

Evolution of diet patterns over time in European countries from 1963 to 2013: an exploratory analysis using PCA for compositional data vectors.

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Introduction:

The study of the evolution of the diet of populations, namely in Europe, is crucial for decision-making by public authorities to promote healthy nutrition and prevent diet related noncommunicable diseases [1]. In fact, for instance, European Commission food law impacts the transformative potential of alternative proteins [2]. Composition food consumption in terms of each macronutrient, i.e., protein (animal and vegetal), fat and carbohydrate, has been usually investigated separately by nutrient [3]. However, these macronutrients constitute the whole diet of each individual and so, add up to a constant sum (in general 100% or 1 if the constitution of macronutrients in a diet is evaluated in percentage or proportion, respectively). Hence, it seems crucial that any analysis of diet compositions must take the constrained or compositional nature of the data. A set of several procedures has been developed in Statistics and applied in many fields to handle with this type of data structure, i.e., with compositional data [4,5].

Compositional data are multivariate observations that represent quantitative descriptions of the parts of some whole, conveying only relative information between parts. In mathematical notation, a p-multivariate observation corresponds to a sample unit described by a vector with p components. When all the components of this vector are positive numbers and contain relative information of parts of a whole (e.g. proportions, percentages), the nature of multivariate observation is compositional. This means that the important information between components is given by ratio rather difference. For instance, while the difference between 0.05 and 0.10 and between 0.45 and 0.50 are the same, in a compositional perspective, the number 0.05 should be considered as half of 0.10 while 0.45 forming a fraction 0.9 of 0.50. In mathematical notation, a D-part compositional observation is expressed as $\mathbf{x} = [x_1 x_2 \cdots x_D]$ with $x_j > 0$, for j = 1, 2, ..., D, and subject to the so-called unit-sum constraint, $\Sigma_{j=1}^{D} x_j = 1$.

In this paper, composition food consumption by both country and decade will be analyzed. In this context, a $(p \times D)$ -multivariate observation is described by p variables each one with D-part compositional components. This type of multivariate observation is referred to as a (*p*-dimensional) compositional data vector as, for instance, the quantity of four macronutrients (p=4: animal protein, vegetal protein, fat, carbohydrate) across six years (D=6: 1963, 1973, 1983, 1993, 2003, 2013) by country). In [6], the operations in the space of the *p*-dimensional compositional data vectors were deduced: \bigoplus for the "addition" (called perturbation) and \bigotimes for the multiplication by a scalar (called powering) (for details see Equation 14 and 15 in [7])

Recently, principal component analysis (PCA) [8] for modeling compositional data vectors was developed in [7]. In this study, this statistical procedure is explored to reduce the dimensionality of the evolution of diet patterns of ten countries in Europe during five decades. The main focus is on a temporal composition of the consumption in the diet of the four macronutrients.

Methods:

Daily caloric supply derived from each of the four macronutrients by country has been produced by Our World in Data and are freely accessible in [9]. For this study, data related to ten European countries (see Table 2) 10-by-10 years from 1963 to 2013 were collected. Hence, the data set includes ten observations featured by four 6-dimensional compositional variables: animal protein (U1), vegetal protein (U2), fat (U3), carbohydrate (U4). Given the structure of the data and following [7], a set of instructions in R were constructed to perform the PCA method on the correlation matrix for compositional data vectors (for more details see [10]).

Keywords: Compositional data vectors, Diet pattern, Macronutrient shares, Principal component analysis.

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Results:

The unit-loading and the variance of each principal component (PC) are displayed in Table 1 and the compositional scores of PC1 for the ten observations are shown in Table 2. PC1 accounts for 60.9% of the total variance and

 $PC1 = 0.475 \bigotimes \mathbf{U1} \bigoplus (-0.526) \bigotimes \mathbf{U2} \bigoplus 0.545 \bigotimes \mathbf{U3} \bigoplus (-0.448) \bigotimes \mathbf{U4}.$

Table 1 - Coefficient	s of the PCs and	d the percentage	of variance explai	ined by each PC.
Components	PC1	PC2	PC3	PC4

PC1	PC2	PC3	PC4
0.475	-0.562	0.285	0.614
-0.526	-0.421	-0.661	0.330
0.545	-0.400	-0.464	-0.572
-0.448	-0.589	0.515	-0.432
60.9	31.7	5.2	2.2
	PC1 0.475 -0.526 0.545 -0.448 60.9	PC1 PC2 0.475 -0.562 -0.526 -0.421 0.545 -0.400 -0.448 -0.589 60.9 31.7	PC1 PC2 PC3 0.475 -0.562 0.285 -0.526 -0.421 -0.661 0.545 -0.400 -0.464 -0.448 -0.589 0.515 60.9 31.7 5.2

Table 2 -	Score	of the	first	PC
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Country	PC1 scores (%)
Finland	[24.0 24.2 21.0 13.6 8.8 8.3]
France	[9.0 18.8 16.7 19.5 22.1 13.9]
Germany	[36.7 19.8 12.4 13.6 9.6 7.9]
Greece	[11.1 12.6 11.9 14.7 18.4 31.4]
Italy	[5.9 6.3 18.5 23.9 21.2 24.2]
Norway	[28.5 36.3 13.2 8.1 7.3 6.5]
Portugal	[3.5 4.7 8.5 13.9 27.3 42.1]
Spain	[2.8 4.3 10.3 18.4 36.2 28.1]
Sweden	[40.1 19.8 14.3 8.7 9.1 8.0]
United Kingdom	[40.8 23.3 15.0 10.4 5.0 5.6]

Then, considering the coefficient signals, PC1 represents a balance of consumption of animal protein and fat comparatively with vegetal protein and carbohydrate such that higher the score of PC1, the greater the fraction consumption of animal protein and fat compared with those of vegetal protein and carbohydrate.

It is clearly observed in Table 2 that there existed a higher percentage of consumption of animal protein and fat in the first decade in Finland, Germany, Norway, Sweden and United Kingdom, while this percentage is higher, in general, for the other countries in the last two decades.

Discussion:

PCA can be applied on data sets that can be arranged into compositional vector. In terms of macronutrient diet pattern, PCA revealed different shares of animal protein and fat between countries further Mediterranean region and further north in Europe from 1963 to 2013.

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