

#### Graph Matching improves the Similarity of Structural Brain Networks

Anna Calissano, Théodore Papadopoulo, Xavier Pennec, Samuel Deslauriers-Gauthier

#### ▶ To cite this version:

Anna Calissano, Théodore Papadopoulo, Xavier Pennec, Samuel Deslauriers-Gauthier. Graph Matching improves the Similarity of Structural Brain Networks. OHBM 2023 - Organization for Human Brain Mapping, Jul 2023, Montreal (Canada), Canada. hal-04390489

HAL Id: hal-04390489

https://hal.science/hal-04390489

Submitted on 12 Jan 2024

**HAL** is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

# Graph Matching improves the Similarity of Structural Brain Networks

Anna Calissano<sup>1</sup>, Theodore Papadopoulo<sup>2</sup>, Xavier Pennec<sup>1</sup>, Samuel Deslauriers-Gauthier<sup>2</sup> Inria Centre at Université Côte d'Azur (1. EPIONE team, 2. Cronos team)

25 Parcels Left Hemisphere



## **ABSTRACT**

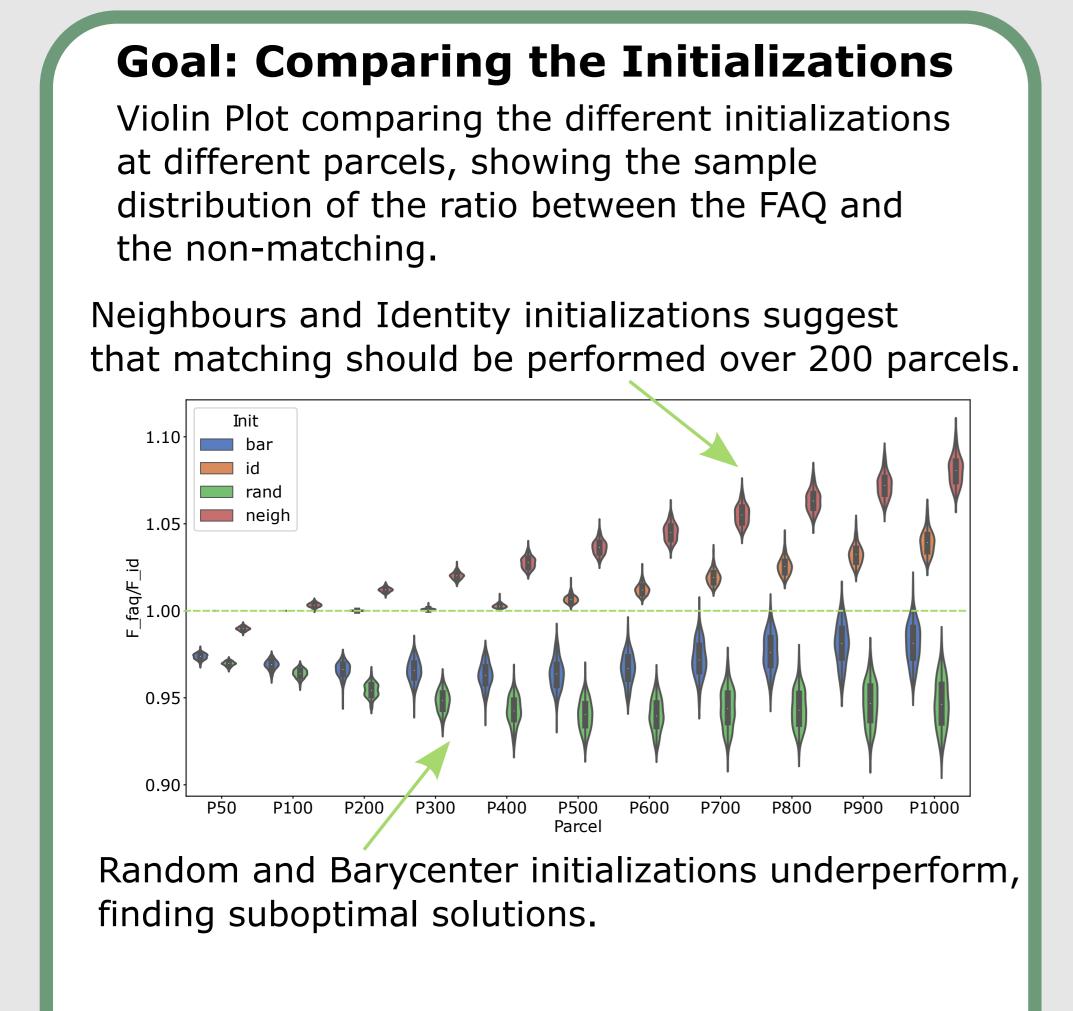
Structural parcellations are often built under the assumption of parcels' correspondences across patients. In the current work, we test the robustness of this hypothesis using graph matching strategies. Given the structural parcellation built in [1], we analyse a set of patients at different parcellation granularity (from 50 up to 1000 parcels). The aim of the analysis is to test via graph matching if the graphs obtained with a given parcellation should be considered as labelled or unlabelled. The results show a node correspondence for low resolution and the requirement of an unlabelled embedding for higher number of parcels. We showcase the graph matching results for the different parcellations as well as the importance of a correct geometrical embedding for unlabelled graphs to compute graph-valued statistics.

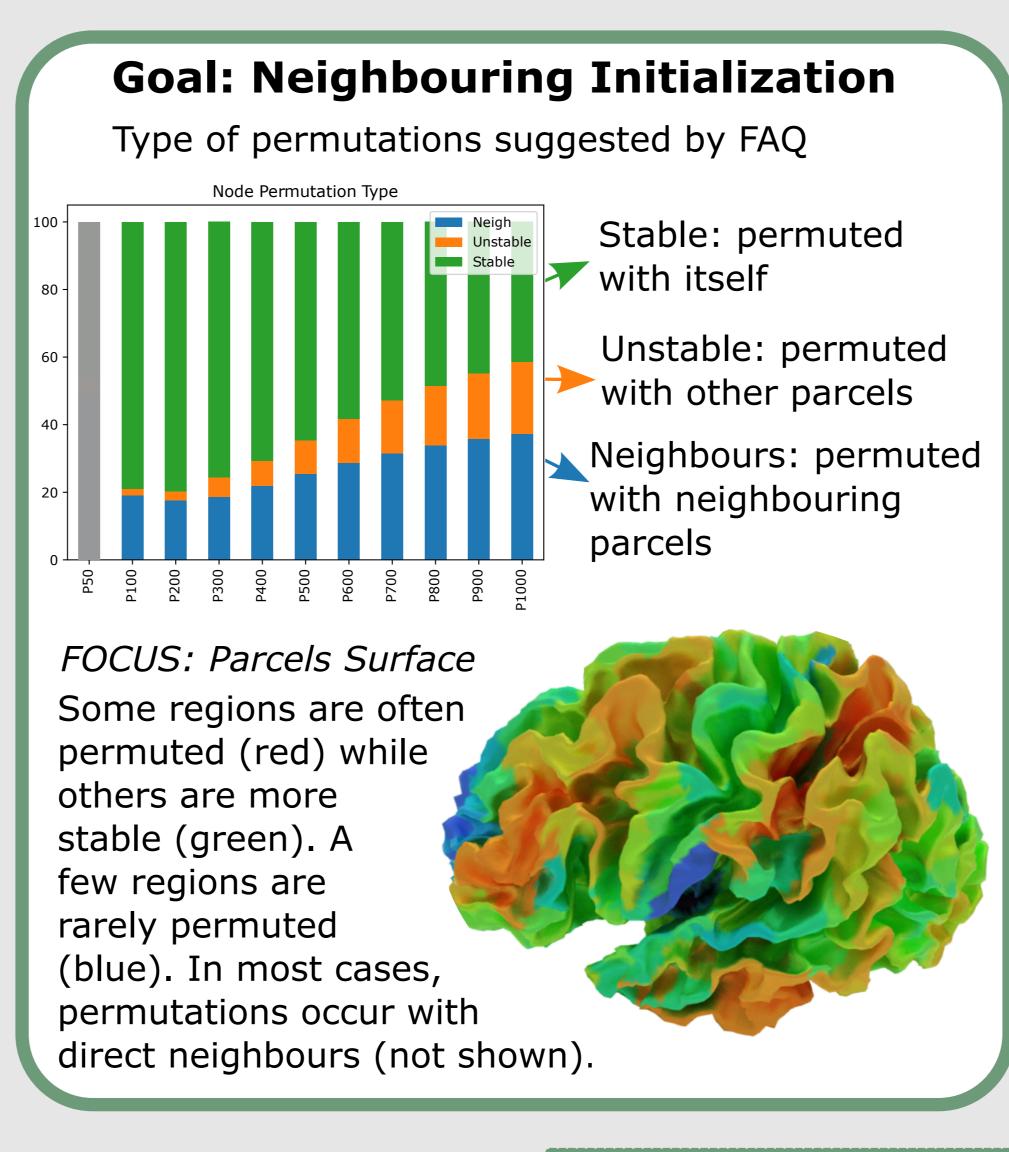
#### MODEL Tractograms (rows) in Subject's Seed-to-Seed 3 Graph Space **Extracted Parcellation** 1 Parcellation $A_i \in \mathbb{R}^{p \times p}$ Every hemisphere and p parcels Gallardo Percellations [1] is Log Weighted Adjacency Matrices built based on 138 tractograms **Graph Space** [2] G = A/P a quotient space obtained by applying the node surfaces from Human permutation P action to an adjacency matrix: $[A] = \{P^TAP : P \in \mathcal{P}\}$ . The Connectome Project. space of A can be equipped with the Frobenius norm $d_A(A_i, A_j) = trace(A_i A_i^T)$ , Each tractogram is which induces the following distance on the quotient: CLUSTERING transformed with a logit model $d_G([A_i], [A_j]) = \min_{P \in \mathcal{D}} d_F(A_i, P^T A_j P)$ the parcels are obtained with Hierachical Clustering. IS PATIENT i ALIGNED with PATIENT j? CLUSTERING $\widetilde{\epsilon}_c \sim \mathcal{N}(0, \sigma_c^2 I) - logit(\widetilde{T}_{sp}) - \widetilde{\epsilon}_s \sim \mathcal{N}(0, \sigma_s^2 I)$ **HP. Node Correspondence** 4 Fast Quadratic Assignment Matching (FAQ) **FAQ** [3] Given two adjacency matrices $A_i, A_j \in \mathbf{R}^{p \times p}$ 2 Structural Brain Data $\min f(P) = \operatorname{trace}(A_i P A_i^T P^T)$ Data: 100 healthy patients from HCP s.t. $P \in \mathcal{D} = \{ P \in \mathbb{R}^{p \times p} : P^T 1 = P 1 = 1, P \ge 0 \}$ Parcels: 25 to 500 nodes per hemisphere Graphs: 100 Structural Brain Connectivity Given $P_0 \in \mathcal{D}$ , FAQ is using Gradient descent in $\mathcal{D}$ .

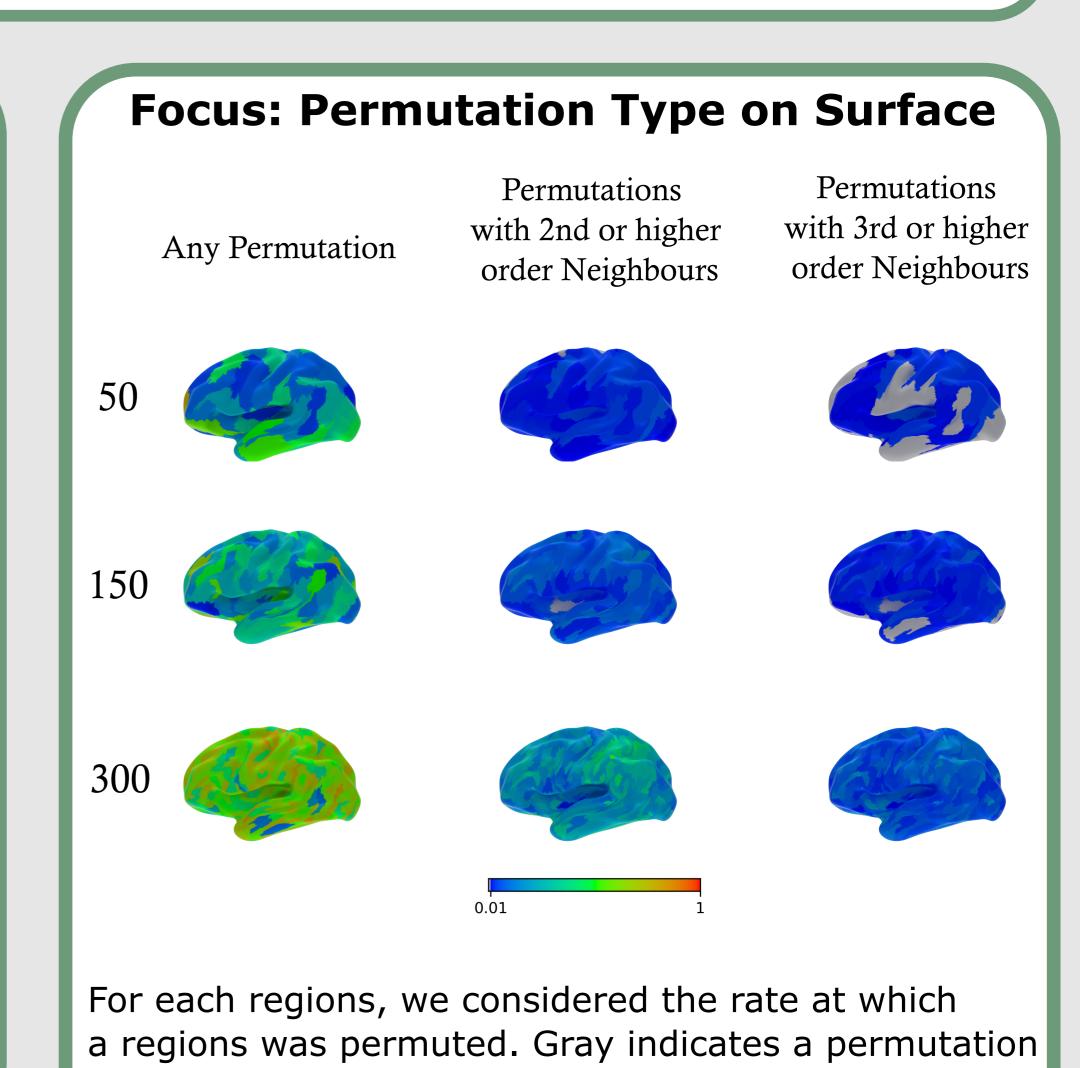
## SIMULATIONS PARAMETERS

Initialization: four different initial values are compared **EVALUATION FUNCTION:** Objective function obtained with FAQ Objective function obtained without matching

500 Parcels Left Hemisphere







### CONCLUSION

### THE ALIGNMENT DEPENDS ON THE PARCELLATIONS

In this work, we studied the alignment across patients for structural brain connectivity network at different granularity of parcellations. We used the FAQ graph matching strategy as state of the art method to perform alignment. As the dimension of the parcels decreases, the parcels start being mis-aligned. The matching procedure shows how the optimal matching is realized with neighbouring parcels. Analyzing graphs without alignment is a model choice, which should be taken with caution when the dimension of the parcels decreases.

### **AKNOWLEDGMENT**

This work has been partially financed by a ERC "Geometric Statistics". The authors are grateful to the OPAL infrastructure from Université Côte d'Azur for providing

rate below 1%.

The solution  $D \in \mathcal{D}$  is projected giving  $P \in \mathcal{P}$ .

## REFERENCES

PLOS one, 10(4), e0121002.

- esources and support.
  - the whole cortex: A logistic random effects model based approach. NeuroImage, 170, 307-320. [2] Jain, B. J., & Obermayer, K. (2009). Structure Spaces. Journal of Machine Learning Research, 10(11). [3] Vogelstein, J. T., Conroy, J. M., Lyzinski, V., Podrazik, L. J., Kratzer, S. G., Harley, E. T., Fishkind, D. E., and Vogelstein, R., & Priebe, C. E. (2015). Fast approximate quadratic programming for graph matching.

[1] Gallardo, G., Wells III, W., Deriche, R., & Wassermann, D. (2018). Groupwise structural parcellation of