

Stress Testing & Reverse Stress Testing

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Defining halfspace depth

Given $\mathbf{y} \in \mathbb{R}^d$, and a direction $\mathbf{u} \in \mathbb{R}^d$, define the closed half space

$$H_{\mathbf{y},\mathbf{u}} = \{\mathbf{x} \in \mathbb{R}^d \text{ such that } \mathbf{u}'\mathbf{x} \leq \mathbf{u}'\mathbf{y}\}$$

and define depth at point \mathbf{y} by

$$\text{depth}(\mathbf{y}) = \inf_{\mathbf{u}, \mathbf{u} \neq \mathbf{0}} \{\mathbb{P}(H_{\mathbf{y},\mathbf{u}})\}$$

i.e. the smallest probability of a closed half space containing \mathbf{y} .

The empirical version is (see Tukey, 1975)

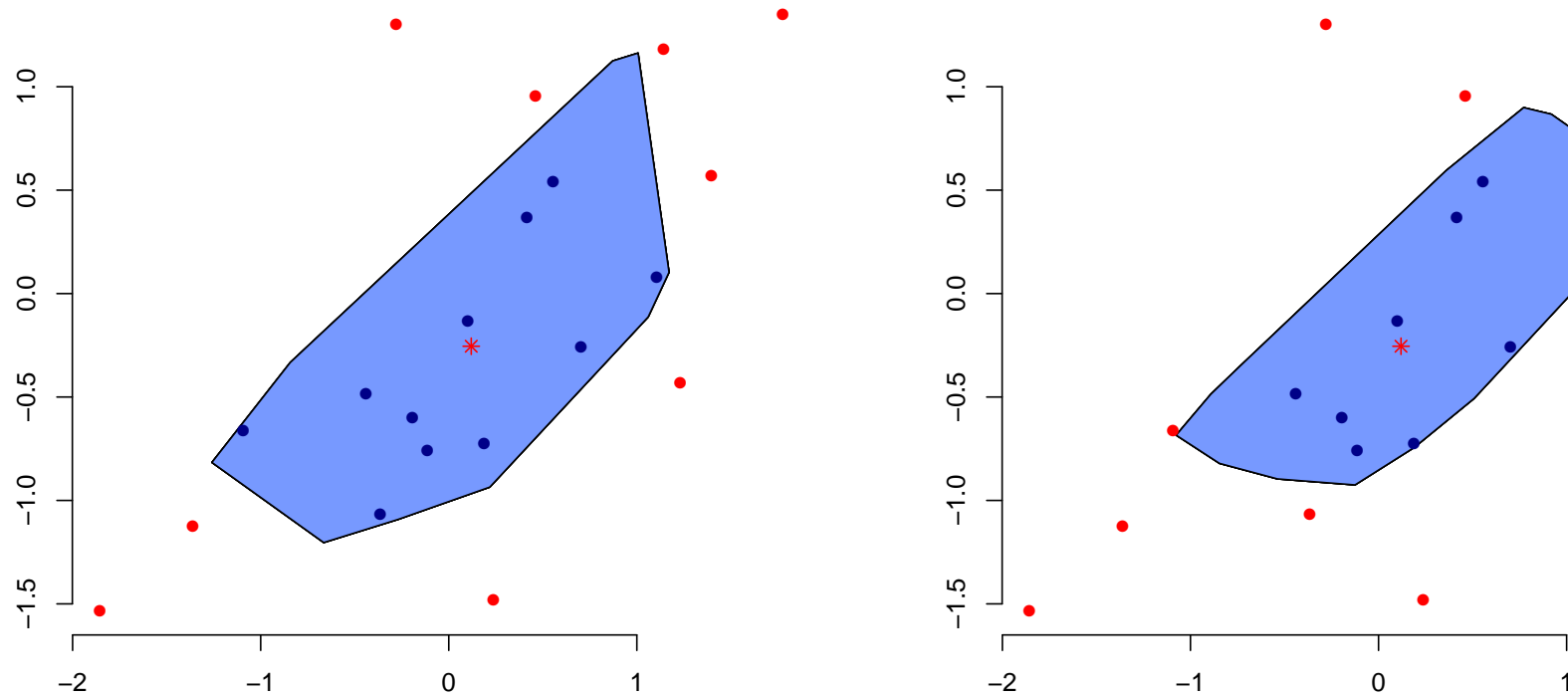
$$\text{depth}(\mathbf{y}) = \min_{\mathbf{u}, \mathbf{u} \neq \mathbf{0}} \left\{ \frac{1}{n} \sum_{i=1}^n \mathbf{1}(\mathbf{X}_i \in H_{\mathbf{y},\mathbf{u}}) \right\}$$

For $\alpha > 0.5$, define the **depth set** as

$$D_\alpha = \{\mathbf{y} \in \mathbb{R} \in \mathbb{R}^d \text{ such that } \geq 1 - \alpha\}.$$

The empirical version is can be related to the **bagplot** (Rousseeuw & Ruts, 1999).

Empirical sets extremely sensitive to the algorithm



where the **blue set** is the empirical estimation for D_α , $\alpha = 0.5$.

The bagplot tool

The **depth** function introduced here is the multivariate extension of standard univariate depth measures, e.g.

$$\text{depth}(x) = \min\{F(x), 1 - F(x^-)\}$$

which satisfies $\text{depth}(Q_\alpha) = \min\{\alpha, 1 - \alpha\}$. But one can also consider

$$\text{depth}(x) = 2 \cdot F(x) \cdot [1 - F(x^-)] \text{ or } \text{depth}(x) = 1 - \left| \frac{1}{2} - F(x) \right|.$$

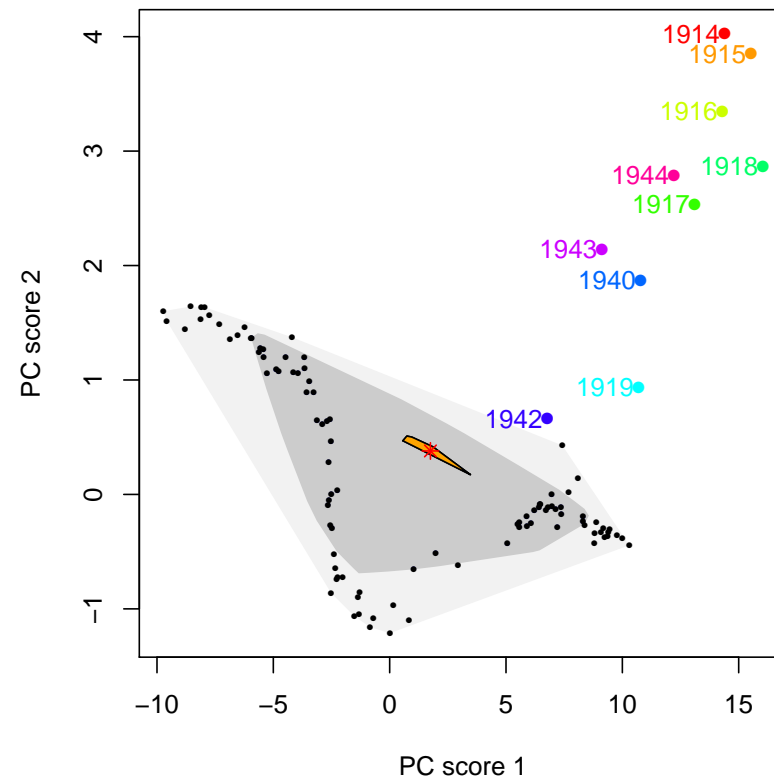
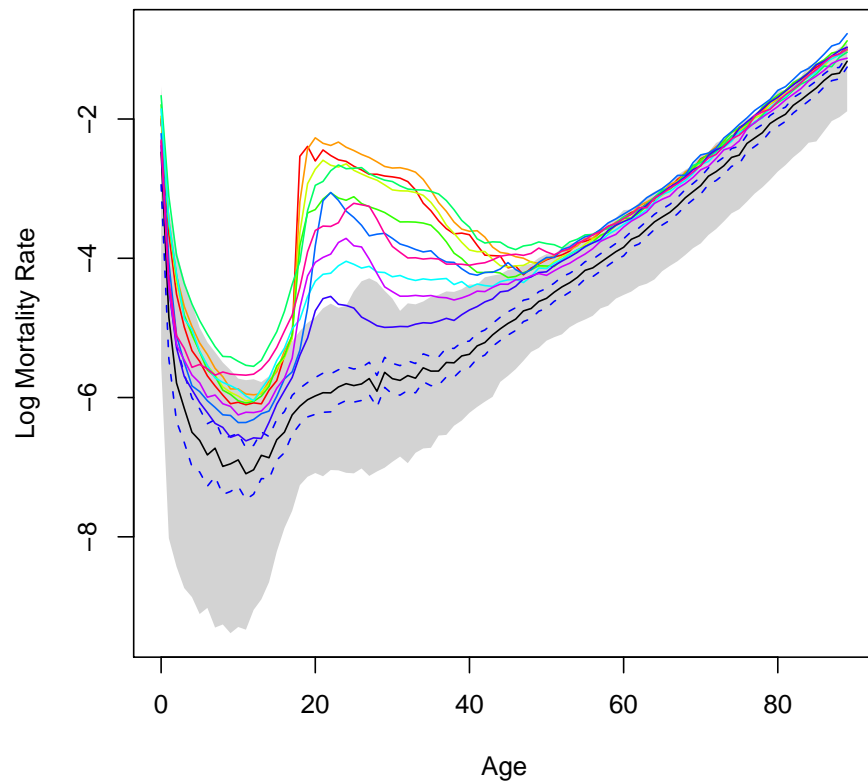
Possible extensions to *functional bagplot*. Consider a set of functions $f_i(x)$, $i = 1, \dots, n$, such that

$$f_i(x) = \mu(x) + \sum_{k=1}^{n-1} z_{i,k} \varphi_k(x)$$

(i.e. principal component decomposition) where $\varphi_k(\cdot)$ represents the eigenfunctions. Rousseeuw et al., 1999 considered bivariate depth on the first two scores, $\mathbf{x}_i = (z_{i,1}, z_{i,2})$. See Ferraty & Vieu (2006).

The bagplot tool for mortality models

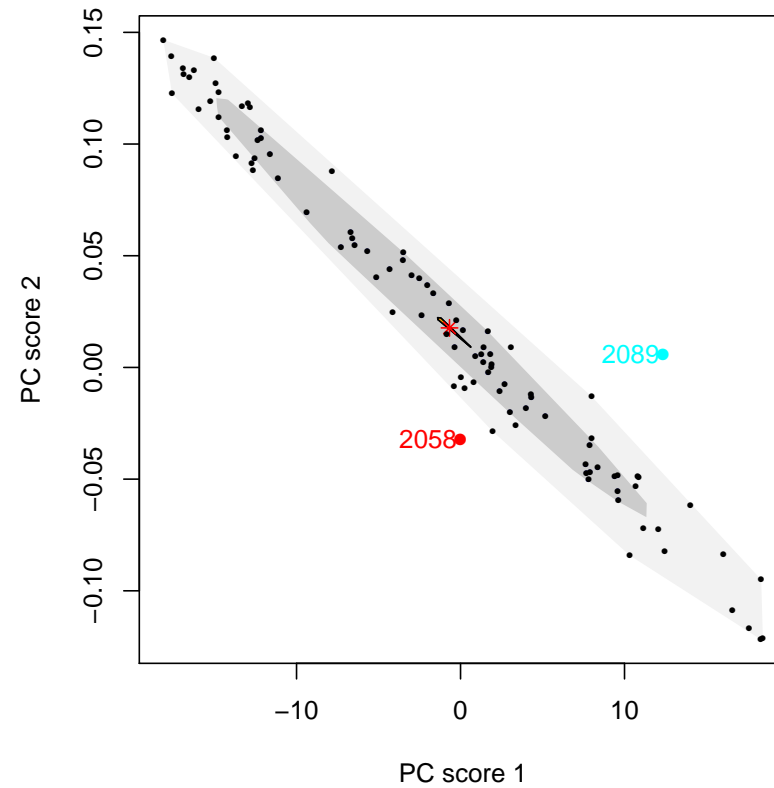
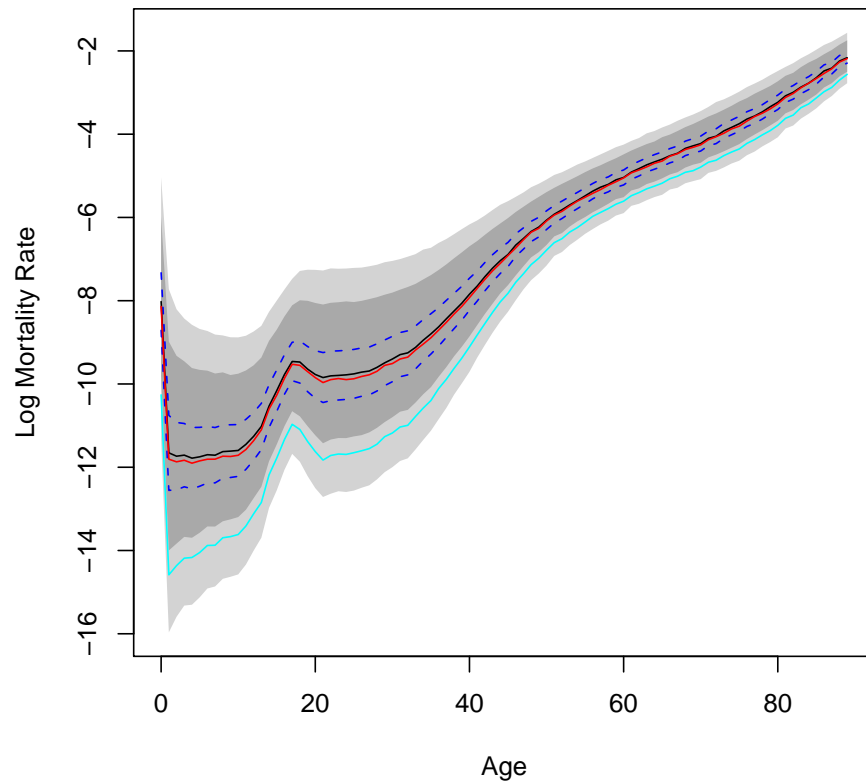
On the a French dataset, we have the following *past* outliers,



(here male log-mortality rates in France from 1899 to 2005).

The bagplot tool for mortality models

Using **functional bagplot techniques** it is also possible to identify outliers in stochastic scenarios,



Further references

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