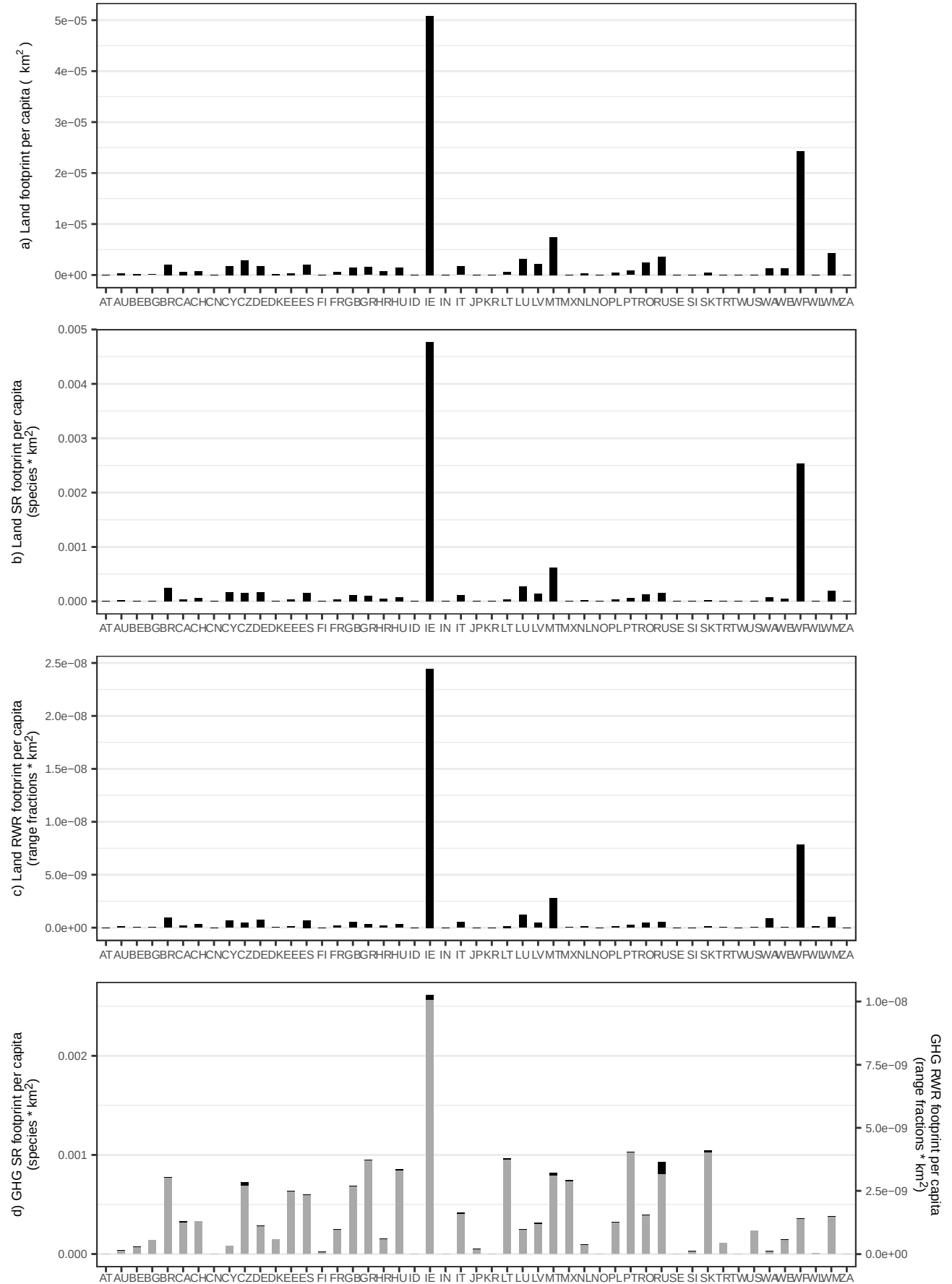
Food waste: composting and land application: consumption-based footprints per capita



Supplementary Figure 8.
Food waste: waste water treatment: consumption-based footprints per capita

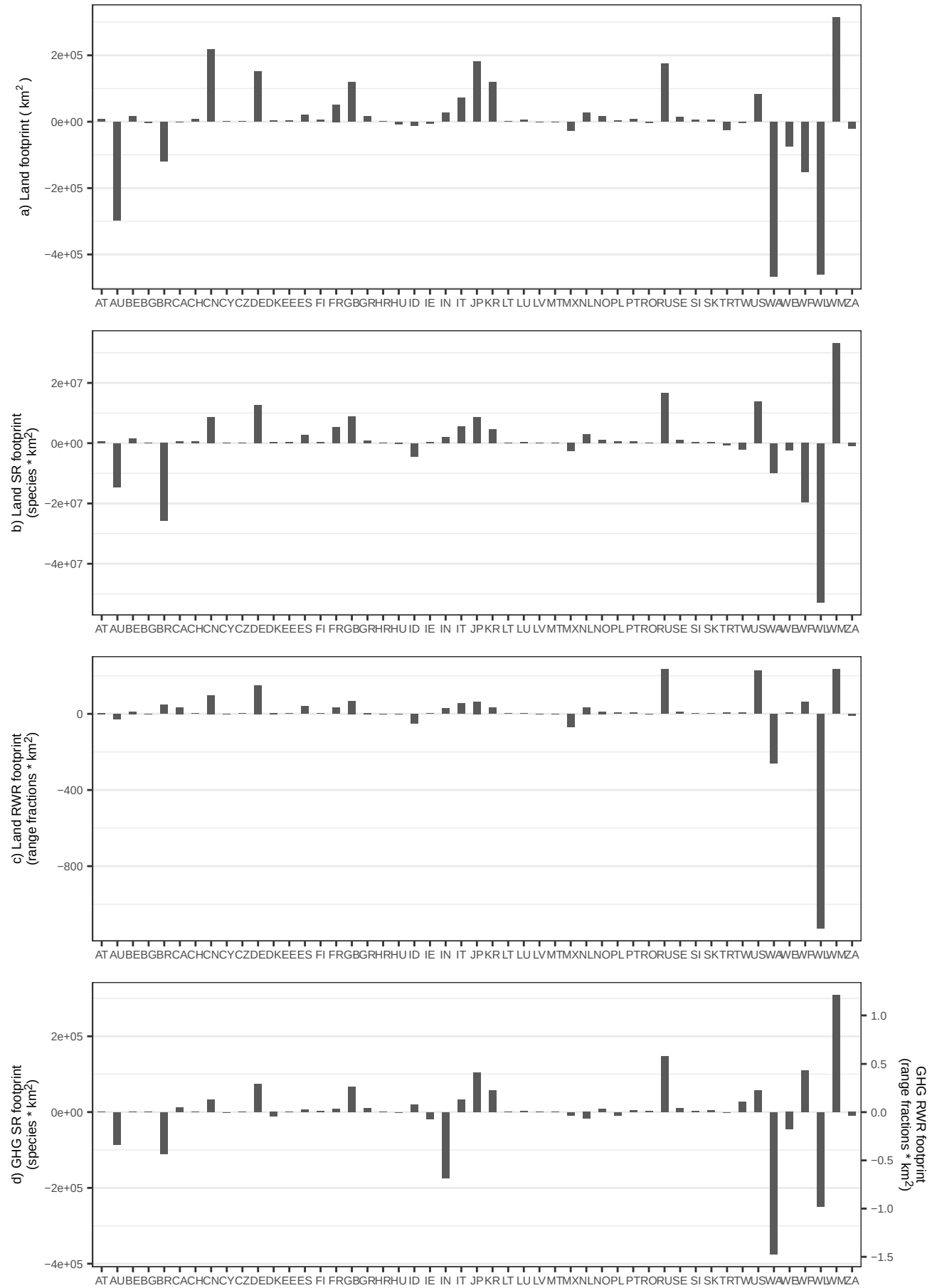


Supplementary Figure 8.
Food waste: landfill: consumption-based footprints per capita



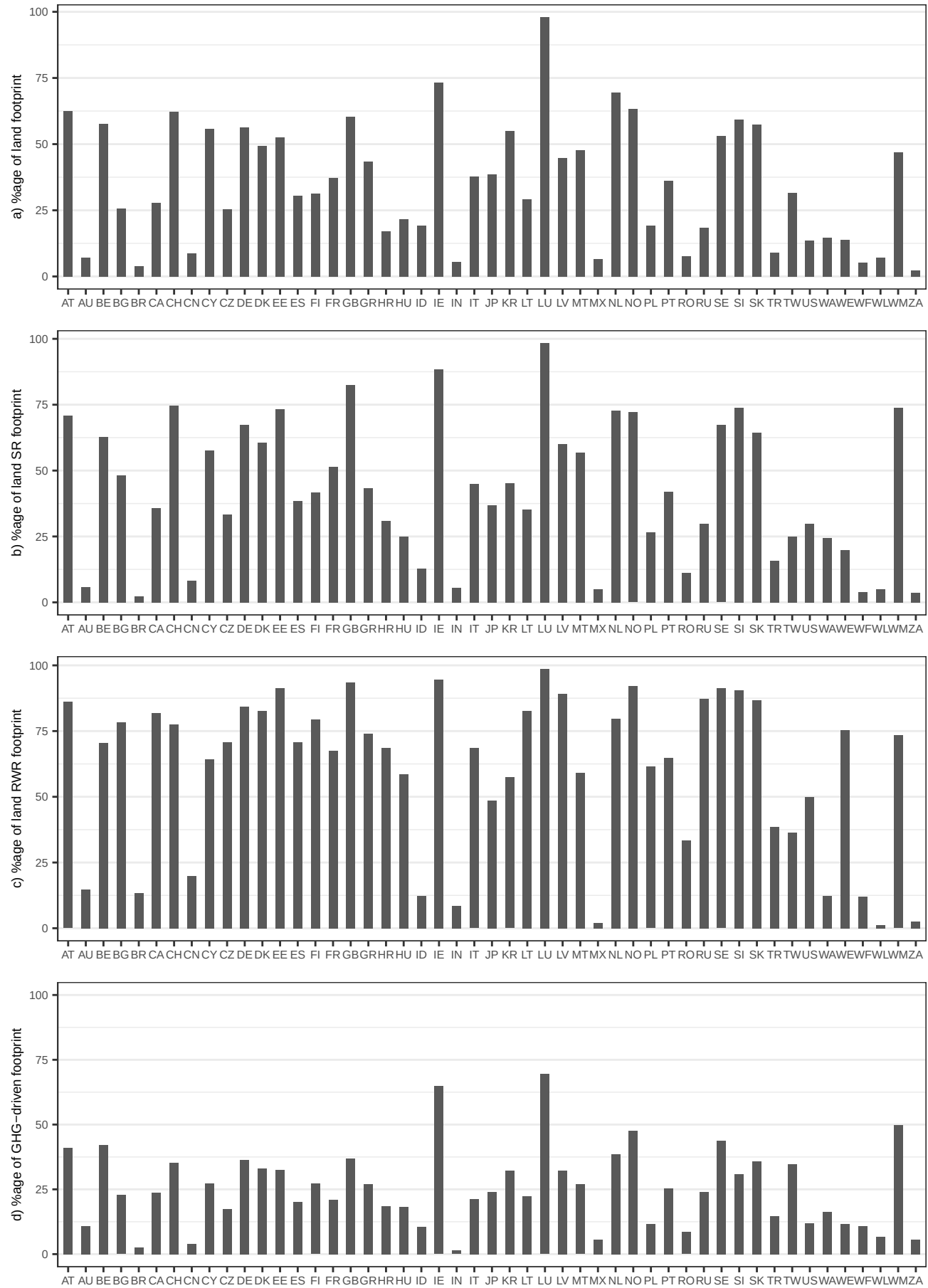
Supplementary Figure 8. Regions' consumption per capita footprints of food-related products. Consumption-based footprints for individual products for each EXIOBASE region for a) land use, b) land-driven species richness loss, c) land-driven rarity-weighted richness loss and d) GHG-driven biodiversity loss (right-hand axis – species richness; left-hand axis – rarity-weighted richness). Results are for the year 2011. AT = Austria, AU = Australia, BE = Belgium, BG = Bulgaria, BR = Brazil, CA = Canada, CH = Switzerland, CN = China, CY = Cyprus, CZ = Czech Republic, DE = Germany, DK = Denmark, EE = Estonia, ES = Spain, FI = Finland, GB = United Kingdom, HR = Croatia, HU = Hungary, ID = Indonesia, IE = Ireland, IN = India, IT = Italy, JP = Japan, KR = South Korea, LT = Lithuania, LV = Latvia, MT = Malta, MX = Mexico, NL = Netherlands, NO = Norway, PL = Poland, PT = Portugal, RO = Romania, RU = Russia, SE = Sweden, SI = Slovenia, SK = Slovakia, TR = Turkey, TW = Taiwan, US = United States, WA = RoW Asia & Pacific, WE = RoW Europe, WF = RoW Africa, WL = RoW CS America, WM = RoW Middle East, ZA = South Africa. Source data are provided in Supplementary Data 10.

Supplementary Figure 9. Net imported footprint: total food-related products



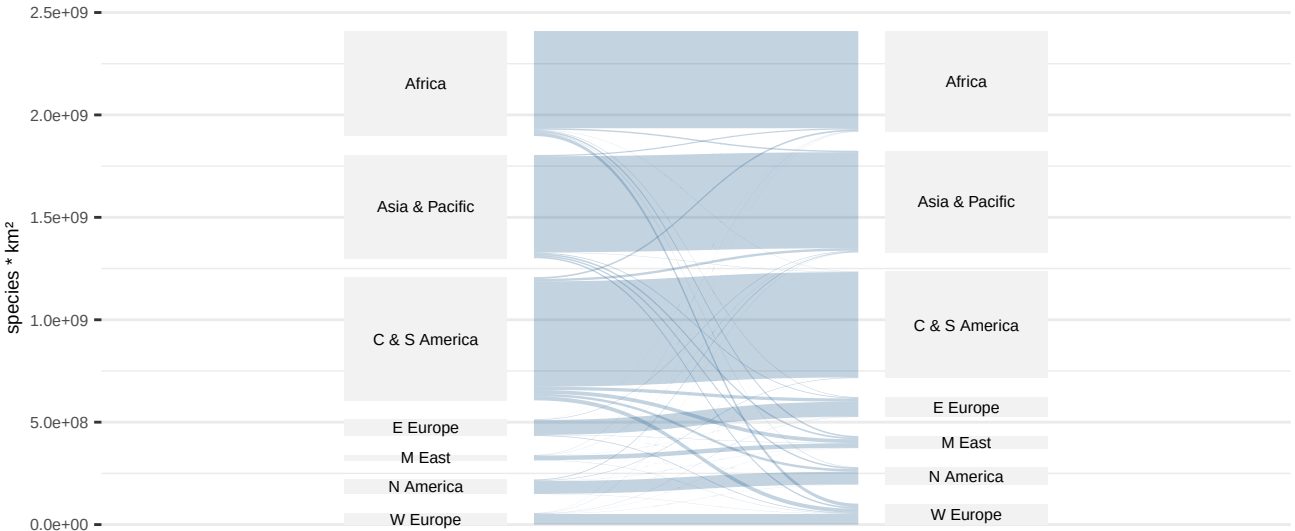
Supplementary Figure 9. Regions' net import/export footprints. Regions' net import/export footprints for a) land use, b) land-driven species richness loss, c) land-driven rarity-weighted richness loss and d) GHG-driven richness loss. Results are for the year 2011. AT = Austria, AU = Australia, BE = Belgium, BG = Bulgaria, BR = Brazil, CA = Canada, CH = Switzerland, CN = China, CY = Cyprus, CZ = Czech Republic, DE = Germany, DK = Denmark, EE = Estonia, ES = Spain, FI = Finland, GB = United Kingdom, HR = Croatia, HU = Hungary, ID = Indonesia, IE = Ireland, IN = India, IT = Italy, JP = Japan, KR = South Korea, LT = Lithuania, LV = Latvia, MT = Malta, MX = Mexico, NL = Netherlands, NO = Norway, PL = Poland, PT = Portugal, RO = Romania, RU = Russia, SE = Sweden, SI = Slovenia, SK = Slovakia, TR = Turkey, TW = Taiwan, US = United States, WA = RoW Asia & Pacific, WE = RoW Europe, WF = RoW Africa, WL = RoW CS America, WM = RoW Middle East, ZA = South Africa. Source data are provided as a source data file.

Supplementary Figure 10. Percentage imported footprint: total food-related products

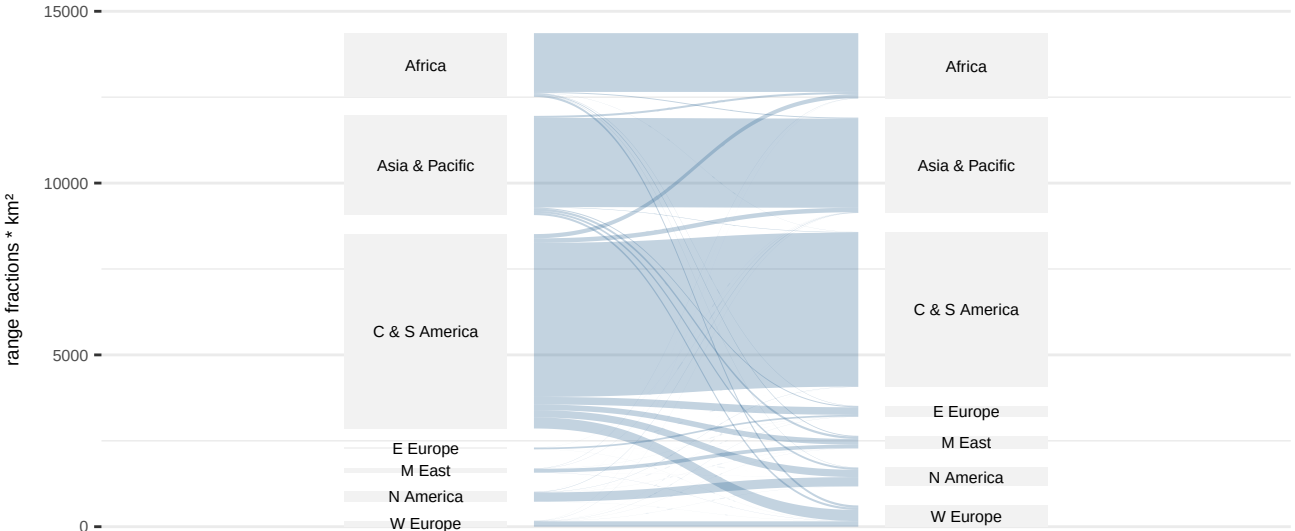


Supplementary Figure 10. The percentage of a region's footprint that is imported. The percentage of a region's footprint that is imported for a) land use, b) land-driven species richness loss, c) land-driven rarity-weighted richness loss and d) GHG-driven richness loss. Results are for the year 2011. AT = Austria, AU = Australia, BE = Belgium, BG = Bulgaria, BR = Brazil, CA = Canada, CH = Switzerland, CN = China, CY = Cyprus, CZ = Czech Republic, DE = Germany, DK = Denmark, EE = Estonia, ES = Spain, FI = Finland, GB = United Kingdom, HR = Croatia, HU = Hungary, ID = Indonesia, IE = Ireland, IN = India, IT = Italy, JP = Japan, KR = South Korea, LT = Lithuania, LV = Latvia, MT = Malta, MX = Mexico, NL = Netherlands, NO = Norway, PL = Poland, PT = Portugal, RO = Romania, RU = Russia, SE = Sweden, SI = Slovenia, SK = Slovakia, TR = Turkey, TW = Taiwan, US = United States, WA = RoW Asia & Pacific, WE = RoW Europe, WF = RoW Africa, WL = RoW CS America, WM = RoW Middle East, ZA = South Africa. Source data are provided as a source data file.

Supplementary Figure 11
a



b

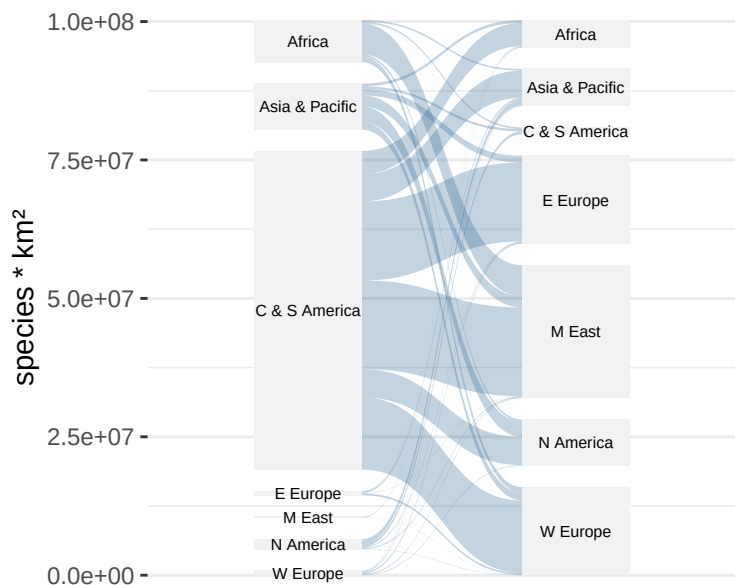


c

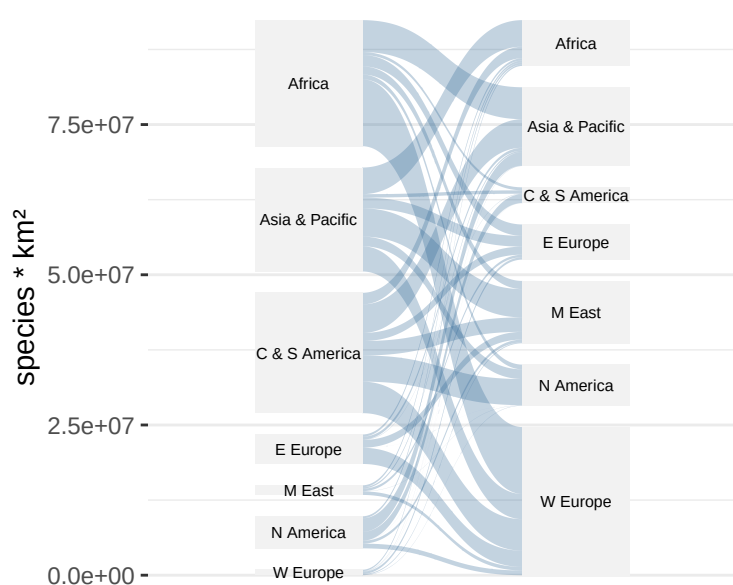


Supplementary Figure 11. Biodiversity impacts embodied within all traded food-related products. Biodiversity impacts embodied within all food-related products for three footprint types for the year 2011: a) land-driven species richness, b) land-driven rarity-weighted richness, c) GHG-driven species richness. The left axis shows the aggregated world region where the impacts are produced and the right axis shows the aggregated world region in which they are consumed. The width of the flows shows the magnitude of the footprints. Domestic production and consumption impacts are included. Source data are provided as a source data file.

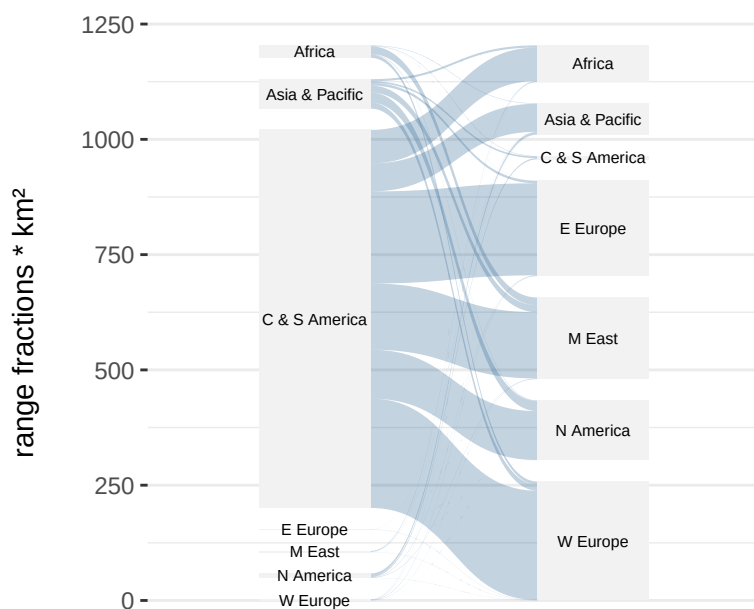
a



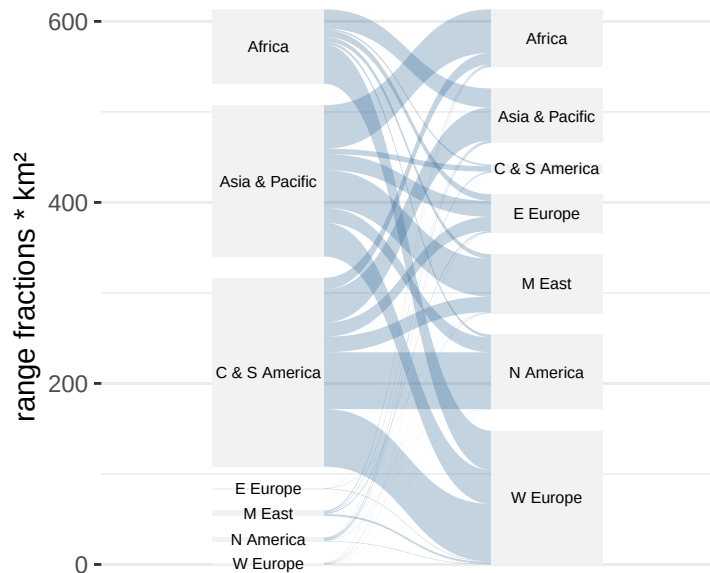
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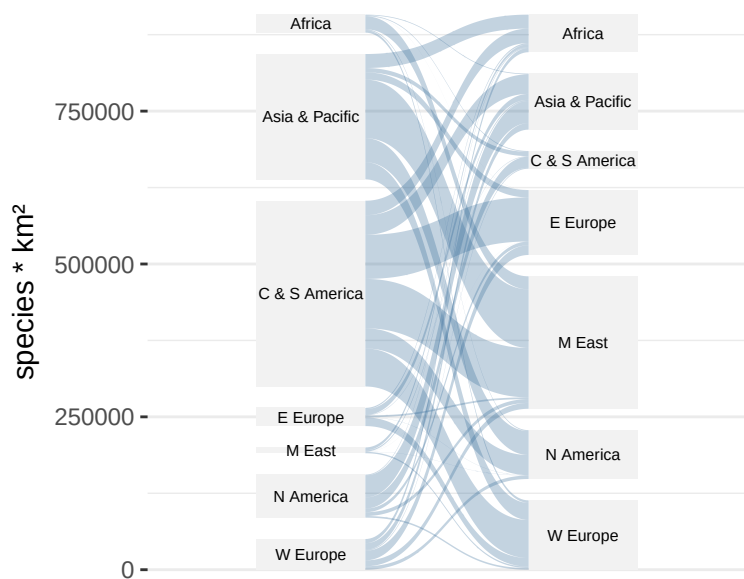
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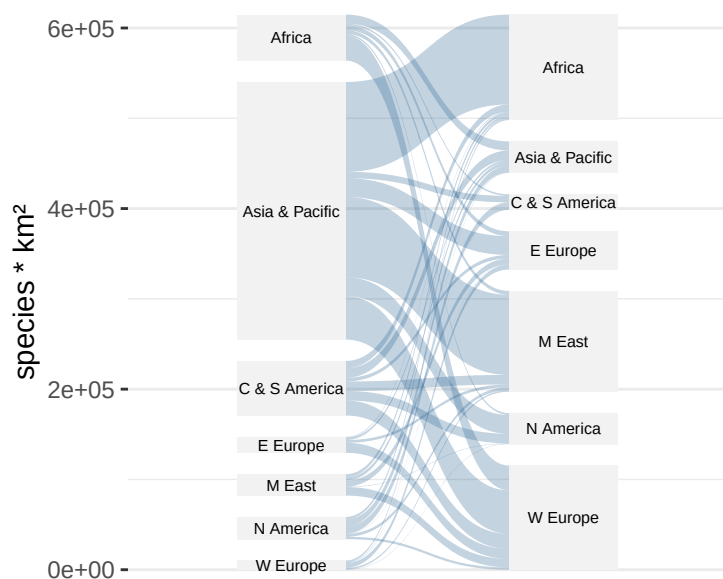
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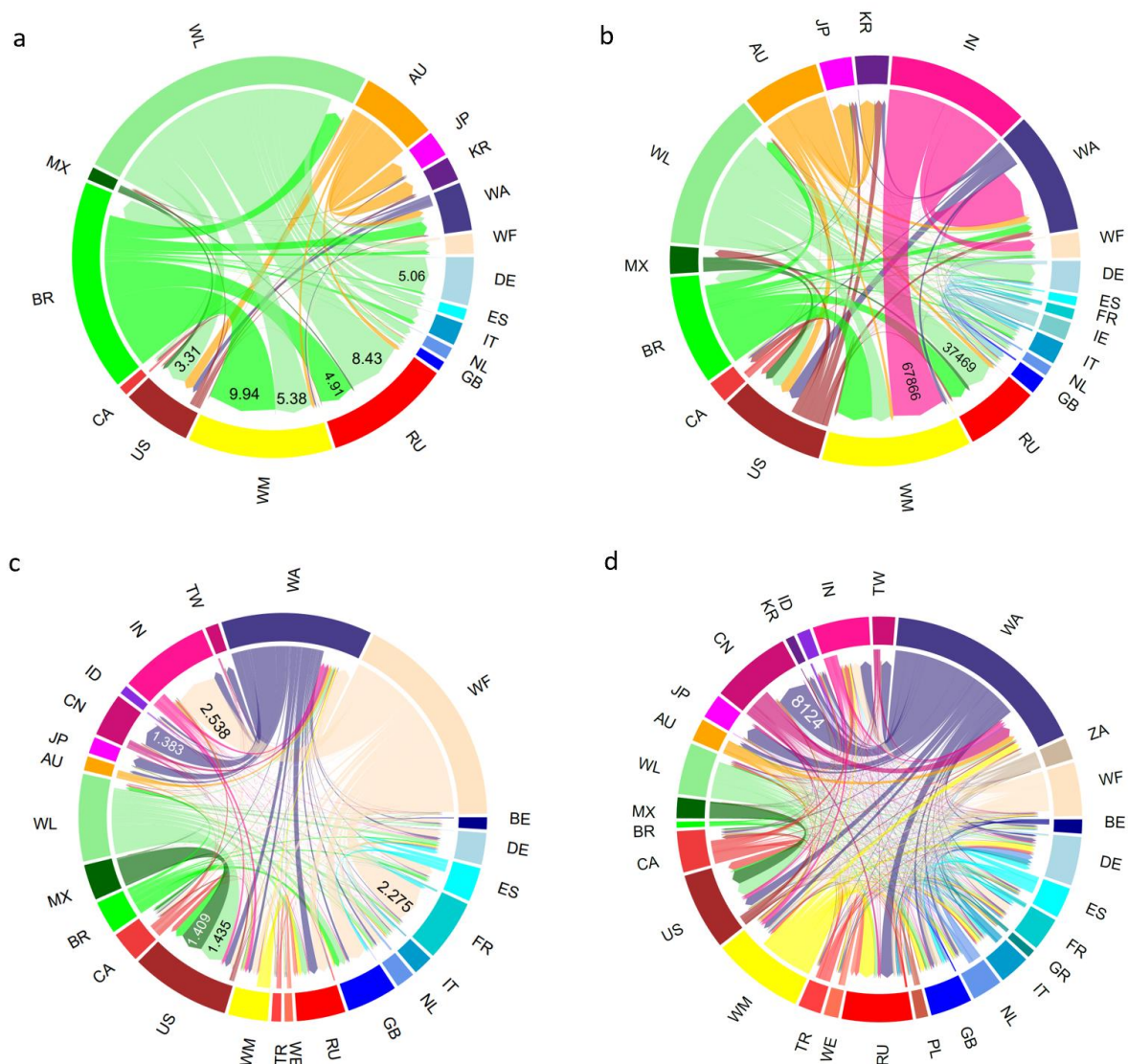
e



f



Supplementary Figure 12. Biodiversity impacts embodied within trade between aggregated world regions. Biodiversity impacts embodied within trade between aggregated world regions for three footprint types for the year 2011: a, b) land-driven species richness, c, d) land-driven rarity-weighted richness, e, f) GHG-driven species richness for a, c, e) animal-derived food products and b, d, f) plant-derived food products. The left axis shows the world region where the impacts are produced and the right axis shows the world region in which they are consumed. The width of the flows shows the magnitude of the footprints. The impacts shown here represent 8%, 17%, 15%, 21%, 6%, 12% respectively of the total impact. It can be seen that CS America dominates the export of animal-derived, land-driven biodiversity footprints whereas the traded land-driven biodiversity footprint of plant-derived products is more evenly distributed across the tropics. Western Europe is the greatest importer of biodiversity loss embedded in crops, with the largest fraction of its species richness footprint coming from Africa but the largest fraction of its rarity-weighted footprint from CS America. The total animal-derived import/export biodiversity footprint is higher than its plant-derived counterpart for land-driven footprints but lower for the GHG-driven biodiversity footprint. Source data are provided as a source data file.



Supplementary Figure 13. Biodiversity footprints embedded within the trade in processed cattle products and fruit/vegetables/nuts. Embedded land-driven species richness loss (a, c) and GHG-driven species richness loss (b, d) footprint in the international trade of processed cattle products (a, b) and fruit/vegetables/nuts (c, d), two of the products with the highest footprints for both land-driven and GHG-driven biodiversity loss. Land-driven footprints are in units of species $\times$ km² $\times 10^6$; GHG-driven footprints are in units of species $\times$ km². Regions with footprints < 0.5% of the circumference of the plot are not shown. Results are for the year 2011. AU = Australia; BE = Belgium, BR = Brazil, CA = Canada, CN = China, DE = Germany, DK = Denmark, ES = Spain, FR = France, ID = Indonesia, IE = Ireland, IN = India, IT = Italy, JP = Japan, MX = Mexico, NL = the Netherlands, GB = United Kingdom, GR = Greece, KR = South Korea, PL = Poland, RU = Russia, SE = Sweden, TR = Turkey, TW = Taiwan, US = United States, WA = RoW Asia, WE = RoW Europe, WF = RoW Africa, WL = RoW Central & South America, WM = RoW Middle East, ZA = South Africa. Source data are provided as a source data file.

The trade flows show two striking differences. (1) The origin of the land-driven species richness footprint differs between cattle products and fruit/vegetables/nuts. Almost half of the land-driven species richness loss from cattle products is exported from only three regions: RoW CS America, Brazil and Australia. In contrast, the land-driven biodiversity footprint of fruit/vegetables/nuts comes

from a wider spread of regions, with the largest share originating from RoW Africa. (2) Cattle products' land-driven and GHG-driven species richness footprint patterns differ markedly, with GHG-driven export footprints being more evenly distributed across regions than land-driven. The United States, India and W European regions, for example, have very low land-driven export footprints but sizable GHG-driven export footprints. The difference between India's land-driven and GHG-driven export footprints are particularly striking.

A similar theme is seen for fruit/vegetables/nuts with W European regions mostly importing land-driven species richness footprint, but both importing and exporting GHG-driven richness footprint.

Supplementary Tables

Supplementary Table 1. The food-related products included in the analysis. The EXIOBASE product code is given in brackets in the first column. Examples of products are given in the second column.

EXIOBASE product name	Examples of products
Paddy rice (PARI)	Paddy rice
Wheat (WHEA)	Wheat
Cereal grains nec (OCER)	Maize, barley, rye
Vegetables, fruit, nuts (FVEG)	Onions, avocados, cashews in shell
Oil seeds (OILS)	Soy beans, ground nuts, palm nuts and kernels
Sugar cane, sugar beet (SUGB)	Sugar cane, sugar beet
Crops nec (OTCR)	Potatoes, yams, coffee, tea, cocoa, spices
Cattle (CATL)	Live cattle, buffalo, other bovine animals
Pigs (PIGS)	Live pigs
Poultry (PLTR)	Live chickens, turkeys, geese, ducks
Meat animals nec (OMEA)	Live goats, sheep, horses
Animal products nec (OANP)	Eggs, honey, snails
Raw milk (MILK)	Raw milk of cattle, sheep, goats, camels
Fish and other fishing products; services incidental of fishing (FISH)	Live fish not for human consumption, fish live fresh or chilled for human consumption, crustaceans, marine molluscs, aquatic invertebrates
Chemical and fertilizer minerals, salt and other mining and quarrying products nec (CHMF)	Natural calcium phosphate, sodium chloride in sea water
Products of meat cattle (PCAT)	Meat of bovine animals – fresh, chilled, frozen, smoked
Products of meat pigs (PPIG)	Meat of pigs – fresh, chilled, frozen, smoked
Products of meat poultry (PPLT)	Meat of poultry – fresh, chilled, frozen, smoked
Meat products nec (POME)	Meat of sheep, goats, horses – fresh, chilled, frozen, smoked
Products of vegetable oils and fats (VOIL)	Lard, fish-liver oil, palm oil, olive oil, coconut oil
Dairy products (DAIR)	Processed liquid milk, cheese, yoghurt, butter, ice-cream
Processed rice (RICE)	Semi or wholly milled rice, husked rice
Sugar (SUGR)	Cane sugar, beet sugar, refined sugar
Food products nec (OFOD)	Frozen vegetables, fruit juices, wheat flour, soups, sandwiches, pizza, plant-based meat substitutes, chocolate, pasta, roasted coffee, processed spices
Beverages (BEVR)	Soft drinks, bottled water, wine, beer
Fish products (FSHP)	Frozen or dried fish, crustaceans and molluscs
N-fertiliser (NFER)	Urea, ammonium sulphate, ammonium nitrate
P- and other fertilizer (PFER)	Superphosphate, potassium chloride, ammonium chloride

Food waste for treatment: incineration (INCF)	
Food waste for treatment: biogasification and land application (BIOF)	
Food waste for treatment: composting and land application (COMF)	
Food waste for treatment: waste water treatment (WASF)	
Food waste for treatment: landfill (LANF)	

Supplementary Table 2. The aggregation of The Nature Conservancy's biomes¹⁴ into the broad biome groupings used in the models of biodiversity responses to land use.

Biome – The Nature Conservancy	Biome – PREDICTS
Tropical and subtropical moist broadleaf forests	Tropical forest
Tropical and subtropical dry broadleaf forests	Tropical forest
Tropical and subtropical coniferous forests	Tropical forest
Temperate broadleaf and mixed forests	Temperate forest
Temperate conifer forests	Temperate forest
Boreal forests/taiga	Temperate forest
Tropical and subtropical grasslands, savannas and shrublands	Tropical grasslands
Temperate grasslands, savannas and shrublands	Temperate grasslands
Montane grasslands and shrublands	Temperate grasslands
Mediterranean forests, woodlands and scrub	Mediterranean
Deserts and xeric shrublands	Drylands
Tundra	Excluded
Mangroves	Excluded
Flooded grasslands and savannas	Excluded
Inland water	Excluded
Rock and ice	Excluded

Supplementary Table 3. The allocation of crops and pasture to EXIOBASE's and PREDICTS' land-use categories. Maps were taken from SPAM¹ with the exception of those with an asterisk which came from EarthStat². In some instances, crops are allocated to more than one EXIOBASE land-use category.

Crop/pasture	Land use category – EXIOBASE	Land use category – PREDICTS
Wheat	Cropland – Paddy rice	Harvested agriculture
Rice	Cropland – Wheat	Harvested agriculture

Maize	Cropland – Other cereals nec	Harvested agriculture
Barley	Cropland – Other cereals nec	Harvested agriculture
Pearl millet	Cropland – Other cereals nec	Harvested agriculture
Small millet	Cropland – Other cereals nec	Harvested agriculture
Sorghum	Cropland – Other cereals nec	Harvested agriculture
Other cereals	Cropland – Other cereals nec	Harvested agriculture
Potato	Cropland – Other crops nec	Harvested agriculture
Sweet potato	Cropland – Other crops nec	Harvested agriculture
Yams	Cropland – Other crops nec	Harvested agriculture
Cassava	Cropland – Other crops nec	Harvested agriculture
Other roots	Cropland – Other crops nec	Harvested agriculture
Bean	Cropland – Vegetables, fruit, nuts; Cropland – Other crops nec	Harvested agriculture
Chickpea	Cropland – Vegetables, fruit, nuts	Harvested agriculture
Cow pea	Cropland – Vegetables, fruit, nuts	Harvested agriculture
Pigeon pea	Cropland – Vegetables, fruit, nuts	Harvested agriculture
Lentil	Cropland – Vegetables, fruit, nuts	Harvested agriculture
Other pulses	Cropland – Vegetables, fruit, nuts	Harvested agriculture
Soy bean	Cropland – Oil seeds	Harvested agriculture
Groundnut	Cropland – Oil seeds	Harvested agriculture
Coconut	Cropland – Oil seeds	Harvested agriculture
Banana	Cropland – Vegetables, fruit, nuts	Harvested agriculture
Plantain	Cropland – Vegetables, fruit, nuts	Harvested agriculture
Tropical fruit	Cropland – Vegetables, fruit, nuts	Harvested agriculture
Temperate fruit	Cropland – Vegetables, fruit, nuts	Harvested agriculture
Vegetables	Cropland – Vegetables, fruit, nuts	Harvested agriculture
Oil palm	Cropland – Oil seeds	Harvested agriculture
Sunflower	Cropland – Oil seeds	Harvested agriculture
Rape seed	Cropland – Oil seeds	Harvested agriculture
Sesame seed	Cropland – Oil seeds	Harvested agriculture
Other oil crops	Cropland – Oil seeds	Harvested agriculture
Sugar cane	Cropland – Sugar cane, sugar beet	Harvested agriculture

Sugar beet	Cropland – Sugar cane, sugar beet	Harvested agriculture
Cotton	Cropland – Oil seeds; Cropland – Plant-based fibers	Harvested agriculture
Other fibre crops	Cropland – Plant-based fibers	Harvested agriculture
Arabica coffee	Cropland – Other crops nec	Harvested agriculture
Robusta coffee	Cropland – Other crops nec	Harvested agriculture
Cocoa	Cropland – Other crops nec	Harvested agriculture
Tea	Cropland – Other crops nec	Harvested agriculture
Rest of Crops	Cropland – Other crops nec	Harvested agriculture
Alfalfa*	Cropland – Fodder crops cattle; Cropland – Fodder crops meat animals; Cropland – fodder crops pigs; Cropland – fodder crops poultry; Cropland – fodder crops raw milk	Harvested agriculture
Beet for forage*	Cropland – Fodder crops cattle; Cropland – Fodder crops meat animals; Cropland – fodder crops pigs; Cropland – fodder crops poultry; Cropland – fodder crops raw milk	Harvested agriculture
Cabbage for forage*	Cropland – Fodder crops cattle; Cropland – Fodder crops meat animals; Cropland – fodder crops pigs; Cropland – fodder crops poultry; Cropland – fodder crops raw milk	Harvested agriculture
Carrot for forage*	Cropland – Fodder crops cattle; Cropland – Fodder crops meat animals; Cropland – fodder crops pigs; Cropland – fodder crops poultry; Cropland – fodder crops raw milk	Harvested agriculture
Clover*	Cropland – Fodder crops cattle; Cropland – Fodder crops meat animals; Cropland – fodder crops pigs; Cropland – fodder crops poultry; Cropland – fodder crops raw milk	Harvested agriculture
Forage nec*	Cropland – Fodder crops cattle; Cropland – Fodder crops meat animals; Cropland – fodder crops pigs; Cropland – fodder	Harvested agriculture

	crops poultry; Cropland – fodder crops raw milk	
Grass nec*	Cropland – Fodder crops cattle; Cropland – Fodder crops meat animals; Cropland – fodder crops pigs; Cropland – fodder crops poultry; Cropland – fodder crops raw milk	Harvested agriculture
Maize for forage*	Cropland – Fodder crops cattle; Cropland – Fodder crops meat animals; Cropland – fodder crops pigs; Cropland – fodder crops poultry; Cropland – fodder crops raw milk	Harvested agriculture
Mixed grass*	Cropland – Fodder crops cattle; Cropland – Fodder crops meat animals; Cropland – fodder crops pigs; Cropland – fodder crops poultry; Cropland – fodder crops raw milk	Harvested agriculture
Oil seed for forage*	Cropland – Fodder crops cattle; Cropland – Fodder crops meat animals; Cropland – fodder crops pigs; Cropland – fodder crops poultry; Cropland – fodder crops raw milk	Harvested agriculture
Rye for forage*	Cropland – Fodder crops cattle; Cropland – Fodder crops meat animals; Cropland – fodder crops pigs; Cropland – fodder crops poultry; Cropland – fodder crops raw milk	Harvested agriculture
Sorghum for forage*	Cropland – Fodder crops cattle; Cropland – Fodder crops meat animals; Cropland – fodder crops pigs; Cropland – fodder crops poultry; Cropland – fodder crops raw milk	Harvested agriculture
Swede for forage*	Cropland – Fodder crops cattle; Cropland – Fodder crops meat animals; Cropland – fodder crops pigs; Cropland – fodder crops poultry; Cropland – fodder crops raw milk	Harvested agriculture

Turnip for forage*	Cropland – Fodder crops cattle; Cropland – Fodder crops meat animals; Cropland – fodder crops pigs; Cropland – fodder crops poultry; Cropland – fodder crops raw milk	Harvested agriculture
Vegetable for forage*	Cropland – Fodder crops cattle; Cropland – Fodder crops meat animals; Cropland – fodder crops pigs; Cropland – fodder crops poultry; Cropland – fodder crops raw milk	Harvested agriculture
Pasture*	Permanent pastures – grazing cattle; Permanent pastures – grazing meat animals nec; Permanent pastures – grazing raw milk	Livestock pasture

Supplementary Table 4. Modelled differences in total sampled species richness among land-use types. All values are expressed as a percentage change relative to primary vegetation as the baseline. Numbers in parentheses are the lower and upper bounds of the 95% confidence limits

Biome	Primary vegetation	Secondary vegetation	Harvested agriculture	Livestock pasture	Urban area
Tropical forest	0 (-10.3 11.5)	-14.2 (-23.1 -4.2)	-31.1 (-38.5 -23.2)	-34.3 (-41.7 -26.2)	-25.8 (-40.0 -8.2)
Tropical grassland	0 (-12.3 14.0)	-11.7 (-23.1 1.3)	-17.0 (-27.2 -5.3)	-36.6 (-45.0 -26.9)	-23.7 (-55.5 30.9)
Drylands	0 (-26.9 36.8)	12.8 (-14.6 48.9)	-11.8 (-36.2 21.9)	11.6 (-14.9 46.5)	-6.9 (-35.9 35.1)
Mediterranean	0 (-12.8 14.7)	-43.2 (-50.6 -34.9)	-31.3 (-40.8 -20.3)	-21.0 (-33.0 -6.9)	-54.8 (-62.9 -44.9)
Temperate/boreal forest	0 (-10.7 12.0)	-1.4 (-12.1 10.7)	-20.3 (-29.4 -10.1)	-2.4 (-13.1 9.6)	-11.3 (-22.3 1.2)
Temperate/montane grasslands	0 (-11.0 12.4)	2.4 (-10.5 17.3)	-27.9 (-36.2 -18.4)	-17.0 (-26.1 -6.7)	-31.6 (-45.2 -14.4)

Supplementary Table 5. Modelled differences in total sampled rarity-weighted species richness among land-use types. All values are expressed as a percentage change relative to primary vegetation as the baseline. Numbers in parentheses are the lower and upper bounds of the 95% confidence limits

Biome	Primary vegetation	Secondary vegetation	Harvested agriculture	Livestock pasture	Urban area
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Tropical forest	0 (-16.9 20.3)	-18.8 (-32.6 -2.2)	-43.5 (-53.4 -31.4)	-56.4 (-64.6 -46.4)	-43.9 (-64.3 -12.1)
Tropical grassland	0 (-20.3 25.5)	-21.1 (-38.4 0.9)	-26.9 (-41.6 -8.5)	-58.7 (-67.1 -48.0)	-70.0 (-83.7 44.0)
Drylands	0 (-52.2 109.1)	-54.7 (-77.5 -8.9)	-23.3 (-69.0 89.8)	-69.4 (-84.1 -41.0)	19.0 (-55.4 217.3)
Mediterranean	0 (-32.4 48.0)	-19.8 (-41.7 10.2)	2.5 (-26.2 42.6)	-23.9 (-47.1 9.7)	-59.1 (-74.1 -35.3)
Temperate/boreal forest	0 (-18.0 22.0)	14.3 (-5.5 38.2)	-12.3 (-28.0 6.7)	2.5 (-16.4 25.8)	-33.3 (-47.4 -15.6)
Temperate/montane grasslands	0 (-26.6 36.3)	-35.3 (-58.6 0.3)	-61.5 (-73.0 -44.9)	-58.0 (-70.9 -39.4)	-69.5 (-83.7 -42.9)

Supplementary Table 6. The ten highest per-capita consumption-based footprints at a product-region level for the different footprint types. (The values of GHG-driven rarity-weighted richness are not given since they are directly proportional to the values of the GHG-driven species richness footprints). Each cell shows the consumption-based footprint per capita, the food product category and the region of consumption. CATL = cattle, DAIR = dairy products, FVEG = fruit/vegetables/nuts, MILK = raw milk, OFOD = food product nec (e.g. soups, pizza, sauces), OILS = oil seeds, OTCR = crops nec (e.g. coffee, cocoa, spices), PCAT = products of meat cattle.

Land use (km ²)	Land-driven species richness (species x km ²)	Land-driven rarity-weighted richness (range fractions x km ²)	GHG-driven species richness (species x km ²)
4.69E-03, PCAT, US	1.15, PCAT, Australia	4.26E-06, PCAT, Australia	1.02E-02, PCAT, Australia
4.50E-03, DAIR, Australia	6.97E-01, PCAT, Brazil	2.71E-06, PCAT, RoW Central & South America	4.50E-03, DAIR, Australia
4.24E-03, OFOD, Australia	3.72E-01, CATL, Brazil	2.11E-06, PCAT, Brazil	3.21E-03, PCAT, Brazil
3.74E-03, PCAT, Brazil	2.47E-01, FVEG, Luxembourg	2.05E-06, OFOD, RoW Central & South America	2.57E-03, OFOD, RoW Central & South America
3.54E-03, MILK, South Africa	2.36E-01, OTCR, Luxembourg	2.03E-06, OFOD, Mexico	2.55E-03, CATL, Luxembourg
3.43E-03, OFOD, Russia	2.16E-01, OFOD, Australia	2.02E-06, MILK, RoW Central & South America	2.36E-03, PCAT, RoW Central & South America
3.09E-03, PCAT, South Africa	2.05E-01, OFOD, Taiwan	1.99E-06, MILK, South Africa	2.07E-03, OFOD, Netherlands
3.04E-03, FVEG, Luxembourg	2.03E-01, DAIR, Brazil	1.89E-06, FVEG, Luxembourg	2.02E-03, OFOD, Australia

2.85E-03, OILS, Luxembourg	1.85E-01, OFOD, Russia	1.78E-06, OTCR, Luxembourg	1.91E-03, OFOD, Belgium
2.85E-03, CATL, Australia	1.81E-01, DAIR, Australia	1.74E-06, PCAT, South Africa	1.89E-03, OTCR, Luxembourg

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