

**Neighbourhood models induced by
the Euclidean distance and
the Kullback-Leibler divergence**

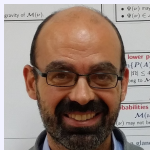
Ignacio Montes
University of Oviedo

Asturias: Natural Paradise



IP group in Oviedo

IP researchers



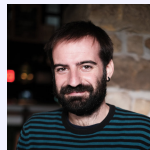
E.Miranda



I.Montes

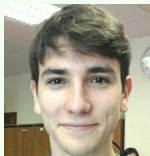


J.J.Salamanca



R.Pérez

Young people



J.Álvarez



D.Nieto



M.Porrón



A.Presa

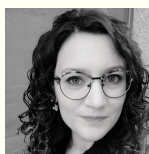
Close reserachers



A.Bouchet



S.Díaz



I.Mariñas

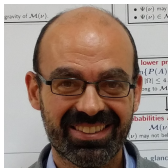


S.Montes

Distortion models







S.Destercke



E.Miranda



P.R.Alonso

-  Unifying neighbourhood and distortion models: Part I. Montes, Miranda, Destercke. IJGS 2020.
-  Unifying neighbourhood and distortion models: Part II. Montes, Miranda, Destercke. IJGS 2020.
-  Processing distortion models: a comparative study. Destercke, Montes, Miranda. IJAR 2022.
-  Distortion models for estimating human error probabilities. Alonso, Montes, Miranda. SS 2023.

Distortion models

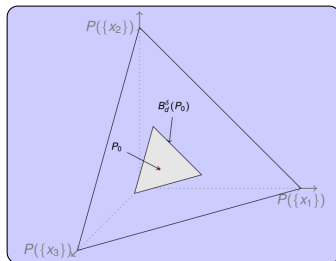
Ingredients

P_0 : probability measure

δ : distortion parameter

d : distorting function

Example



The model

$$B_d^\delta(P_0) = \{P \mid d(P, P_0) \leq \delta\}$$

$$\underline{P}_d(f) = \inf\{P(f) \mid P \in B_d^\delta(P_0)\}$$

$$\mathcal{M}(\underline{P}_d) = \{P \mid P(f) \geq \underline{P}_d(f) \forall f\}$$

d continuous and convex

Particular models

Pari Mutuel Model

$$\underline{P}_{PMM}(A) = \max\{0, (1 + \delta)P_0(A) - \delta\}$$

Pelessoni et al., 2010
Montes et al., 2019
Walley, 1991

Linear vacuous model

$$\underline{P}_{LV}(A) = (1 - \delta)P_0(A) \quad A \neq \mathcal{X}$$

Huber, 1981
Walley, 1991

Constant odds ratio

$$\underline{P}_{COR}(A) = \frac{(1-\delta)P_0(A)}{1-\delta P_0(A)}$$

Benavoli and Zaffalon, 2013
Walley, 1991

Total Variation model

$$\underline{P}_{TV}(A) = \max\{0, P_0(A) - \delta\} \quad A \neq \mathcal{X}$$

Montes et al., 2020
Herron et al., 1997

Vertical barrier models

$$\underline{P}_{VB}(A) = \max\{bP_0(A) + a, 0\} \quad \begin{array}{l} a \leq 0, b > 0 \\ a + b \in (0, 1) \end{array}$$

Pelessoni et al., 2021
Corsato et al., 2019

Increasing transformation

$$\underline{P}(A) = g(P_0(A))$$

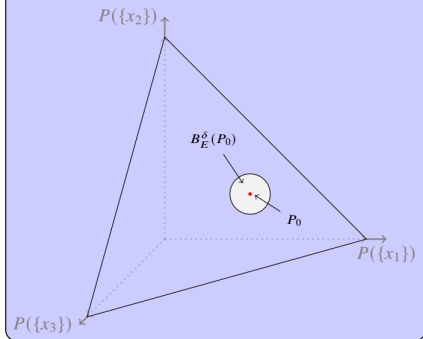
$g : [0, 1] \rightarrow [0, 1]$ increasing
 $g(0) = 0, g(1) = 1$

Bronevich, 2005

Aim of the paper

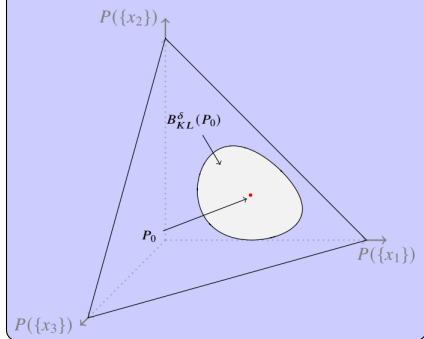
Euclidean model

$$d_E(P, Q) = \sqrt{\sum_{i=1}^n (p_i - q_i)^2}$$



Kullback-Leibler model

$$D_{KL}(P, Q) = \sum_{i=1}^n p_i \log \left(\frac{p_i}{q_i} \right)$$



Euclidean model

Expression for \underline{P}_E :

$$\underline{P}_E(f) = ??$$

$$\underline{P}_E(A) = ??$$

Extreme points:

$$\text{ext}(B_E^\delta(P_0)) = ??$$

Properties:

2-monotonicity ??

Complete-monotonicity ??

Conditioning ??

KL-model

Expression for \underline{P}_{KL} :

$$\underline{P}_{KL}(f) = ??$$

$$\underline{P}_{KL}(A) = ??$$

Extreme points:

$$\text{ext}(B_{KL}^\delta(P_0)) = ??$$

Properties:

2-monotonicity ??

Complete-monotonicity ??

Conditioning ??

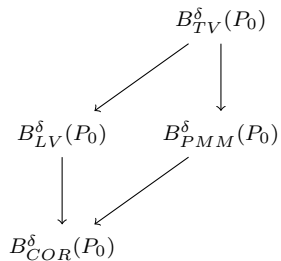
Comparison: properties

	2-monot.	∞ -monot.	Prob.interval	Conditioning
Euclidean-model...				
...on gambles	??	??	??	??
...on events	??	??	??	??
KL-model...				
...on gambles	??	??	??	??
...on events	??	??	??	??
PMM	✓	✗	✓	✓
LV	✓	✓	✓	✓
COR...				
...on gambles	✗	✗	✗	✓
...on events	✓	✓	✗	✓
TV	✓	✗	✓	✓
Vertical Barrier models	✓	✗	✗	✓

Comparison: amount of imprecision

$$B_{KL}^{\delta}(P_0) \quad ??$$

$$B_E^{\delta}(P_0) \quad ??$$



1. WHAT CAN I DO?

I HAVE A PROBABILITY MEASURABLE MODELS BUT I WANT TO R-ORDER MY MODEL.

WHY DO NOT YOU USE DISTANCE MODELS?

1. DISTANCE MODELS

1. PROBABILITIES
 P_i probability measure
 A decision parameter
 d distorting function

2. THE MODEL
 $H_i^d(P_i) = (P \mid d(P, P_i) \leq \delta)$
 $E_d(f) = \inf\{E(f) \mid P \in H_i^d(P_i)\}$
 $M(E_d) = \{P \mid E_d(f) \geq E(f) \forall f\}$

3. PARTICULAR CASES
 3.1 Known D -models
 $E_d(f) = \max\{(1-\delta)P_i(A) - \delta\}$
 $E_d(A) = (1-\delta)P_i(A)$
 $E_d(A) = \frac{1-\delta P_i(A)}{1-\delta}$
 3.2 Distance-based models
 $E_d(f) = \max\{0, d(A) - \delta\}$
 3.3 Vertical barrier models
 $E_d(f) = \max\{d(P_i(A) + \delta, 0)\}$
 3.4 Transformations and distortions
 $E_d(A) = g(P_i(A))$

2. WHY DO NOT YOU CONSIDER THE EUCLIDEAN DISTANCE?

BUT IF YOU USE A DISTANCE, YOU ARE RIGHT.

1. EUCLIDEAN MODEL
 $d_E(P, P_i) = \sqrt{\sum_{j=1}^n (p_j - p_i^j)^2}$
 d_E continuous and convex

2. Lower prevision
 $f = \sum_{j=1}^n a_j x_j$
 $H_i^d(P_i) = \{x \mid \sum_{j=1}^n (x_j - p_i^j)^2 \leq \delta^2\}$
 $E_d(f) = P_i(f) - \delta \sqrt{\sum_{j=1}^n a_j^2}$

3. Lower probability
 $E_d(A) = P_i(A) - \delta \sqrt{\sum_{j=1}^n a_j^2}$

4. Properties
 1) E_d is not 2-monotone...
 2) ...but it is 2-monotone on events.
 3) the model is not generated under conditioning.
 4) extreme points:
 $\text{ext}(H_i^d(P_i)) = \{P \mid d_E(P, P_i) = \delta\}$

5. Example

3. WOULDN'T IT BE BETTER TO USE A DIVERGENCE?

BUT FOR COMPARING PROBABILITIES, NONE POINT.

1. KL-DIVERGENCE
 $D_{KL}(P, P_i) = \sum_{j=1}^n p_j \ln\left(\frac{p_j}{p_i^j}\right)$
 D_{KL} continuous and convex

2. Lower prevision
 $f = \sum_{j=1}^n a_j x_j$
 $E_d(f) = \min_{x \in H_i^d(P_i)} \sum_{j=1}^n a_j x_j$
 subject to
 $\sum_{j=1}^n a_j p_i^j = 1, \sum_{j=1}^n a_j p_j \ln(p_j) = d$

3. Lower probability
 $E_d(A) = \alpha P_i(A)$
 where $\alpha \in (0, 1]$ is the unique solution of
 $\alpha P_i(A) \ln(\alpha) + (1 - \alpha P_i(A)) \ln\left(\frac{1 - \alpha P_i(A)}{1 - P_i(A)}\right) = d$

4. Properties
 1) D_{KL} is not 2-monotone...
 2) ...but it is 2-monotone on events.
 3) the model is not generated under conditioning.
 4) extreme points:
 $\text{ext}(H_i^d(P_i)) = \{P \mid D_{KL}(P, P_i) = \delta\}$

5. Example

4. WHICH ONE SHOULD I USE?

AMONG ALL THE MODELS, IT DEPENDS ON WHAT YOU ARE LOOKING FOR.

1. PROPERTIES

	continuous	convex	2-monotone	2-monotone on events	generated under conditioning
Euclidean model	✓	✓	✗	✓	✗
KL model	✓	✓	✗	✓	✗
Vertical barrier	✗	✓	✓	✗	✗
Known D-model	✗	✓	✓	✗	✗
Distance-based	✗	✓	✓	✗	✗
Vertical barrier	✗	✓	✓	✗	✗

2. Amount of dispersion

3. Example

5. DOES ANYBODY CARE ABOUT DISTANCE MODELS?

MANY PEOPLE!

1. REFERENCES

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 [2] Montes, Miranda, Detecke. Unifying neighbourhood and distortion models. Paris I (LACS, 2020)

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 [4] Montes, Miranda, Detecke. The maximal probability in an distortion model. IFAC, 2018.
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 [6] Walley. Statistical Reasoning with Imprecise Probabilities. 1991.

Distance-based models: [7] Hesse, Seidenfeld, Wasserman. Decision conditioning: further results on dilation. PDSG, 1997.

Vertical barrier models: [8] Corcuera, Pálsson, Vígg. Nearly-Linear uncertainty measures. IJAR, 2019.

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KL-divergence in IP: [10] Moral. Dissolving imprecise probability: The Mathematics of the uncertain. 2018.