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The measurement of dietary species richness reveals that a higher consumption of dietary fibre, fish, fruits and vegetables, is associated with greater food biodiversity in UK diets

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Abstract

Objective: We determined whether dietary species richness (DSR) (i) can be robustly measured using 4-day food intake data, (ii) is dependent on socio-demographic characteristics and (iii) is associated with diet quality. Design: The National Diet and Nutrition Survey (NDNS) nutrient databank 2018-2019 was expanded to include FoodEx2 food classifications, ingredients, the number and identity of unique species, Nutrient Rich Food 8-3 (NRF 8-3) Index scores and greenhouse gas emissions. Four-day food intake data and socio-demographic variables were used to calculate diet quality and DSR on the food and diet level. Setting: The United Kingdom (UK). Participants: Participants from NDNS 9-11 (2016-2019). Results: Composite dishes had the highest DSR (median 8 (Q1 = 4, Q3 = 12)), followed by seasoning, sauces and condiments (median 7, (Q1 = 4, Q3 = 10)) and, grains and grain-based products (median 5, (Q1 = 2, Q3 = 10)) Q3 = 7)). Median DSR over 4 days was 49 (Q1 = 43, Q3 = 56; range 14–92), with the first 2 days achieving 80 % of DSR measured over 4 days. DSR was significantly higher in those who were younger, those with a higher household income or those with a lower level of deprivation (all P < 0.001). Higher DSR was associated with a small but significant improvement in nutritional quality (P < 0.001). Also, adherence to dietary guidelines such as fibre, fruits and vegetables and fish was associated with significantly higher DSR (all P < 0.001). Conclusions: We successfully established DSR based on 4-day food intake data. We also identified opportunities to improve DSR by increasing the consumption of fruits, vegetables, fibre and fish.

Diet diversity is a key element of healthy diets. A wide variety of foods, between and within food groups, is associated with an increased intake of essential nutrients and bioactive components, helping to meet micronutrient requirements and a lowered risk of mortality and diet-related non-communicable diseases⁽¹⁾. Typically, dietary diversity has been calculated by counting food groups consumed over a given period and used as a measure of diet quality in especially low and middle-income countries⁽²⁾. However, food group scores do not necessarily capture the variability in species and nutrients across diets, and more recently, the concept of food biodiversity has been introduced to measure the diversity of plants, animals and other organisms (e.g. fungi, insects) used for food⁽³⁻⁵⁾. Food biodiversity has already been associated with total and cause-specific mortality across European countries⁽⁵⁾, and it is also associated with planetary health: food species biodiversity reduces pressures on single species and supports food and nutrition security in the face of anthropogenic challenges⁽⁶⁾. Agriculture is being singled out as a direct threat to 86 % of species facing extinction, primarily due to land conversion, particularly for industrial mono-crop and animal agriculture⁽⁷⁾. Estimates are that we have access to 300 000 edible plant species, but that around half of the dietary calories we consume globally are met by only four crops: rice, potatoes, wheat and maize, whilst these are beef, wheat, pork and potato in Europe⁽⁸⁾. Consequently, current food systems accelerate biodiversity loss and malnutrition⁽⁹⁾.

The measurement of food biodiversity can be divided into three main components: richness, evenness and disparity⁽⁴⁾. Richness considers the total number of distinct edible species consumed over a specific period⁽⁴⁾, evenness considers the number of unique species and the evenness of their quantities in a diet, measured as the probability that two randomly selected food items belong to the same species⁽¹⁰⁾ and disparity measures the 'entropy of disorder' or the differences in the functional traits or ecological roles of the species in the diet⁽¹¹⁾. All of these indices of diversity measure different features of a diet. Dietary species richness (DSR) has been highlighted as a novel, comprehensive and simple metric for the simultaneous measurement of food biodiversity between and within food groups and the nutritional quality of human diets, capturing both agricultural and wild food biodiversity⁽¹²⁾. Recently, DSR was inversely



associated with total and cause-specific mortality within the European Prospective Investigation into Cancer and Nutrition (EPIC) cohort. This effect was independent of socio-demographic, lifestyle and other known dietary risk factors⁽⁵⁾. Such findings advocate for food biodiversity to be included in public health strategies as a new metric for healthy and sustainable foods and diets, linking the fields of ecological, agricultural and food biodiversity with human and planetary health outcomes. However, how DSR can be best measured, the exact relationship between DSR and diet quality, and what DSR score is required for optimal health, are currently unknown. Such information is required to explore meaningful opportunities for increasing food biodiversity in our diets.

Using National Diet and Nutrition Survey (NDNS) data, we quantified food biodiversity for foods and diets consumed in the UK using the DSR metric. We assessed (i) whether a 4-day food diary is appropriate to capture DSR, (ii) whether there are variations in DSR over 4 days for different consumer segments based on sex, age group, BMI, ethnicity, Index of Multiple Deprivation (IMD), household income and adherence to different dietary guidelines and (iii) whether DSR is associated with diet quality.

Methods

Database development

An expanded version of the NDNS nutrient databank 2018-2019 was used in this analysis. This databank included nutritional composition, level of processing (NOVA categories)⁽¹³⁾, greenhouse gas (GHG) emissions (farm-to-fork)^(14,15) and current price (September 2023 retail prices without adjusting for inflation), for nearly 6000 commonly consumed foods and drinks in the $UK^{(16-19)}$. The Nutrient Rich Food 8.3 (NRF8.3) index scores were also calculated for all NDNS nutrient databank items on a per 418 kJ (100 kcal) basis^(20,21). NRF index scores are diet quality indices based on the nutrient density of each food item, accounting for beneficial nutrients, nutrients to limit or a combination of both⁽¹⁷⁾. In addition, each item was categorised according to the European Food Safety Authority's (EFSA) FoodEx2 food classification, which is a comprehensive food classification and description system designed to standardise how food is described and classified across different food safety domains⁽²²⁾. To calculate DSR, the ingredients for each food and drink item, including composite dishes, were identified using online data from a single major UK retailer⁽²³⁾. Ingredients for homemade dishes were obtained from the BBC Good Food website⁽²⁴⁾. In the first instance, the list of unique species was taken from the food biodiversity codes assigned to the EPIC cohort's food list⁽⁵⁾ and expanded with twenty-one additional species, resulting in 269 unique species (see online supplementary material, Supplementary Table 1). From this, 216 unique species were found to be present in the foods and drinks of the NDNS nutrient databank. Generic species names and synonyms were produced for each of these species. We then matched the list of species and synonyms to the ingredients list of the NDNS nutrient databank using an R algorithm whilst manually checking for inconsistencies. This approach yielded the total number and type of unique species (i.e. DSR) for each of the >6000 food or drink items in the NDNS nutrient databank. When considering the total list of 216 unique species for the UK database, the highest proportion was provided by fish, fish products and any other marine and freshwater food products (29%), fruit (24%) and vegetable species (22%).

NDNS analysis

Data were obtained from the annual rolling cross-sectional UK National Diet and Nutrition Survey (NDNS) waves 9-11, which comprise data gathered between 2016–2017 and 2018–2019⁽²⁵⁾. A 4-day estimated food diary was used, where participants were asked to keep a record of foods and drinks consumed over four consecutive days⁽²⁵⁾. Only data from participants who completed 3 or 4 d were included in this analysis. We used socio-demographic data on age, sex, BMI, ethnicity, household income and IMD, a widely used metric to classify the relative deprivation of small areas in the UK⁽²⁶⁾, which are provided in the NDNS. In addition, in the absence of a validated healthy diet index for UK Diets, we used NRF 8.3 index scores estimated at an individual level as a proxy of diet quality. To estimate adherence to dietary guidelines, each participant was categorised as adhering (or not) to the following guidelines: consuming less than 11 % of total energy from saturated fats⁽²⁷⁾, less than 5 % of total energy from free sugars⁽²⁸⁾, less than 70 g of red meat per day⁽²⁹⁾, less than 6 g of salt per day⁽³⁰⁾, more than 30 g of fibre per day⁽²⁸⁾, more than 400 g of fruits and vegetables per day⁽³¹⁾ or more than 280 g of fish a week⁽³²⁾. Moreover, it was estimated that ultra-processed foods account for almost 60 % of total energy intake in UK diets⁽³³⁾. Based on this, we categorised those consuming ≥ 60 % of total energy from processed or ultra-processed foods as 'high processed food eaters'.

Statistical analysis

For the food-level analysis, Shapiro–Wilk tests were conducted to check for the normality of data in the NDNS nutrient databank. Because of non-normality, median values were used to plot the distributions. Correlations between DSR and different food characteristics (e.g. energy density, GHG emissions and current price) were evaluated using Spearman correlations and confidence intervals (95 %), adjusted for multiple testing using a Bonferroni method.

For the individual diet-level analysis, to estimate the DSR, we considered the absolute number of unique biological species consumed per day per person, across foods and drinks and across the one, two, three and four food diary days. We included spices, extracts and flavourings in foods, considering they are common in composite dishes and that bioactives can have benefits to health even if consumed in small amounts^(34,35). However, we excluded extracts and flavourings coded or presented in an unknown nomenclature. Differences in DSR across 4 d and between socio-demographic characteristics were tested with Kruskal–Wallis tests and adjusted for multiple testing using a Bonferroni method.

To estimate if adherence to different nutritional guidelines is associated with DSR, simple and multiple linear regression models were fitted using adherence as the predictor variable and DSR as an outcome. Only those socio-demographic variables that showed statistically significant different DSRs across categories were included in the regression models. Finally, to evaluate the association of DSR with the nutritional quality of individual diets, simple and multiple regressions were fitted using DSR as a predictor, and average NRF8.3 index scores as an outcome. Linearity and residual distributions were visually assessed, and collinearity (through tolerance level to ensure variables were not closely related) was evaluated before modelling the regressions. F-statistics were used to test the significance of each term included in the regression models. Analysis was done using R (version 4.3.3), with the following libraries: "*dplyr*", "ggpubr", "reshape2" and "ggstatsplot"⁽³⁶⁾. "ggplot2", "tidyverse",



Figure 1. Median Dietary Species Richness for food groups including food and drinks in the NDNS nutrient databank.

Results

Across EFSA FoodEx2 food groups, and on a per food item basis, composite dishes had the highest DSR (median of 8 ($Q_1 = 4$, $Q_3 = 12$), followed by seasoning, sauces and condiments (median of 7, $(Q_1 = 4, Q_3 = 10))$ and grains and grain-based products (median of 5, $(Q_1 = 2, Q_3 = 7)$), meat and dairy substitutes (median of 4, $(Q_1 = 3, Q_3 = 7)$) and confectionery including chocolate (median DSR of 4, $(Q_1 = 2, Q_3 = 6)$). Food products in the other groups had a median DSR of 1 or 2 (Fig. 1). Within the food group of composite dishes, meat-based dishes had the highest DSR (median of 12, $(Q_1 = 7, Q_3 = 14)$), followed by legumes-based dishes (median of 10, $(Q_1 = 7, Q_3 = 12)$) and vegetable-based dishes (median of 9, $(Q_1 = 7, Q_3 = 15)$). Within the food group of seasoning, sauces and condiments, savoury sauces had the highest DSR (median of 8, $(Q_1 = 6, Q_3 \ 10)$). Within the food group of grains and grain-based products, cereal bars had the highest DSR (median of 10, $(Q_1 = 5, Q_3, 12)$) (see online supplementary material, Supplementary Fig. 1).

Across all foods, foods with a higher DSR had a higher energy density ($\rho = 0.19$, *P*-value <0.001), a higher price ($\rho = 0.35$, *P*-value <0.001) and higher GHG emissions ($\rho = 0.28$, *P*-value <0.001) (Fig. 2(a)). However, such correlations appeared food group specific. For example, within the food group of composite dishes, foods with a higher DSR had a lower energy density ($\rho = -0.13$, P < 0.001), higher GHG emissions ($\rho = 0.20$, P < 0.001) and a higher price ($\rho = 0.24$, P < 0.001) (Fig. 2(b)). Within the food group of sauces, seasonings and condiments, foods with a higher DSR had a lower price (-0.19, P = 0.03) (Fig. 2(c)). Within the food group of grains and grain based-dishes, foods with a higher DSR had a significantly higher energy density ($\rho = 0.15$, P < 0.001), higher GHG emissions ($\rho = 0.39$, P < 0.001) and a higher price ($\rho = 0.41$, P < 0.001) (Fig. 2(d)).

The median DSR measured over 4 days was 49 ($Q_1 = 43$, $Q_3 = 56$; range 14–92). The median daily DSR, measured across 3558 participants, was 29 ($Q_1 = 24$, $Q_3 = 35$) for the first reporting day, with a further nine additional unique species ($Q_1 = 6$, $Q_3 = 13$) for the second reporting day, a further five additional

unique species $(Q_1 = 3, Q_3 = 8)$ for the third reporting day and a further four additional unique species $(Q_1 = 2, Q_3 = 6)$ for the fourth reporting day (Fig. 3). The first 2 days covered 80 % of the median DSR measured over 4 days. Initial analysis revealed that DSR over the four reporting days significantly differed between age categories (children, adolescents, adults and elders; P < 0.001), household income (low, middle and high tertile; P < 0.001), IMD (the most deprived living, 1, to the least deprived areas, 5; P < 0.001) and marital status (single, married, civil partnership, separated or divorced, widow; P < 0.001), but not between sexes, ethnic groups (White, Mixed, Black or Black British, Asian or Asian British, other groups; P = 0.123), BMI categories (underweight, normal, overweight, obesity and morbid obesity; P = 0.382) or sex (female or male, P = 0.438). Differences, if any, were determined after the first day of measuring food intake (see online supplementary material, Supplementary Table 2).

When fitting a simple regression analysis, the variables, age, household income and IMD each explained a significant amount of variance in DSR, with lower age, higher household incomes and those in less deprived IMD categories predicted to have a significantly (all P < 0.001) higher median DSR over 4 days (Table 1). All 'adherence to dietary recommendation' variables also explained significant variance in median DSR over 4 days. Those consuming more than 30 g of fibre per day, more than 280 g of fish a week or at least 400 g of fruits and vegetables per day were predicted to have a significantly (all P < 0.001) higher median DSR over 4 days of 7, 7 and 3, respectively. Those complying with all the healthy dietary guidelines were also predicted to have a higher median DSR over 4 days of 6; however, because of low numbers (only 5 out of 3558 participants achieved this), this was not statistically significant (Table 1, see online supplementary material, Supplementary Table 3). On the other hand, the median DSR of those consuming less than 5 % of total energy from free sugars or those consuming less than 6 g of salt per day over 4 days were predicted to be 5 and 3 lower, respectively. Also, the median DSR of those consuming less than 60 % of total energy from processed or ultra-process foods was predicted to be 2 lower. In both simple and multiple regression models, age, household income, IMD and all



Figure 2. Association between Dietary Species Richness and energy density, greenhouse gas emissions and price, across foods and within the food groups composite dishes, seasoning, sauces, condiments and grains and grain-based products. All analyses were adjusted using the Bonferroni method. ρ = Spearman correlation coefficient; GBP = Great British Pounds price in September 2023, CO2e = gCO2-equivalents.



Figure 3. Median Dietary Species Richness for each of the four food intake assessment days.

dietary quality variables were still significant and hence included in the analysis. Overall, these predictor variables explained significant variance in median DSR over 4 days (Table 1).

When fitting regression analysis to evaluate the association of DSR with the nutritional quality (measured through NRF 8·3 index scores per 100 kcal) of individual diets, we found that DSR could predict a small but significant increase in NRF8·3 index scores per 100 kcal (estimate 168·9, SE 8·01, *t*-value 21·07, P < 0.001), which remained significant after adjusting for age, household income and IMD (estimate 159·3, SE 8·71, *t*-value 18·13, P < 0.001) (see online supplementary material, Supplementary Fig. 2). This means that with every unit increase in DSR, the average NRF8·3 index score per 100 kcal would increase to a small extent, showcasing a better nutritional quality.

Discussion

Here, we show that DSR can be robustly measured using NDNS 4day food intake data and that DSR can be used as an indicator of food biodiversity in UK diets. We identified 216 relevant, unique species across foods, food groups and diets. Median DSR was 49 over 4 days, with 80 % of the unique species consumed over 4 days being captured in the first 2 days of recall. DSR was significantly higher in those who were younger, those who had a higher household income and those residing in the least deprived areas, but DSR did not differ between sex, ethnic groups or BMI categories. Composite dishes, especially meat-based dishes, mainly contributed to DSR in UK diets, followed by seasoning, sauces and condiments and grains and grain-based products. Adherence to different dietary guidelines, especially in relation to fibre, fruits and vegetables and fish consumption, was also associated with a significantly higher DSR. However, those with a lower consumption of dietary saturated fats, free sugars and processed or ultraprocessed foods were predicted to have a significantly lower DSR. A higher DSR was generally associated with a higher diet quality, but the effect estimate was small.

Food biodiversity is a relatively novel concept, and there is currently no standardised methodology to calculate DSR. Most

Table 1. Regression models to identify significant sociodemographic and dietary quality predictor variables for DSR

Predictor variables	п	Estimate	SE	<i>t</i> -value	Р			
Total number of participants	3558							
Simple linear regression models using cut-offs based on dietary guidelines								
Sociodemographic predictor variables								
Age (y)	-	-0.025	0.007	-3.604	<0.001			
Household income (tertiles)	_	3.445	0.220	15.61	<0.001			
IMD (categories)	-	1.544	0.120	12.84	<0.001			
Adherence to dietary recommendations for								
Saturated fats	1291	-0.847	0.366	-2.337	0.019			
Free sugars	416	-4-663	0.543	-8.577	<0.001			
Low consumption of processed or ultra-process foods	1634	-1.836	2.551	-5.205	<0.001			
Red meat	2660	0.984	0.406	2.425	0.0154			
Salt	2946	-2.84	0.465	-6.11	<0.001			
Fibre	1898	7.033	0.335	21.09	<0.001			
Fruits and vegetables	235	6.673	0.701	9.509	<0.001			
Fish	898	2.796	0.568	4.921	<0.001			
Meeting all the above dietary quality variables	5	6.460	4.710	1.379	0.168			
Multiple linear regression models using cut-offs based on di Multiple R-squared: 0·2223, adjusted R-squared: 0·2196, F-sta	nd 3063 DF, <i>P</i> < 0∙001							
Socio-demographic predictor variables								
Age	-	-0.064	0.007	-8-639	<0.001			
Household Income	-	2.24	0.219	10.201	<0.001			
IMD	-	0.781	0.125	6-242	<0.001			
Adherence to dietary recommendations for								
Saturated fats	1291	-1.483	0.353	-4.200	<0.001			
Free sugars	416	-4-432	0.531	-8.347	<0.001			
Low consumption of processed or ultra-process foods	1634	-0.967	0.336	-2.877	<0.001			
Red meat	2660	1.478	0.403	3.661	<0.001			
Salt	2946	-1.203	0.477	-2.522	0.011			
Fibre	1898	5.737	0.372	15-410	<0.001			
Fruits and vegetables	235	5.013	0.706	7.096	<0.001			
Fish	898	1.753	0.560	3.129	0.001			

Four-day dietary data was used to perform the regression analysis. People categorised as IMD 1 represent 20% of the most deprived, while those categorised as IMD 5 are 20% of the least deprived. Household income was categorised into tertiles; those in the first tertile group had the lowest income, and those in the third tertile group had the highest incomes. Cut-offs for dietary quality predictors follow national dietary guidelines; see the methods section for more details.

studies assessing DSR have been conducted in low- and middleincome countries. These studies used 24-hour diet recalls or ecological assessments to identify DSR across diets (ranging from 40 to 234), with DSR or animal protein species richness ranging from 8 to 24, or from 44 to 52, respectively, depending on whether DSR or animal protein species richness was assessed during the dry or wet season in some of the studies^(12,37,38). An analysis of DSR in the EPIC cohort found a median DSR of 68 per person per year, calculated from FFQ, out of a list of 248 unique species⁽⁵⁾. In our study, using NDNS survey data, we found a median DSR of 49 from four daily food diaries on consecutive days, including weekdays and a weekend day. Whilst DSR would have increased with more recording days, we found that 80 % of the unique species consumed were reported within the first 2 days of recording food intake, suggesting that a 2-day food diary captured the majority of species diversity when calculating DSR from 4-day NDNS food diaries. However, more days over different periods may be required for low- and middle-income countries where dietary diversity depends more on seasons^(12,37,38). The higher value for DSR in the study by Hanley-Cook⁽⁵⁾, compared with our study, may reflect a much wider geographical sampling area, and the FFQ covering a much longer period of dietary intake which would have captured a higher number of less frequently consumed food items. On the other hand, that study was based on an observational cohort using FFQ, and exposure misclassification and residual confounding cannot be ruled out⁽⁵⁾.

It is essential to understand how food biodiversity is associated with dietary quality and health outcomes to justify its use as a meaningful new metric that can link diets to human and planetary health, complementing existing indicators for healthy and sustainable diets. Associations between dietary diversity indicators and health outcomes, such as body weight and non-communicable diseases, have been largely inconsistent⁽²⁾. However, Hanley-Cook⁽⁵⁾ showed, in the largest study of its kind, that DSR was inversely associated with total mortality and mortality due to cancer, heart disease, digestive disease and respiratory disease, independent of socio-demographic, lifestyle and other known dietary risk factors such as intake of energy, meat and fibre. This suggests that DSR may provide health benefits beyond dietary quality alone. Absolute death rates among participants in the highest and lowest fifth of DSR were 65.4 and 69.3 cases/10,000 person-years, respectively (hazard ratio and 95 % CI: 0.63 (0.59, 0.66)), providing a powerful association between low DSR and disease outcomes across nine European countries⁽⁵⁾. In another study, DSR was linked to a higher diet quality (e.g. mean adequacies of vitamin A, vitamin C, folate, Ca, Fe and Zn) and a higher diet diversity score in women and young children in rural areas from seven low- and middle-income countries, in both the wet and the dry $season^{(12)}$. Our study explored the link between DSR and dietary quality using the nationally representative UK NDNS database. We found that DSR was driven by composite dishes on the food level. Furthermore, we found that those consuming at least 30 g of fibre per day, at least 400 g of fruits and vegetables per day or more than 280 g of fish a week, all representing important dietary guidelines, were predicted to have a significantly higher DSR. The highest proportion of species was provided by different fish, fish products and any other marine and freshwater food products (29 %), fruit (24 %) and vegetable species (22%). These categories also showed the largest amount of species diversity, indicating opportunities for increasing DSR in existing products through reformulation.

Increasing the DSR of our diets would arguably improve human health and benefit the environment, linking in-farm and on-plate biodiversity⁽³⁾ and serving an ecological as well as societal role. Moreover, diversifying consumption of crops, fruits, livestock and aquatic species would strengthen nutrition security. This is particularly important in low- and middle-income countries, where higher agricultural biodiversity has been associated with more diverse diets through subsistence and income-generating pathways. Greater crop species richness has been associated with small but positive increments in child height for age outcomes⁽¹¹⁾. In our study, a higher DSR was associated with a higher value for the NRF8.3 index score, but the estimate for change in diet quality was relatively small. In contrast, adhering to current dietary guidelines for fruits and vegetables and fish consumption indicated an effective approach to increasing DSR. Together, this implies that the number of species alone does not necessarily predict dietary quality, but that the level of intake of unique species from 'healthier' food groups, such as fruit and vegetables and fish, as stipulated in the dietary guidelines, also plays an important role. This is an important finding, especially in the UK, where adherence to dietary guidelines is low - only 26 % and 17 % of the UK population adhere to recommendations for fruits and vegetables and oily fish, respectively. Intermediate to high adherence to Eat Well Guide recommendations, especially those for fruit and vegetable consumption, has been associated with a 10% reduction in the risk of mortality and a lower carbon footprint⁽³⁹⁾.

Recently, it was established that aquatic species richness in the ocean is critical for the ecosystem's multifunctionality and provides significant nutritional benefits in relation to recommended nutrient intakes for humans, as nutrient concentrations vary substantially across aquatic food species⁽⁴⁰⁾. Interestingly, the benefits of aquatic species richness for human consumption exceeded the diversity effects of plant and forest species richness⁽⁴⁰⁾. The diversity of the UK fruit and vegetable supply has increased significantly in the past decades, with an increased contribution of tropical fruits, but a declining contribution of more traditional vegetables, such as cabbages and carrots⁽⁴¹⁾. Currently, most of the fruits, vegetables and fish we consume in the UK are imported, but the UK's fruit and vegetable supply is increasingly dependent on imports from climate-vulnerable producing countries^(41,42). For instance, it has been reported that in the UK only 7 % of fruits are produced domestically, with the rest imported, largely (70%) from outside of Europe⁽⁴²⁾. In addition, it is estimated that meeting the recommendations for fruit and vegetables, or oily fish, is 16-17 % more expensive than the costs of an average 2000 kcal diet in the UK⁽⁴³⁾. Therefore, to achieve impact, higher DSR foods and diets must not only be within planetary boundaries, but also affordable⁽⁴⁴⁾. Indeed, our analysis showed that household income and the deprivation level of participants were strong and significant predictors of DSR in the UK diet, with those with higher incomes and least deprived having a significantly higher DSR. To increase DSR in UK diets in a just manner, we will need to ensure that we conserve natural biodiversity worldwide through our dietary choices and maintain the affordability of high-DSR foods and diets, which requires integration of environmental and public health policies^(45,46).

The most important strength of this study is the robust bottom-up analysis of DSR in UK diets, using 4-day food intake data collected from 3558 participants across sex, age, ethnicity and socio-economic groups over a recent period of 3 years. Another strength is using an in-house NDNS nutrient databank, allowing us to link nutrient composition, FoodEx2 food classification, ingredients, number and identity of unique species, GHG emissions and cost to calculate dietary quality indicators and DSR on the food and diet level. In this study, we included unique species from all foods apart from extracts and flavourings with unknown or unfamiliar nomenclature. There may be reasons to exclude certain foods and/or ingredients such as herbs, spices, flavourings and extracts, but including all foods v. excluding the lowest 5 % or 10 % species intake from each of the food groups did not substantially change the protective effect of DSR on all-cause mortality⁽⁵⁾. Limitations include our current inability to link DSR to important health outcomes using the NDNS database and the inability to measure DSR over a period of more than 4 days, which would likely have resulted in higher DSR values. Also, DSR does not assess 'evenness', which is a key component of diversity. Therefore, DSR gives rare edible species the same weight as more common species, in kcal/day. Finally, we did not correct for potential under-overreporting, as the outcome (DSR) relates to the number of unique species consumed rather than its amount. Underreporting may have led to fewer foods being reported and, therefore, to a lower DSR, which means that real DSR values may have been underestimated for some participants.

In conclusion, we established DSR in UK diets, based on fourday food intake data, as a case study for countries that have a relatively high consumption of processed foods as part of Western-style diets. We found that DSR is significantly higher in those adhering to dietary guidelines for fruit and vegetables, fish and fibre intake. DSR is also higher in younger people, those with higher household incomes and those living in the least deprived areas. Composite dishes contributed mostly to DSR and, therefore, offer an opportunity to increase DSR through food reformulation, although our data show the importance of the level of intake of unique species from 'healthier' food groups, such as fruit and vegetables and fish, as already set out in the dietary guidelines.

Supplementary material. For supplementary material accompanying this paper visit https://doi.org/10.1017/S1368980025000473

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Competing of interests. E.H.Z. and A.J.W. are employed by the Unilever Foods Innovation Centre Wageningen, which markets food products. The remaining authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

Authorship. B.R., E.H.Z. and A.J.W. formulated the research question(s) and designed the study. M.A.M., A.L. and C.F.M.G. retrieved and cleaned data. M.A.M., A.L. and B.R. analysed the data and wrote the article. All the authors revised the submitted draft.

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Supplementary Data

Supplementary Table 1. List of 269 unique species identified.

Supplementary fable 1. List 0
Abalone
Akee apples
Alfalfa
Alfonsino
Almond, Almonds
Amberjack
Anchovies
Anise seed
Apples
Apricots
Arctic char
Asian rice
Asparagus
Atlantic halibut
Atlantic mackerel
Atlantic pomfret
Atlantic salmon
Aubergine
Avocado
Bamboo
Barley
Barnacle
Basil
Beets, chards
Black eyed peas
Blackberry
Blackcurrant
Blue shark
Blue whiting
Blueberries
Bonito
Borage
Brazil nuts
Breadfruits
Broad beans
Buckwheat
Buffalo
Burr oherkin
Butternut sauashes
Button mushrooms
Caner
Cardamon Fruit
Carobs
Carrots
Cashew nut
Cassaya
Cassava

de species identified.
Celeriac, celery
Chayote fruits
Cherimoya
Chestnuts
Chicken
Chickling vetches
Chickpea
Chili peppers
Chinese cabbages, turnip
Chive
Cinnamon bark
Cloudberry
Coalfish
Cockles
Cocoa bean
Coconut
Cod
Coffee bean
Common banana, plantain
Common beans
Common dab
Common mussel
Common nottle
Common neaches
Common periwinklo
Common periwinkle
Common shrimns
Common shrimps
Common shrimps Common skate
Common shrimps Common skate Common wheat, spelt
Common shrimps Common skate Common wheat, spelt Coriander leaves
Common shrimps Common skate Common wheat, spelt Coriander leaves Courgette, pumpkin, squash
Common shrimps Common skate Common wheat, spelt Coriander leaves Courgette, pumpkin, squash Cow
Common shrimps Common skate Common wheat, spelt Coriander leaves Courgette, pumpkin, squash Cow
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Common shrimps Common skate Common wheat, spelt Coriander leaves Courgette, pumpkin, squash Cow Cowberries, lingonberries Cranberries Cucumber, gherkin
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Common shrimps Common skate Common wheat, spelt Coriander leaves Courgette, pumpkin, squash Cow Cowberries, lingonberries Cranberries Cucumber, gherkin Cumin Seeds Curly endives
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Common shrimps Common skate Common skate Coriander leaves Courgette, pumpkin, squash Cow Cowberries, lingonberries Cranberries Cranberries Cucumber, gherkin Cumin Seeds Curly endives Curly endives Cuttlefishes Date Dill Duck Durum wheat
Common shrimps Common skate Common skate Coriander leaves Courgette, pumpkin, squash Cow Cowberries, lingonberries Cranberries Cranberries Cucumber, gherkin Cumin Seeds Curly endives Curly endives Cuttlefishes Date Dill Duck Durum wheat Edible crab
Common shrimps Common skate Common skate Coriander leaves Courgette, pumpkin, squash Cow Cowberries, lingonberries Cranberries Cranberries Cucumber, gherkin Cumin Seeds Curly endives Cuttlefishes Date Date Dill Duck Durum wheat Edible crab
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Common shrimps Common skate Common skate Coriander leaves Courgette, pumpkin, squash Cow Cowberries, lingonberries Cranberries Cranberries Cucumber, gherkin Cumin Seeds Curly endives Cuttlefishes Date Date Date Dill Duck Durum wheat Edible crab European conger European eel

European freshwater
European moose
European ovster
European perch
European plaice
European sardine
European spider crab
European sprat
Fennel
Fig
Flounders
Gages
Garden snail
Garfish
Garlic
Giant tiger prawn
Gilthead seabream
Ginger
Globe artichoke, cardoon
Goat
Goiiberry
Goose
Gooseberry
Grapefruits
Grapes
Groupers
Guava
Haddock
Hakes
Hare
Hazelnut
Head cabbages
Head lettuces
Hemp seeds
Herrings
Hibiscus infusion flowers
Hijiki
Horse
Horse mackerel
Horseradish root
Jerusalem artichoke
Kaki
Kiwi
Kumquats
Leek

Lemon
Lentil
Lime
Ling
Linseed
Litchis
Loquats
Lumpfish
Macadamias
Maize
Mandarins, clementine
Mango
Marble goby
Marjoram Fruit
Medlar
Megrims
Melons
Mints
Monkfish, anglerfish
Mulherries (black and
white)
Mullets
Mung beans
Mustard seeds
Nectarines
Northern pike
Northern pike Northern prawn
Northern pike Northern prawn Norway pout
Northern pike Northern prawn Norway pout Norwegian lobster
Northern pikeNorthern prawnNorway poutNorwegian lobsterNutmeg seed
Northern pikeNorthern prawnNorway poutNorwegian lobsterNutmeg seedOat
Northern pikeNorthern prawnNorway poutNorwegian lobsterNutmeg seedOatOctopuses
Northern pikeNorthern prawnNorway poutNorwegian lobsterNutmeg seedOatOctopusesOil palms
Northern pikeNorthern prawnNorway poutNorwegian lobsterNutmeg seedOatOctopusesOil palmsOkra
Northern pikeNorthern prawnNorway poutNorwegian lobsterNutmeg seedOatOctopusesOil palmsOkraOlive
Northern pikeNorthern prawnNorway poutNorwegian lobsterNutmeg seedOatOctopusesOil palmsOkraOliveOnions
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Northern pikeNorthern prawnNorway poutNorwegian lobsterNutmeg seedOatOatOctopusesOil palmsOkraOliveOnionsOrangesOregano
Northern pikeNorthern prawnNorway poutNorwegian lobsterNutmeg seedOatOctopusesOil palmsOkraOliveOnionsOrangesOreganoPangas catfishes
Northern pikeNorthern prawnNorway poutNorwegian lobsterNutmeg seedOatOctopusesOil palmsOkraOliveOnionsOrangesOreganoPangas catfishesPanavas
Northern pikeNorthern prawnNorway poutNorwegian lobsterNutmeg seedOatOctopusesOil palmsOkraOliveOnionsOreganoPangas catfishesPapayasParsley
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Northern pikeNorthern prawnNorway poutNorwegian lobsterNutmeg seedOatOctopusesOil palmsOkraOliveOnionsOreganoPangas catfishesPapayasParsleyPartridge fresh meat
Northern pikeNorthern prawnNorway poutNorwegian lobsterNutmeg seedOatOctopusesOil palmsOkraOliveOnionsOreganoPangas catfishesPapayasParsnip rootPartridge fresh meatPassionfruit
Northern pikeNorthern prawnNorway poutNorwegian lobsterNutmeg seedOatOctopusesOil palmsOkraOliveOnionsOrangesOreganoPangas catfishesPapayasParsleyParsnip rootPartridge fresh meatPassionfruitPeanut
Northern pikeNorthern prawnNorway poutNorwegian lobsterNutmeg seedOatOctopusesOil palmsOkraOliveOnionsOreganoPangas catfishesPapayasParsnip rootPartridge fresh meatPear
Northern pikeNorthern prawnNorway poutNorwegian lobsterNutmeg seedOatOatOctopusesOil palmsOkraOliveOnionsOreganoPangas catfishesPapayasParsleyParsnip rootPartridge fresh meatPassionfruitPeanutPearl millet
Northern pikeNorthern prawnNorway poutNorwegian lobsterNutmeg seedOatOctopusesOil palmsOkraOliveOnionsOreganoPangas catfishesPapayasParsleyPartridge fresh meatPassionfruitPearPearl milletPass
Northern pikeNorthern prawnNorway poutNorwegian lobsterNutmeg seedOatOatOctopusesOil palmsOkraOliveOnionsOreganoPangas catfishesPapayasParsleyParsnip rootPartridge fresh meatPassionfruitPeanutPearl milletPeasPasapa

Peppercorn
Pheasant
Pig, boar
Pigeon
Pike-perch
Pine nuts
Pineapple
Pistachio
Plum
Pollock
Pomegranate
Poppy seeds
Potato
Prickly pear
Ptarmigan
Purple urchin
Quail
Quinces
Quinoa grain
Rabbit
Radicchio, Belgian endives
Radish, daikon
Raspberries
Razor clam
Red mustard leaves
Redcurrant
Rhubarb
Rockweed
Roe deer
Roman rocket and similar-
Rose hips
Rosemary
Rye
Safflower seeds
Saffron
Sage
Sago plant
Salsify leaves
Sapodillas
Scallops, pectens
Scorpion fish
Sea bass
Sea bream
Sea catfish
Sea lettuce
Sesame
Shea nuts
Sheep
Smalt

Smooth hounds
Snappers
Sole
Soybean
Spinach
Spiny dogfish
Squids
Strawberry
Sturgeon
Sunflower
Swede, rapeseed
Sweet cherry
Sweet corn
Sweet pepper
Sweet potato
Swordfish
Tamarind
Tarragon
Tea leaves
Thyme
Tomatoes
Trout
Tuna
Turbot
Turkey
Walnut
Watercress
Watermelon
Whelks
White lupine
Whitefish, Coregonus
Whiting
Witloofs
Wolffish
Yam
Yeast cultures

*Those species highlighted in orange were identified in the EPIC cohort's food list (See Hanley-Cook et al. 2021) but were not identified across the NDNS food and drink products. **Supplementary Figure 1.** Median Dietary Species Richness (DSR) across composite dishes (A), seasoning, sauces, and condiments (B), and grains and grain-based products (C).



Supplementary Table 2. Differences in median Dietary Species Richness (DSR) across sociodemographic characteristics.

Median DSR (IQR)						
		1	<u> </u>	ay 3	4	-
Sex	Women	29.5 (12)	9 (7)	6 (6)	4 (4)	40- 58 30- 20-
	Men	29 (11)	9 (6)	5 (5)	4 (4)	
		<i>p</i> =0.438				1 2 3 4 Food Diary Day
	Children	31 (10)	9 (6)	5 (5)	4 (4)	40-
	Adolescents	28 (10)	9 (7)	6 (6)	4 (4)	डा प्रा 30- अ डा
Age	Adults	29 (13)	9 (7)	5 (5)	4 (5)	
	Elders	29 (12)	9 (7)	5 (5)	4 (4)	
		p <0.001				1 2 3 4 Food Diary Day
	Lowest income tertile	27 (10.5)	8 (7)	5 (5)	4 (4)	40-
income	Middle income tertile	30 (12)	9 (7)	5 (5)	3 (4)	30- 20- 20- 20-
ehold	Highest income tertile	32 (11)	10 (8)	6 (6)	4 (5)	
Hous	NA	28 (10.5)	9 (7)	5 (5)	4 (4)	
		P <0.001				1 2 3 4 Food Diary Day
	1	28 (11)	8 (7)	5 (5)	3 (4)	
	2	28 (12)	9 (7)	5 (5)	3 (4)	40-1 s v
_	3	30 (11)	9 (6)	5 (5)	4 (4)	4 30- %
IMD	4	30 (11)	9 (8)	6 (6)	4 (4)	
	5	31 (12)	10 (6)	6 (5.5)	4 (5)	
		P <0.001				0- 1 2 3 4 Food Diary Day
Ethni	White	29 (11)	9 (7)	5 (5)	4 (4)	

	Mixed ethnic group	30 (10.5)	11 (9.5)	5 (5.5)	5 (4)	40- I
	Black or Black British	28 (13)	9 (7)	5 (5)	4 (4)	
	Asian or Asian British	29 (10)	8 (6)	6 (6)	3 (5)	
	Any other group	26 (13.5)	10 (7)	5 (4)	4 (6)	
	NA	34 (10)	8 (5)	3 (4)	2 (3)	Food Diary Day
		<i>p</i> =0.123				
	Single	28 (12)	9 (7)	5 (5)	4 (4)	40 I
	Married	30 (12)	9 (7)	6 (6)	4 (5)	
tatus	Civil partnership	26 (12.5)	9 (7)	5 (7)	3 (4)	
ital st	Separated/ divorced	26 (13.75)	9 (7)	5 (5)	3 (4)	
Mar	Widow	25 (11)	9 (8)	5 (6)	3 (5)	
	NA	30 (12)	9 (7)	5 (7)	3 (4)	1 2 3 4 Food Diary Day
		p<0.001				
	Underweight	30 (11)	7 (10)	5 (5)	3 (5)	
	Normal weight	29 (14)	9 (8)	6 (6)	4 (5)	
	Overweight	29 (13)	9 (7.5)	6 (5)	3 (4)	
BMI	Obesity	28 (12)	9 (6)	5 (6)	4 (4)	
	Morbid Obesity	28 (9)	8 (7)	5 (5)	4 (5)	
	NA	30 (11)	9 (10)	5 (5)	4 (5)	
		<i>p</i> =0.382				Food Diary Day

Supplementary Table 3. Differences in median Dietary Species Richness (DSR) across dietary quality categories

Variables Consuming more		Median DSR (IQR)					
		1	Da	y		Graph	
		I	2	3	4		
+ -	than 11% of total			- /->		40 T	
	energy from	30 (11)	9 (7)	5 (5)	4 (4)	40 - %	
d fa tio	saturated fats					بلغ 30-	
ated mp	Consuming less					1 5 20- -	
nra	than 11% of total	29 (12)	9(7)	5(5)	4(4)		
Sat	energy from	_> (1_)	- (1)	0 (0)	. (.)		
	saturated fats						
		<i>p</i> =0.151				Food Diary Day	
	Consuming more						
_	than 5% of total	30 (12)	9(7)	6 (5)	4 (4)		
ar ion	energy from free		- (.)			49 30 -	
npf	Consuming less						
ee .	than 5% of total						
Fr	energy from free	26 (12)	9 (7)	5 (6)	3 (4)		
	sugars					0- <u>±±</u> ±	
		p<0.000			-	1 2 3 4 Food Diary Day	
-	Consuming more						
ssec	than 60% of total						
DCe	energy from	30 (11)	9 (7)	6 (5)	4 (4)	40 - T	
-pro tion	ultra-processed and						
tra- mp	food						
lu l nsu	Consuming less						
ed and ood cor	than 60% of total						
	energy from	29 (12)	9(7)	5 (5)	4 (4)	D	
esse	processed and	-> ()	- (1)	0 (0)	. (.)		
roc	ultra-processed					Food Diary Day	
Р	1000	n<0.002					
	Consuming more	<u> </u>					
ion	than 280g of red	29 (11)	9 (7)	5 (5)	4 (4)	40	
npt	meat across four					see	
uns	days.						
con	Consuming less						
at e	than 280g of red	29 (12)	9 (7)	5 (5)	4 (4)		
me	davs.						
led		0.262			1		
H		<i>p=0.363</i>				Food Diary Day	
	Consuming more					т.	
ion	than 24g of salt	32 (11)	9 (7)	6 (5)	4 (4)	40- 56 80	
sumpti	across tour days.					4 230-	
	than 24 g of solt	29 (11)	9(7)	5(5)	4(A)	39 20-	
con	across four days	<i>27</i> (11)		5(5)	- (+)		
alt				I			
Ň		p<0.000					
_	Consuming lass					Food Diary Day	
re 1mj	than 120 g of			.			
Fib inst tio	fibre across four	27 (10)	9(7)	5 (5)	4 (4)		
CO CO	days.						

	Consuming more than 120 g of fibre across four days.	32 (11)	9 (7)	6 (6)	4 (4)	40- 599 40- 40- 599 50 50 50 50 50 50 50 50 50 50 50 50 50
		p<0.000				$ \begin{array}{c} $
egetable otion	Consuming less than 1600 g of fruit and vegetables across four days.	29 (11)	9 (7)	5 (5)	4 (4)	s 40-
Fruit and v consum	Consuming more than 1600 g of fruit and vegetables across four days.	33 (12)	10 (8)	6 (6)	3 (5)	Image: Second
		p<0.000				
Fish consumption	Consuming less than 160g of fish across four days.	29 (11)	9 (7)	5 (5)	4 (4)	40- 50- 50- 50-
	Consuming more than 160g of fish across four days.	30.5 (12)	9 (8)	6 (6)	4 (4)	\vec{v}_{0}^{2} 20- \vec{v}_{0}^{2} 10- \vec{v}_{0}^{2} 10- \vec{v}_{0}^{2} \vec{v}_{0}^{2} \vec{v}_{0
		<i>p=0.049</i>				rood blary bdy
the above- idelines.	Not meeting one or more guidelines.	29 (11)	9 (7)	5 (5)	4 (4)	40- sep 230- 2 se
ollowing all 1 enlisted gui	Meeting all guidelines.	35 (8)	9 (6)	5 (5)	4 (5)	
H		p=0.288				Food Diary Day



Supplementary Figure 2. Regression analysis to evaluate the association of DSR with the nutritional quality (measured through NRF 8.3 as a proxy) of individual diets