

Nonparametric bayesian modelling of individual co-exposures to various pesticides to determine cocktails

Amélie Crépet and Jessica Tressou

ANSES, French agency for food, environmental and occupational health safety



INRA UR-1204, Mét@risk, Paris, France



The Hong Kong University of Science and Technology, Hong Kong

Applibugs, Paris, 26 November, 2010

- For statisticians

What are the cocktails of pesticides simultaneously present in the diet?

- Co-exposure assessment
- Co-exposure clustering

- For toxicologists

What are the possible combined effects of multiple residues of pesticides?

- **Contamination levels :** (DGCCRF, DGAL, SISE-EAUX)
 - $p = 1, \dots, P = 79$ pesticides
 - $a = 1, \dots, 121$ commodities
 - $H : 0$ if the value $< LOD$
- **Consumption :** the National French survey (INCA 2, 2006)
 - consumed quantities c_{ia} of $n = 3,337$ individuals
 $n = 1,439$ children (3-17 years) and $n = 1,898$ adults
 - 7 days detailed

Empirical exposure distribution x_{imp}

$$x_{imp} = \log_{10} \left(\sum_{a=1}^{A_p} c_{ia} \times q_{pam} / w_i \right)$$

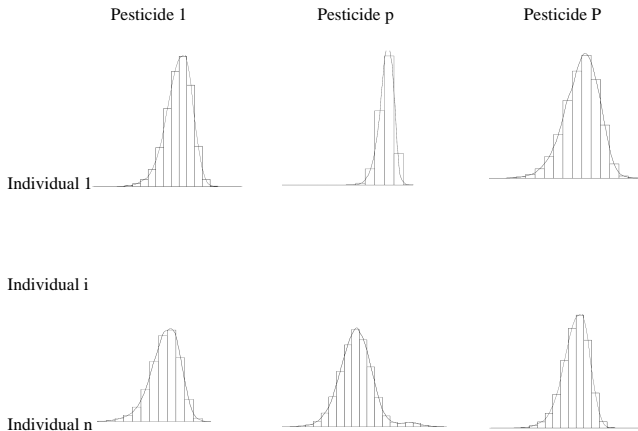
$M=100$ values computed by

- randomly sampling for each individual i and pesticide p
 - consumptions of food a observed on a day : c_{ia}
 - pesticide residue levels q_{pam} of the different commodities A_p
- dividing by the associated individual body weight w_i
- normalized the logarithm of exposures

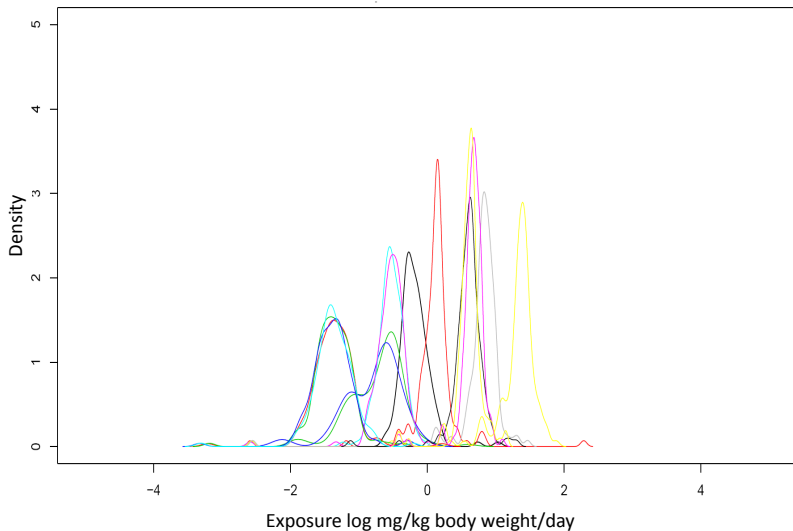
Exposure Structure

Co-exposures, $\{x_{imp}, i : \text{individuals} ; m : \text{residues levels} ; p : \text{pesticides}\}$

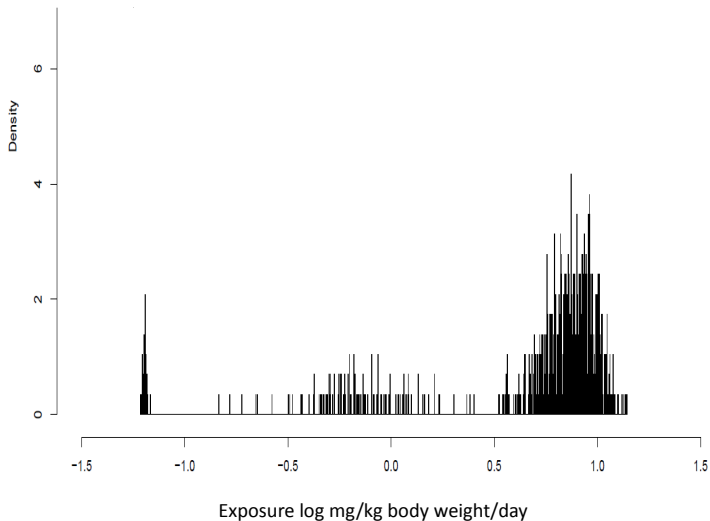
- $M = 1$, 95th percentile of exposure to each pesticide P
- $M = 100$, account for contamination variability



Co-exposures of 15 children to 79 pesticides



Children exposure to a single pesticide



Infinite Mixture Models

Non-parametric Bayesian Models

- A way of getting very flexible models
- No assumption on the number of mixture components
- No prior parametric assumption
- Infer an adequate model (size/complexity) without doing Bayesian model comparison (AIC, DIC...)
- Derived from finite parametric model with number of parameters going to infinity

Let $x = (x_1, \dots, x_n)$, with x_i a P dimensional vector $x_i = (x_{i1}, \dots, x_{iP})$ distributed with a density of probability :

$$f(x) = \int_{\Theta} k(x|\theta)G(d\theta)$$

where

- $k(\cdot|\theta)$ the known density of the mixture components,
- $\theta \in \Theta$ is a latent variable,
- G the unknown mixture distribution, infinite dim. parameter

How to place an appropriate prior on G ?

Define *a priori* on a unknown probability distribution

- $G|\gamma, H \sim \mathcal{D}(dG|\gamma, H)$ is a Dirichlet process with parameters
 - a real $\gamma > 0$
 - a probability measure H

- if and only if for any partition B_1, \dots, B_K of Ω ,

$$(G(B_1), \dots, G(B_K)) \sim \text{Dir}(\gamma H(B_1), \dots, \gamma H(B_K))$$

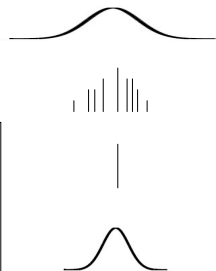
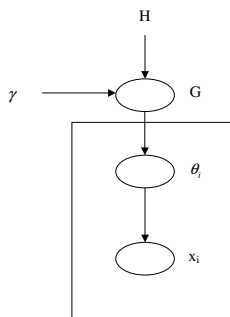
- $\mathbb{E}[G(B)] = H(B)$ $\mathbb{V}[G(B)] = H(B)(1 - H(B))/(1 + \gamma)$

Dirichlet Process Mixture Models - DPM

$$G \sim DP(\gamma, H)$$

$$\theta_i | G \sim G$$

$$x_i | \theta_i \sim k(\cdot | \theta_i)$$

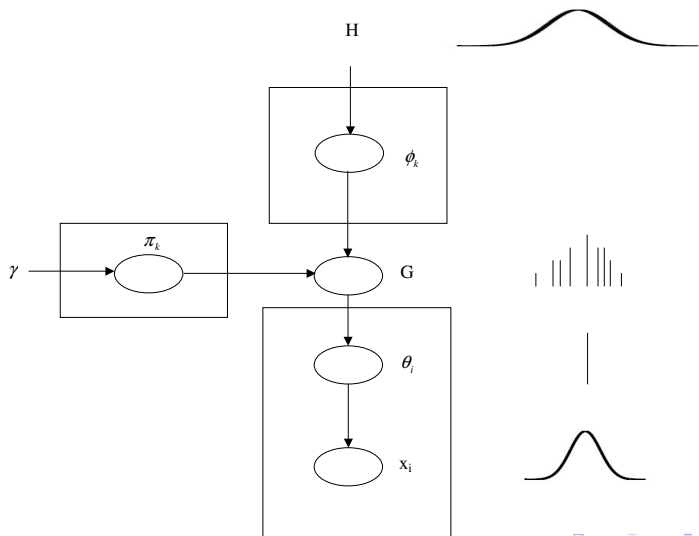


$$\text{Sample } G \sim DP(\gamma, H) \Leftrightarrow G = \sum_{k=1}^{\infty} \pi_k \delta_{\phi_k}$$

- $\phi_k \sim H$
- Infinite mixing proportions, $\sum_{k=1}^{\infty} \pi_k = 1 : \pi_k = \beta_k \prod_{l=1}^{k-1} (1 - \beta_l)$
- Infinite sequence of Beta r.v. : $\beta_k \sim \text{Beta}(1, \gamma)$
- Good quality of approximation for reasonable $K < \infty$

$$G = \sum_{k=1}^K \pi_k \delta_{\phi_k} \quad (\text{Ishwaran and James 2001})$$

Stick-Breaking Representation - SB



Modeling co-exposure to pesticides $M = 1$

$x = (x_1, \dots, x_n)$, with x_i a P dimensional vector $x_i = (x_{i1}, \dots, x_{iP})$

$$x_i | \theta_i \sim k(\cdot | \theta_i) \quad (1)$$

$$\theta_i | G \sim G \quad (2)$$

$$G \sim DP(\gamma, H) \quad (3)$$

- kernel k : a MultivariateNormal $\mathcal{N}(\mu, \tau), \phi = (\mu, \tau)$
- base probability measure H : the conjugate Normal-Wishart

Model for cocktails of pesticides $M = 100$

One level of hierarchy

$x_{im} = (x_{pim}, p = 1, \dots, P)$ with $i = 1, \dots, n$ and $m = 1, \dots, M$

$$x_{im} | \theta_{im} \sim k(\cdot | \theta_{im}) \quad (4)$$

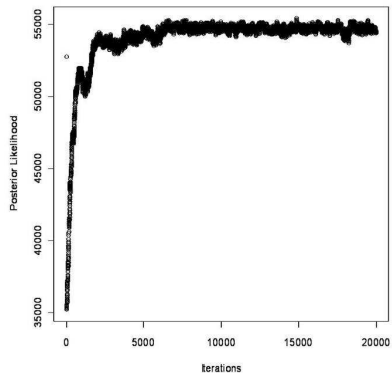
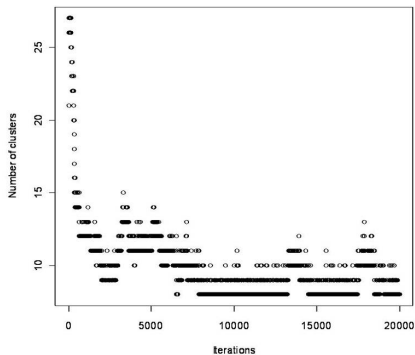
$$\theta_{im} | G_i \sim G_i \quad (5)$$

$$G_i \sim DP(\alpha_i, G_0) \quad (6)$$

$$G_0 \sim DP(\gamma, H) \quad (7)$$

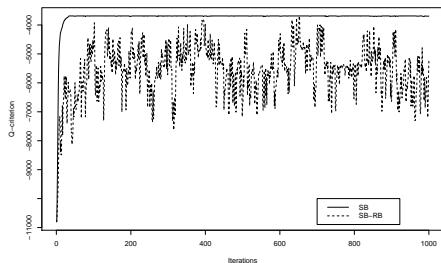
- Gibbs sampler based on the Stick-Breaking priors
- Random Block to reduce the heavy computational burden
 - at each Gibbs cycle select $d < P$ pesticides
 - a subset of random observations $x_i = (x_{i1}, \dots, x_{id})$ is used instead of $x_i = (x_{i1}, \dots, x_{iP})$

Example of convergence checking

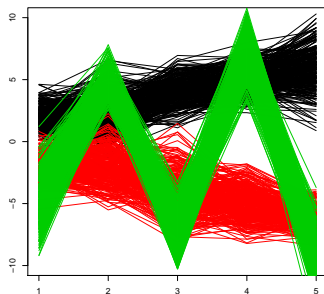


Validation data

Comparison between Stick-breaking (SB) and “Random-Block Stick-Breaking” (RB-SB) algorithms ($N = 30$ atoms, 30000 iterations)



Q-criterion of the first 1,000 iterations performed with the SB and the SB-RB algorithms

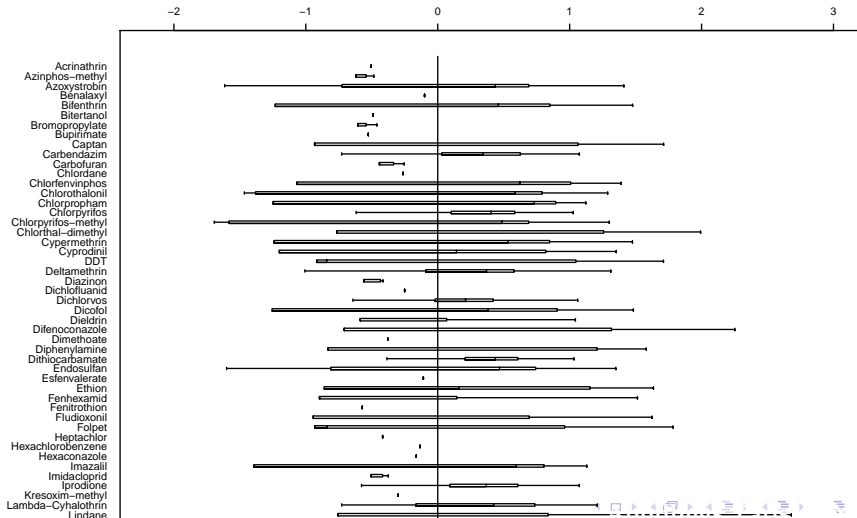


Optimal partition obtained with the SB algorithm : mixture of 3 gaussian distributions

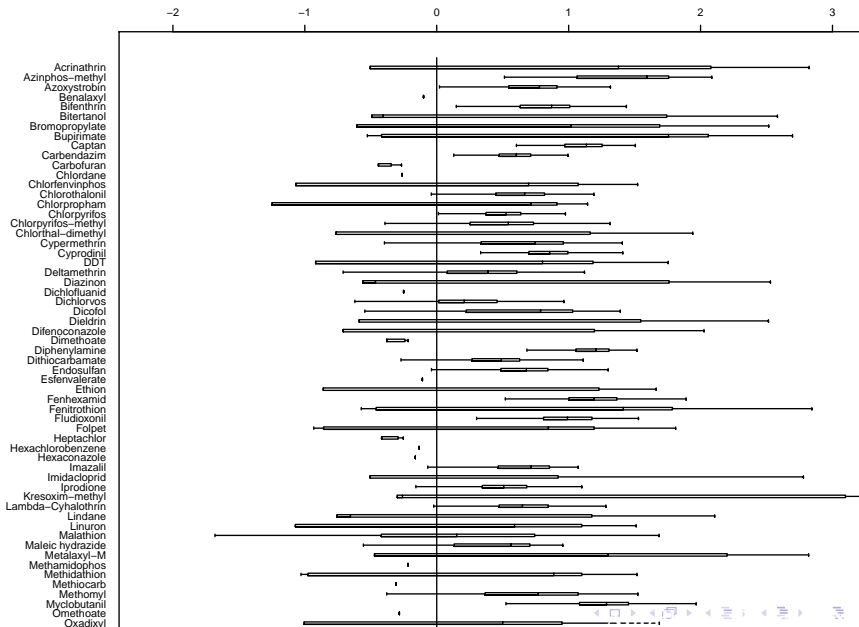
Children co-exposure to pesticides

Three main clusters of children with similar co-exposures

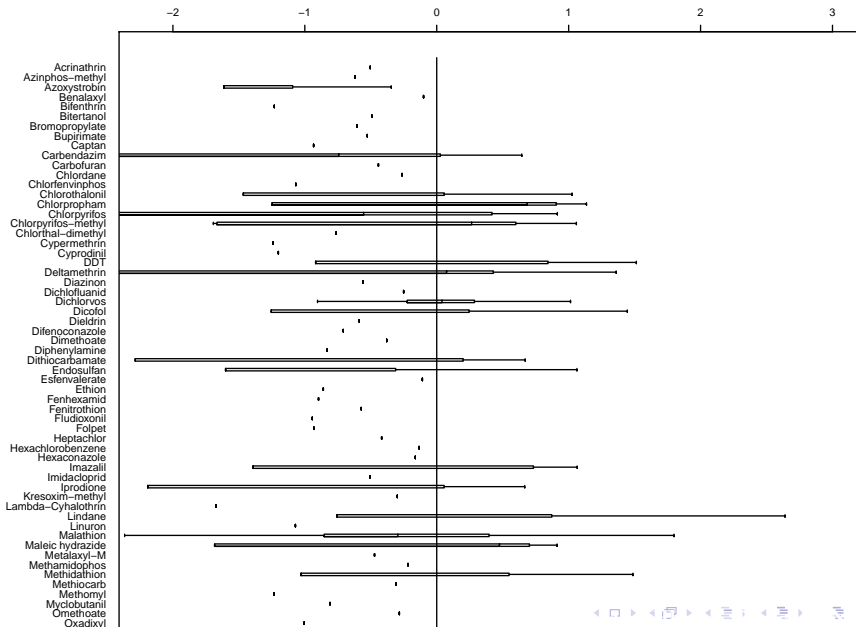
Box-plot of the cluster 1 : 699 children



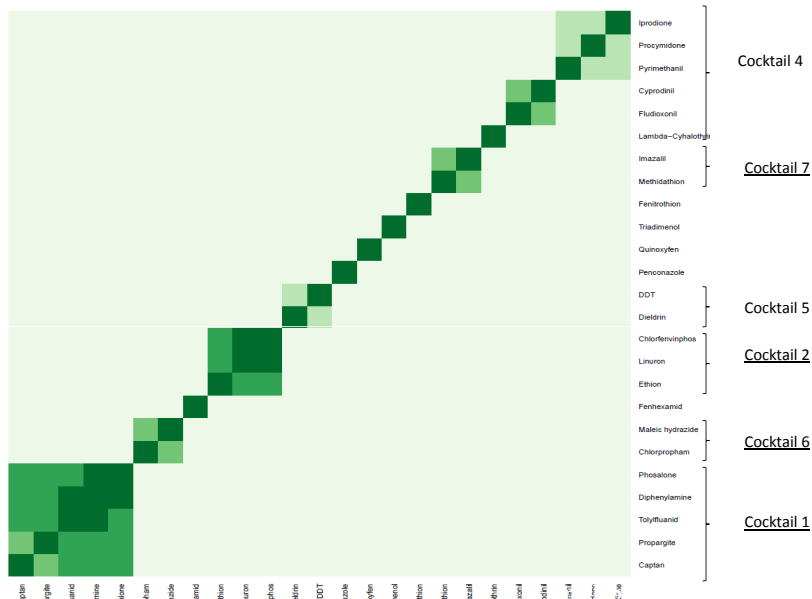
Box-plot of the cluster 2 : 238 children



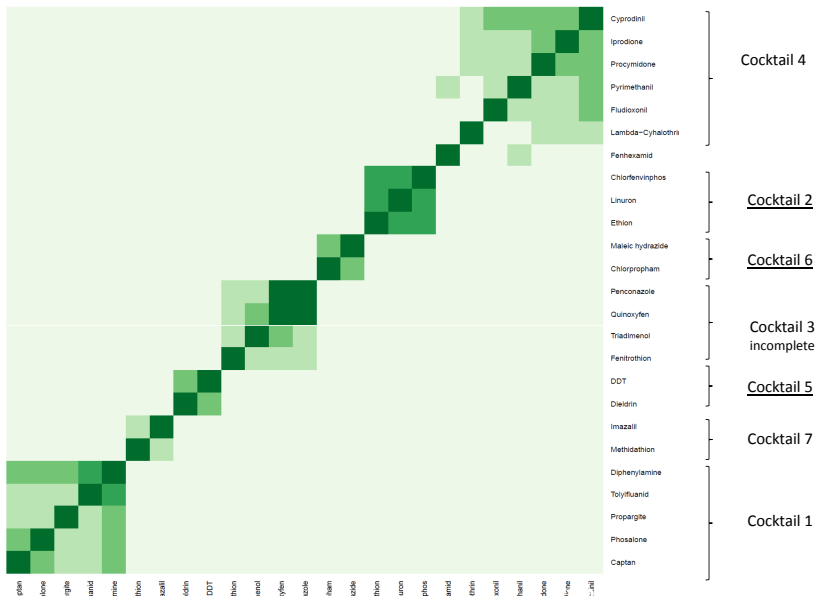
Box-plot of the cluster 3 : 239 children



Exposures correlations of cluster 1 : 699 children



Exposures correlations of cluster 2 : 238 children



- Prior distribution on γ : $\text{Gamma}(a_\gamma, b_\gamma)$
- Use Poisson-Dirichlet process instead of DP : slow convergence
 - a_γ : $\text{Uniform}[0, 1]$
 - b_γ : $\text{Gamma}(0.01, 0.01)$
- Hierarchical model not really adapted
- Pareto kernel for k
- DP for censored data
- 3 of the 7 defined cocktails have toxicological effects

Thanks

- F. Héraud, JC Leblanc and JL Volatier, AFSSA
- ORP (Pesticide Residues Observatory) for financial support
- ANR (National Research Agency) for financial support
- Thank **you** for your attention

