



# Dissimilarity for functional data clustering based on smoothing parameter commutation

研究方法（一）期末報告

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# 介紹

- We propose a novel and easy method to implement **dissimilarity measure for functional data clustering based on smoothing splines and smoothing parameter commutation.**
- Our method takes into account the estimation uncertainty using smoothing parameter commutation and is not strongly affected by outliers. It can also be used for **outlier detection.**

# Functional Data Clustering Methods

Functional data clustering :

1. dissimilarity-based methods
2. decomposition-based methods
3. model-based methods

Dissimilarity-based methods : using pointwise dissimilarities between pairs of subjects, are the most straightforward approach.

# Smoothing Spline

Assume that the curve of the  $i$ -th subject is observed as a set of measurements  $\{y_{i1}, y_{i2}, \dots, y_{iK_i}\}$  contaminated by noises at distinct finite time points  $\{t_{i1}, t_{i2}, \dots, t_{iK_i}\}$  in an interval  $[T_L, T_U]$  according to the model

$$y_{ik} = f_i(t_{ik}) + \varepsilon_{ik}, k = 1, 2, \dots, K_i, i = 1, 2, \dots, n$$

A reasonable estimation of  $f_i$  (a smoothing spline  $\hat{f}_i(\cdot; \lambda)$ ) is to

$$\min \frac{1}{K_i} (y_i - f_i)^T (y_i - f_i) + \lambda \int_{T_L}^{T_U} (f_i''(t))^2 dt$$

# Find $\lambda$

Mixed-effects model to find  $\lambda$ :

$$y_i = X_i\beta_i + u_i + \varepsilon_i$$

Under the **Gaussian assumption** for  $\varepsilon_i$  and  $u_i$ , the two variance components  $\sigma_u^2$  and  $\sigma^2$  can be determined based on the restricted maximum likelihood method (REML), so that  **$\lambda$  is also determined.**

# Smoothing Parameter Commutation dissimilarity

Concept : If the 'true'  $f_i$  and  $f_j$  are similar, it is expected that  $\hat{f}_i$  and  $\hat{f}_j$  should be close, given an identical smoothing parameter  $\lambda$ .

The dissimilarity between subjects  $i$  and  $j$ :

$$d_{i,j} = \frac{1}{2} \left\{ \left[ \int_{T_L}^{T_U} (\hat{f}_i(t; \hat{\lambda}_i) - \hat{f}_j(t; \hat{\lambda}_i))^2 dt \right]^{\frac{1}{2}} + \left[ \int_{T_L}^{T_U} (\hat{f}_i(t; \hat{\lambda}_j) - \hat{f}_j(t; \hat{\lambda}_j))^2 dt \right]^{\frac{1}{2}} \right\}$$

we call it a Smoothing Parameter Commutation dissimilarity

# 結論

1. data observed at irregular time points can be handled directly.
2. The concept of the proposed dissimilarity measure is simple and easy to implement.
3. The dissimilarity also serves as a useful tool for outlier detection.





*Thank  
you*