

Dietary protein consumption profiles show contrasting impacts on environmental and health indicators

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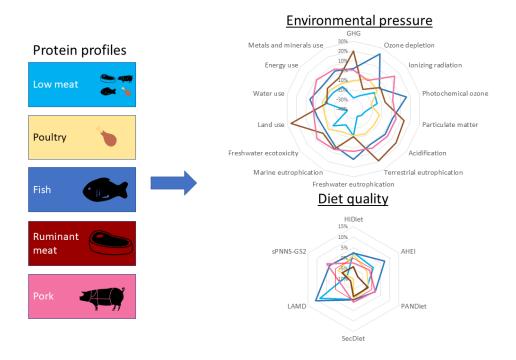
23 Abstract

Patterns of protein intake are strong characteristics of diets, and protein sources have been linked to the environmental and nutrition/health impacts of diets. However, few studies have worked on protein profiles, and most of them have focused on specific diets like vegetarian or vegan diets. Furthermore, the description of the environmental impact of diets has often been limited to greenhouse gas emissions (GHGe) and land use. This paper analyzes the alignment of environmental pressures and nutritional impacts in a diversity of representative protein profiles of a western population.

Using data from a representative survey in France (INCA3, n = 1,125), we identified protein profiles using hierarchical ascendant classification on protein intake (g) from main protein sources (refined grains, whole grains, dairy, eggs, ruminant meat, poultry, pork, processed meat, fish, fruits & vegetables, pulses). We assessed their diet quality using 6 dietary scores, including assessment of long-term risk for health, and associated 14 environmental pressure indicators using the Agribalyse database completed by the SHARP database for GHGe.

Five protein profiles were identified according to the high contributions of ruminant meat, pork, poultry, fish, or, conversely, as low contribution from meat. The profile including the lowest protein from meat had the lowest impact on almost all environmental indicators and had the lowest long-term risk. Conversely, the profile with high protein from ruminant-based foods had the highest pressures on most environmental indicators, including GHGe.

We found that the protein profile with low contribution from meat has great potential for human health and environment preservation. Shifting a large part of the population toward this profile could be an easy first step toward building a more sustainable diet.



- 46 List of abbreviation and acronyms
- 47 ADEME: Agence de la transition écologique
- 48 AHEI-2010: Alternative Healthy Eating Index
- 49 CIQUAL: French Centre d'Information sur la Qualité des Aliments
- 50 CTUe: Comparative Toxic Unit for ecosystems
- 51 DHA: docosahexaenoic acid
- 52 EPA: eicosapentaenoic acid
- 53 GHGe: greenhouse gas emission
- 54 INCA3: Third individual and national study on food consumption survey
- 55 LAMD: Literature-Based Adherence Score to the Mediterranean Diet
- 56 LCA: life cycle analyses
- 57 LCI: life cycle inventory
- 58 NMF: Non-negative Matrix Factorization
- 59 PM: particulate matter PUFA: polyunsaturated fatty acids
- 60 sPNNS-GS2: simplified *Programme National Nutrition Santé* Guideline Score 2
- 61

62

63 1 Introduction

64 Chronic diseases strongly influenced by diets are leading causes of morbidity and mortality around the 65 world (GBD 2013 Mortality and Causes of Death Collaborators, 2015). The food systems underlying 66 these diets are responsible for a third of global greenhouse gas emissions (GHGe) (Crippa et al., 2021) and influence several other environmental indicators (Willett et al., 2019). More and more studies are 67 showing the significant environmental pressures of meat and dairy food production compared to plant 68 69 food production (González et al., 2011; Nijdam et al., 2012). For example, beef produces 7.1 g protein /kg CO₂ eq compared to beans producing 246 g protein/kg CO₂ eq (González et al., 2011). The same 70 relationship was seen for land use with beef protein needing between 15 m². y⁻¹ .kg⁻¹ in industrial 71 systems to 420 m². y⁻¹ .kg⁻¹ for extensive pastoral systems in contrast to pulses needing between 3 to 8 72 m^2 .v.kg⁻¹ for the production of 1kg of protein (Niidam et al., 2012). The amount of water used for 73 74 protein production shows similar extremes of variation, with beef using 728 L/100 grams of protein, 75 and peas and other pulses using around 200 L/100 grams of protein (Ritchie and Roser, 2020).

Protein profiles largely define diets (De Gavelle et al., 2018; Marini et al., 2021), since protein food sources contribute other nutrients that tend to cluster together as a "protein package", e.g. SFA and absorbed iron with red meat and fiber and PUFA with plant protein sources (Mariotti, 2019). They are also associated with other types of foods as part of dietary patterns (De Gavelle et al., 2018; de Gavelle et al., 2019; Mariotti, 2019). Protein profiles in diets are thus key to both environmental and health issues (Clarys et al., 2014; González-García et al., 2018; Seconda et al., 2018).

Several diets have been modeled to improve the population's health while concomitantly considering the environmental issues. These diets display drastic differences in protein profile compared to current westernized diets (Kramer et al., 2017; Seconda et al., 2021; Willett et al., 2019; Wilson et al., 2019). For example, the EAT-Lancet diet includes higher consumption of fruit, vegetable, and whole grains, and is very low in red and processed meat (Willett et al., 2019). This shift in plant and animal sources tend to reverse the plant-animal protein ratio compared to the one observed in modern diets. In the EPIC Oxford study, when compared to a high meat diet, a low meat diet (<50 g/day), was associated

89 with a lower carbon footprint by 920 kg CO₂eq every year (Scarborough et al., 2014). Similarly, the association between the consumption of food group sources of protein and long-term health has been 90 widely studied (Micha et al., 2017a; Turner and Lloyd, 2017). For example, red and processed meats 91 92 have been associated with several non-communicable diseases like colorectal carcinoma, type 2 93 diabetes, and cardiovascular disease (Boada et al., 2016; Micha et al., 2010). In contrast, plant protein 94 intakes have been associated with long-term health benefits (Mariotti, 2019; Naghshi et al., 2020). 95 Thus, compliance with dietary guidelines (Herforth et al., 2019), including those from France (Chaltiel 96 et al., 2019), implies a shift of the dietary protein patterns to diets that include lowering meat 97 consumption in some subgroups of the population in Western countries (Brunin et al., 2021; Dagevos, 98 2021). Many healthy dietary patterns promoted via states official national guidelines include specific 99 recommendations regarding protein sources (Canada, 2020; Chang and Koegel, 2017; ODPHP, 2021). 100 However, contemporary data specifying the link between diet and its impact on both the health and 101 environment is primarily based on single food groups rather than dietary profiles, and does not focus 102 specifically on protein patterns.

103 It is important to note that studies that intended to depict the sustainability of diet profiles rather than 104 of isolated food groups have focused mostly on restrictive diets consumed by a minority, such as 105 vegetarian or vegan diets, which were compared with the whole omnivore population (Chai et al., 106 2019; Fresán and Sabaté, 2019; Rabès et al., 2020; Rosi et al., 2017). Still, no literature to date has 107 studied the intermediate proteins profiles that represent most of the profiles seen in the population.

Furthermore, while several criteria are used to estimate the nutritional or health value of diets, very few environmental indicators are reported. Indeed, most studies limit their indicators to GHGe and land occupation (Aleksandrowicz et al., 2016; Jones et al., 2016). Although these two indicators are important, often a markers of the whole environmental footprint (Kramer et al., 2017), they remain limited in depicting the environmental impact along several dimensions.

In order to inform the future diet transition, it is essential to advance accurate knowledge on the relationship between dietary profiles as naturally occurring in the diversity of people diets, and environmental pressures, and nutritional/health quality. In the present study, we aimed to analyze the

- 116 potential alignment between environmental pressures and nutritional/health impacts of the main
- 117 protein profiles identified in a French representative population.

118 2 Methods

119

120 2.1 Studied population and dietary data

121 This study used data from the third individual and national study on food consumption survey (INCA3) performed in France between 2014 and 2015 (Dubuisson et al., 2019; French Agency for 122 123 Food, Environmental and Occupational Health Safety (ANSES), 2017). 347 participants identified as 124 under-reporters were excluded using the basal metabolic rate as estimated by the Henry equation (Henry, 2005), using the cut-off values recommended by Black (Black, 2000). Since nutrient 125 126 requirements for older adults differ from younger ones, the 649 older adults (above 54 years old for 127 women and 64 years old for men) were excluded. The final sample was composed of 1125 adults, 128 including 564 men and 561 women.

Data were collected by three non-consecutive 24h-dietary recalls (two during the week and one at the
weekend) over a 3-week period. The nutrient content values of the food were extracted from the 2016
database of the *French Centre d'Information sur la Qualité des Aliments* (CIQUAL) (ANSES, 2016).

132

133 2.2 Diet quality assessment

134 Six scores were used to compare the different impacts of the diet on health and nutritional status.

Three of them evaluated the adherence to a specific diet or dietary recommendations were calculated. The Literature-Based Adherence Score to the Mediterranean Diet (LAMD) assesses adherence to the Mediterranean diet. The LAMD score includes three negative components (meat, dairy products, and alcohol) and six positive components (fruits, vegetables, legumes, grains, fish, and olive oil). The consumption of each food group is classified into three categories using cutoff points (Supplemental

140 Table 1)(Sofi et al., 2014). The Alternative Healthy Eating Index (AHEI-2010) is a modified Healthy 141 Eating Index, assessing the adherence to Dietary Guidelines for Americans, improving target food 142 choices and macronutrient sources associated with reduced chronic disease risk (Chiuve et al., 2012). 143 Since information on *trans* fatty acids was not available, only ten parameters were considered, namely 144 vegetables, fruit, whole grains, sugar-sweetened beverages and fruit juice, nuts and legumes, 145 red/processed meat, long-chain (n-3) fatty acids (EPA + DHA), polyunsaturated fatty acids (PUFA), 146 sodium, and alcohol. All AHEI-2010 components were scored from 0 (worst) to 10 (best), resulting in 147 a total AHEI-2010 score ranging from 0 (non-adherence) to 100 (perfect adherence). The scoring criteria are described in Supplemental Table 2. The sPNNS-GS2 (simplified Programme National 148 149 Nutrition Santé Guideline Score 2) assesses the adherence to the French Food-based Dietary 150 Guidelines (Chaltiel et al., 2019). Components, scoring and weighting are shown in Supplemental 151 Table 3.

152 The PANDiet evaluates the probability of adequate nutrient intake. This score is a 100-point 153 probabilistic score evaluating adequate overall nutrient intake. It combines an adequacy sub-score and 154 a moderation sub-score. The adequacy sub-score is calculated as the average probability of adequacy 155 of nutrients for which the usual intake should be above a reference value, multiplied by 100. For a given nutrient, the probability of adequacy was determined from either the Estimated Average 156 157 Requirement or the Adequate Intake, and its variability. The moderation sub-score was calculated as 158 the average of probabilities of inadequacy of six nutrients for which an upper bound reference value 159 exists, together with penalty values. For the six nutrients, the probability of adequacy was calculated, 160 using the upper bound of the acceptable macronutrient distribution range (De Gavelle et al., 2018).

161 The SecDiet evaluates nutrient deficiency risk. This score is based on the intake of twelve critical 162 nutrients for nutritional risk of overt deficiency (Salomé et al., 2021). For each nutrient, we calculated 163 the probability of having sufficient-enough intake to avoid overt nutrient deficiency. We used the 164 probability distribution of the standard normal distribution for nutrient requirements while taking into 165 account the mean intake, the day-to-day intake variability, the inter-individual variability and the 166 nutrient deficiency threshold.

The HiDiet was used to assess the diet impact on long-term mortality and morbidity (Perraud et al., 2022). The HiDiet score is based on the principle of the Comparative Risk Assessment but applied to the risk of one individual and so is an individual version of the EpiDiet (Evaluate the Potential Impact of a Diet) (Dussiot et al., 2022; Kesse-Guyot et al., 2020). The conceptual basis and methodological foundation of the two models are the same which are laid out in the Comparative Risk Assessment framework. It allows for the evaluation of the potential impacts of dietary changes on the long-term morbidity and mortality caused by some diet-related diseases.

In this study, we set up the HiDiet model with values reported in a series of validated international meta-analyses published by a European team (Bechthold et al., 2019; Schwingshackl et al., 2019, 2018, 2017a, 2017b). As in our previous study (Dussiot et al., 2022), we selected 12 diet-related factors (including consumption of fruit, vegetables, nuts or seeds, whole grains, unprocessed red meats, processed meats, sugar-sweetened beverages, fish, dairy products, eggs, refined grains, and legumes) and 4 diet-related diseases (including coronary heart disease, stroke, type 2 diabetes, and colorectal cancer).

181 The reference population considered was the French adult population (18 and 64 years old) in 2014, 182 stratified into sub-populations by 5-year age-bands and sex. The population demographics and national disease-specific deaths were provided by the National Institute of Statistics and Economic 183 Studies (INSEE, 2017), and by the Epidemiological Centre on Medical Causes of Death (CépiDc, 184 2018), respectively. For each individual, two dietary scenarios were built. Here, the two scenarios that 185 186 are compared are the baseline one, where the individual daily intakes (of each food and beverage 187 group) is the average daily intake in the general population but keeping the individual characteristics 188 (age, sex), and the counterfactual scenario is the present situation of the individual, i.e. its actual 189 consumption. Using these two scenarios, it is possible to assess the positive or negative impact of a 190 dietary change from the first one to the other.

191

193 2.3 Protein profile

Protein foods were defined as foods having more than 10% of their energy as protein and providing at least 5g/d of protein in high consumers (i.e. at the 90th percentile of the INCA3) (De Gavelle et al., 2018). Protein foods were then classified into twelve groups of protein sources (refined grains, whole grains, dairy, eggs, ruminant meat, poultry, pork, processed meat, fish, fruits – vegetables, pulses, and others) exhibiting differences in nutritional properties and according to classical dietary classification. Each individual daily intake of protein from a protein group was then calculated by summing the protein intake from the consumption of each food of that group.

201

202 2.4 Agribalyse database and environmental database

203 The Agribalyse database was used to assess environmental indicators of the diet. This database was 204 developed by the Agence de la transition écologique (ADEME) using life cycle analyses (LCA) to 205 assess the environmental pressure of foods eaten in France between 2005 and 2009 with a scope "from field to plate" (ADEME, 2020a). Agribalyse is a database reporting 14 environmental indicators 206 207 (Supplemental Method 1): greenhouse gas emissions in carbon dioxide equivalent (kg CO2 eq); 208 exposure ionizing radiation in equivalent of kilobecquerels of Uranium 235 (kg U235 eq); 209 photochemical ozone (O3) formation in equivalent of kilograms of non-methane volatile organic 210 compounds (kg NMVOC eq); ozone depletion in equivalent of kilograms of trichlorofluromethane 211 (Freon-11); emission of particulate matter in change in mortality due to particulate matter emissions; 212 acidification in equivalent of moles hydron (mol H+ eq); terrestrial eutrophication in equivalent of 213 moles of nitrogen (mol N eq); freshwater eutrophication in equivalent of kilograms of phosphorus (kg 214 P eq); marine eutrophication in equivalent of kilograms of nitrogen (kg N eq); freshwater ecotoxicity 215 in Comparative Toxic Unit for ecosystems (CTUe) an indicator based on a model called USEtox; 216 water use in cubic meters of water; land use in loss of soil organic matter content in kilograms of 217 carbon deficit (kg C deficit); fossils resource use in MJ; and metals and minerals resource use in equivalent of kilograms of antimony (kg Sb eq). Agribalyse assessed the environmental pressures of 218

both single and multi-ingredients foods. Further description of the Agribalyse method is presented inSupplemental Method 1.

Here the Agribalyse database was used to assess the food eaten as listed in INCA3 (n= 1761) (Supplemental Figure 1). Two different methods were used to complete the database resulting in 46 missing data: one specifically for GHGe and a second one for the other environmental impacts in Agribalyse (the method is described in detail in Supplemental Figure 2 and Supplemental Figure 3).

Data from the SHARP database were used to complete missing values for the GHGe indicator (Mertens et al., 2019). This database also includes data on land use but because the unit is different, we were unable to impute the data. Supplemental method 2 explains the process of merging the Agribalyse and SHARP database to the foods eaten in the population.

229 2.5 Statistical analysis

230 We performed a cluster analysis, using a data driven approach to group the INCA3 individual participants into separate protein profile. First, we used a Non-negative Matrix Factorization (NMF) 231 on 11 main protein groups. The protein group composed of the varied other types of protein foods was 232 not included to keep factors interpretability. NMF is a method for identifying independent factors 233 adapted for non-negative data with excess zeros (Lee and Seung, 1999) and has already been used by 234 our team to identify dietary profiles in another survey (De Gavelle et al., 2018). Specifically, the Ls-235 nmf method (Wang et al., 2006) was used, and decomposition in six factors was conserved since it had 236 237 the best R² value. Then we used hierarchical ascendant classification on the factors to identify protein 238 profiles in the population. The number of clusters chosen was based on both the elbow and silhouette 239 methods (Kaufman and Rousseeuw, 2005). The consumption of the 12 proteins groups between all 240 profiles was compared using an ANOVA. The environmental indicators and diet quality scores of the 241 protein profiles were compared using an ANCOVA adjusted for energy intake. Two by two 242 comparison were made using Tuckey adjustment. Statistical analyses were performed using SAS software (version 9.4, SAS Institute Inc, Cary, NC, USA) and R version 4.0.3. 243

245 3 Results

246 3.1 Food group consumption and environmental and nutritional indicators

247 3.1.1 Consumption

The hierarchical ascendant classification identified five distinct profiles (figure 1). The characteristics of the population and the different profiles are described in Table 1. Profile 1 (Low meat profile) represents consumers with a low protein intake from meat (9 g/day vs. 20.8 g/day in the total population), who also have a relatively higher intake of protein from dairy products (13.9 g/day vs. 9.8 g/day in the total population). Supplemental Table 5 shows that the higher dairy protein intake mostly comes from the higher consumption of milk (88 g/day vs. 80 g/day in the total population) and cheese (54 g/day vs. 42 g/day in the total population).

Profile 2 (Poultry profile) represents people with high intakes of protein from poultry (10.8 g/day vs. 5.3 g/day in the total population), while profile 3 (Fish profile) represents individuals with the highest fish protein intake (9.6 g/day vs. 4.4 g/day in the total population). Profile 4 (Ruminant meat profile) represents individuals whose protein intake is largely contributed by ruminant meat proteins (12.4 g/day vs. 6.1 g/day in the total population), and profile 5 (Pork profile) represents people with high intakes of protein from pork (9.6 g/day vs. 2.7 g/day in the total population).

261 The consumption of 33 food groups in grams per day is also presented in Supplemental Table 5.

262 3.1.2 Environmental indicators

Figure 2 presents the environmental indicators for each protein profile (see Supplemental Table 6). The profile with high intake of protein from ruminant has the highest environmental pressure for 5 of the 14 indicators. Notably, this profile also has the highest GHGe (7.7 kg CO2 eq/day vs. 6.4 kg CO2 eq/day in the total population). Compared to the total population, the land use, the emission of particulate matter, the acidification, and the terrestrial eutrophication are also at their highest for this profile, being 26%, 14%, 16%, and 19% higher, respectively.

269 The profile with high intake of protein from fish and the one with the highest pork protein intake have the next most elevated GHGe (6.6 kg CO2 eq/day and 6.4 kg CO2 eq/day). For the other indicators, 270 these two profiles are different. The profile with high intake of protein from pork contributes the most 271 272 to freshwater ecotoxicity (164 CTUe/day vs. 151 CTUe/day in the total population), marine eutrophication (26 kg N eq/day vs. 25 kg N eq/day in the total population), ionizing radiation (1.7 kg 273 274 U235eq/day vs. 1.5 kg U235eq/day in the total population), energy use (67 MJ/day vs. 62 MJ/day in 275 the total population), and metals and minerals use (10.3 kg Sb eq/day vs. 9.7 kg Sb eq/day in the total 276 population). For profile with high intake of protein from fish, it contributes the most to ozone 277 depletion (0.73 Freon-11/day vs. 0.59 Freon-11/day in the total population), photochemical ozone (20 kg NMVOC/day vs. 17 kg NMVOC/day in the total population), freshwater eutrophication (1.15kg P 278 279 eq/day vs. 1.03 kg P eq/day in the total population) and water use (7.1m3/day of water vs. 6.7 1m3/day 280 of water in the total population).

The profiles with the lowest environmental impacts are the poultry profile and the low meat profile. These profiles emit the smallest amount of GHGe with 5.6 kg CO2 eq/day for the high consumers of poultry protein and 4.6 for the low meat consumers. This last profile has the lowest impact for most environmental indicators (11 from the 14 indicators) and is the second-best profile for the three others.

Supplemental Figures 3 to 16 show the variability of the environmental indicators for each food group,
and Supplemental Table 7 describes each food group's contribution to every environmental indicator
for each profile.

288 3.1.3 Nutritional quality

Figure 3 shows the nutritional and health scores of the different profiles. The profile with high fish consumers has the highest scores for the AHEI, the LAMD, and the second-best score for the PNNSGS2, the SecDiet, the PANDiet, and the HiDiet. The profile with the lowest meat protein consumption has the highest HiDiet score, and the one with the highest ruminant meat protein intake has the lowest HiDiet score (respectively +3% and -4% compared to the total population). The profile with the lowest meat protein consumption also has the lowest PANDiet (-3%) and a neutral SecDiet (0%) score. The value of the health scores are presented in Supplemental Table 8.

296 4 Discussion

In this study, we have identified five distinct profiles based on protein consumption, which show the variety of protein consumption patterns. These patterns are easy to interpret, and display contrasting characteristics on both environmental and nutritional/health dimensions.

300

301 4.1 Environmental differences between profiles

302 The high ruminant meat protein consumption profile was one of the worst regarding health scores and 303 environmental indicators. This is in line with the modeled diets and dietary guidelines featuring low 304 red meat intake (Kesse-Guyot et al., 2020; Willett et al., 2019). Although ruminant meat is well known 305 for its high environmental pressures from GHGe and land use (Clark et al., 2018; Poore and Nemecek, 306 2018), our study has shown that actual protein intake from meat is also associated with other 307 environmental pressures. The particulate matter (PM) production is significantly higher in the diet of people with high intake of protein from ruminant meat. This is in part ascribed to ruminant meat with 308 a high production of PM coming from manure (Garcia et al., 2013). It could be argued that this is not 309 310 an important issue given the small percentage of particulate matter coming from agriculture compared 311 to other sources like transport (Cambra-López et al., 2010). However, this percentage is growing, with PM emissions from agriculture being estimated at about 25% of total emissions in the Netherlands 312 313 (Cambra-López et al., 2010). The Global Burden of Disease estimated that air pollution is responsible for about 1.2% of premature deaths and 0.5% of lost life years (Cohen et al., 2005). It is considered as 314 the ninth larger risk factor contributing to deaths in the United States (The US Burden of Disease 315 Collaborators, 2018). Acidification causing soil degradation (Goulding, 2016) and acid rain are other 316 317 environmental impact of animal production. The dominant sources of acidification in agriculture come 318 from ammonia emissions released used for feed production and manure production. With a higher feed 319 conversion ratio and an important manure production, ruminants thus have a higher acidification 320 power per meat produced (Röös et al., 2013).

321 The low meat profile had the smallest environmental pressure. This is in line with the literature as most publications pointed out that meat, especially from ruminant and to a lesser extent from 322 323 monogastric, has by far the most significant environmental impact, specifically for GHGe (Poore and 324 Nemecek, 2018). Indeed, low GHGe in this profile should be ascribed to the very low consumption of 325 ruminant meat. This is similar with other studies showing plant-based diets having lower GHGe 326 compare to omnivorous diet (Rabès et al., 2020). Note that dairy intake was relatively high for this 327 profile – in fact the highest of the 5 profiles – which may be considered a practical conflict between the production of ruminant meat and the production of milk. This is because of the co-production 328 329 factor; in France, if the bovine dairy protein intake is above 0.43 times the bovine meat protein intake, there is an excess meat production (Barré et al., 2018). In the case of the low meat consumer profile, 330 331 the bovine dairy protein consumption exceeded this number. However, this issue only occurred in the hypothetical situation where the entire population follows this diet, with no meat exportation or dairy 332 importation. Moreover, changes in the dairy systems, like a change in the life span of dairy cows, 333 could potentially strongly modify this factor. 334

335 It is also important to note that a high share of red meat consumed comes from dairy cows in France 336 (Assmann, 2020). Thus, the allocation of the environmental pressure using biophysics systems in the life cycle inventory (LCI) is considered. The literature showed different types of allocations 337 338 underlying different choices. These choices came from 4 criteria: compliance with recommendations, 339 stability in time and space, and practical aspects (Wilfart et al., 2021). Agribalyse allocation for dairy and ruminant meat is based on the metabolic energy required to produce each co-product 340 (Supplemental Method 1). On a food basis, the indicators already studied, like GHGe, have a similar 341 order of magnitude of allocations as other studies (Clune et al., 2017), indicating the choices made in 342 Agribalyse (ADEME, 2020b) are similar to that of most published studies. 343

This low-meat protein profile also has drawbacks on a nutritional aspect, with the lowest PANDiet values of all profiles. This is mainly explained by the low intake of nutrient-dense food sources, in particular red meat, which is not offset by higher intakes of fruits and vegetables. It should be noted that even if it is the most plant-based profile, the consumption of pulses, whole grains and other

348 recommended plant food groups is far from what is recommended in optimization studies (Dussiot et 349 al., 2022; Willett et al., 2019). Moreover, the nutrient security in this low-meat profile was not lower 350 than that of the general population, confirming that even if there are some slightly lower intakes of 351 some nutrients, there is no overall increase in the risk of overt deficiency. As far as long-term health is 352 concerned, this profile has the lowest estimated risk of long-term morbidity or mortality from coronary 353 heart disease, stroke, type 2 diabetes, and colorectal cancer, as assessed using the HiDiet score. Given 354 that the small increase in nutrient inadequacy does not increase the risk of overt deficiency, this profile 355 shows strong co-benefits for environmental and human health.

Compared to the other profiles high in animal flesh (meat or fish), the profile with high intakes of protein from poultry has a lower impact on most of the environmental indicators. In fact, GHGe are specifically one of the advantages of this profile. Similar results from previous studies show that poultry produces the lowest GHGe among all types of meat (González et al., 2011; Nijdam et al., 2012).

Surprisingly, this profile exhibited poor dietary scores with values under the population average for all scores except the HiDiet. This latter score was the only one in line with the epidemiological data showing that poultry consumption was not associated with a higher risk of mortality or morbidity (Du et al., 2020; Makiuchi et al., 2020; Micha et al., 2017a, 2017b). The low scores for the other nutritional and health scores may be explained by the low fish consumption and the fact that other food categories with a positive weight in these health and nutritional scores are low in the poultry profile.

The profile with a high fish consumption has high values for all health and nutritional scores. This was 367 expected as fish consumption is positively weighted in all of the scores considered in this study. Fish 368 is a very good contributor to bioavailable nutrients and has well-known health benefits against chronic 369 370 disease (Bogard et al., 2019). The high intake of indispensable nutrients associated with the higher 371 level of protein for fish but also an overall dietary profile of higher quality explains the high PANDiet and SecDiet score. In contrast, this high fish intake has some important drawbacks on the 372 environmental indicators. As described before, water use and photochemical ozone formation are two 373 374 important issues of fish consumption (Ruiz-Salmón et al., 2021). It is important to note that there is

375 significant variability between fish species, and it may be possible to improve the health and 376 environmental impact by consuming different fish species (ADEME, 2020a). However, these results 377 need to be verified with other LCA as environmental pressures from fish are not precise, being mostly 378 measured indirectly.

379 4.2 Limits and strengths

380 First, it should be noted that INCA3 survey was conducted between 2014 and 2015, so diets may have 381 changed since then, and the sample of 1125 individuals that we used is small. However, it is the latest French representative survey to date. The environmental database Agribalyse has also some 382 limitations. For example, the quantification of soil carbon storage and removal is not considered in the 383 calculation of the GHGe. Biodiversity and the impact of phytosanitary products are also not yet 384 385 calculated. Land use is often a proxy for lost biodiversity. However, in the Agribalyse database, land use is expressed in the loss of Soil Carbon and cannot be used similarly to when expressed in hectares 386 or m². Another drawback is that there is no information about farming practices (such as organic 387 production) for foods production that are consumed in the survey, and the environmental indicators are 388 389 only for mean of consumed foods differing in farming practices. However, this database includes a 390 large variety of environmental indicators covering the entire food chain from production to the plate 391 and virtually all the foods consumed by the French population since it was created for this purpose. 392 Though, it should be noted that several of these indicators are strongly linked to fuel use.

Finally, the major strength of this study is the variety of indicators and food described in a representative survey. Using this information, it was possible to precisely describe 14 environmental indicators and 6 nutritional and health dietary scores for five protein profiles in a representative sample of the French population

397 5 Conclusion

In the present study, we showed that the profiles of protein intake of the population are varied and have contrasting associations with health and environmental impacts. Using the Agribalyse study, we could assess 14 environmental pressures, which we brought together with 6 health indicators. The

401 protein profile marked by ruminant meat had the worst scores on both health and environmental aspects, especially regarding GHGe, land use, and particulate matter emissions, and long-term health. 402 Conversely, the most environment-friendly protein profile is very low in meat, and this profile also 403 had the lowest risk for long-term morbidity and mortality. Apart from the protein profile low in meat, 404 the profile with high intakes of protein from poultry has a lower impact on most of the environmental 405 406 indicators, including GHGe. The fish protein profile had overall the best diet quality indicators but mixed values for environmental indicators, with high water use and photochemical ozone formation. 407 As we studied real profiles identified in the general population, the differences in health and 408 409 environmental impacts between profiles may be useful to consider realistic targets for acceptable 410 changes in the diet. These changes may be more practical than those identified by modeling studies. 411 Our results thus support the importance of protein profiles for health and environmental impacts

Acknowledgments and statement of authors' contributions to manuscript

EP, FM, and EK-G: designed the research; EP: conducted the research; EP, FM, and EK-G analyzed the data; MS, and J-FH provided methodological support; EP drafted the first version of the manuscript, EP, FM, and EK-G wrote the manuscript and all authors provided critical comments on the manuscript. EP, FM, and EK-G had primary responsibility for the final content, and all authors read and approved the final manuscript.

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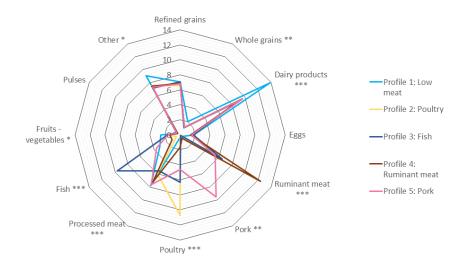
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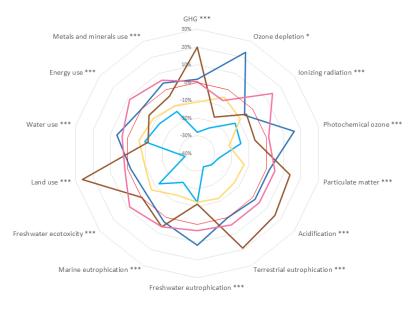
666

LEGEND FOR FIGURES



667 **Figure 1** Consumption (g/day) of protein groups in individuals in the 5 protein profile as identified in

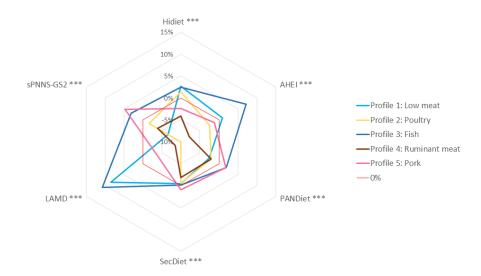
668 the dietary survey (INCA3). * P < 0.05, ** P < 0.01, *** P < 0.001





— Profile 1: Low meat — Profile 2: Poultry — Profile 3: Fish — Profile 4: Ruminant meat — Profile 5: Pork — 0%

671 Figure 2. Environmental indicators associated with the diet consumed by individuals in the 5 protein profiles as identified in the dietary survey (INCA3). Data are in percentage of difference to the mean 672 value of the total population. GHGe is the emission of greenhouse gases measured in kg CO₂ eq. The 673 ozone depletion is in kg CFC-11eq. The photochemical ozone formation is in kg of Non-Methane 674 Volatile Organic Compounds eq. Particulate matter is in kg of PM_{2.5} emitted. The acidification is in 675 mol H^+ eq. The terrestrial eutrophication is in mol N eq, the freshwater eutrophication is in kg P eq, 676 and the marine eutrophication is in kg N eq. The freshwater ecotoxicity is based on the USEtox model. 677 Land use is in kg C deficit, water use in m³, fossils resource use in MJ, and Metals and minerals use in 678 kg SB eq. * *P* < 0.05, ** *P* < 0.01, *** *P* < 0.001 679



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682 Figure 3. Nutritional and health indicators by profile. All data are in percentage of difference to the 683 mean value of the population. The HiDiet was used to assess the diet impact on long-term mortality 684 and morbidity (variation between -1 and 1). The Alternative Healthy Eating Index (AHEI-2010) is a modified Healthy Eating Index, assessing the adherence to Dietary Guidelines for Americans, 685 686 improving target food choices and macronutrient sources associated with reduced chronic disease risk (maxpoint = 100). The PANDiet evaluates the probability of adequate nutrient intake (maxpoint = 687 688 100). The SecDiet evaluates the nutrient risk of overt deficiency (maxpoint = 1). The Literature-Based 689 Adherence Score to the Mediterranean Diet (LAMD) assesses adherence to the Mediterranean diet 690 (maxpoint = 16). sPNNS-GS2 the adherence to the French Food-based Dietary Guidelines (maxpoint = 10.5). * P < 0.05, ** P < 0.01, *** P < 0.001691