



INTRODUCTION

In many applications, individual data are not widely available, if at all, especially when small geographic areas or vulnerable data are involved. It is more frequent to find **aggregate** data instead. Two problems arise :

- Can these aggregate data be used to **infer** individual behavior (we will use the term of **ecological inference**)?
- How to manipulate categorical variables (which become compositional variables once aggregated)?

COMPOSITIONAL DATA

A composition of *D* components is a vector \mathbf{x} of the simplex S^D defined as

$$S^{D} = \left\{ \mathbf{x} = [x_1, x_2, \dots, x_{D}]; \ x_i > 0, i = 1, 2, \dots, D; \ \sum_{i=1}^{D} x_i = 1 \right\}$$
(1)

This sample space is a (D-1)-dimensional subset of \mathbb{R}^D . For example, S^3 is a triangle (ternary diagram).



Picpus (12^e) : 22% BEP, 29% BAC, 79% SUP Aitchison (1986) introduces operations on the simplex (sum and scalar multiplication). The **inner product** of two compositions is defined as :

$$\langle \mathbf{x}, \mathbf{y} \rangle_a = \sum_{i=1}^D \log \frac{x_i}{g(\mathbf{x})} \log \frac{y_i}{g(\mathbf{y})} = \frac{1}{D} \sum_{i < j} \log \frac{x_i}{x_j} \log \frac{y_i}{y_j}$$
 (2)

The **additive** log-ratio transformation :

$$alr(\mathbf{x}) = \left[\log\frac{x_1}{x_D}, \log\frac{x_2}{x_D}, \dots, \log\frac{x_{D-1}}{x_D}\right]$$
 (3)

The **centered** log-ratio transformation :

$$clr(\mathbf{x}) = \left[\log \frac{x_1}{g(\mathbf{x})}, \log \frac{x_2}{g(\mathbf{x})}, \dots, \log \frac{x_D}{g(\mathbf{x})}\right]$$
 (4)

The **isometric** log-ratio transformation :

$$ilr(\mathbf{x}) = [\langle \mathbf{x}, e_1 \rangle_a, \langle \mathbf{x}, e_2 \rangle_a, \dots, \langle \mathbf{x}, e_{D-1} \rangle_a]$$
 (5)

Where $e_1, e_2, ..., e_{D-1}$ represent an orthonormal basis of S^D and $g(\mathbf{x}) =$ $(x_1x_2...x_D)^{1/D}$ is the geometric mean.

Regression analysis is generally used in statistics to study the relation between a response Y and a set of explanatory covariate $\mathbf{x} = (x_1, \ldots, x_D)$ as

The *ilr* transformation is an isometric transformation from the simplex to the real space with usual Euclidean geometry. It allows the application of standard models to compositions as

Using the estimation of β , it is possible to find b with the ilr-inverse transformation. Composition *b* can be interpreted as the slope parameter of the standard regression. If x differs by $\frac{b}{\|b\|}$ in the direction of b, then y differs by $\|b\|$.



Aitchison, J. (1986). The Statistical Analysis of Compositional Data. Chapman and Hall London. Duncan, O., & Davis, B. (1953, dec). An Alternative to Ecological Correlation. American Sociological Review, 18(6), 665-666. Goodman, L. (1953, dec). Ecological Regressions and Behavior of Individuals. American Sociological Review, 18(6), 663-664.

Aggregate data and compositional variables

Enora Belz^{*a*}, Arthur Charpentier^{*b*}

Faculté des Sciences Économiques, Université de Rennes 1 ^{*a*}enora.belz@univ-rennes1.fr, ^{*b*}arthur.charpentier@univ-rennes1.fr

ECONOMETRICS WITH COMPOSITIONAL DATA

$$\mathbf{y} = \beta_0 + \beta^T \mathbf{x} + \epsilon$$

The standard linear model is **unreasonable** with compositional data.

• Compositions contain only relative information : least squares methods examine continuous variables that are linked with an absolute relationship

• When one part is altered, another is altered : the interpretation of the linear regressions coefficients assumes that "all other things remaining equal"

• Composition covariates involve **multicollinearity** problems

• The simplex has particular geometric properties and operations : most common statistical procedures are developed in the usual Euclidean geometry

$$\mathbf{y} = \beta_0 + \beta^T i lr(\mathbf{x}) + \epsilon \tag{7}$$

AN EXAMPLE

Analysis of median income according to diploma levels in Paris by iris.



A-Levels), BAC (Baccalaureate, A-Levels) and SUP (Bachelor, Master, Ph.D.)

Ecological inference consists of making inference of **individual** behaviour using **aggregated** group-level data. Group-level data are generally less reliable and subject to biases and imprecision that can lead to mistake of inference. Robinson (1950) warms to be cautious when using aggregate data to study individuals ("ecological fallacy").



The aim is to find the probability of being republican conditionally of being a man (β_i^0) or a woman (β_i^1) in the area j. Ecological regression (Goodman, 1953) assumes that these two probabilities are **constant** over area and are estimated by least squares as

and maximum of the probability with

(6)

 \max

may

King's solution (King, 1997) is an improvement in ecological inference by combining the Goodman method and the information of the bounds to improve inference. β_i^0 and β_i^1 are linked by **tomography line** within the unit square as follows

King suggests three assumptions :

REFERENCES

King, G. (1997). A Solution to the Ecological Inference Problem. Princeton University Press. Robinson, W. (1950, jun). Ecological Correlation and the Behavior of Individuals. American Sociological *Review*, 15(3), 351-357.







ECOLOGICAL INFERENCE

| | Republican | Democrate | Total |
|-----|------------|-------------|--------------|
| len | ? | ? | ϕ_j |
| n | ? | ? | $1 - \phi_j$ |
| al | p_j | $1 - p_{j}$ | 1 |

$$p_j = \beta^1 \phi_j + \beta^0 (1 - \phi_j) \tag{8}$$

The method of bounds (Duncan & Davis, 1953) is to find the minimum

$$\left\{0; \frac{p_j - (1 - \phi_j)}{\phi_j}\right\} \le \beta_j^1 \le \min\left\{\frac{p_j}{\phi_j}; 1\right\}$$
(9)

$$\operatorname{x}\left\{0;\frac{p_{j}-\phi_{j}}{1-\phi_{j}}\right\} \leq \beta_{j}^{0} \leq \min\left\{\frac{p_{j}}{1-\phi_{j}};1\right\}$$
(10)

$$\beta_{j}^{0} = \frac{p_{j}}{1 - \phi_{j}} - \frac{\phi_{j}}{1 - \phi_{j}}\beta_{j}^{1}$$
(11)

• β_i^0 and β_i^1 are in a single cluster that is generated by a truncated nor**mal** bivariate distribution conditional of ϕ_i

• Absence of **spatial autocorrelation** : the number of exposed cases of one area is not related to the number of cases in the other areas

• Absence of **aggregation bias** : independence between the regressors ϕ_j and the parameters β_i^0 and β_i^1