

# COMPATIBILITY OF CONDITIONAL DISTRIBUTIONS

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Pisa, December 7, 2014

## THE PROBLEM

Up to technicalities, the problem can be stated as follows

Let  $X$  and  $Y$  be real random variables. Suppose we are given the (measurable) kernels

$$\alpha = \{\alpha(x, \cdot) : x \in \mathcal{R}\} \quad \text{and} \quad \beta = \{\beta(x, \cdot) : x \in \mathcal{R}\}$$

where  $\alpha(x, \cdot)$  and  $\beta(x, \cdot)$  are probability measures on  $\mathcal{R}$ .

Is there a joint distribution for  $(X, Y)$  such that

$$P(Y \in \cdot \mid X = x) = \alpha(x, \cdot) \quad \text{and} \quad P(X \in \cdot \mid Y = y) = \beta(y, \cdot)$$

for almost all  $(x, y) \in \mathcal{R}^2$  ?

## MORE GENERALLY

The previous problem is actually a special case of the following

Let  $X_1, \dots, X_k$  be real random variables and

$$Y_j = (X_1, \dots, X_{j-1}, X_{j+1}, \dots, X_k), \quad 1 \leq j \leq k.$$

Suppose we are given the (measurable) kernels

$$\alpha_j = \{\alpha_j(y, \cdot) : y \in \mathcal{R}^{k-1}\}$$

where  $\alpha_j(y, \cdot)$  is a probability measure on  $\mathcal{R}$ .

Is there a joint distribution for  $(X_1, \dots, X_k)$  such that

$$\boxed{P(X_j \in \cdot \mid Y_j = y) = \alpha_j(y, \cdot)}$$

for all  $1 \leq j \leq k$  and almost all  $y \in \mathcal{R}^{k-1}$  ?

If yes, the kernels  $\alpha_1, \dots, \alpha_k$  are **compatible (or consistent)**.

## MOTIVATIONS AND REMARKS

**(i)** We focus on real random variables for the sake of simplicity. Nothing important changes if each  $X_j$  takes values in a measurable space  $(\mathcal{X}_j, \mathcal{B}_j)$ . Similarly,  $(X_1, \dots, X_k)$  can be replaced by an infinite sequence  $X_1, X_2, \dots$

**(ii)** Let  $\mathcal{C}$  be a collection of probability measures on  $\mathcal{R}^k$ . The kernels  $\alpha_1, \dots, \alpha_k$  are  **$\mathcal{C}$ -compatible** if they are the conditional distributions induced by some member of  $\mathcal{C}$ .

**(iii)** Conceptually, we are just dealing with a **coherence** problem within the Kolmogorovian setting. Indeed, those people familiar with de Finetti's coherence can think of this problem as

Coherence + Disintegrability

Technically, however, the problem is much harder than Finetti's coherence

(iv) In some statistical frameworks, such as

Gibbs sampling, Multiple data imputation, Spatial statistics

the conditional distributions of

$X_j$  given  $Y_j = (X_1, \dots, X_{j-1}, X_{j+1}, \dots, X_k)$

are needed. Sometimes, such conditional distributions are not derived from a joint distribution for  $(X_1, \dots, X_k)$ . Rather:

One first selects the kernels  $\alpha_1, \dots, \alpha_k$ , with each  $\alpha_j$  regarded as the conditional distribution of  $X_j$  given  $Y_j$ , and then checks compatibility of  $\alpha_1, \dots, \alpha_k$ .

Note that, if  $\alpha_1, \dots, \alpha_k$  fail to be compatible, **any subsequent analysis does not make sense !!!**

(v) Compatibility issues also occur in

Statistical mechanics, Machine learning, Bayesian statistics

## RESULTS

In BDR (Stochastics, 2013; J. Multivariate Analysis, 2014) compatibility of  $\alpha_1, \dots, \alpha_k$  is characterized under the assumptions that

**(i)** Each  $X_j$  has compact support

or

**(ii)** Each kernel  $\alpha_j$  has a density with respect to some reference measure  $\lambda_j$ , namely  $\alpha_j(y, dx) = f_j(x | y) \lambda_j(dx)$

Furthermore,

**(iii)**  $\mathcal{C}$ -compatibility of  $\alpha_1, \dots, \alpha_k$  is characterized for

$\mathcal{C} = \{\text{exchangeable laws on } \mathcal{R}^k\}$

$\mathcal{C} = \{\text{identically distributed laws on } \mathcal{R}^k\}$

We next focus on (iii)

**THM1:** Let  $\mathcal{C} = \{\text{exchangeable laws on } \mathcal{R}^k\}$ . Suppose

$$\alpha_1 = \dots = \alpha_k = \alpha \quad \text{and} \quad \alpha(y, \cdot) = \alpha[\pi(y), \cdot]$$

for all permutations  $\pi$ . Then,

$$\boxed{\alpha_1, \dots, \alpha_k \text{ are compatible} \Leftrightarrow \alpha_1, \dots, \alpha_k \text{ are } \mathcal{C}\text{-compatible}}$$

**THM2:** Let  $\mathcal{C} = \{\text{identically distributed laws on } \mathcal{R}^k\}$ . Suppose

$k = 2$ ;  $X_1, X_2$  take values in a finite set;  $\alpha_1$  irreducible

Then,  $\alpha_1, \alpha_2$  are  $\mathcal{C}$ -compatible if and only if

$$\alpha_1(x, y) > 0 \Leftrightarrow \alpha_2(y, x) > 0$$

$$\prod_{i=1}^n \alpha_1(x_{i-1}, x_i) = \prod_{i=1}^n \alpha_2(x_i, x_{i-1}) \quad \text{if } x_n = x_0$$

where  $\alpha_i(x, y) = \alpha_i(x, \{y\})$ . Such characterization easily extends to  $X_1, X_2$  taking values in a countable set

## EXAMPLE

Suppose

$$\alpha_1 = \dots = \alpha_k = \alpha \quad \text{with} \quad \boxed{\alpha(y, \cdot) = N(b\bar{y}, 1)}$$

where  $b \in \mathcal{R}$  and  $\bar{y} = (1/(k-1)) \sum_{i=1}^{k-1} y_i$  is the sample mean.

By THM1, those values of  $b$  which make  $\alpha_1, \dots, \alpha_k$  compatible can be determined. For instance,

For  $k = 2$ :  $\alpha_1, \alpha_2$  are compatible  $\Leftrightarrow b \in (-1, 1)$

For  $k = 3$ :  $\alpha_1, \alpha_2, \alpha_3$  are compatible  $\Leftrightarrow b \in (-2, 1)$

and so on. Also, for any  $k$ ,  $\alpha_1, \dots, \alpha_k$  are compatible if and only if they are induced by an exchangeable law on  $\mathcal{R}^k$



## EXAMPLE

Suppose  $k = 2$  and  $X_1, X_2$  take values in  $Z = \{\dots, -1, 0, 1, \dots\}$

Let  $\alpha$  be the kernel of the symmetric random walk on  $Z$ :

$$\alpha(i, i - 1) = \alpha(i, i + 1) = 1/2$$

for each  $i \in Z$ , where  $\alpha(i, j) = \alpha(i, \{j\})$ .

Are there a kernel  $\beta$  on  $Z$  and a joint distribution for  $(X_1, X_2)$  such that

$$X_1 \sim X_2 \quad \text{and}$$

$$P(X_1 = j \mid X_2 = i) = \alpha(i, j), \quad P(X_2 = j \mid X_1 = i) = \beta(i, j) ?$$

The answer is NO, as a consequence of THM2