

TIME INTENSITY curves: statistical treatment

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El Mostafa QANNARI*



OVERVIEW

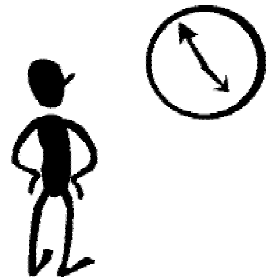
- **Time intensity curves**
 - **Description**
 - **A high variability between assessors meaning difficulties for the analysis**
- **Statistical processing of time intensity data**
 - Pretreatment and average among repetitions
 - The ‘ skeleton ’, a tool to describe the ‘signature’
 - Non linear transformation of time in order to reduce the ‘ signature ’ effect
 - Characterization of the products



TIME INTENSITY CURVES :

Description

- Different evolution of the intensity overtime from one product to another

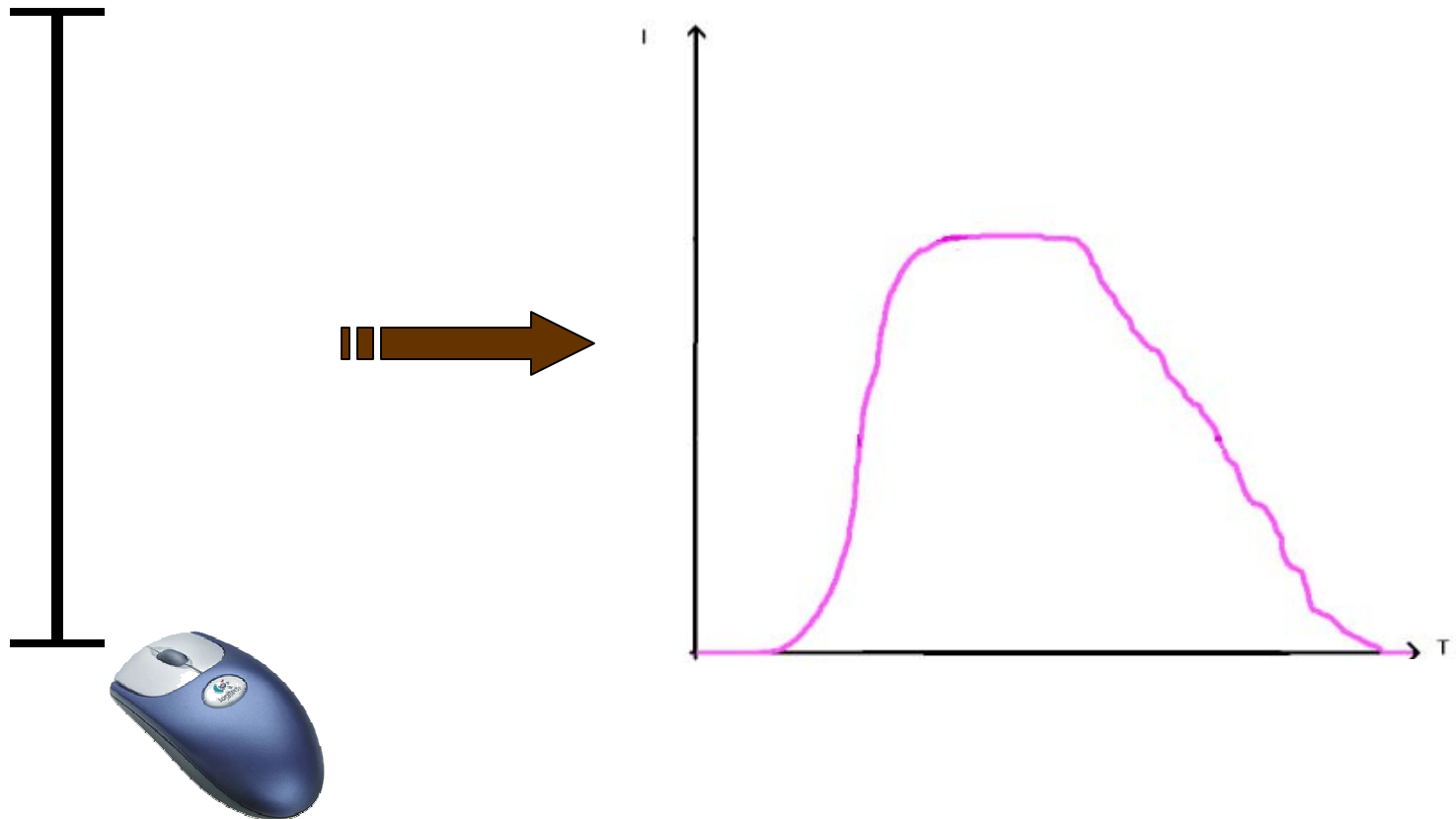


⇒ More complete sensory characterization



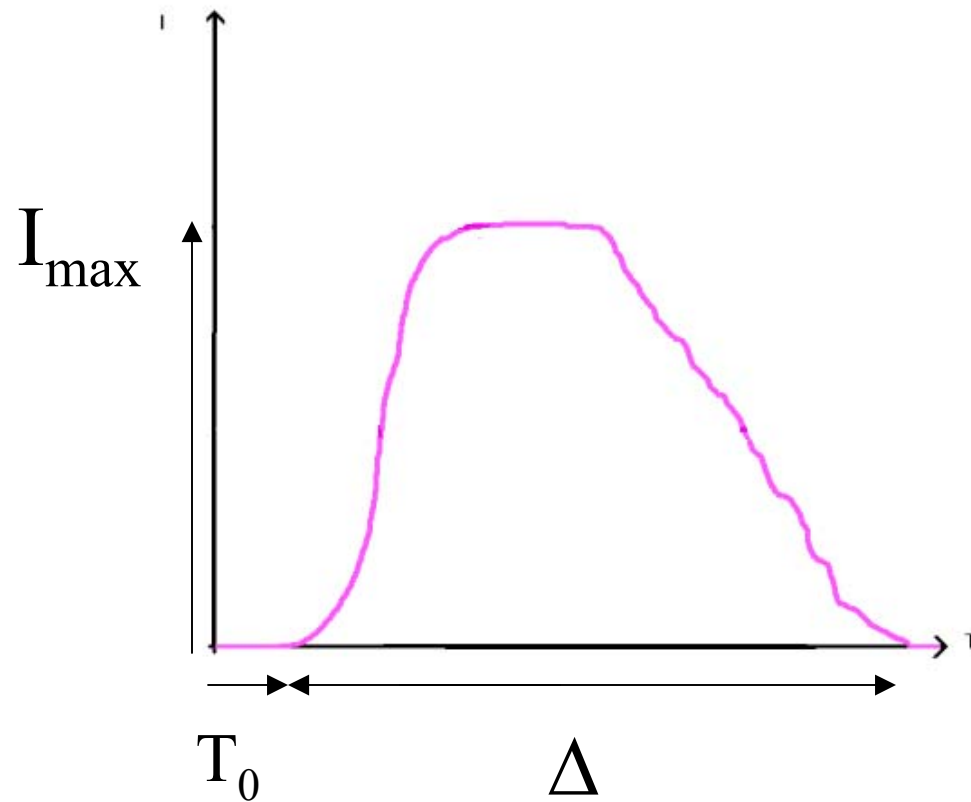
TIME INTENSITY CURVES :

Description



TIME INTENSITY CURVES :

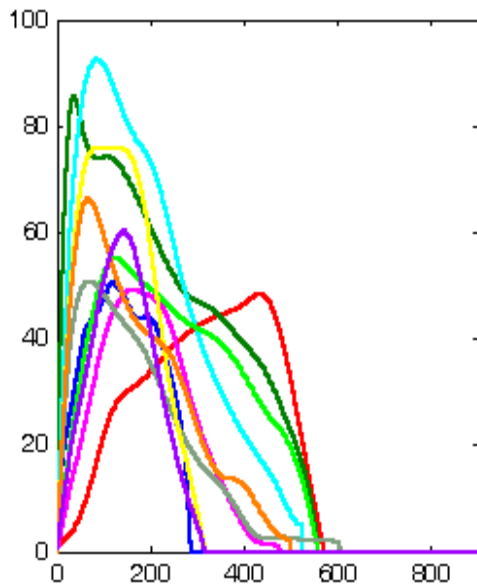
Description



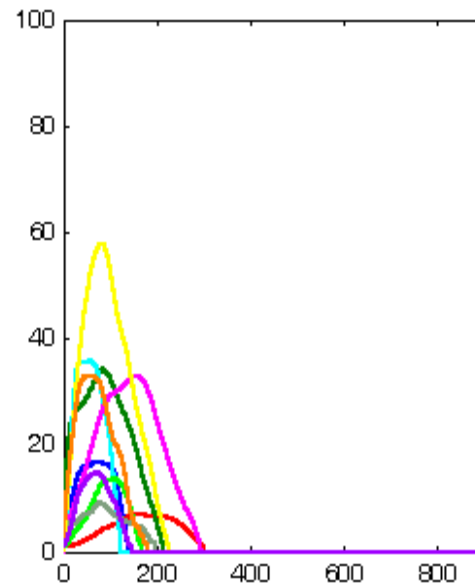
TIME INTENSITY CURVES :

Difficulties to Analyze the Curves...

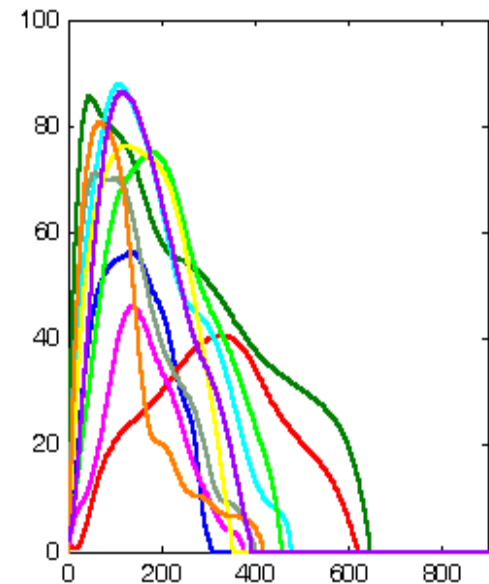
all assessors
product 2



all assessors
product 3



all assessors
product 4

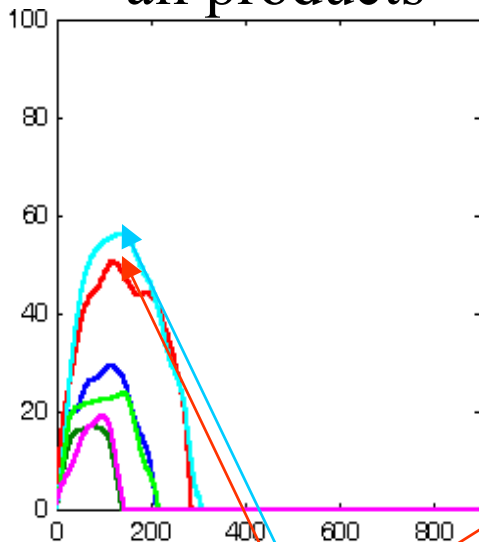


TIME INTENSITY CURVES :

...because of a high variability between assessors

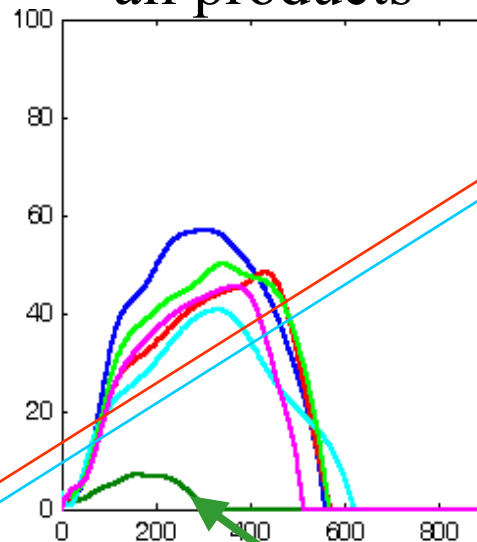
assessor 1

all products



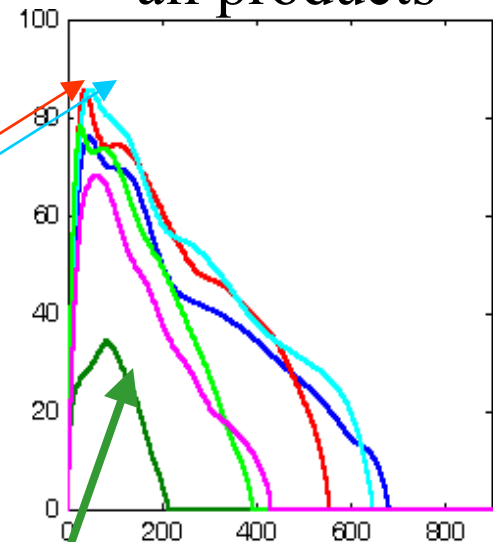
assessor 2

all products



assessor 3

all products



Products 2 and 4 have a large perceived intensity with a fast increase.

The product 3 has a small perceived intensity with a slow increase and a small duration



SAMPLES

Sweetness of Six chocolate drinks

- 10 experienced assessors
- 3 repetitions
- The time of recording = 90 seconds
(with a point of measurement all the tenths of a second).

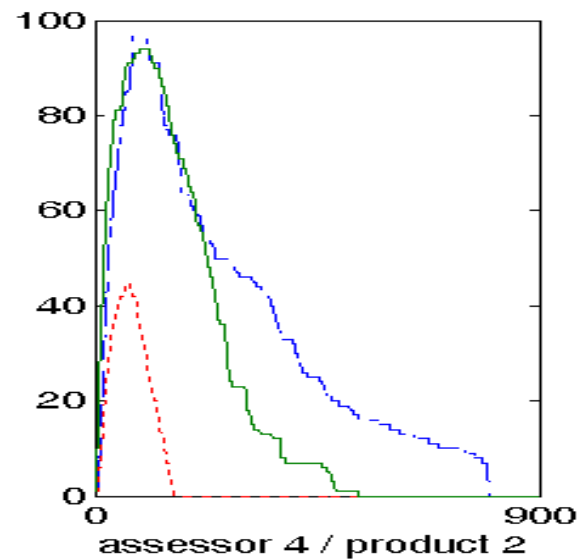
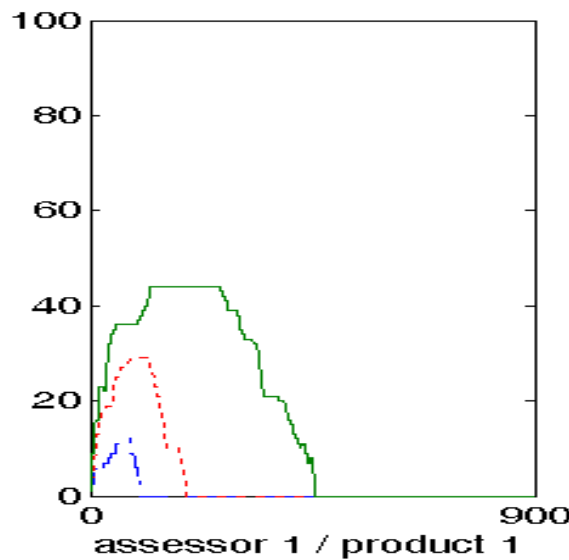
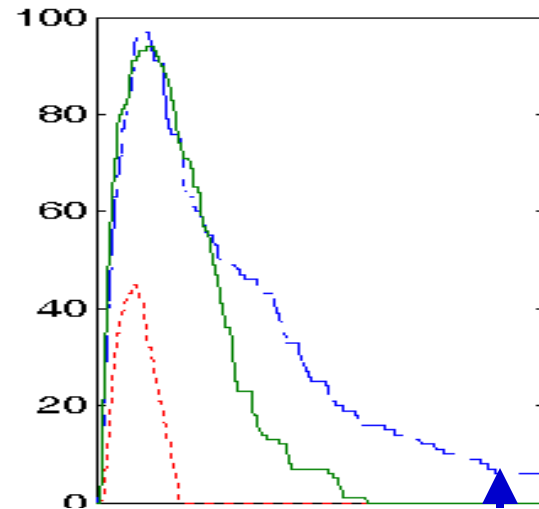
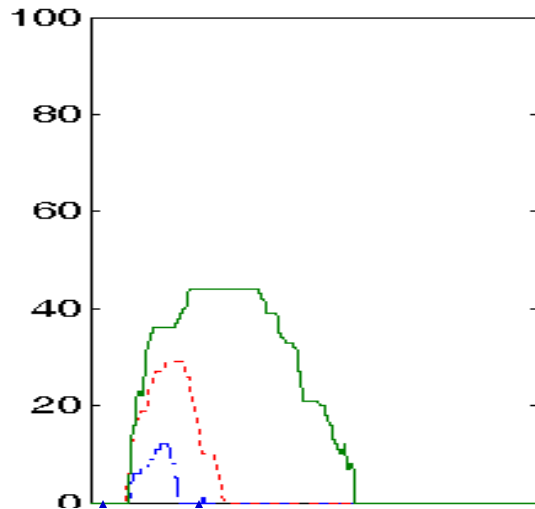


OVERVIEW

- Time intensity curves
 - Description
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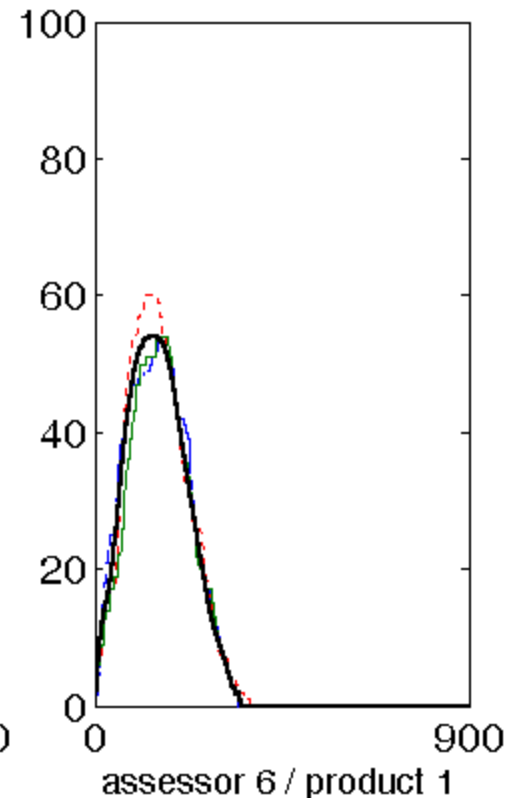
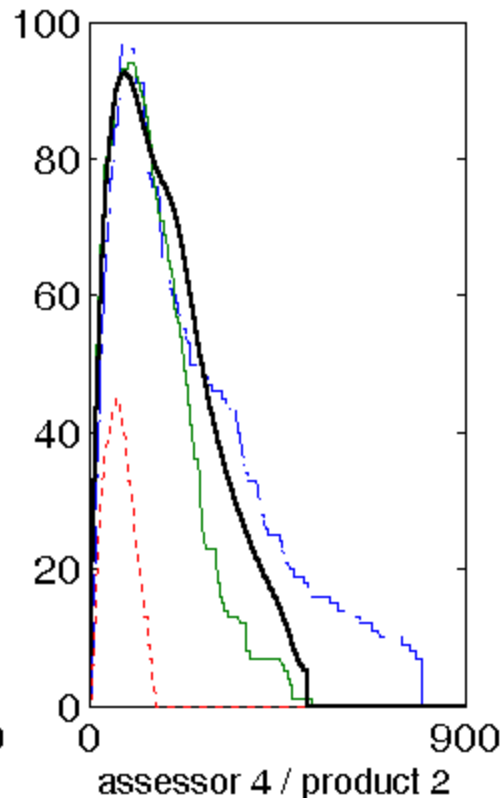
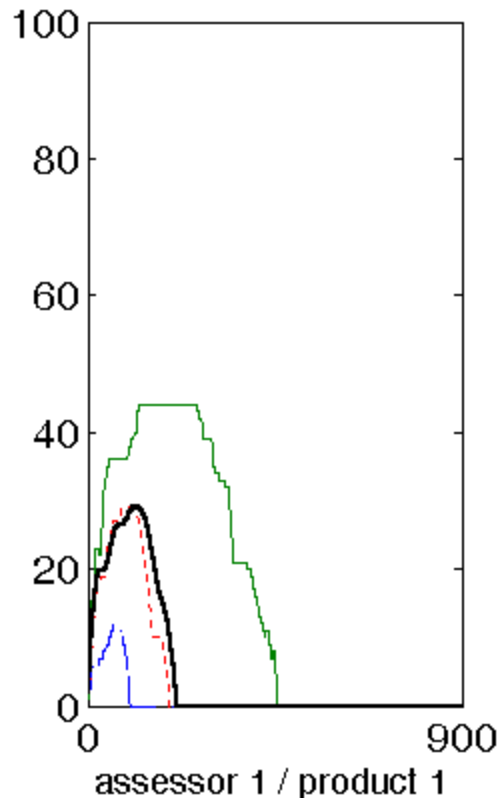
PRETREATMENT



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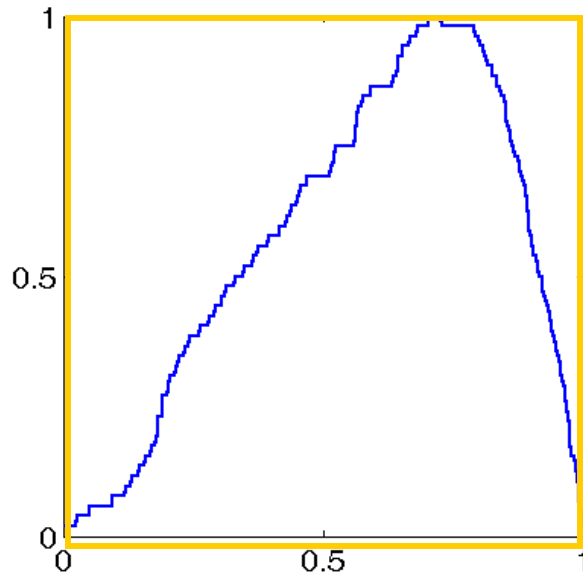
WEIGHTED AVERAGE OVER REPETITIONS



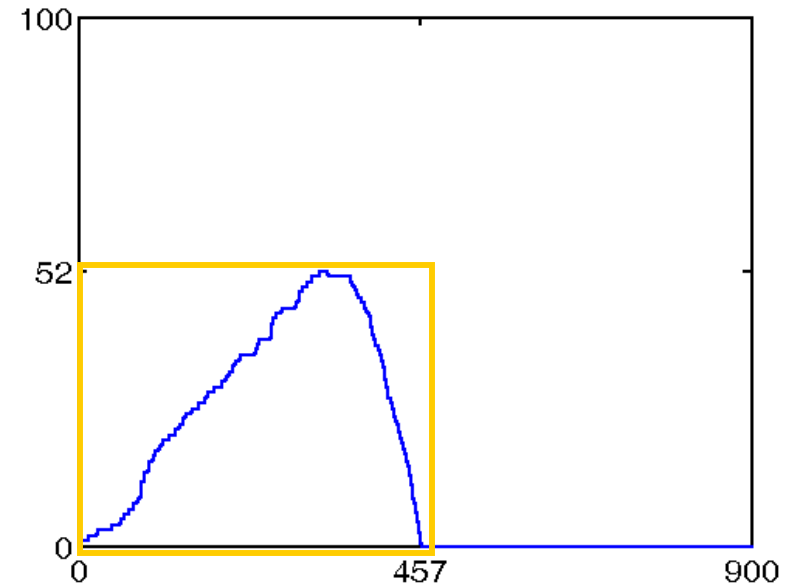
THE ' SKELETON '

a Tool to Describe Shape of Curves

- Select I_{\max} and Δ
 $I_{\max} = 52$ $\Delta = 457$
- Keep only the shape:
 $[0,1] \times [0,1]$



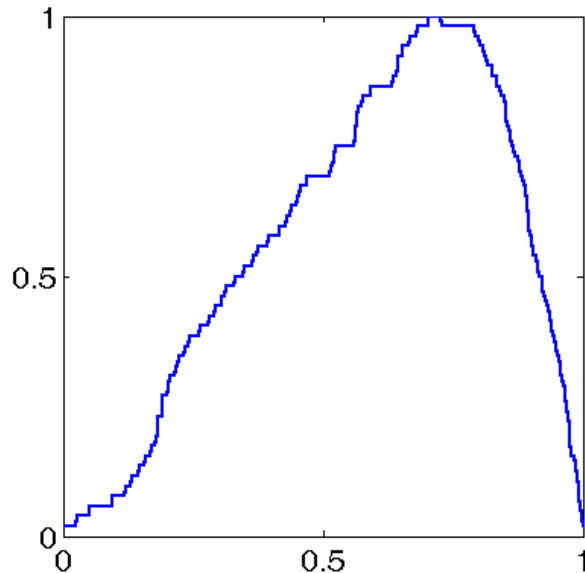
skeleton



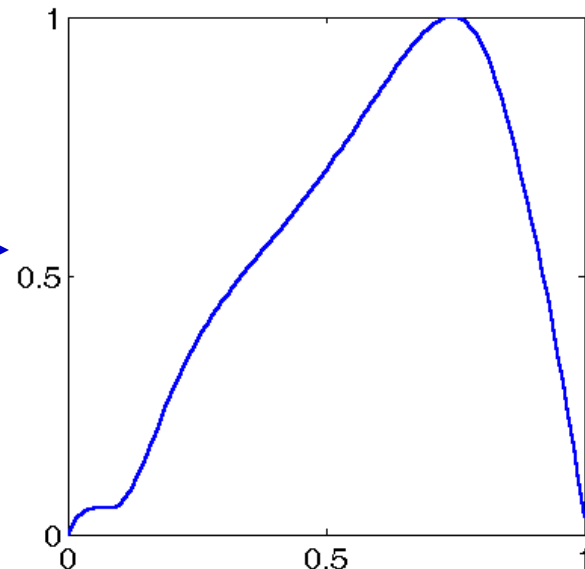
THE ' SKELETON '

a Tool to Describe Shape of Curves

- Select I_{\max} and Δ
 $I_{\max} = 52$ $\Delta = 457$
- Keep only the shape:
 $[0,1] \times [0,1]$
- Project in a basis
of spline functions



skeleton



c =

0.0
0.1
0.0
0.3
0.5
0.6
0.7
0.8
1.0
1.0
0.6
0.3
0.0



THE ' SKELETON '

a Tool to Describe Shape of Curves

Formally, the standardization is :

skeleton

$$y_{ij}(t) = \text{Imax}_{ij} \left(Z_{ij} \left(\frac{t}{\Delta_{ij}} \right) \right) + \varepsilon_{ij}(t)$$

And the decomposition on a B-splines basis:

$$Z(t^*) = c_1 \Phi_1(t^*) + c_2 \Phi_2(t^*) + \dots + c_S \Phi_S(t^*)$$

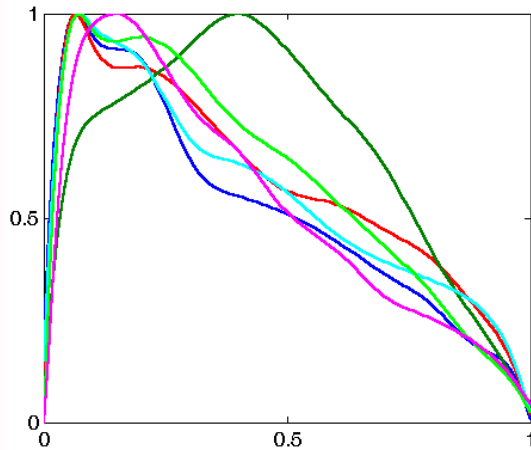
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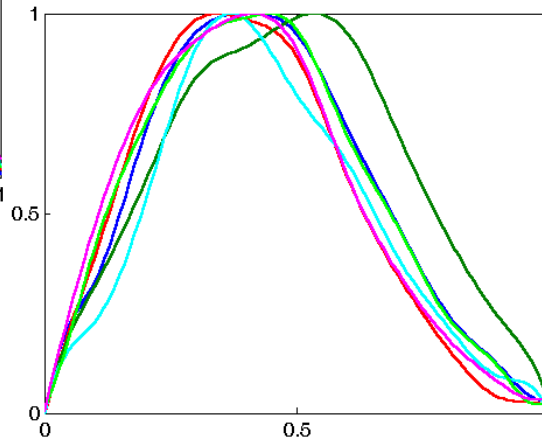


THE 'SIGNATURE'

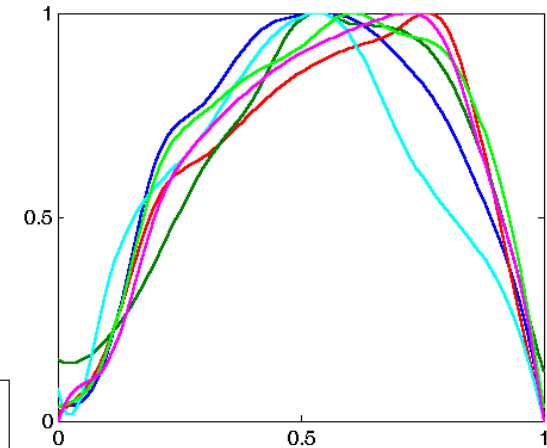
The skeleton reflects the signature of each assessor



assessor 3
all products



assessor 6
all products



assessor 2
all products



OBJECTIVES

- ⇒ To remove the variability between assessors with the aim of better discriminating the products.



METHOD

Assessor effect for the parameters I_{max} and Δ :

$I_{max_{ij}}$ for each product i and assessor j

I_{max_j} : average for assessor j

$I_{max_{..}}$: global average

$$\Rightarrow I_{max_{ij}}' = I_{max_{ij}} + (I_{max_{..}} - I_{max_j})$$

$$\Rightarrow \Delta_{ij}' = \Delta_{ij} + (\Delta_{..} - \Delta_j)$$



METHOD

Assessor effect for the ' skeleton ':

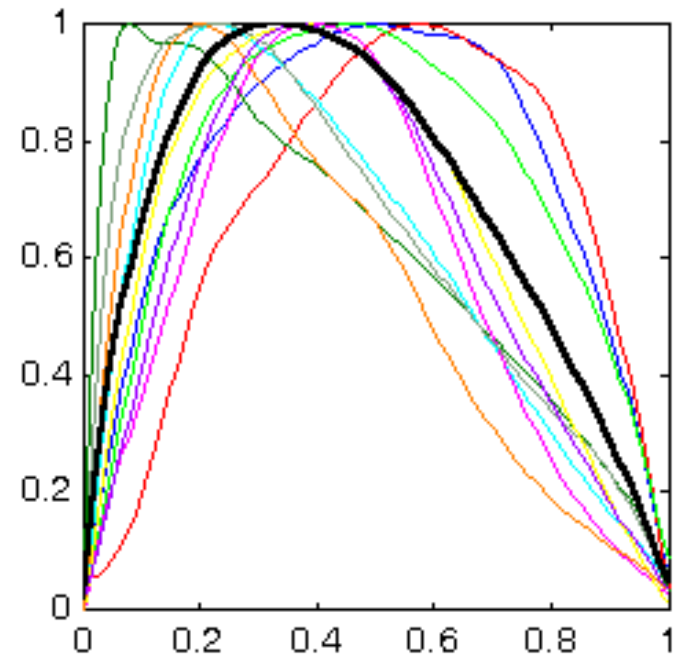
Z_{ij} for each product i and assessor j

Z_j : average for assessor j

$Z_{..}$: global average

$Z_{ij}' = ?$

$Z_{ij}'(t) = Z_{ij}(u_j(t))$



References

« Landmarks based registration of curves »

Kneip and Gasser (1992)

« Dynamic time warping »

Wang and Gasser (1997), Ramsay and Silverman (1997)



Formalization of DTW

For each assessor j , we have a function Z_j representing his or her signature (average skeleton). We are looking for the transformation u_j so that:

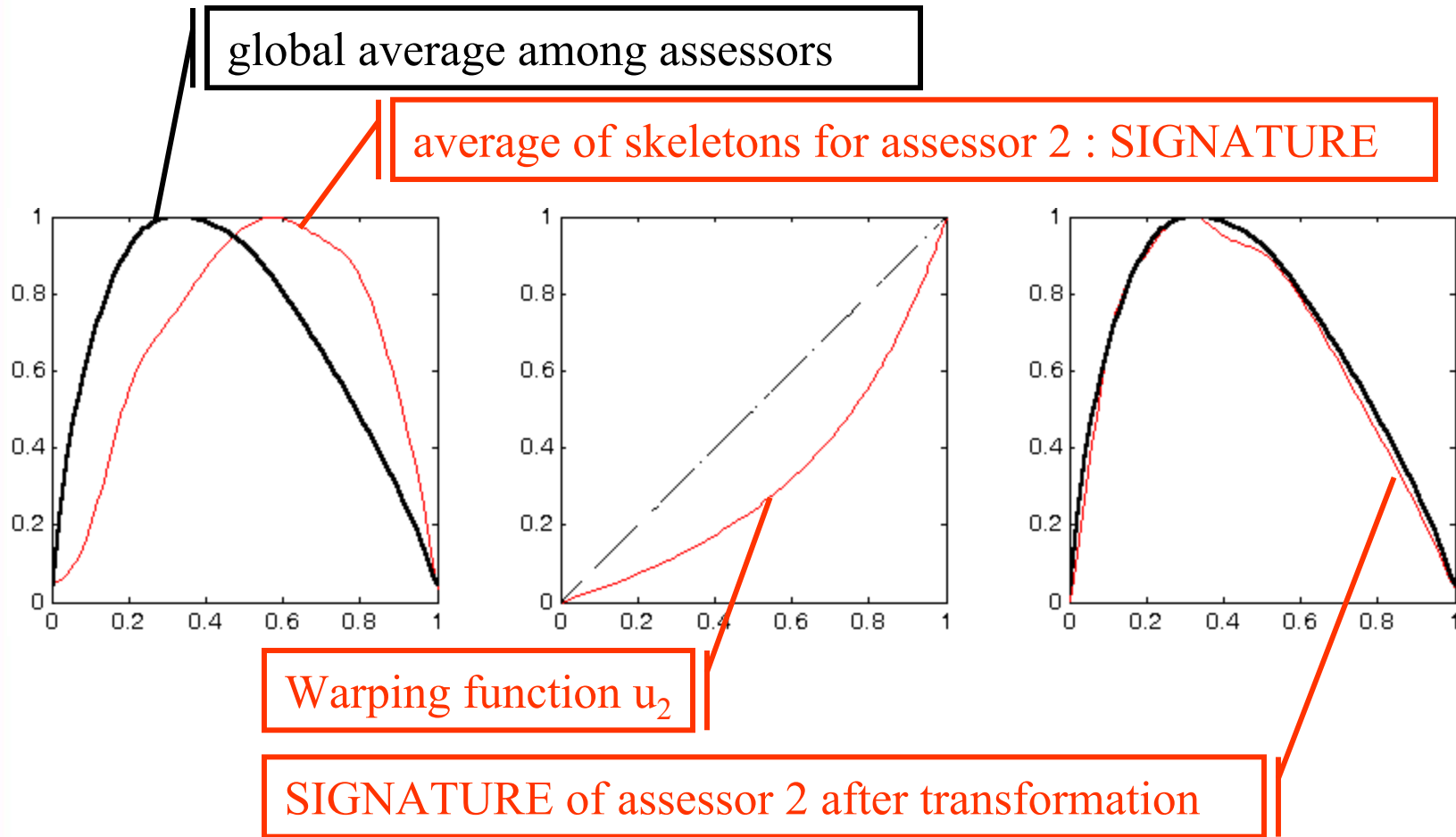
$$Z_{.j}(u_j(t)) = Z_{..}(t)$$

where $Z_{..}$ is the global average skeleton.

The principle of DTW:
$$\text{Min}_{u_j, j=1..m} \sum_{j=1}^m \|Z_{.j}(u_j(t)) - Z_{..}(t)\|^2$$

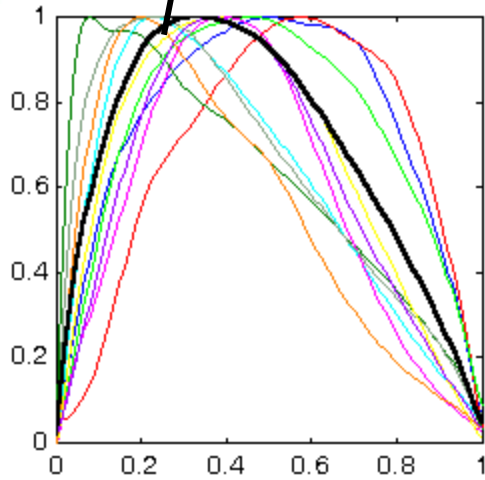


RESULTS : assessor 2

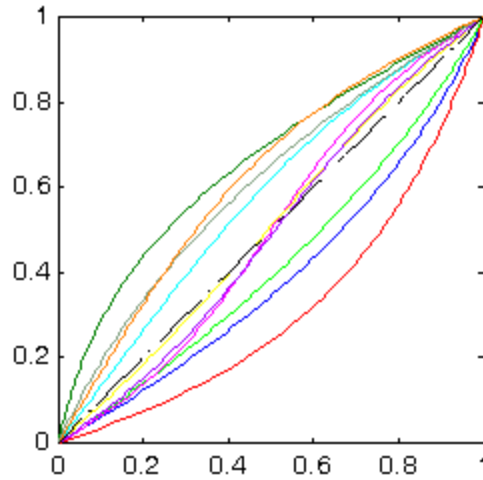


RESULTS : all assessors

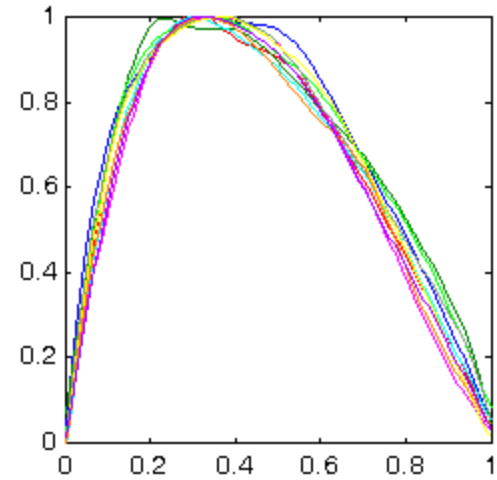
global average among assessors



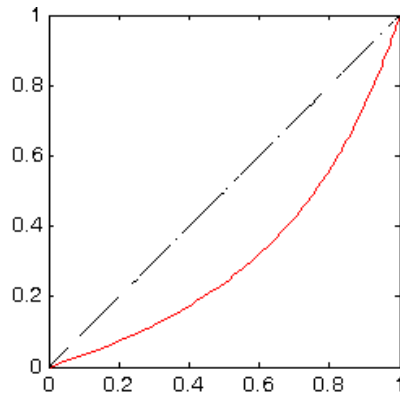
Warping functions u_j



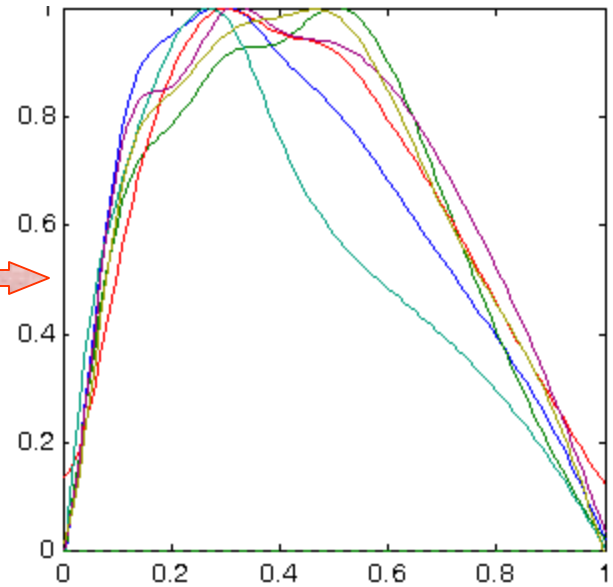
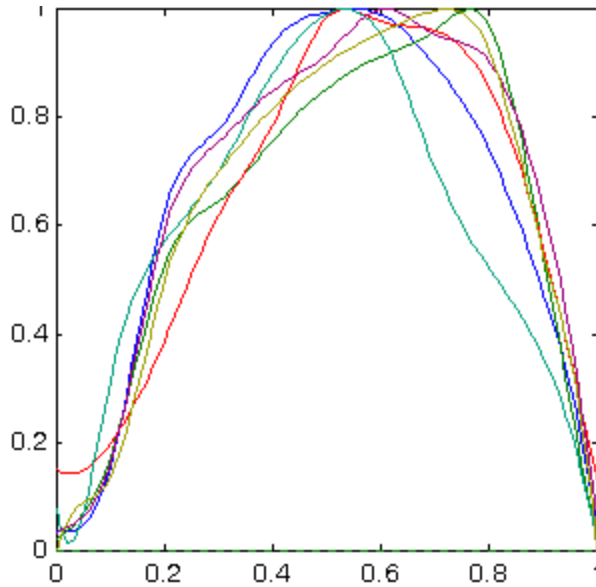
average of skeletons
after transformation

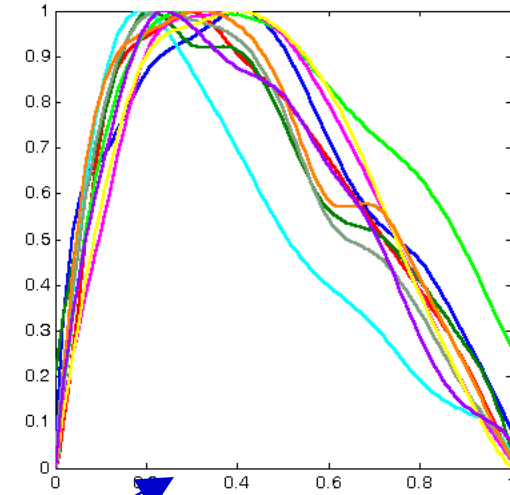
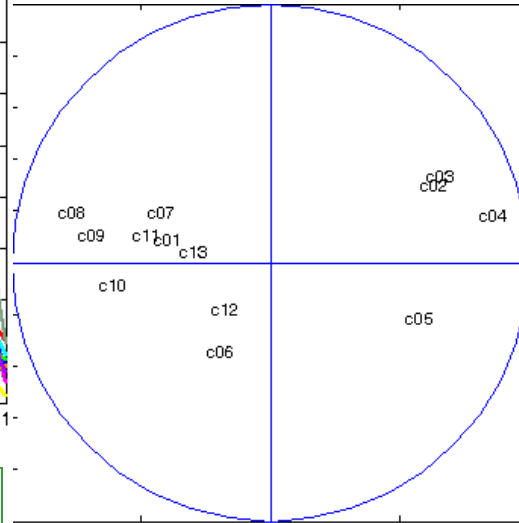
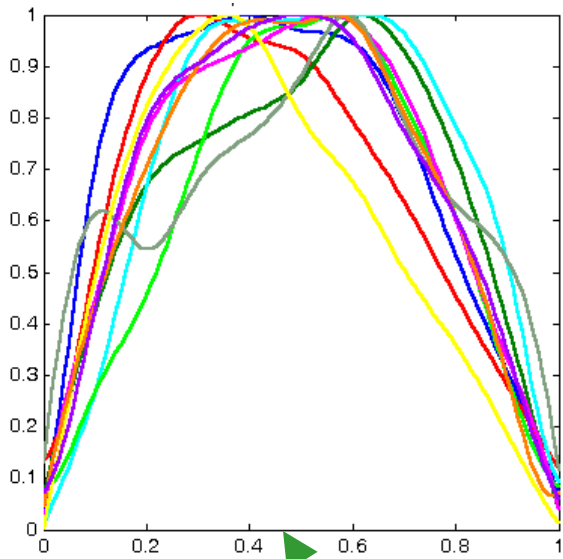


RESULTS : assessor 2



Warping function u_2 is applied to all skeletons associated with this assessor

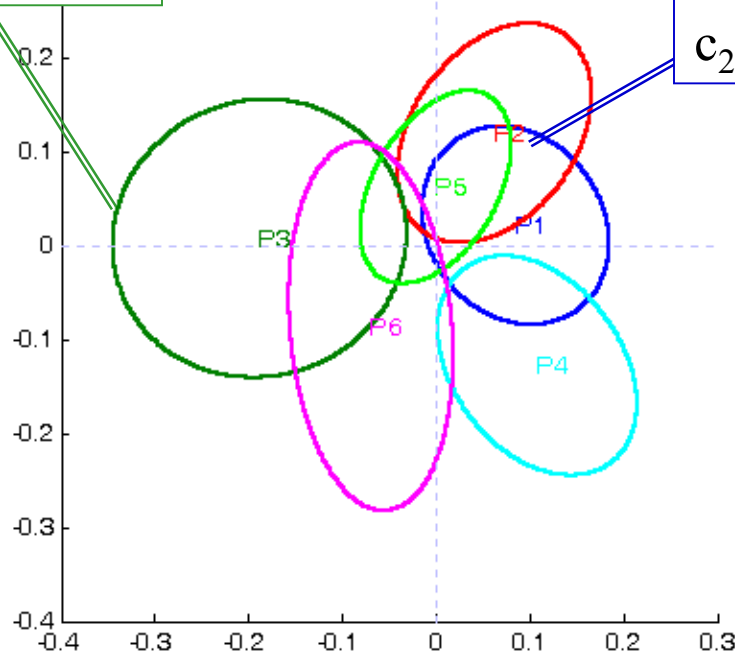


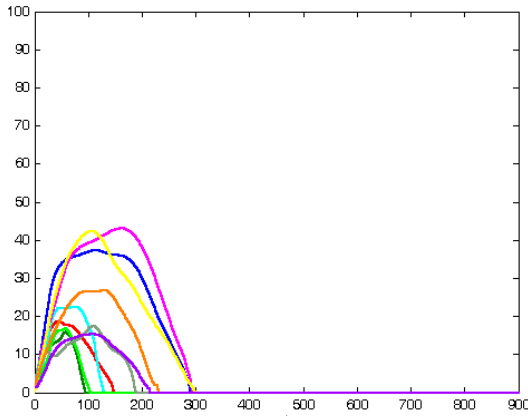


c_2, c_3, c_4 and c_5 small

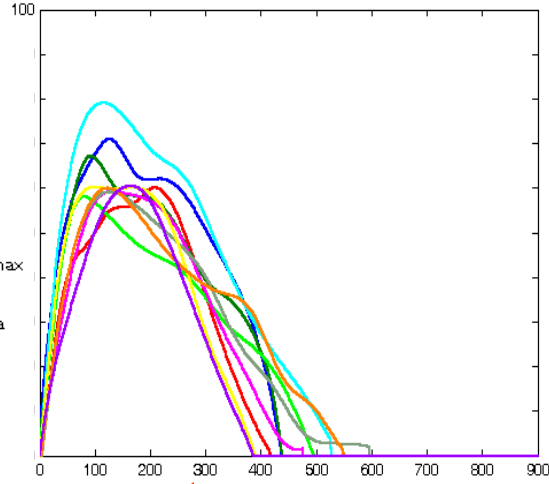
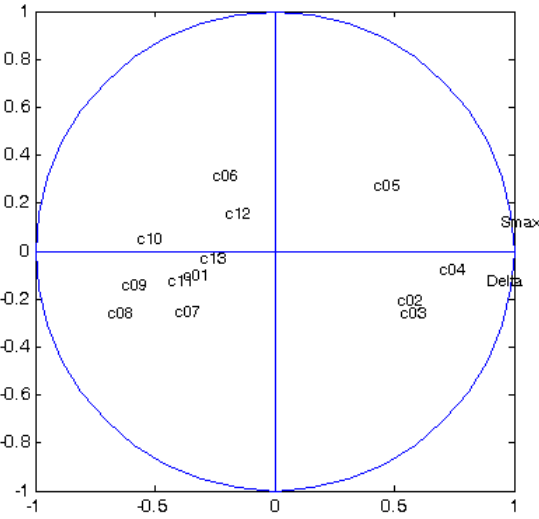
c_2, c_3, c_4 and c_5 high

Fisher's Discriminant Analysis

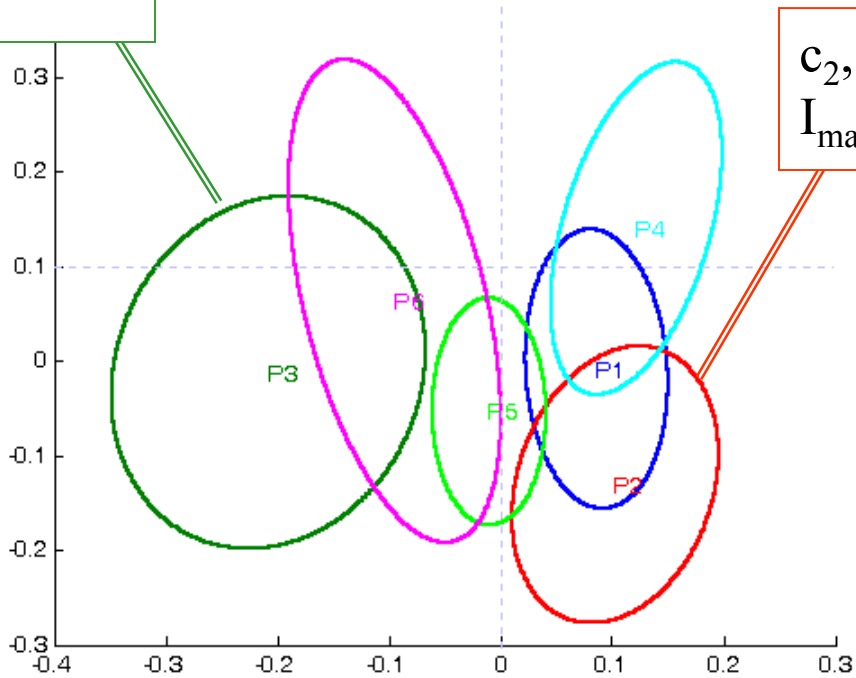




$c_2, c_3, c_4, c_5, \Delta$ and I_{\max} small



c_2, c_3, c_4, Δ and I_{\max} high



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CONCLUSION

- reduce assessor effect
 - variability in I_{\max}
 - variability in Δ
 - variability in the shape of curves (skeleton)
- improve the characterization of the products
- compute an estimation of prototype curve by product



References

- DIJKSTERHUIS G.B. & EILERS P. (1997). Modelling time-intensity curves using prototype curves.
- KNEIP A. & GASSER T. (1992). Statistical tools to analyze data representing a sample of curves.
- RAMSAY J.O. (1998). Curve registration.
- VIGNEAU E., LEDAUPHIN S. & CAUSEUR D. (2004). Approche fonctionnelle de l'analyse des signaux temps intensité.
- WANG K. & GASSER T. (1997). Alignment of curves by dynamic time warping.



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- Time intensity curves
description [\(d.6\)](#), A great variability between assessors [\(d.7\)](#)
- Statistical processing of temporal data
pretreatment [\(d.11\)](#), ‘skeleton’ [\(d.13\)](#), ‘signature’ [\(d.16\)](#),
non linear transformation of times [\(d.20\)](#), dtw [\(d.23\)](#)
characterization of the products [\(d.25\)](#)

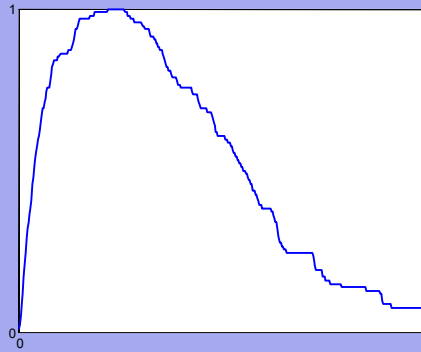


Questions...

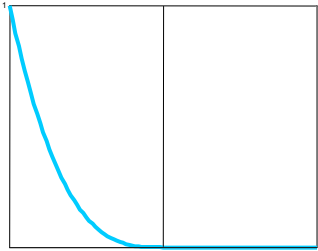
- **Spline basis**
- **Compromise among repetitions**
- **Repeatability**
- **Analysis of variance**
- **Fisher's discriminant analysis**



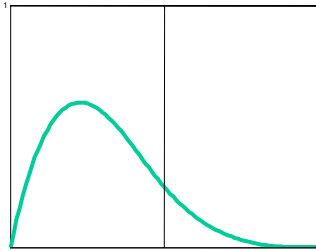
Spline Basis



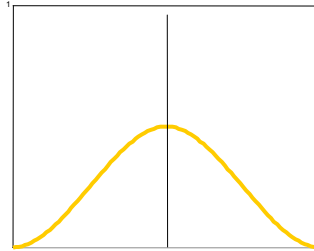
B-splines : degree = 3, 1 knot \rightarrow 5 basis functions



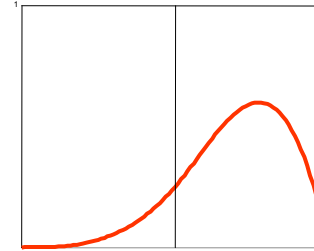
0.21



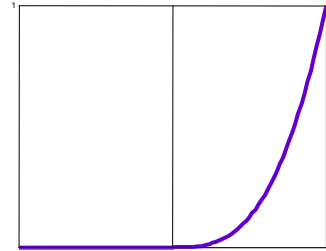
1.53



0.31

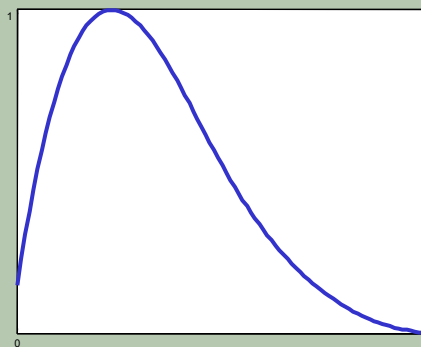


0.09



0.09

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Compromise Among Repetitions

Weighted average of B-splines coefficients by
taking account of a matrix of similarity

Weighted average of the parameters I_{\max} and Δ

Computation of the compromise



Matrix of Similarity coefficients of the B-splines basis

$$m_{kh}^{1/2} = 2 \star \frac{\langle c_{ijk}, c_{ijh} \rangle}{\|c_{ijk}\|^2 + \|c_{ijh}\|^2}$$

For each assessor j and product i :

m_{kh} are indices of similarity between repetitions k and h



Weight for each repetitions

For each assessor j and product i , for a parameter X (I_{\max}, Δ), the weight of repetition k is given by:

$$p_{ijk}^{(X)} = \frac{1}{K-1} \left(1 - \frac{(X_{ijk} - \bar{X}_{ij})^2}{\sum_{h=1}^K (X_{ijh} - \bar{X}_{ij})^2} \right)$$

The weighted average for the parameter X is:

$$X_{ij} = \sum_{k=1}^K p_{ijk}^{(X)} X_{ijk}$$



Repeatability

- Repeatability indices for the parameters:

$$R_j^{\Delta} = \sum_{i=1}^p \sum_{k=1}^K \frac{|\Delta_{ij.} - \Delta_{ijk}|}{p * K * \Delta_{ij.}} \quad R_j^{\text{Imax}} = \sum_{i=1}^p \sum_{k=1}^K \frac{|\text{Imax}_{ij.} - \text{Imax}_{ijk}|}{p * K * \text{Imax}_{ij.}}$$

- Repeatability indices for the skeleton:

$$R_j^{\text{Sk}} = \sum_{i=1}^p \sum_{k=1}^K \sum_{t=0}^1 \frac{(Z_{ij.}(t) - Z_{ijk}(t))^2}{p * K * 101}$$

- ⇒ Global repeatability indices: $R_j = R_j^{\text{Sk}} * R_j^{\Delta} * R_j^{\text{Imax}}$
 $0 \leq R_j \leq 1$



Analysis of variance

For each assessor j and each parameters X (S, To, Δ), the ANOVA model is:

$$X_{ik}^{(j)} = \mu^{(j)} + \alpha_i^{(j)} + \varepsilon_{ik}^{(j)}$$

We are looking for the significance of the product effect



Fisher 's Discriminant Analysis

$$z_1 = Xa_1$$

$$a_1 = \text{Max} \frac{a^T B a}{a^T T a}$$

Between-class Variance

Total Variance

First Discriminant Axis

