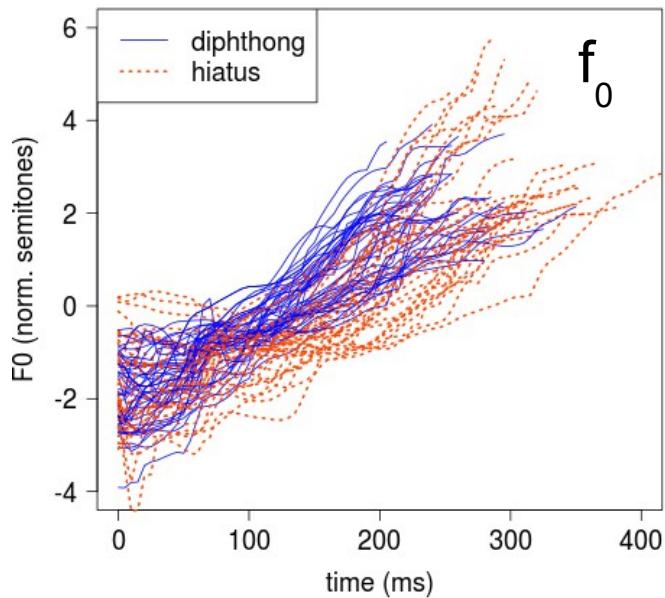
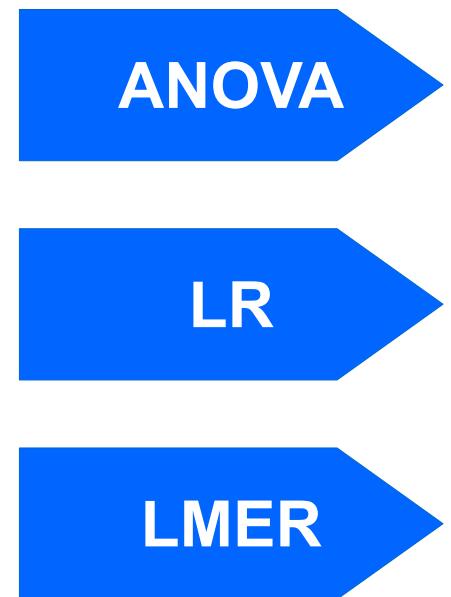
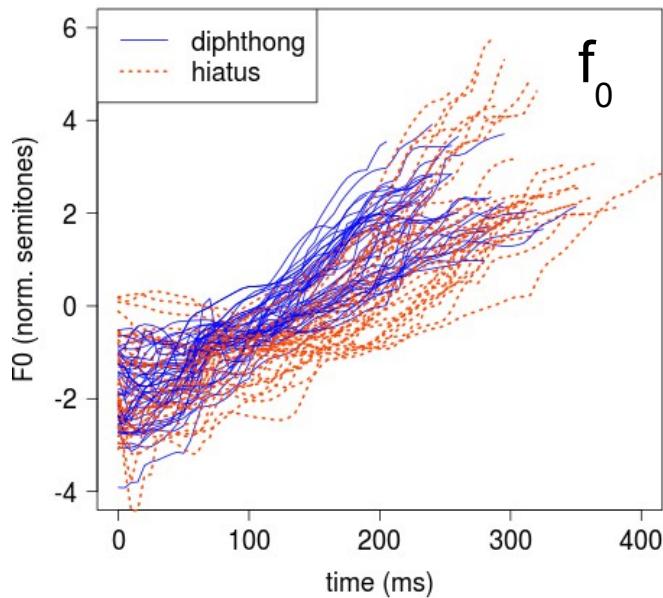


Functional Data Analysis (FDA) for phonetic research

Michele Gubian
University of Bristol, UK

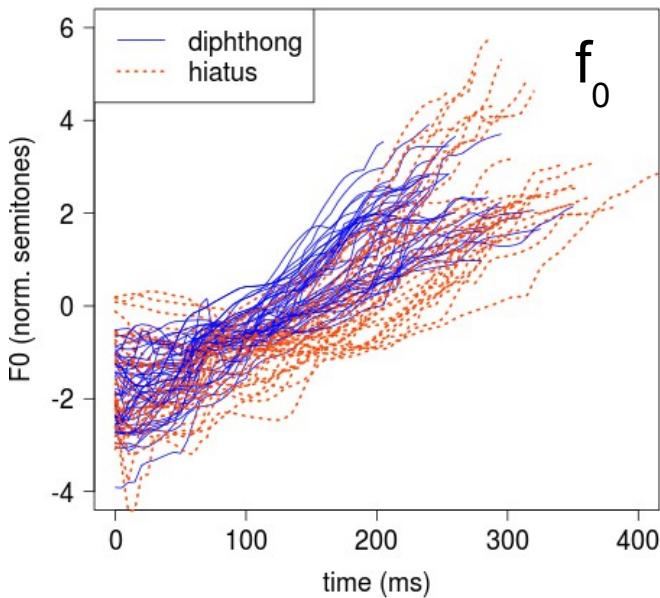


- European Spanish
- **Diphthong**: /ja/
- **Hiatus** /i.a/
- Rising pitch accent
- Tonal alignment?



- European Spanish
- **Diphthong**: /ja/
- **Hiatus** /i.a/
- Rising pitch accent
- Tonal alignment?

CURVES



- European Spanish
- **Diphthong**: /ja/
- **Hiatus** /i.a/
- Rising pitch accent
- Tonal alignment?

NUMBERS

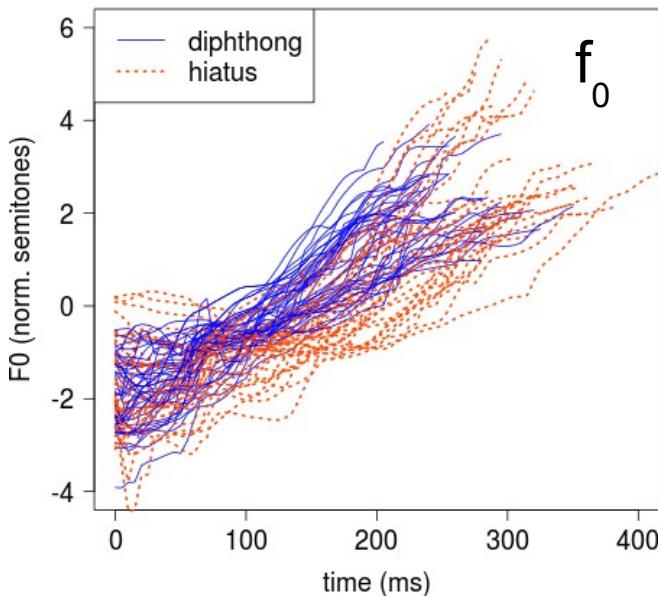
ANOVA

LR

LMER

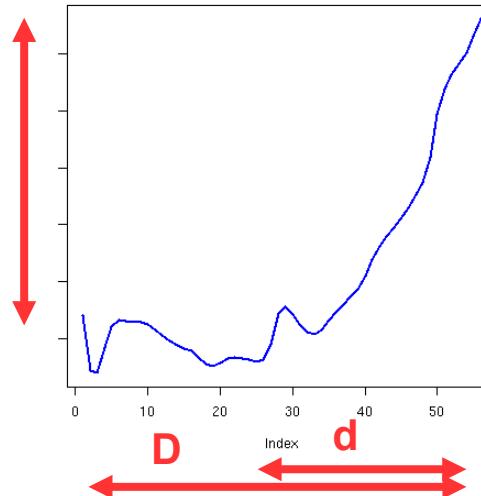
MIND THE GAP

CURVES



- European Spanish
- **Diphthong**: /ja/
- **Hiatus** /i.a/
- Rising pitch accent
- Tonal alignment?

NUMBERS



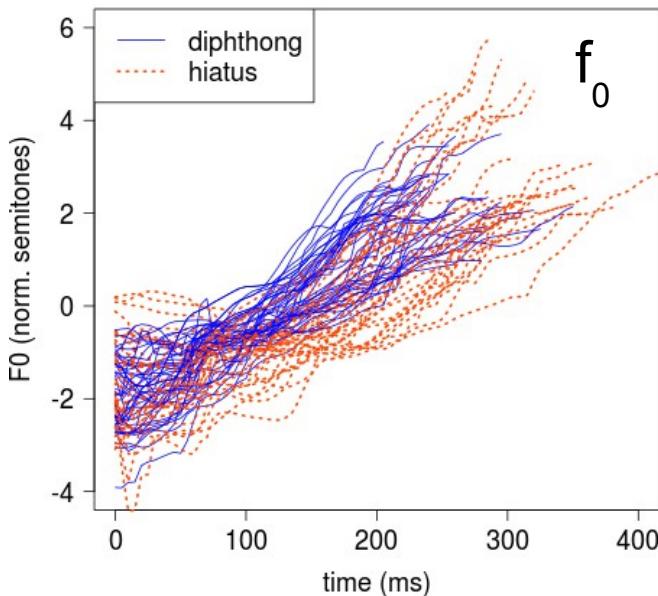
ext (st)	d/D	Cat.
5.3	0.9	D
4.6	0.7	H
....

ANOVA

LR

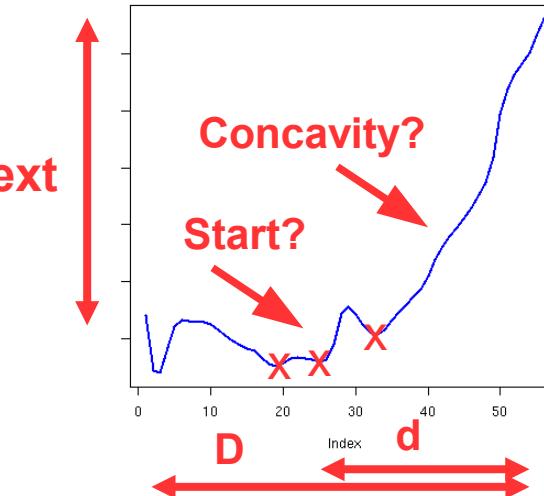
LMER

CURVES



- European Spanish
- **Diphthong**: /ja/
- **Hiatus** /i.a/
- Rising pitch accent
- Tonal alignment?

NUMBERS



ext (st)	d/D	Cat.
5.3	0.9	D
4.6	0.7	H
....

ANOVA

LR

LMER

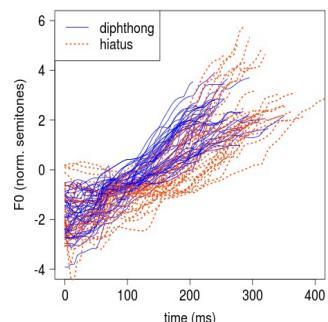
MISSION

automate curve parametrisation

- Data driven
- Few parameters
- Interpretable

Road map

CURVES



Interpolate using a function basis

Dimensionality reduction tool

- Data driven
- Few parameters
- Interpretable

NUMBERS

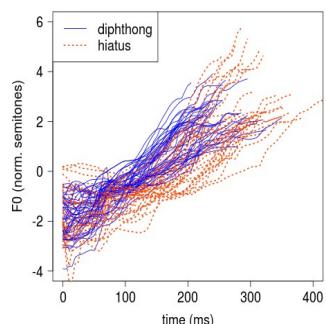
ANOVA

LM

LMER

Road map

CURVES



Interpolate using a function basis

Dimensionality reduction tool

- Data driven
- Few parameters
- Interpretable

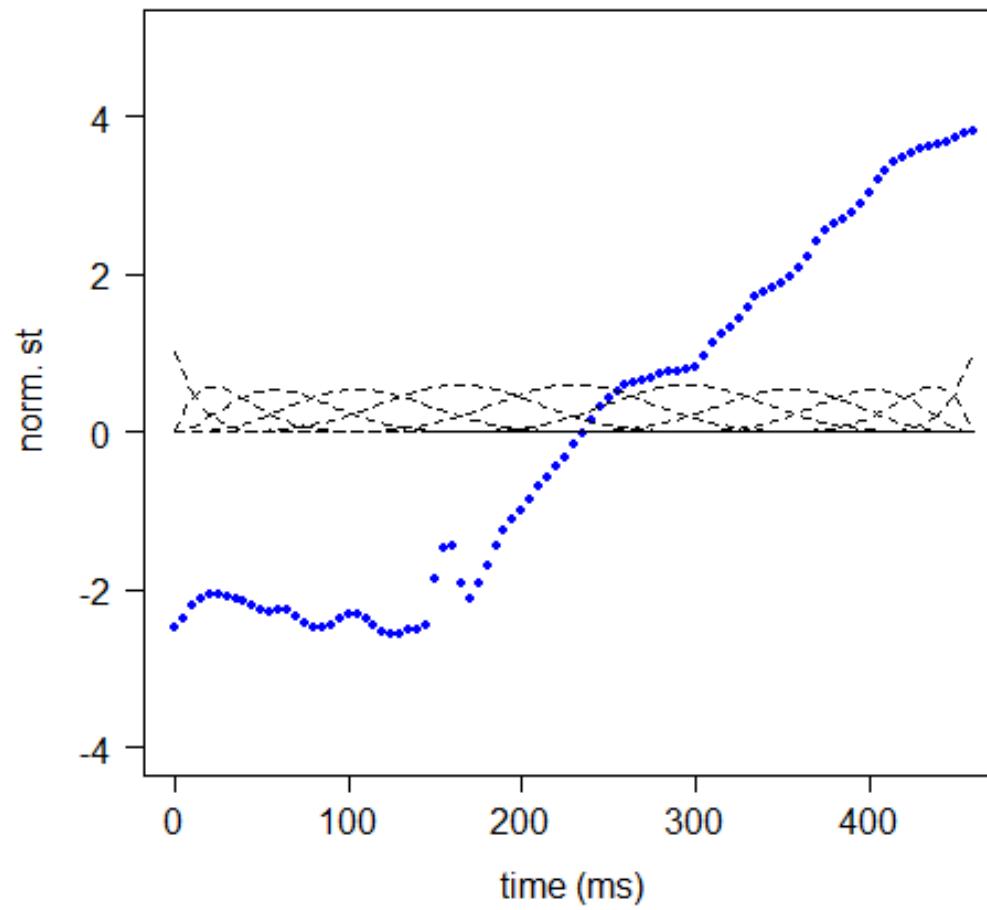
NUMBERS

ANOVA

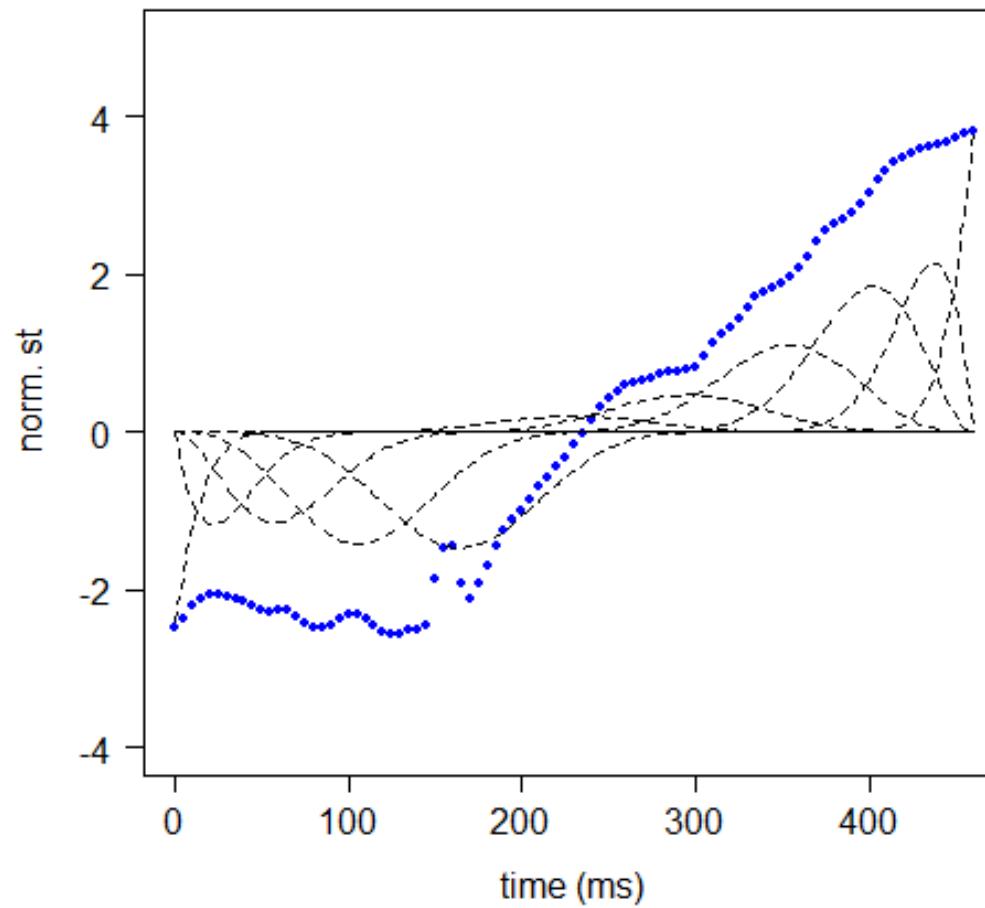
LM

LMER

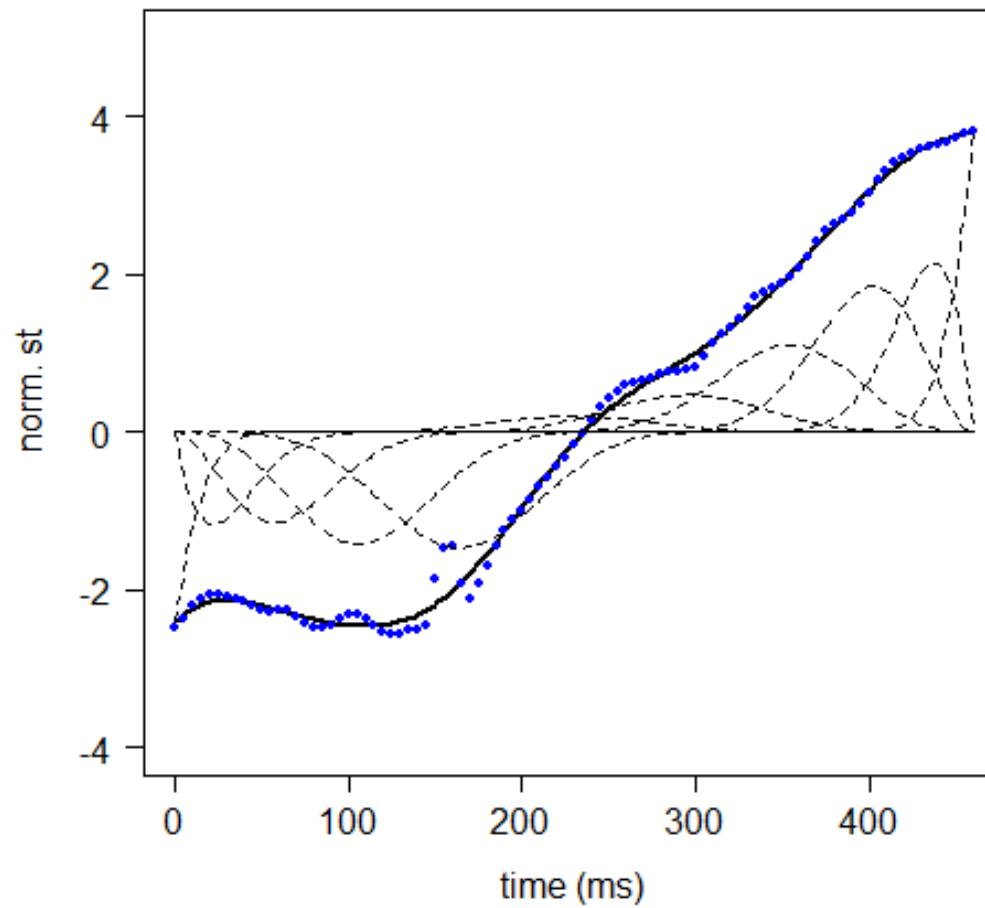
Interpolation with B-splines



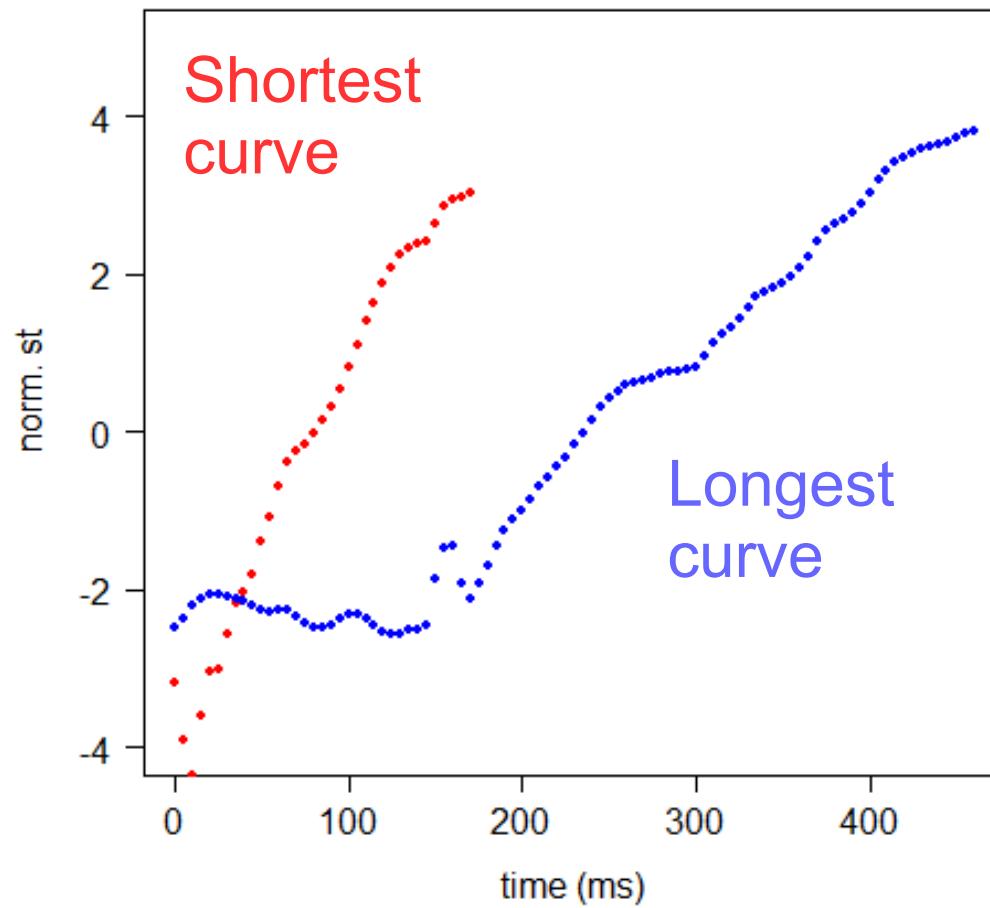
Interpolation with B-splines



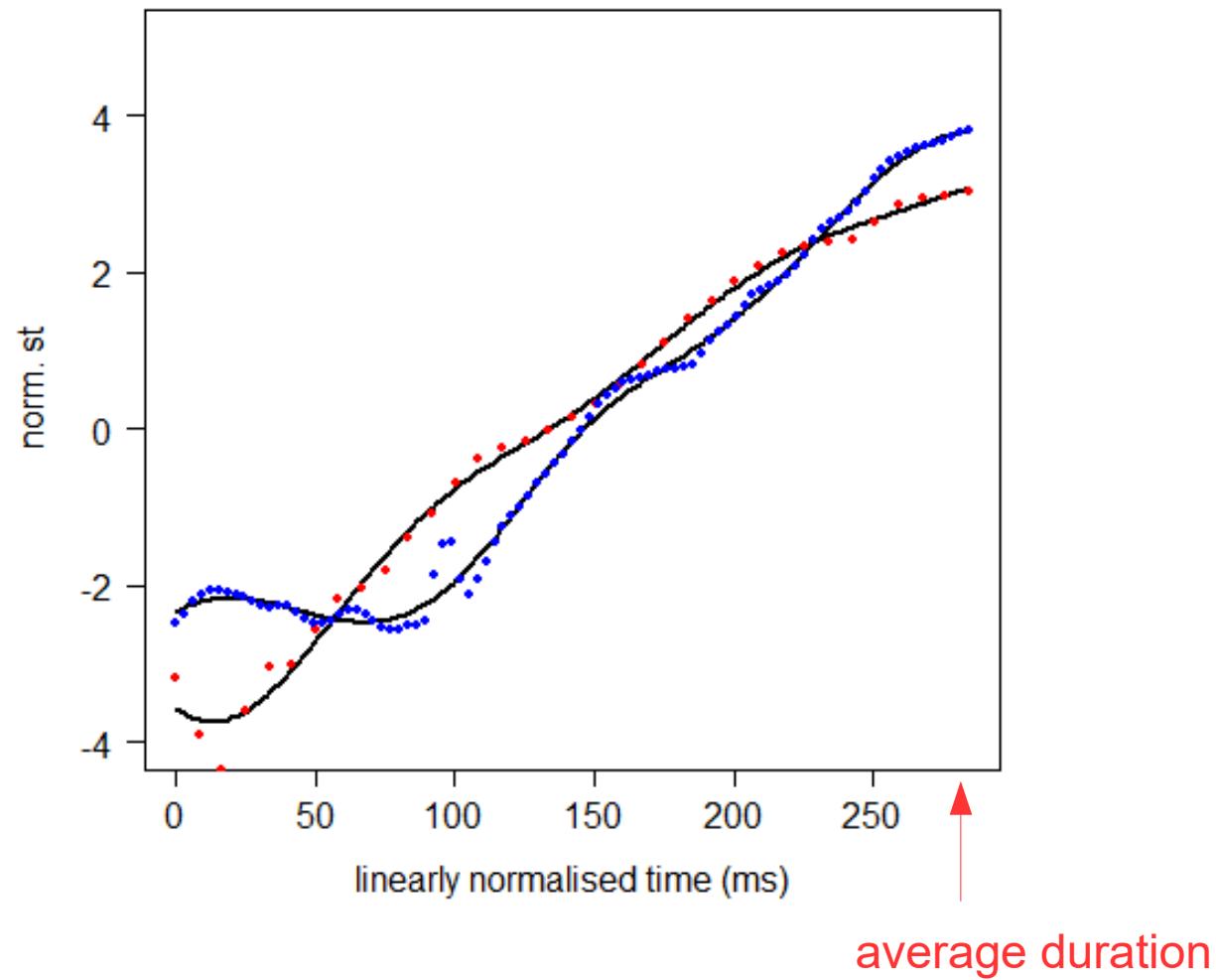
Interpolation with B-splines



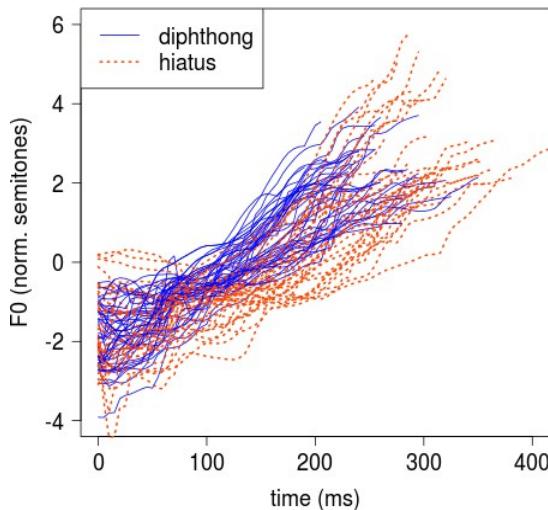
Different durations



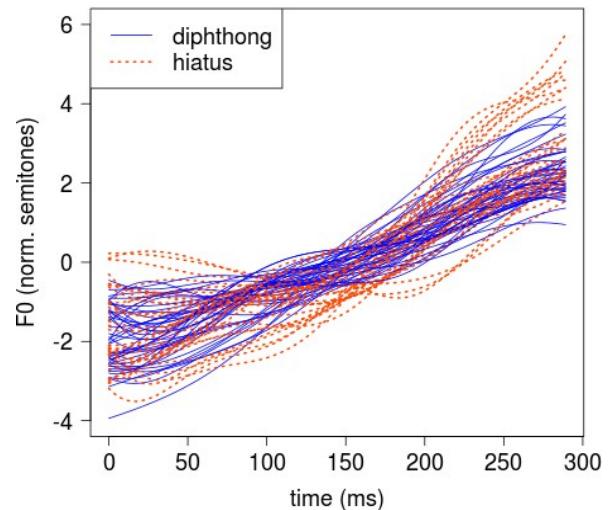
Linear time normalisation



Linear time normalisation



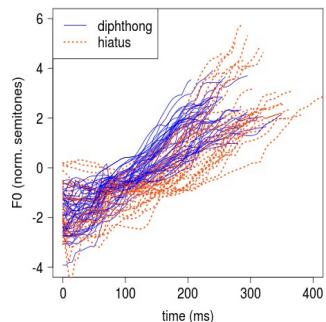
Interpolate on a
common time
interval



- We must use a common time interval
- This induces a linear time normalisation
- Durations have to be reintroduced at the end of the analysis

Road map

CURVES



Interpolate on a
common time
interval

Dimensionality
reduction tool

- Data driven
- Few parameters
- Interpretable

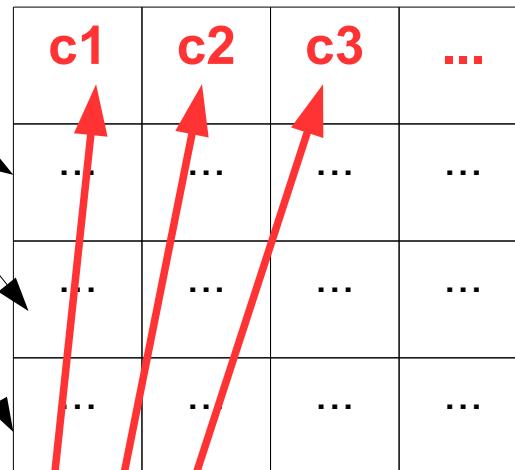
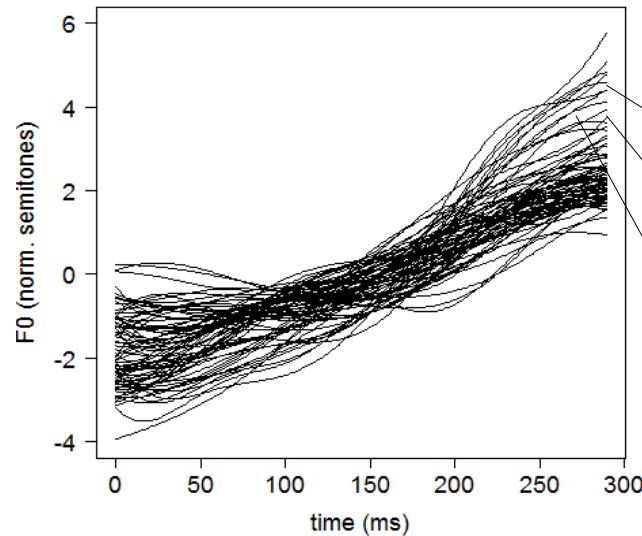
NUMBERS

ANOVA

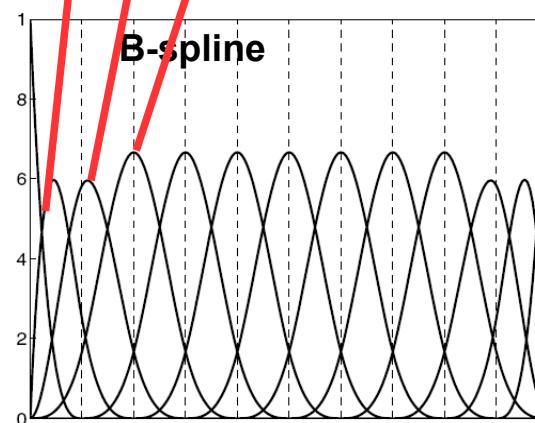
LM

LMER

Principal Component Analysis



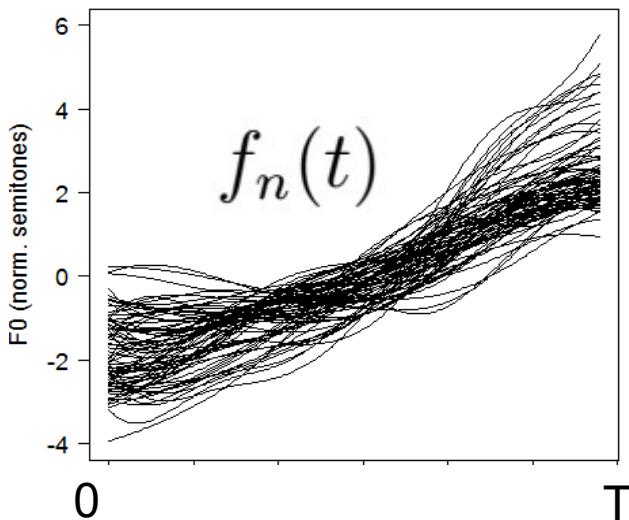
PCA



PCA limitations

- PCA does not use any explicit information related to the curve shapes or the B-splines shapes
- e.g. the sequence of coefficients c_1, c_2, \dots reflects time adjacency of polynomial components, i.e. overlapping 'hills'

Functional PCA



$$\max \left\{ \text{var}_n \left(\int_0^T PC1(t) f_n(t) dt \right) \right\}$$

$$\text{subject to } \int_0^T PC1^2(t) = 1$$

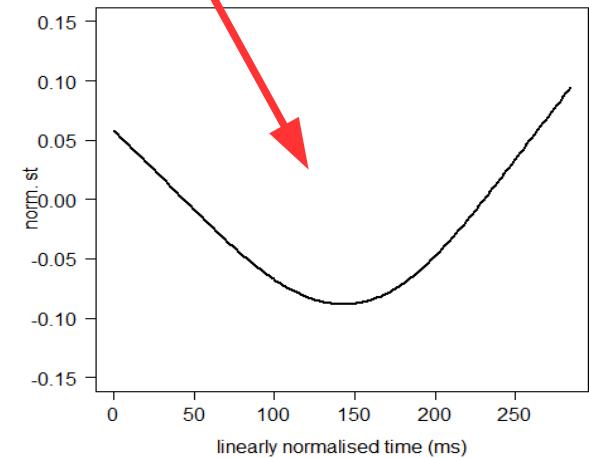
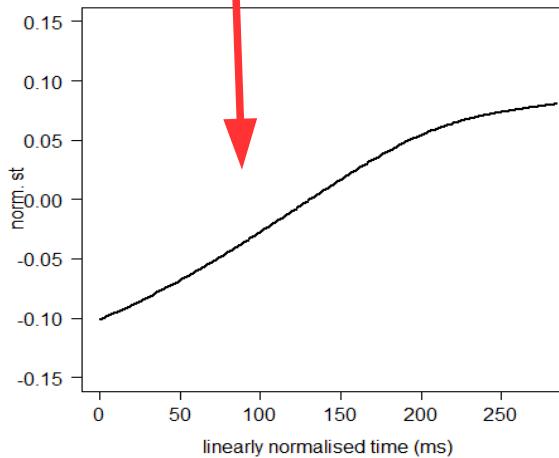
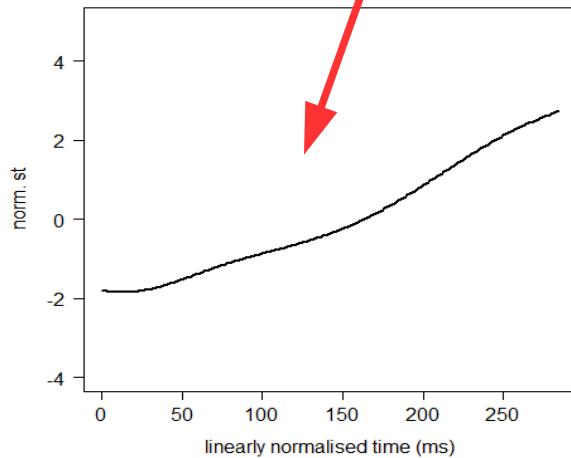
- FPCA definition uses the input curves $f_n(t)$
- FPCA is independent of the B-splines used to smooth $f_n(t)$

Functional PCs

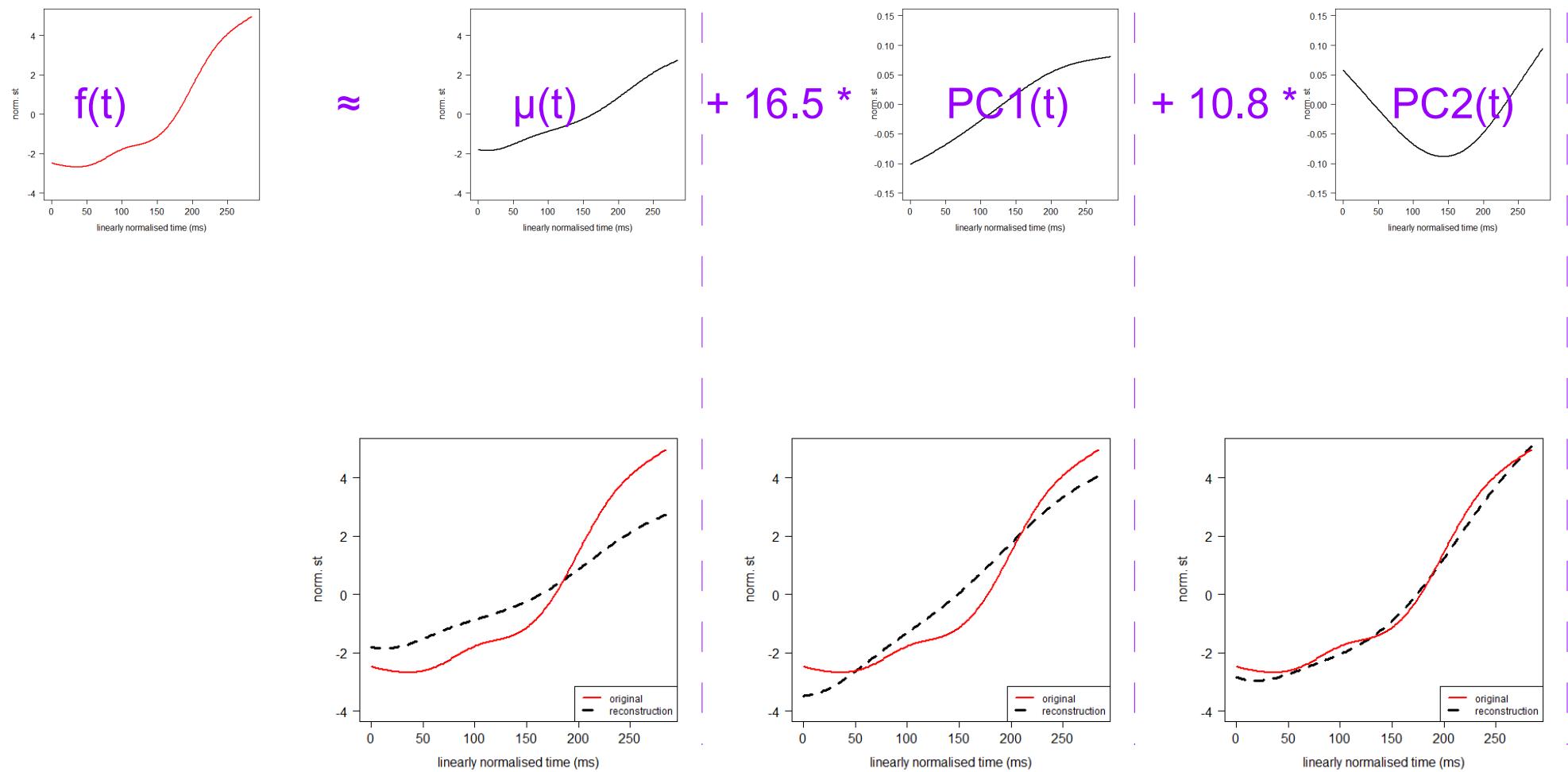
$$f(t) \approx \mu(t) + s_1 \cdot PC1(t) + s_2 \cdot PC2(t) + \dots$$

PC1 score

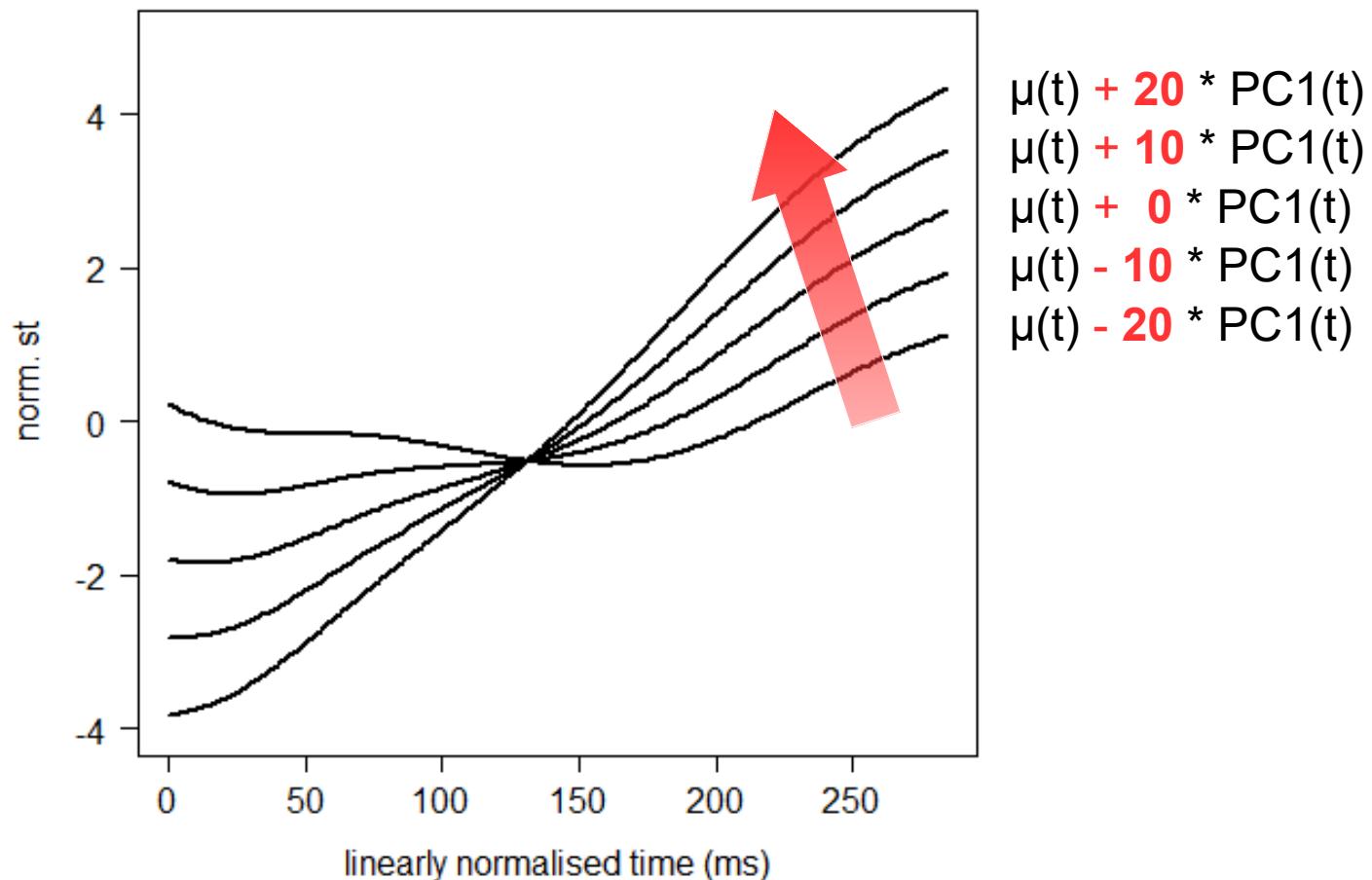
PC2 score



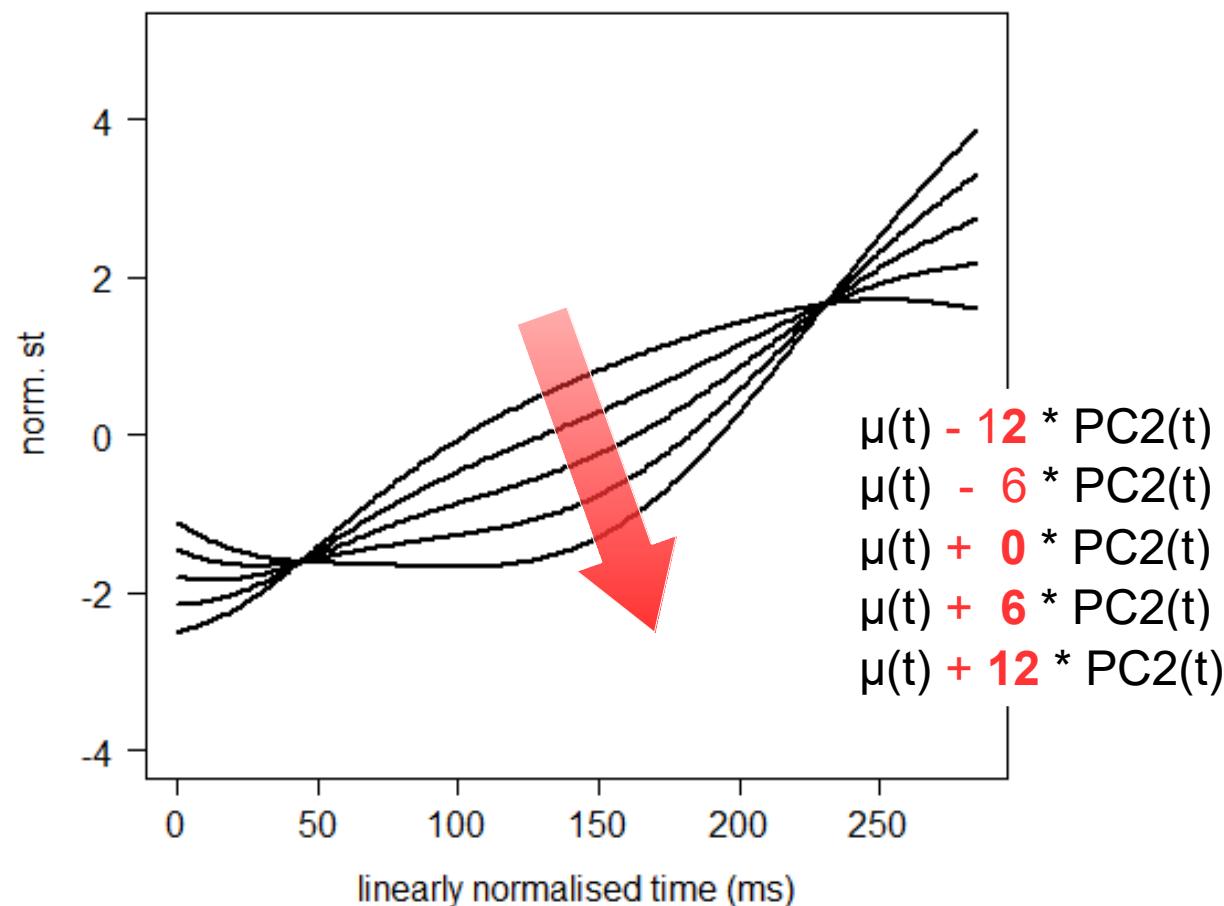
Curve reconstruction



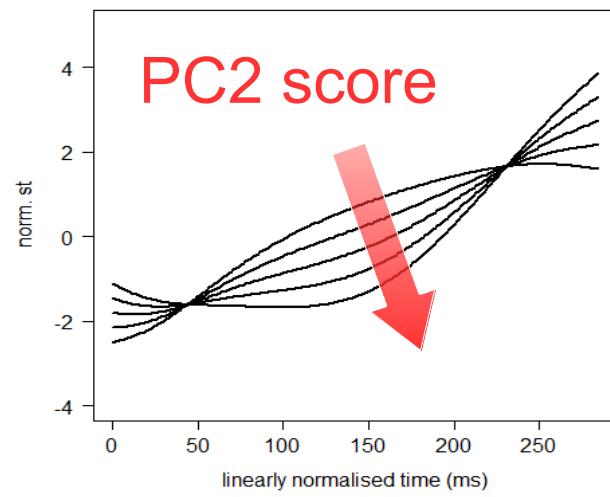
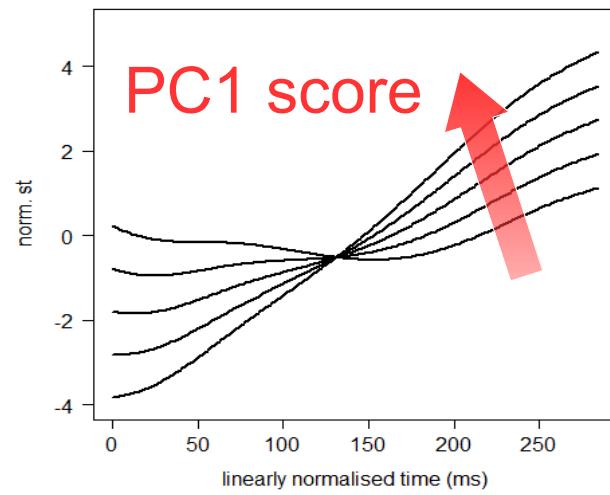
PC1 scores



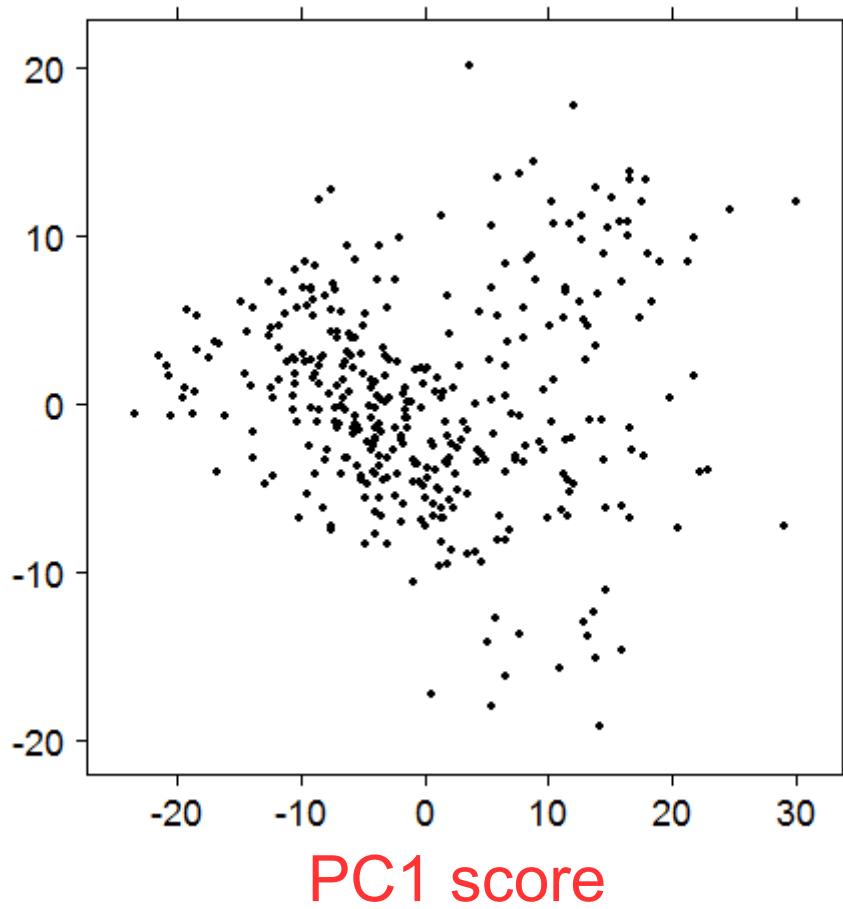
PC2 scores



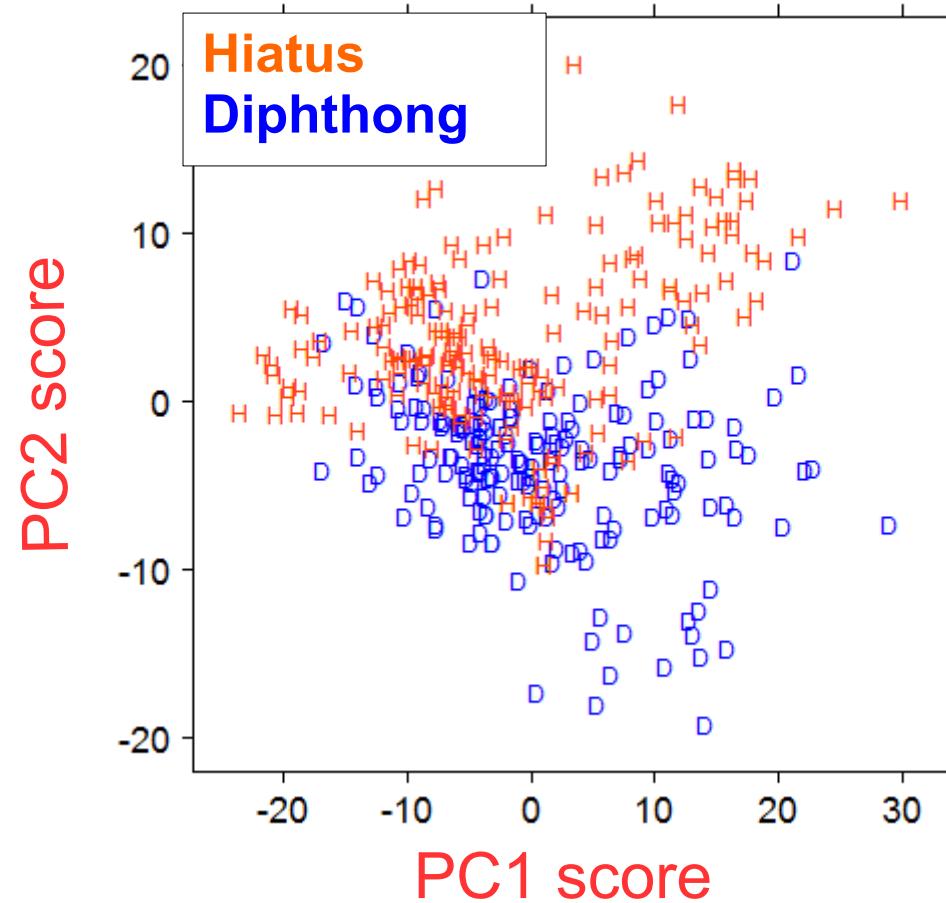
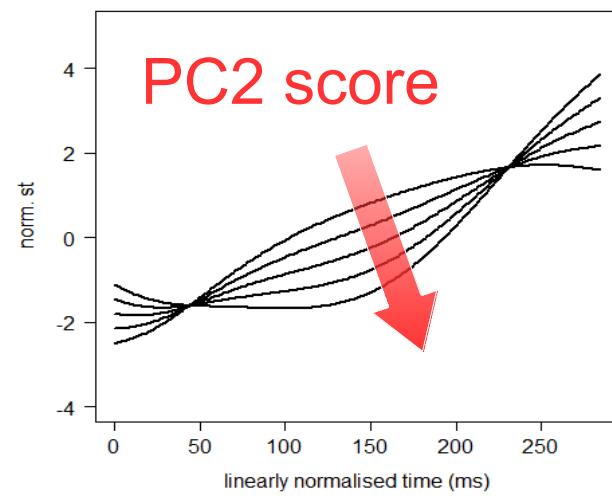
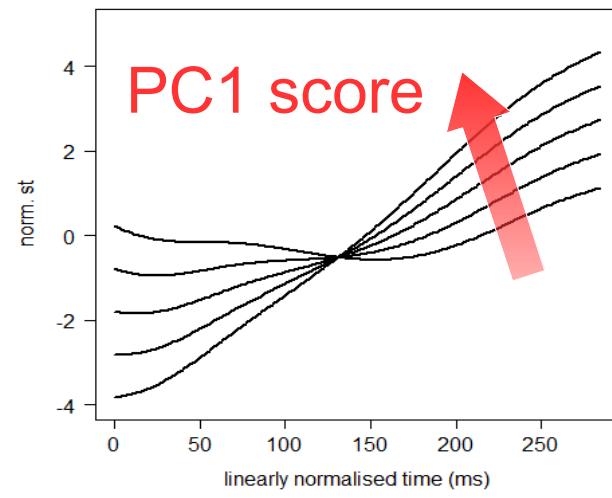
Curve parametrisation



PC2 score

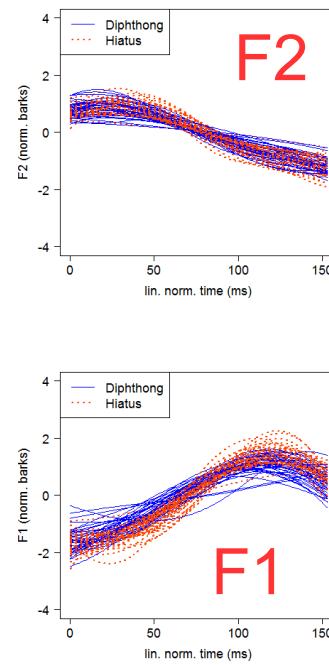
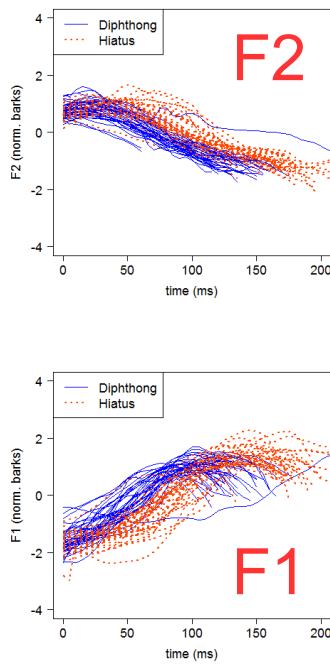


Curve parametrisation



Formants

2D CURVES



NUMBERS

FPCA

ANOVA

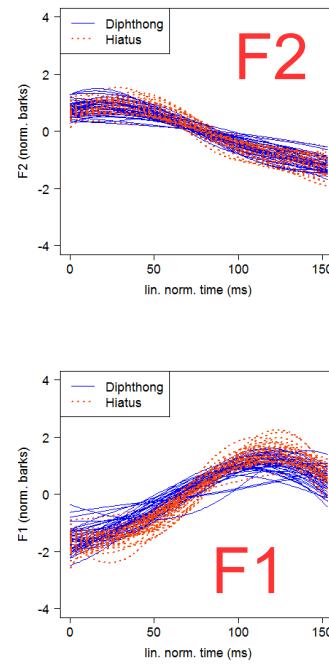
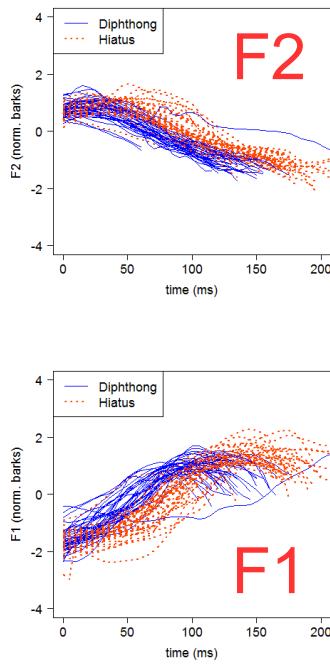
LM

LMER

FPCA

Formants

2D CURVES



2D
FPCA

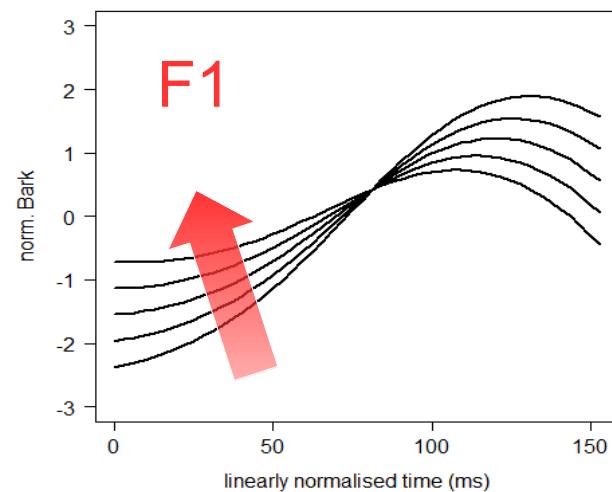
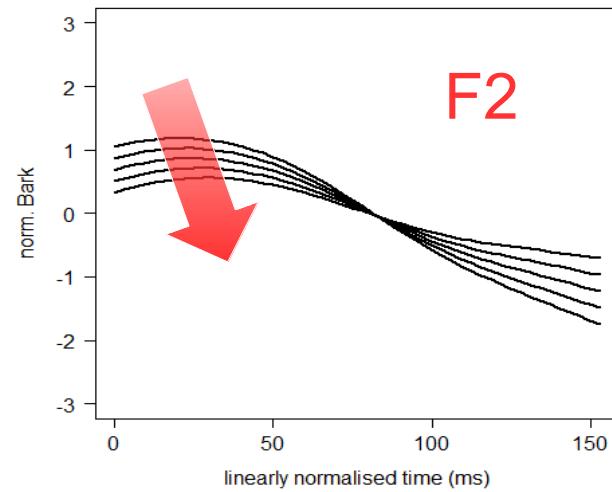
NUMBERS

ANOVA

LM

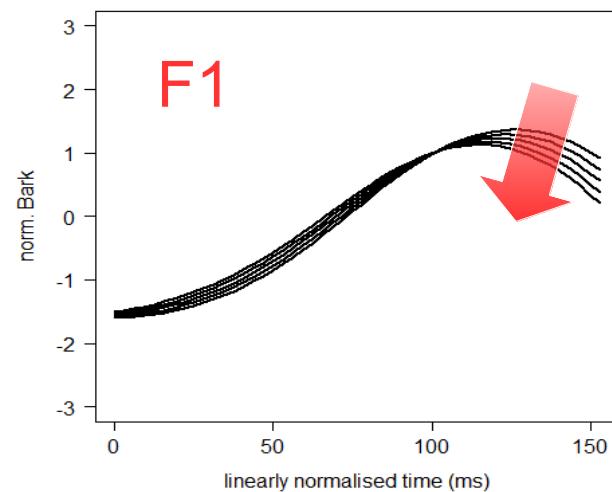
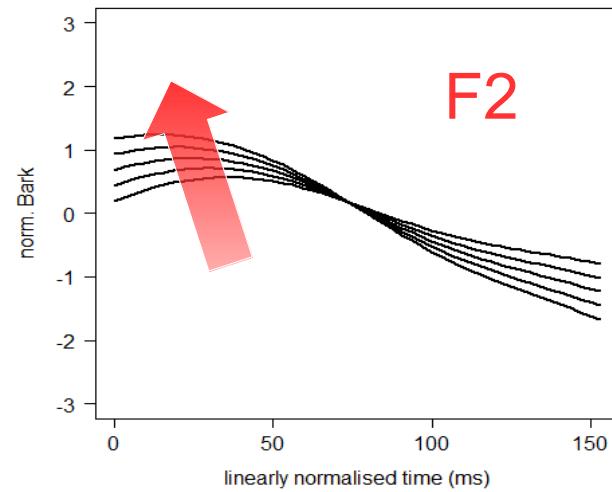
LMER

PC1 scores



- $\mu(t) + 8 * \text{PC1}(t)$
- $\mu(t) + 4 * \text{PC1}(t)$
- $\mu(t) + 0 * \text{PC1}(t)$
- $\mu(t) - 4 * \text{PC1}(t)$
- $\mu(t) - 8 * \text{PC1}(t)$

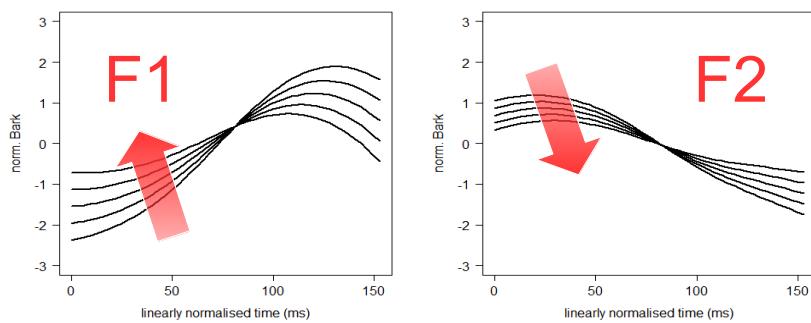
PC2 scores



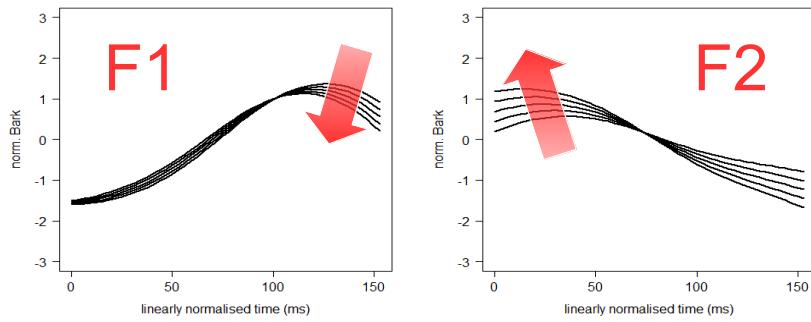
- $\mu(t) + 4 * \text{PC1}(t)$
- $\mu(t) + 2 * \text{PC1}(t)$
- $\mu(t) + 0 * \text{PC1}(t)$
- $\mu(t) - 2 * \text{PC1}(t)$
- $\mu(t) - 4 * \text{PC1}(t)$

2D curve parametrisation

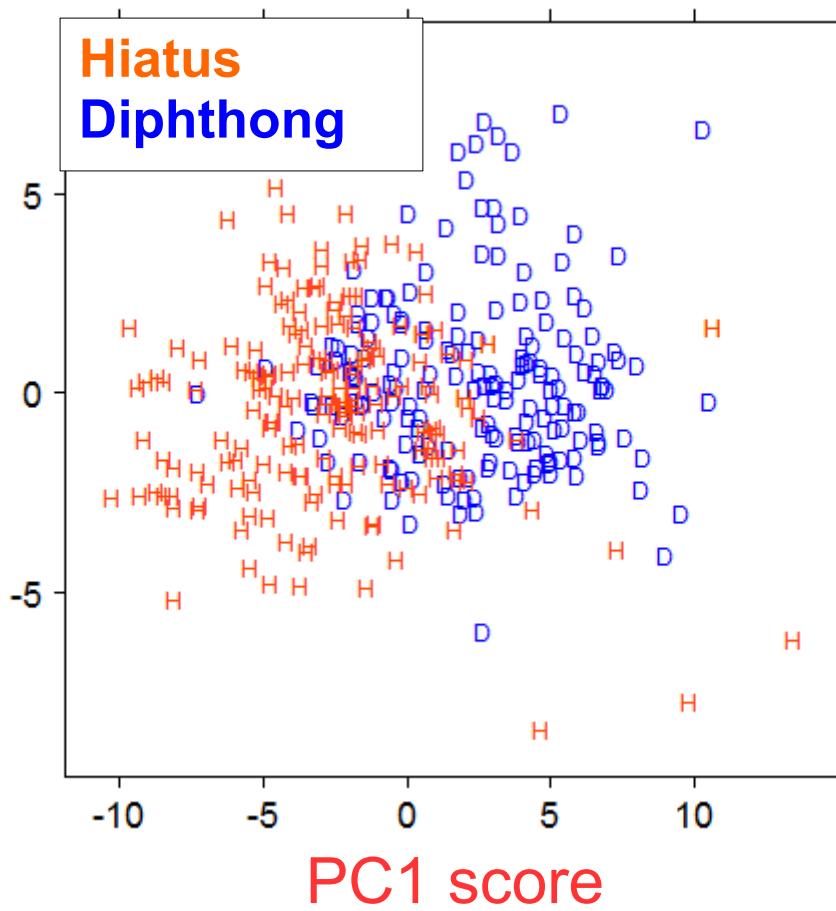
PC1 score



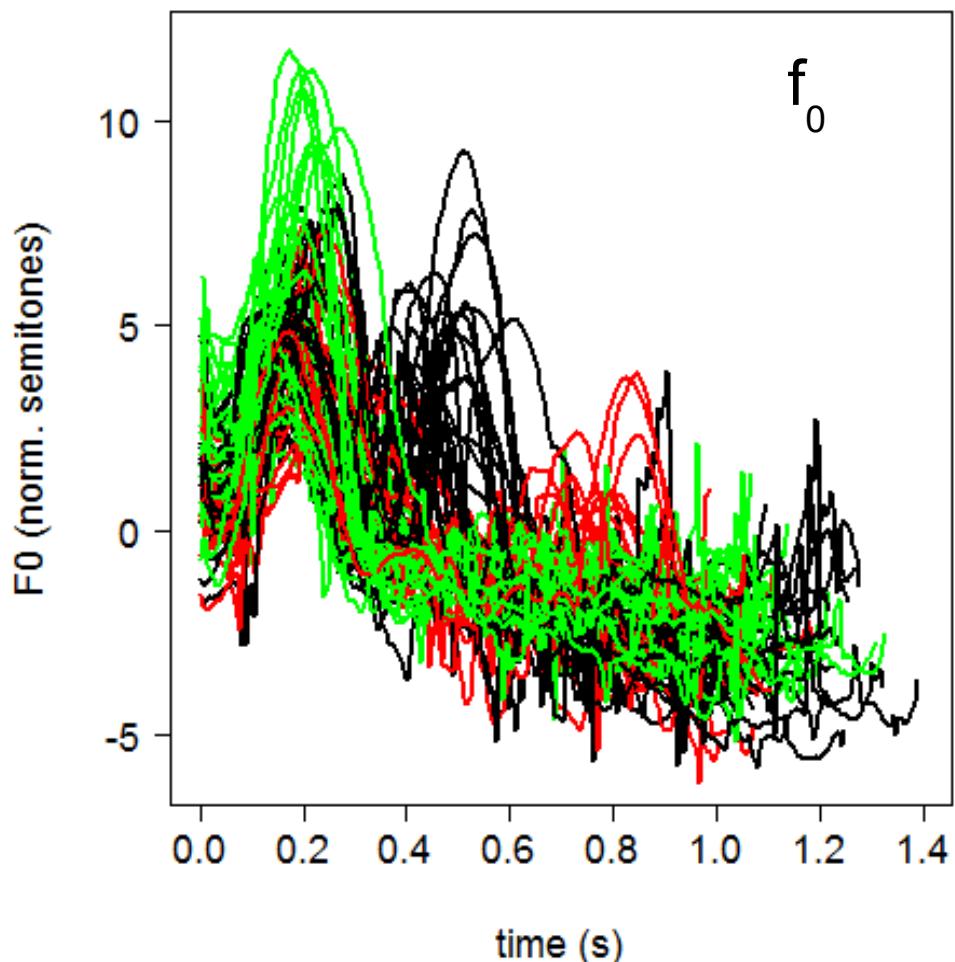
PC2 score



PC2 score



Many segments

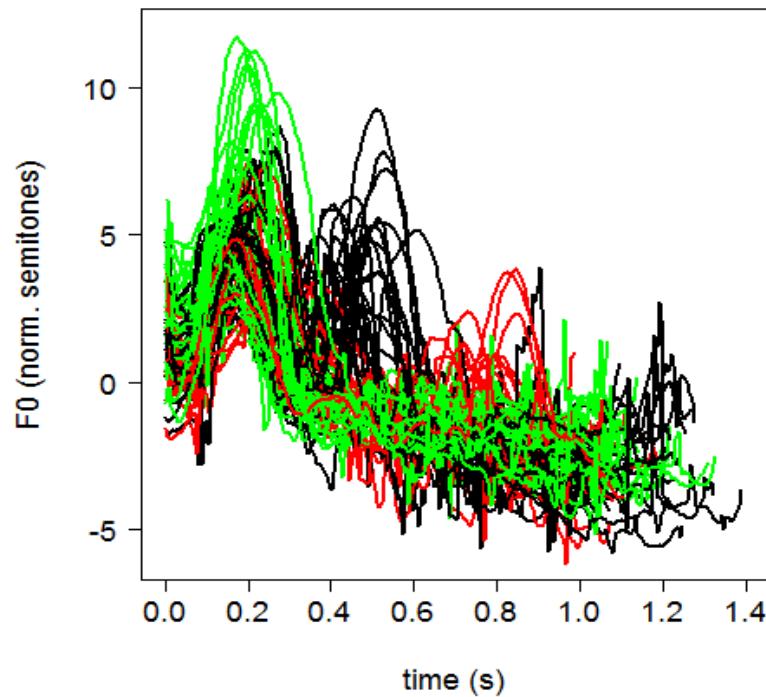


- Narrow focus in Neapolitan Italian
- Focus on
Subject, Verb or Prop. Phrase
Danilo vola da Roma
(*Danilo flies from Rome*)
- 8 CV syllables
first C was excluded (too short)
VCVCVCV CV CVCV****

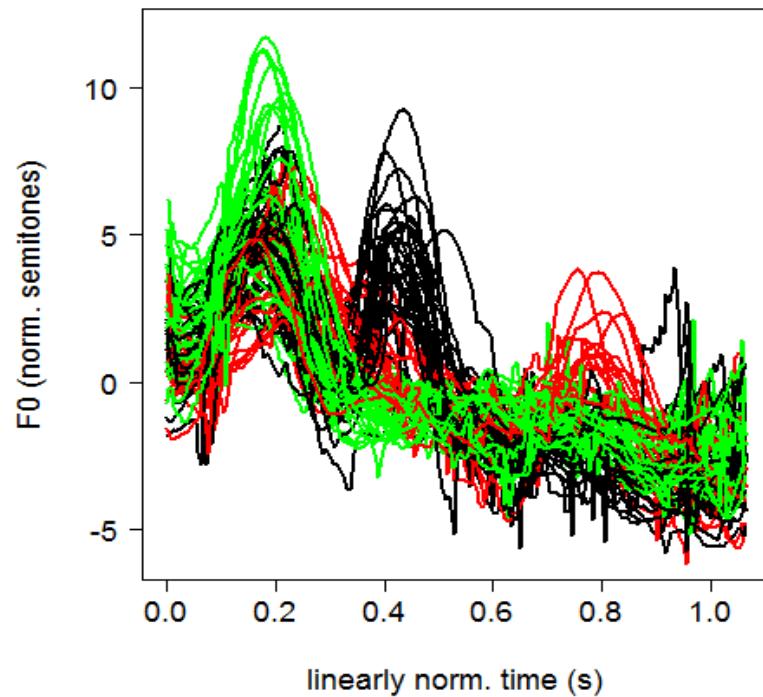
... 15 segments!

Linear time normalisation

BEFORE

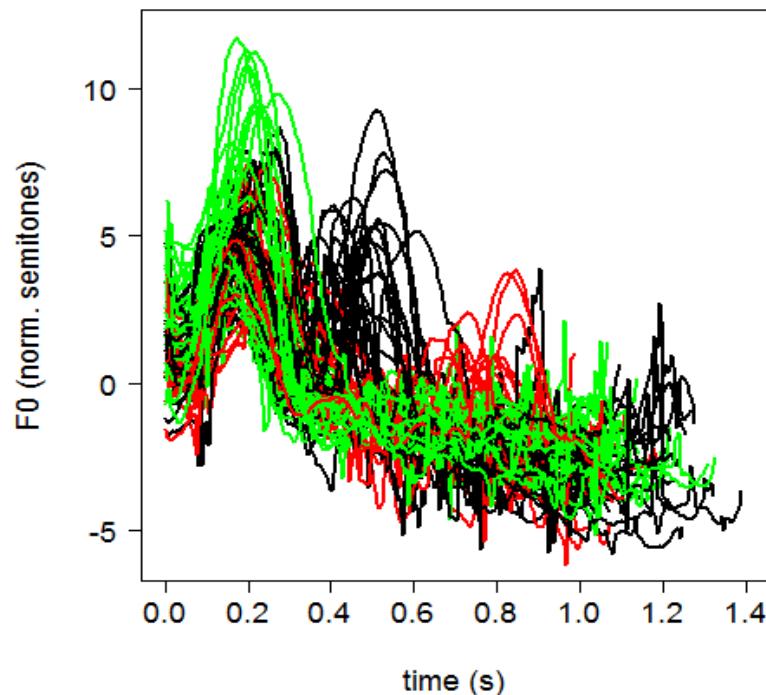


AFTER

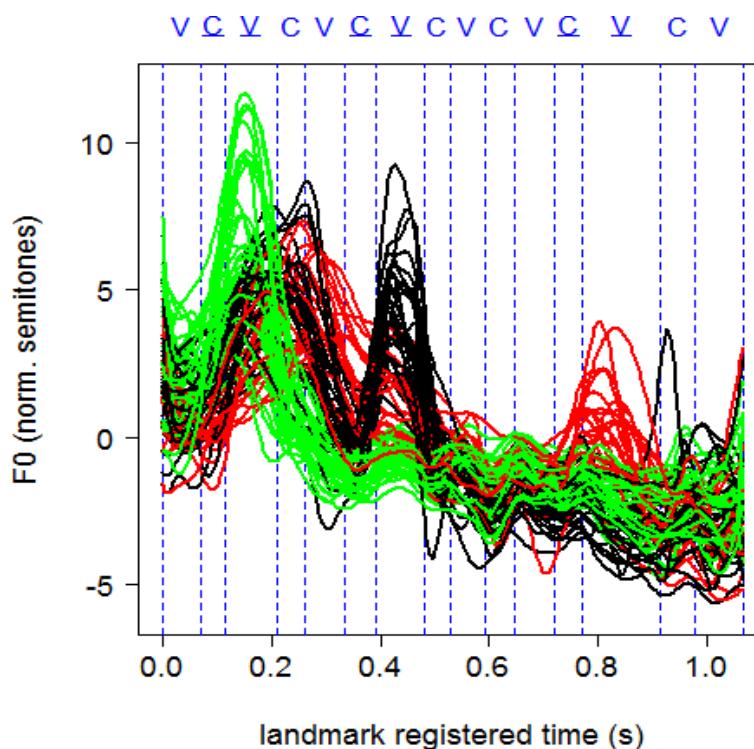


Landmark registration

BEFORE

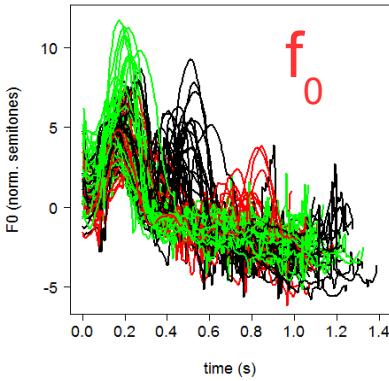


AFTER



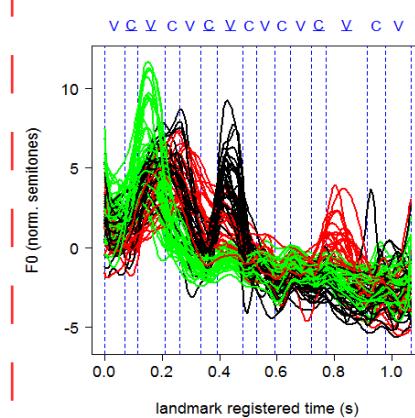
Using landmark registration

CURVES



segment durations

d1	d2	...	d15
...
...
...



NUMBERS

FPCA

ANOVA

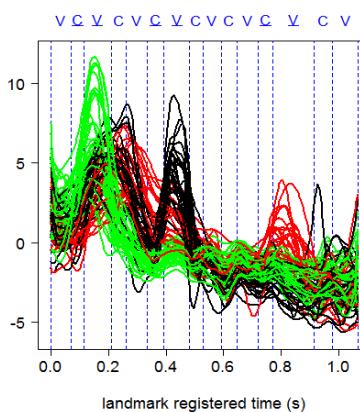
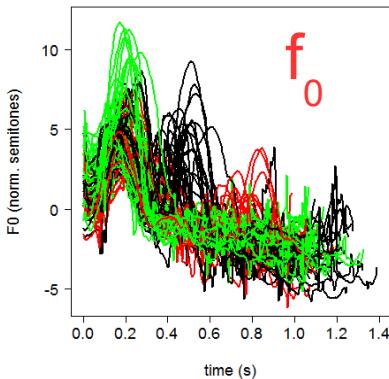
LM

LMER

PCA

Using landmark registration

CURVES



segment
durations

d1	d2	...	d15
...
...
...

NUMBERS



FPCA

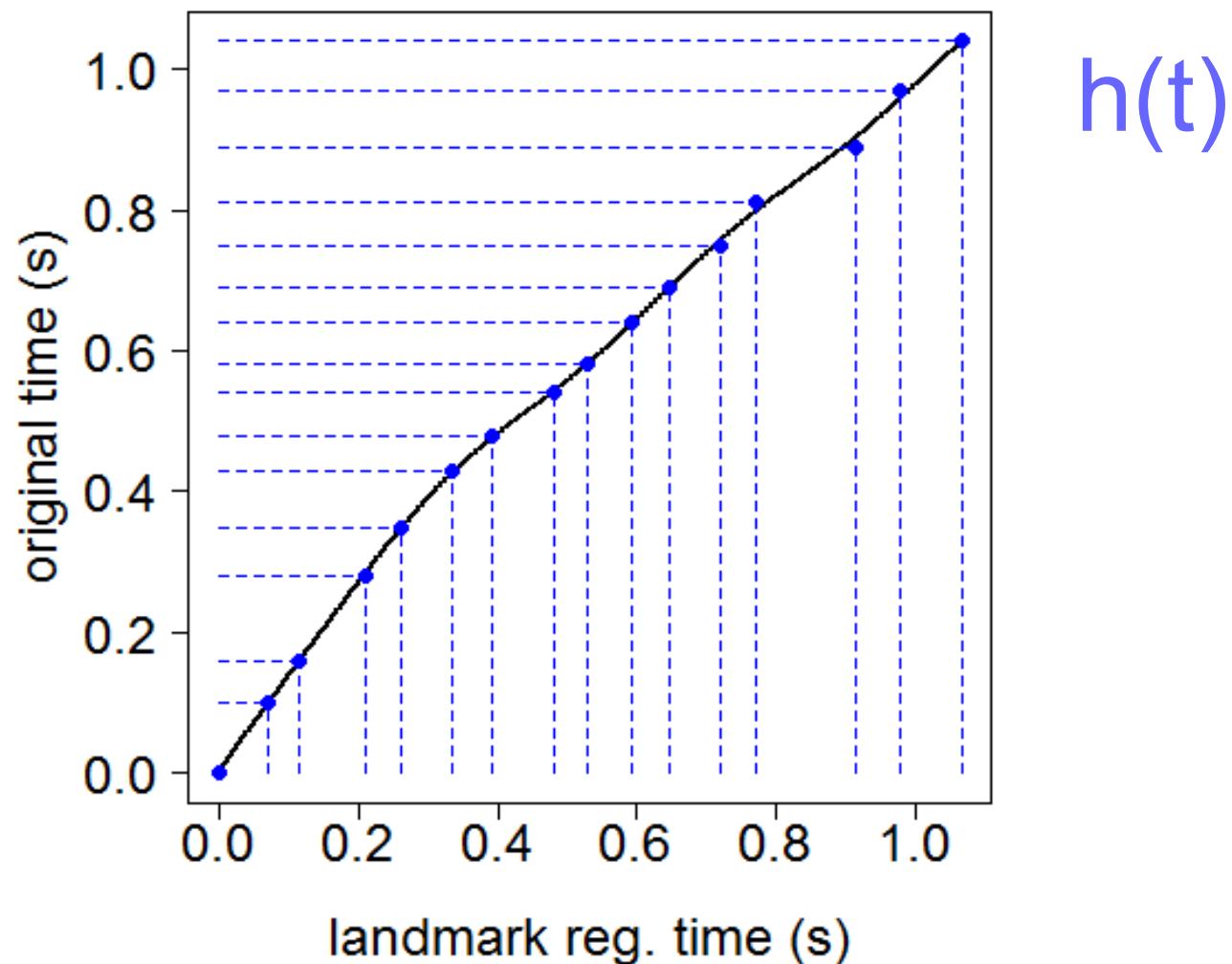


ANOVA

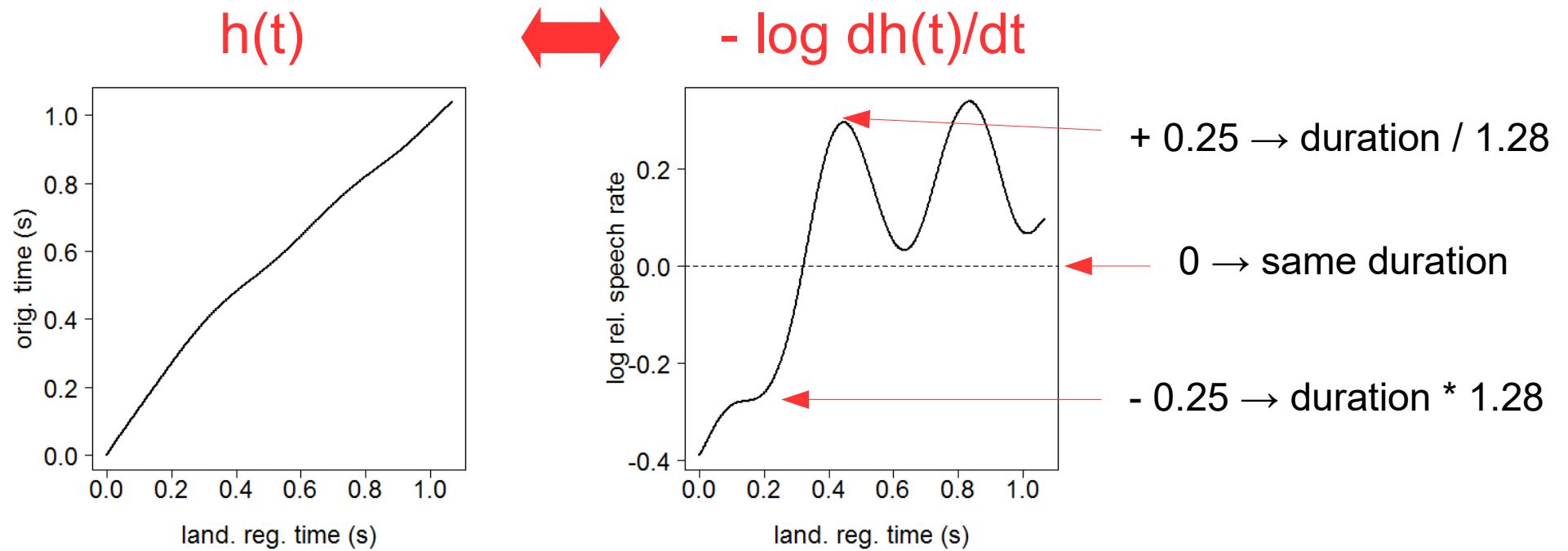
LM

LMER

Inside landmark registration

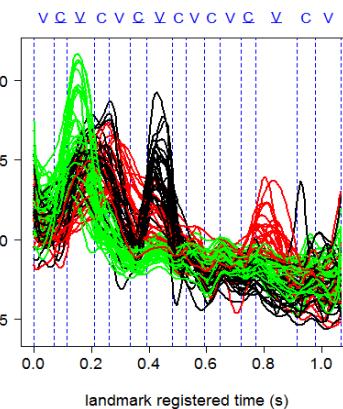
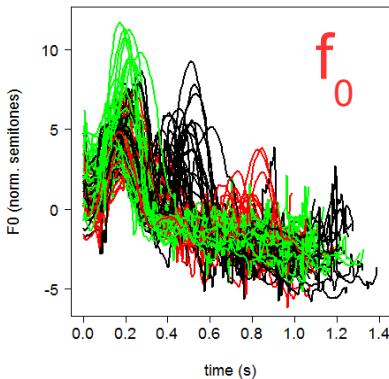


Relative log rate

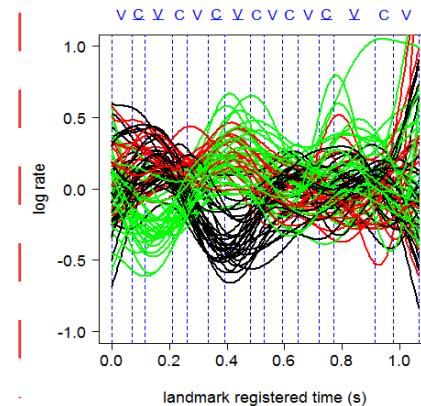


Using log rates

CURVES



log rates



2D
FPCA

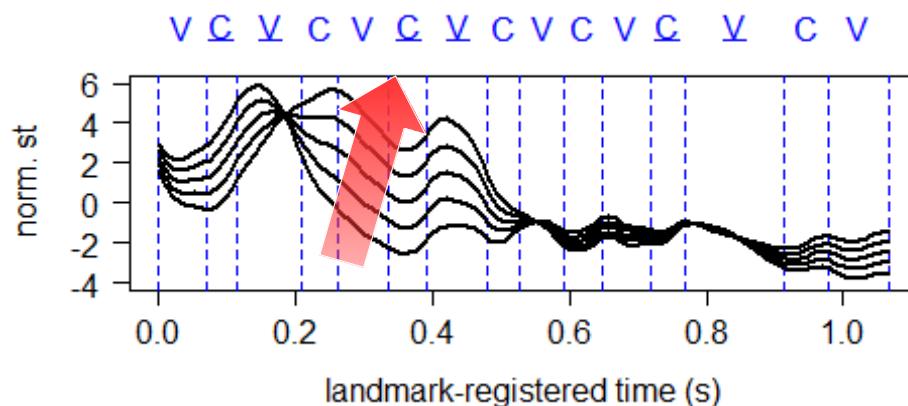
NUMBERS

ANOVA

LM

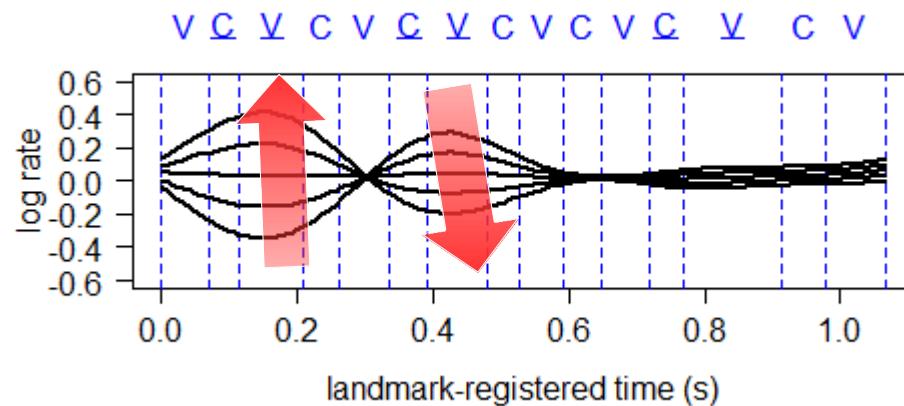
LMER

PC1 scores



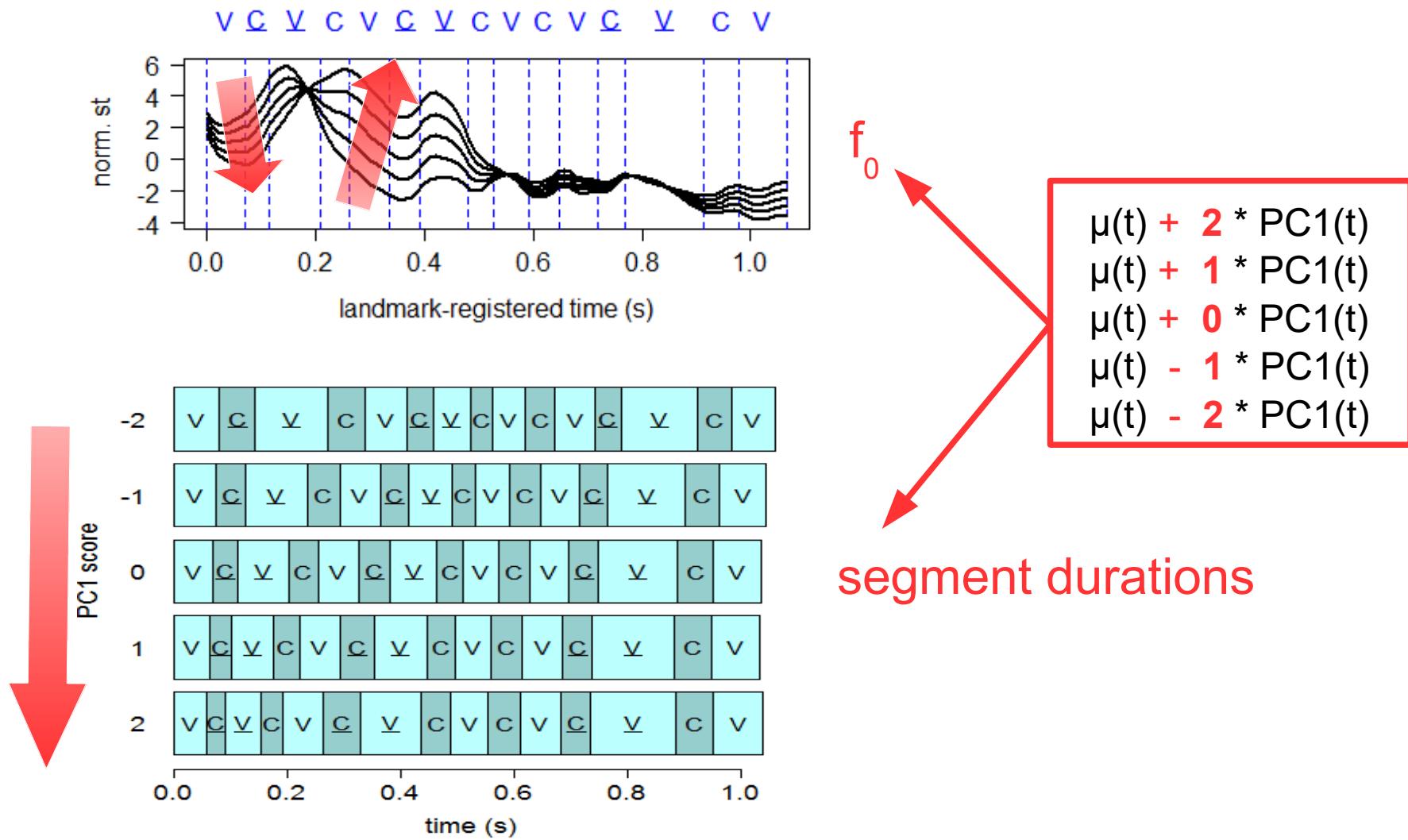
f_0

- $\mu(t) + 2 * \text{PC1}(t)$
- $\mu(t) + 1 * \text{PC1}(t)$
- $\mu(t) + 0 * \text{PC1}(t)$
- $\mu(t) - 1 * \text{PC1}(t)$
- $\mu(t) - 2 * \text{PC1}(t)$

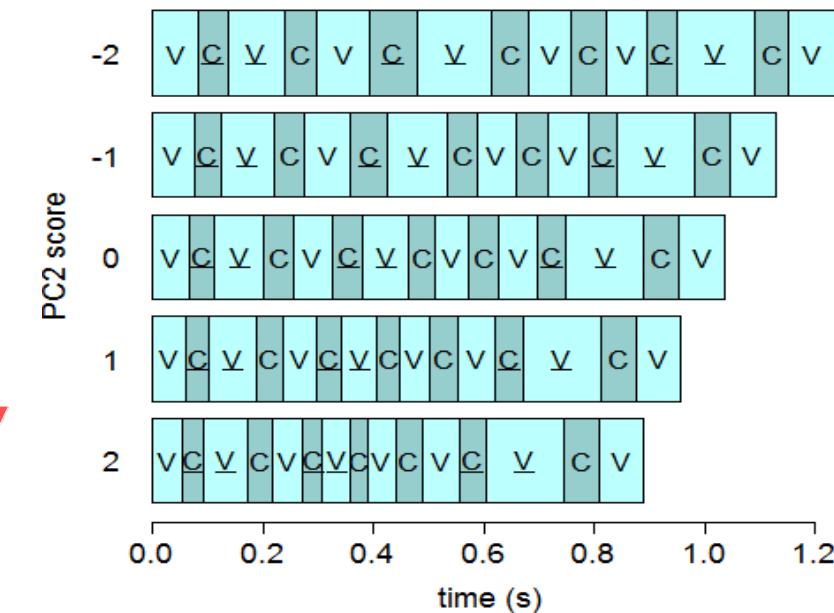
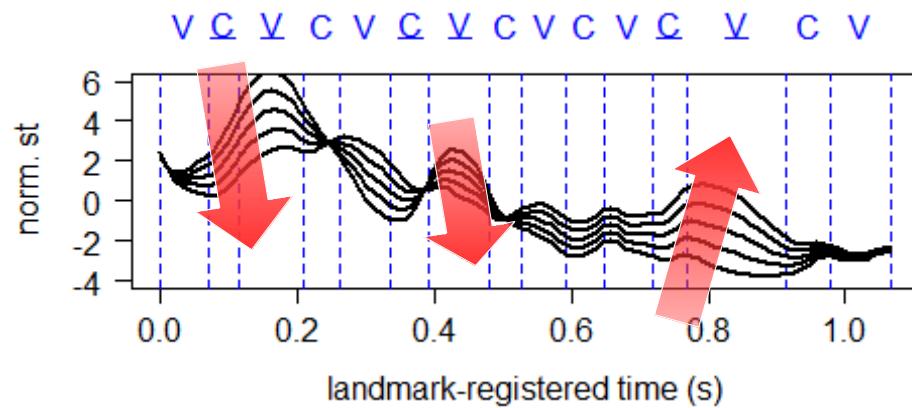


log rates

PC1 scores



PC2 scores



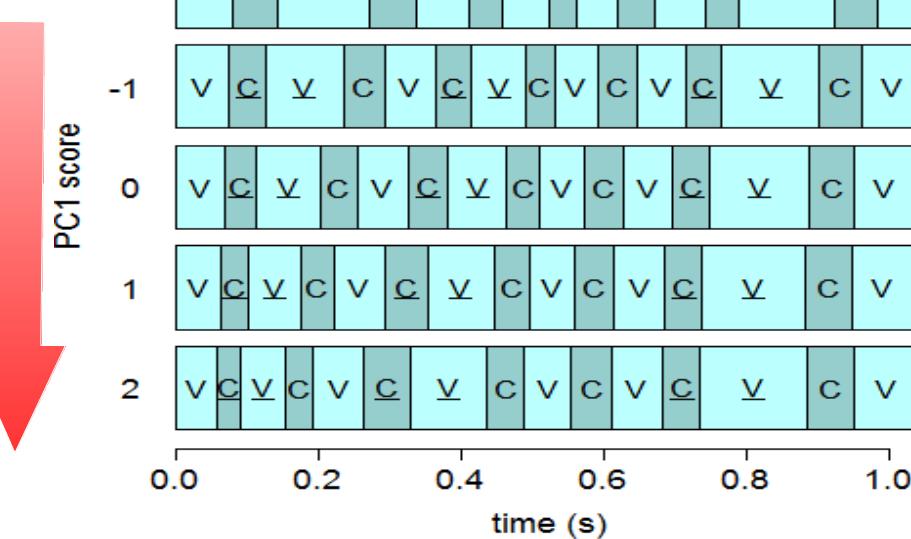
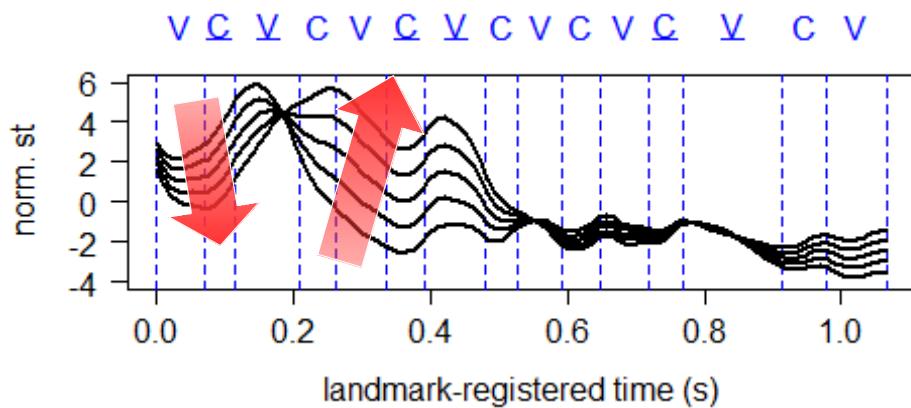
f_0

- $\mu(t) + 2 * \text{PC1}(t)$
- $\mu(t) + 1 * \text{PC1}(t)$
- $\mu(t) + 0 * \text{PC1}(t)$
- $\mu(t) - 1 * \text{PC1}(t)$
- $\mu(t) - 2 * \text{PC1}(t)$

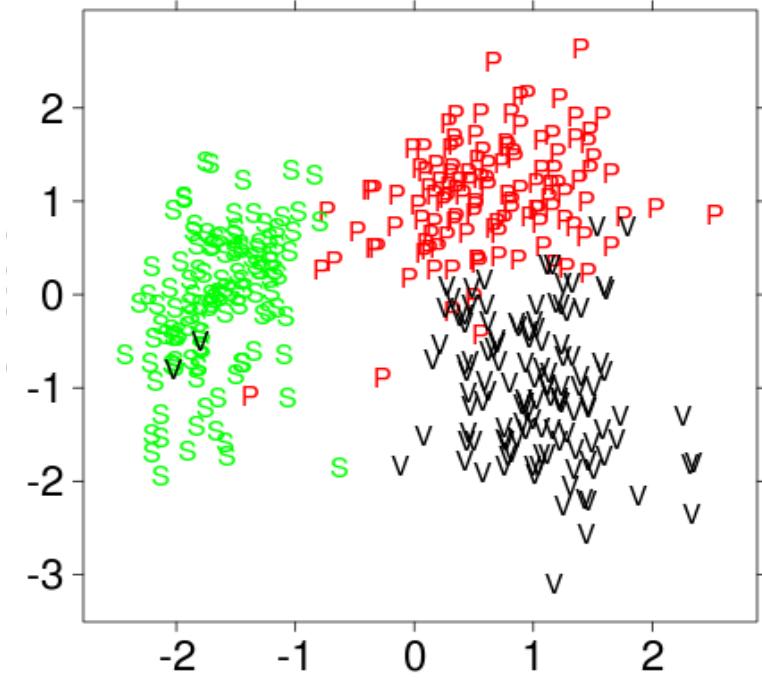
segment durations

multi-segment curve parametrisation

PC1 score



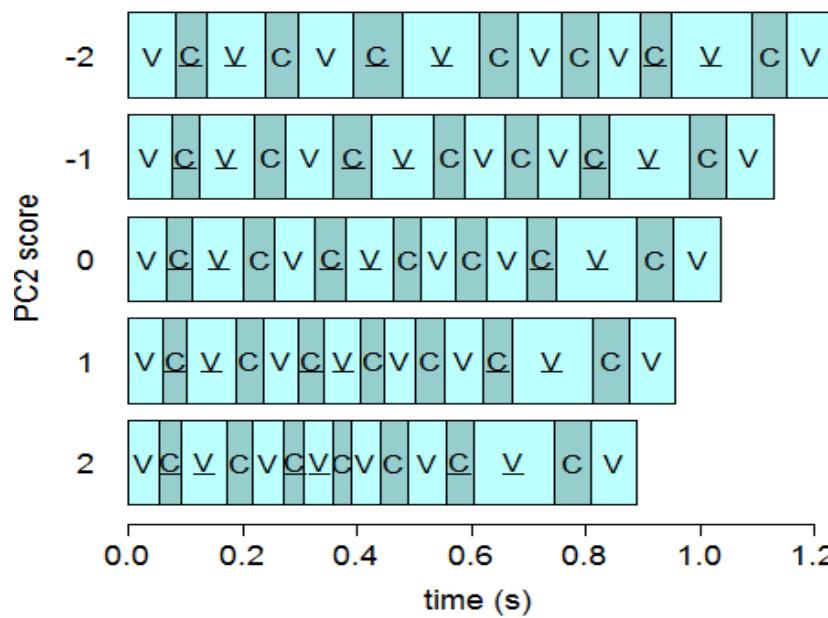
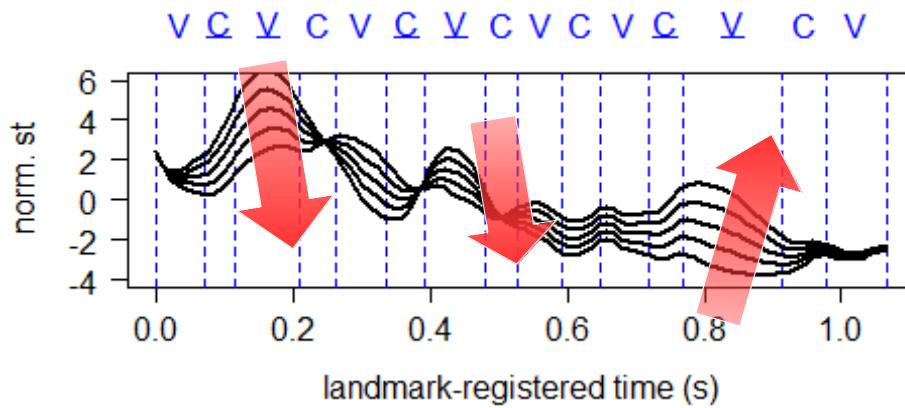
PC2 score



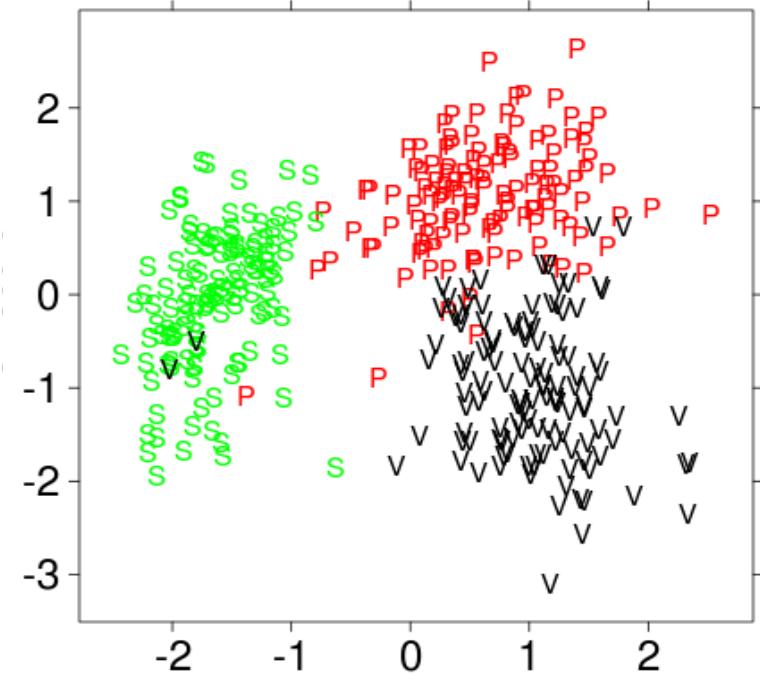
PC1 score

multi-segment curve parametrisation

PC2 score



PC2 score



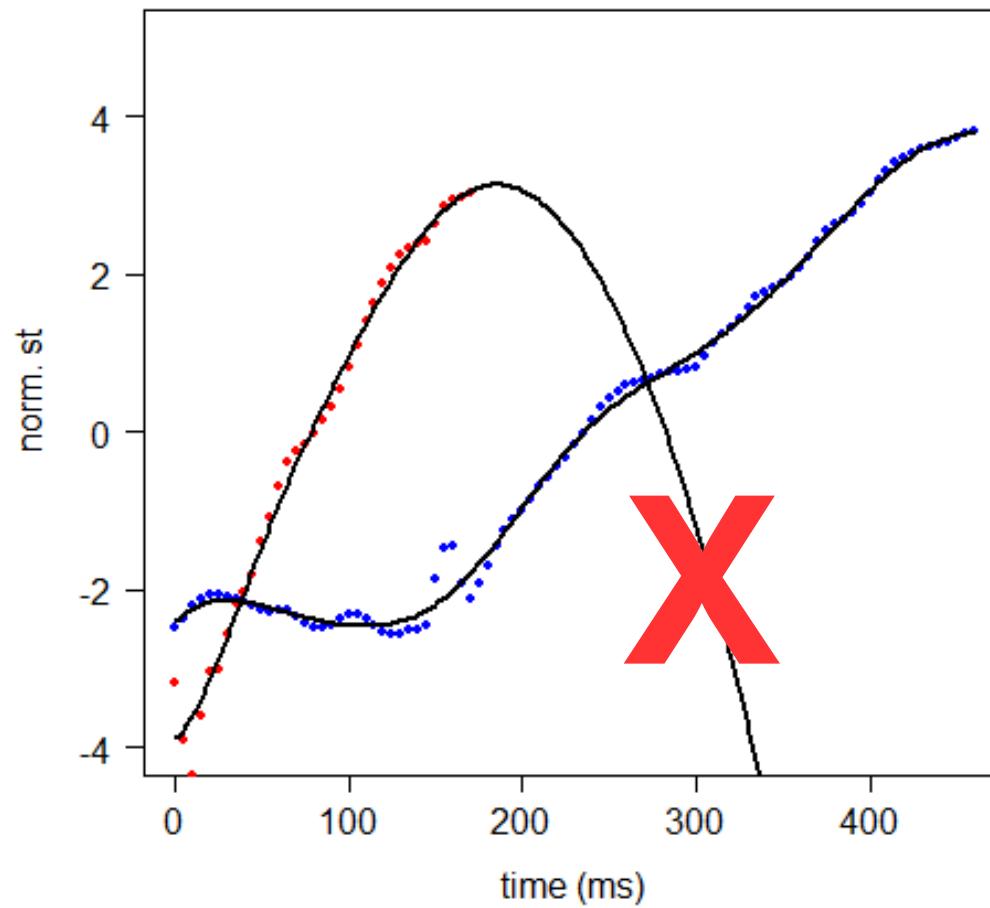
PC1 score

Try this at home?

- R package **fda**
- my website: <http://lands.let.ru.nl/FDA/>
where you find papers, R scripts, tutorials, etc.
- **There is more!**
 - FPCA as re-synthesis tool for perceptual experiment
 - FPCA as map to save on manual (ToBI) annotation

appendix

Take longest duration



Take shortest duration

