

Manifold learning reveals anomalies of language and memory processing in patients with temporal lobe epilepsy

Sonja Banjac, Felix Renard, Elise Roger, Arnaud Attye, Emilie Cousin, Cédric Pichat, Laurent Lamalle, Lorella Minotti, Chrystèle Mosca, Alexandre Krainik, et al.

▶ To cite this version:

Sonja Banjac, Felix Renard, Elise Roger, Arnaud Attye, Emilie Cousin, et al.. Manifold learning reveals anomalies of language and memory processing in patients with temporal lobe epilepsy. OHBM, Jun 2020, Virtual conference, Canada. hal-03248011

HAL Id: hal-03248011 https://hal.univ-grenoble-alpes.fr/hal-03248011

Submitted on 3 Jun 2021

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.





Manifold learning reveals anomalies of language and memory processing in patients with temporal lobe epilepsy

Sonja Banjac¹, Félix Renard¹, Elise Roger¹, Arnaud Attyé², Emilie Cousin¹, Cédric Pichat¹, Laurent Lamalle³, Lorella Minotti⁴, Chrystèle Mosca⁴, Alexandre Krainik³, Philippe Kahane⁴, Monica Baciu¹

¹LPNC, Univ. Grenoble Alpes, CNRS LPNC UMR 5105, F-38000 Grenoble, France ² School of Biomedical Engineering, University of Sydney, Australia















Context and objectives

- Temporal lobe epilepsy (TLE) patients show neural reorganization of language and memory networks^{1,2} in addition to alteration of different resting-state networks³.
- These two functions have mainly been studied separately, but there is growing evidence about their dynamic interaction^{4,5,6}.
- We aim to explore abnormal patterns of activation within the joint language and episodic memory network in TLE patients during the tasks that engage one or the other function or both simultaneously.

Methods

- 21 right-handed participants (9 females, M_{age} = 21) and 12 left TLE patients (7 females, $M_{age} = 35$) performed sentence generation (GE), visual recognition (RECO) and recall with sentence generation (RA) tasks of GE2REC protocol⁷.
- The ROIs were defined in the Brainnetome atlas⁸ as those regions that had more than 40% of the voxels activated (p <.001, uncorrected; k > 20) in the control group during the tasks.
- Mean beta values were extracted using Rex toolbox⁹ for each of the ROIs for each participant and patient for each task.

1. Dimension reduction

By using the Umap¹⁰ a set of standardized ROI beta values was converted into a unique point in a manifold subspace representing one participant during one task.

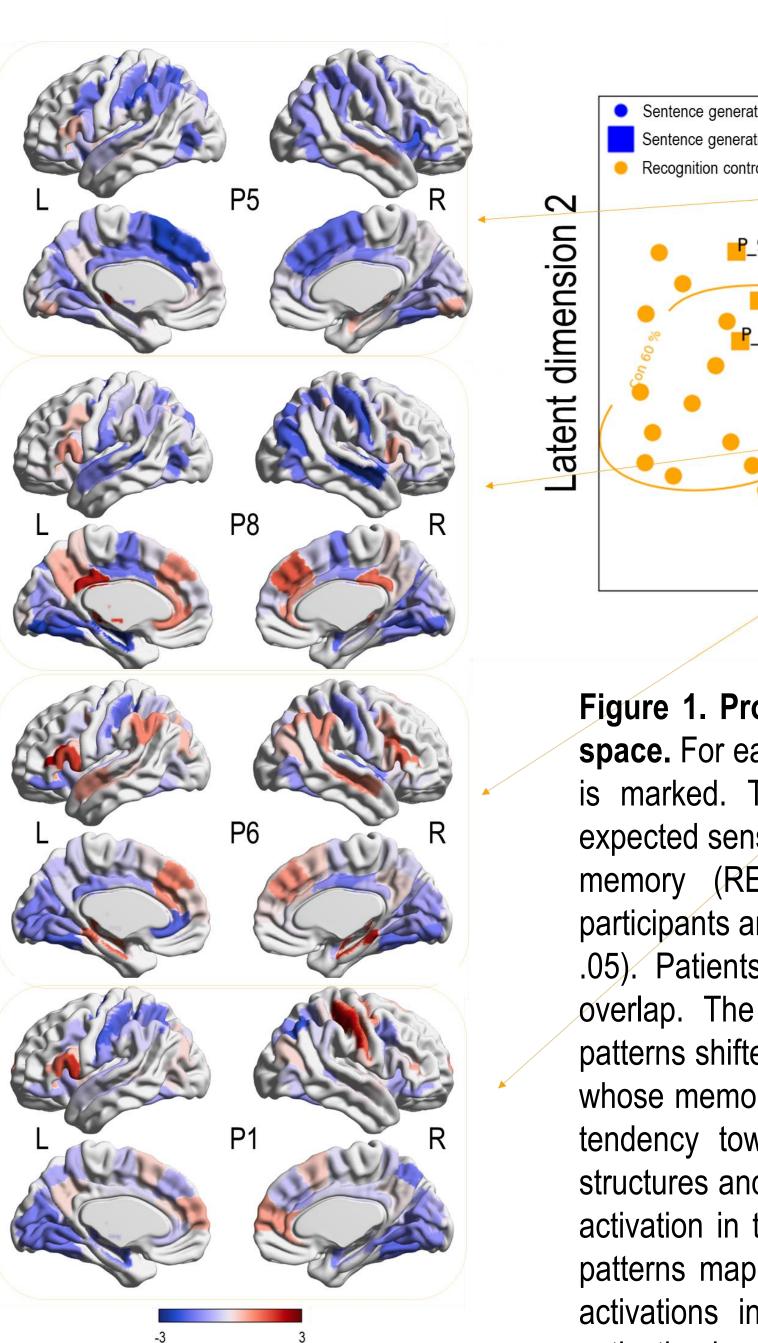
3. TLE patients analysis

Patients are projected onto the learned manifold space in the point they would hold if they were healthy participants.

Pipeline

ullet depicts the abnormalities in the patient's dataset in the case when it is greater than the model variability obtained by the leave-one-out procedure.

Results



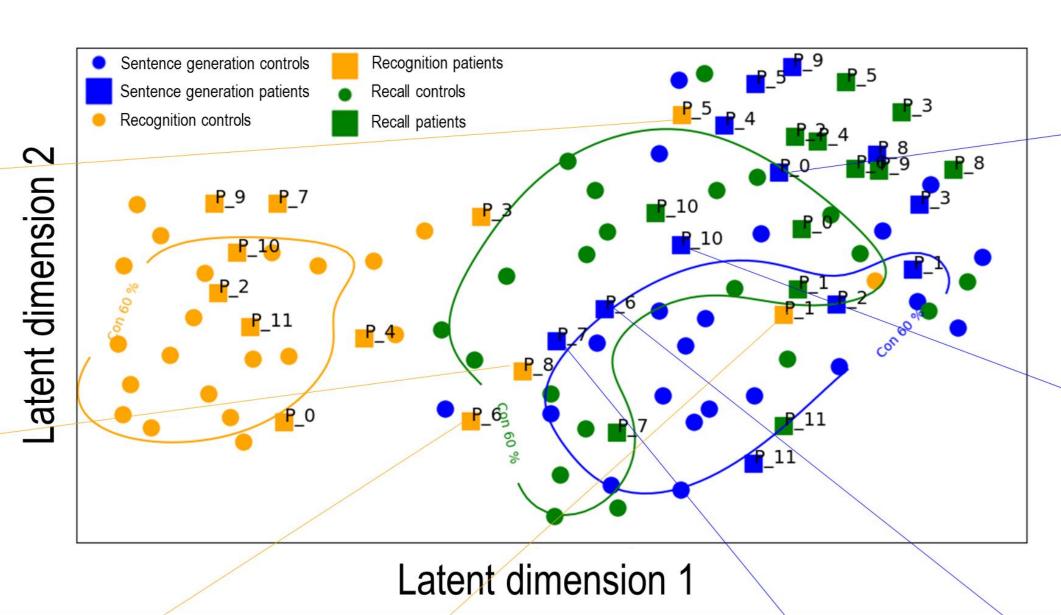


Figure 1. Projection of the patients onto the reconstructed manifold **space.** For each task the place where 60% of healthy subjects are situated is marked. The reconstructed manifold differentiated the tasks in an expected sense, with mixed task (RA) situated between language (GE) and memory (RECO) tasks. The highest dissimilarity between healthy participants and patients was found for the mixed task (RA) ($D_{KI} = 1.37$, p < 1.37.05). Patients also showed a higher tendency for dispersion and task overlap. The z-scores of patients whose language and memory task patterns shifted towards the mixed (RA) task are illustrated. Those patients whose memory patterns mapped on to the mixed task generally showed a tendency towards lower than expected activation in the hippocampal structures and the fusiform gyrus and some of them tended to have higher activation in the inferior frontal region. While those whose language task patterns mapped on to the mixed task showed a tendency towards less activations in some parts of inferior frontal gyrus and towards more activation in medial and lateral parietal regions.

Figure 2. Residual map for the language task (GE) for a

2. Model estimation

Y – values of all ROIs in real space

x – matching point in the reduced space

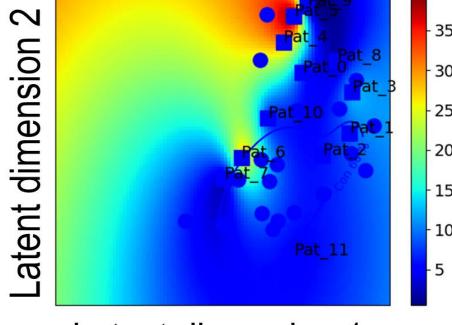
f(x) – non-linear kernel regression function

 $Y = f(x) + \varepsilon$

patient (P6) with lexical access (DO80¹¹). Residuals above/below 2SD were taken as a measure of deviation. Tendency for lower than predicted values was found in the right frontal orbital cortex, while a tendency towards higher values was found in the bilateral precuneus, bilateral middle and posterior cingulum and right angular gyrus.

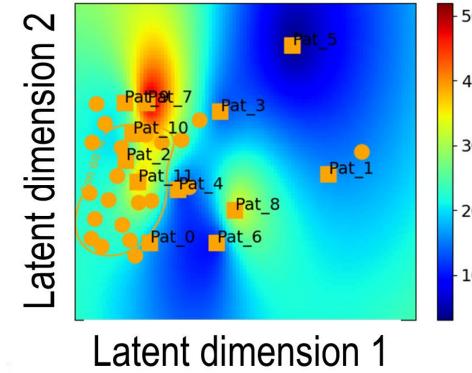
Figure 3. Residual map for memory task for a patient (P0) with difficulties with verbal episodic memory (GB¹²). Residuals above/below 2SD were taken as a measure of deviation. Lower than predicted values are found in the left parahippocampus, bilateral postcentral and bilateral middle occipital, while higher were found in left and right anterior cingulum, right frontal inferior opercularis, triangularis and orbitalis.

Manifold GE with duration

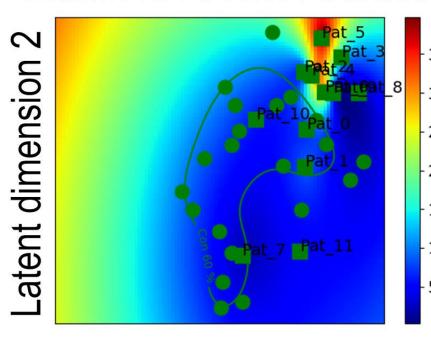


Latent dimension 1

Manifold RECO with AOS



Manifold RA with duration



Latent dimension 1

Figure 4. Projection of epilepsy duration and age of onset (ASO) onto the manifold space. The duration of the epilepsy could help to explain the positioning of the patients in the manifold space. TLE patients with longer duration of the illness tended to map further away from the healthy subjects regarding language (GE) and mixed language and memory task (RA). While TLE patients with earlier start of the epilepsy tended to map further from the healthy participants with respect to the recognition task (RECO).

References

1. Baciu, M., & Perrone-Bertolotti, M. (2015). Reviews in the Neurosciences, 26(3), 323-341. 2. Sidhu, et al. (2013). Brain, 136(6), 1868-1888. 3. Kramer, M. A., & Cash, S. S. (2012). The Neuroscientist, Cognitive Science, 8(4), 722-736. 18(4), 360–372.

67(1), 105–134. 5. Brown-Schmidt, S., & Duff, M. C. (2016). Topics in

6. Larsen, et al. (2002). Memory, 10(1), 45-54.

7. Banjac, et al. (In press). Brain Imaging and behavior.

4. Moscovitch et al. (2016). Annual Review of Psychology, 8. Fan, et al. (2016). Cerebral Cortex, 26(8), 3508–3526. 9. REX toolbox - https://www.nitrc.org/projects/rex/ 10. McInnes et al. (2018). arXiv.

11. Metz-Lutz, et al. (1991). Revue de Neuropsychologie, *1*(1), 73–95. 12. Grober et al. (2008). *J Am Geriatr Soc*, *56*(5), 944–946.

Conclusions

- The non-linear manifold learning for fMRI data dimension reduction can be used as a solution to the problem of high dimensionality of the data.
- This approach enables comparing all the regions of interest simultaneously taking into account their relationship.
- Importantly, using the presented approach we can obtain the measure of deviation.
- The proposed approach is especially relevant for single case and small sample studies.