

The quest for consistency: How to *not* construct functional brain networks

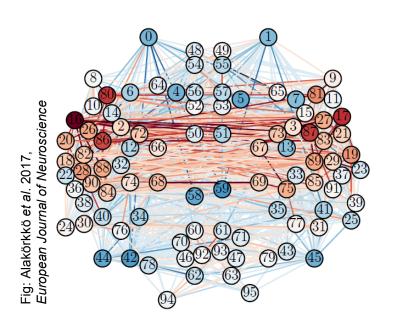
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3.6.2020

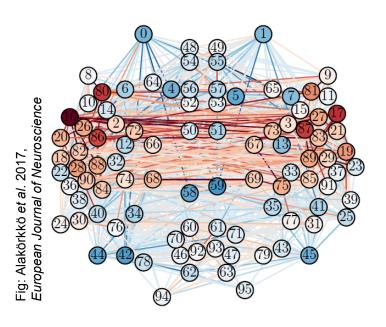
Slides: https://github.com/onerva-korhonen/presentations/blob/master/wins-030620.pdf

Why is the brain a network?



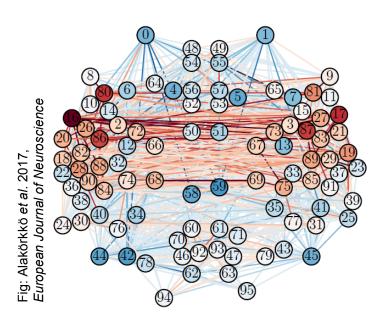
- Brain = a system of neurons
 - Separated neurons tend to reconnect

Why is the brain a network?



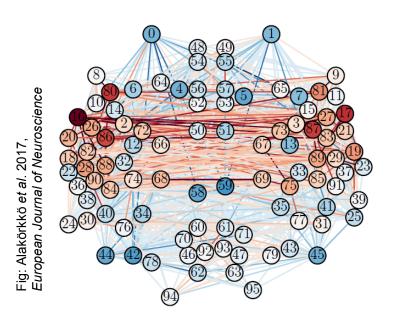
- Brain = a system of neurons
 - Separated neurons tend to reconnect
- Axon bundles connect brain areas

Why is the brain a network?



- Brain = a system of neurons
 - Separated neurons tend to reconnect
- Axon bundles connect brain areas
- Cognitive tasks require collaboration of brain areas

Network neuroscience



 Network neuroscience (Muldoon & Bassett 2016, Bassett & Sporns 2017)
 = applying network tools on the brain

Two aims:

- 1. Understand the healthy brain
- 2. Find causes of diseases

Broad scales:

- Molecule neuron brain area human
- Milliseconds years

Different brain networks:

- Structural: anatomic connections
- Functional: temporal coactivation
- Effective: causality

Functional networks: fMRI



Fig: OpenStar/Wikimedia Commons, under CC BY 4.0

Based on magnetic properties of haemoglobin

- Oxygen-rich and oxygen-poor haemoglobin behave differently in a (strong) magnetic field
 - => oxygen-rich areas localized
- Brain function requires oxygen
 - => high oxygen level = high activity
- Measurement unit = voxel

: Aalto Magnetic Imaging (AMI Center), by permission

High (~mm) spatial resolution, low (~s) temporal resolution

Functional networks: EEG & MEG

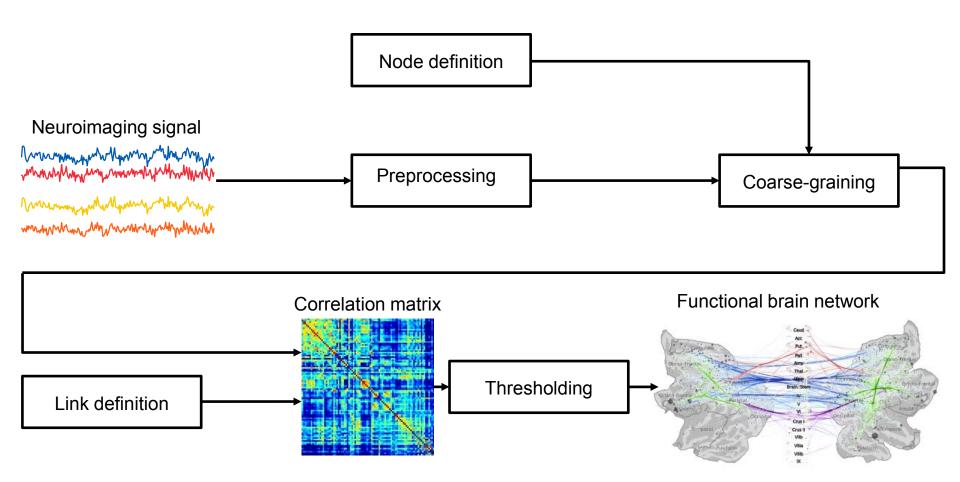


- Neurons interact electrically
- EEG: the electrical field of the brain
- MEG: the magnetic field of the brain
- Inverse model: time series of brain surface vertices



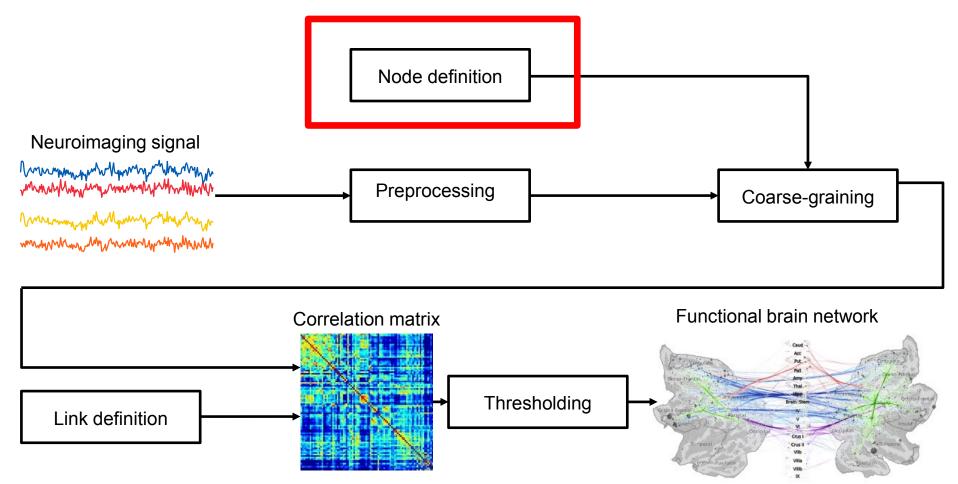
Excellent (~ms) temporal resolution, lower spatial resolution

Functional networks: how-to?



Network from Nummenmaa et al. 2014, *Neurolmage*, by permission

Functional networks: how-to?



Network from Nummenmaa et al. 2014, *Neurolmage*, by permission

Voxels vs ROIs

Voxels:

- fMRI imaging resolution
- noisy signals?
- ~10.000 nodes
- large computational load

ROIs:

- collections of voxels
- defined by anatomy, function, connectivity, ...
- homogeneous (= all voxels have same dynamics)?
- ROI time series to represent voxel dynamics:

$$X_I = \frac{1}{N_I} \sum_{i \in I} x_i$$

Violent?

Consistency of Regions of Interest as nodes of fMRI functional brain networks

Korhonen, O., Saarimäki, H., Glerean, E., Sams, M., & Saramäki, J. 2017. *Network Neuroscience*

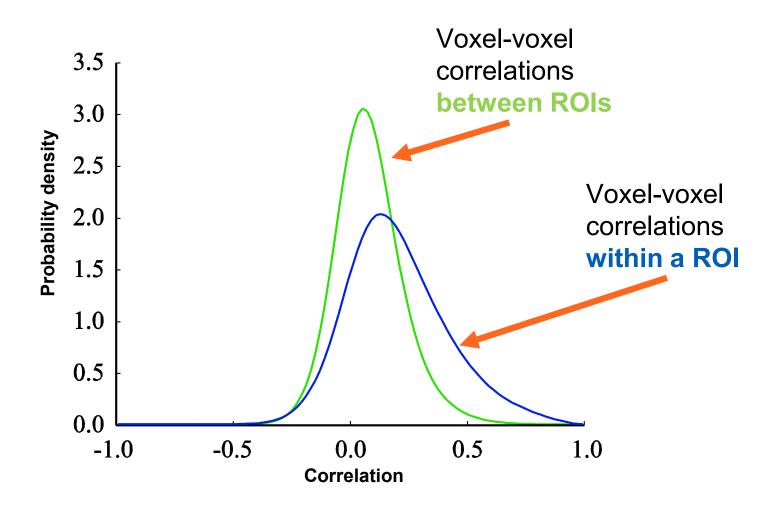
Research questions

- What should nodes of brain networks depict?
 - ROIs or voxels?
- Are ROIs functionally homogeneous?

Methods

- Two sets of resting-state fMRI data:
 - 13 in-house subjects
 - 28 subjects from ABIDE I initiative
- 215 time points (~6 min)
- ROIs from three atlases:
 - HO: anatomical
 - AAL: anatomical
 - Brainnetome: connectivity-based
- Connectivity investigated at voxel and ROI levels

How correlated are voxels of a ROI?

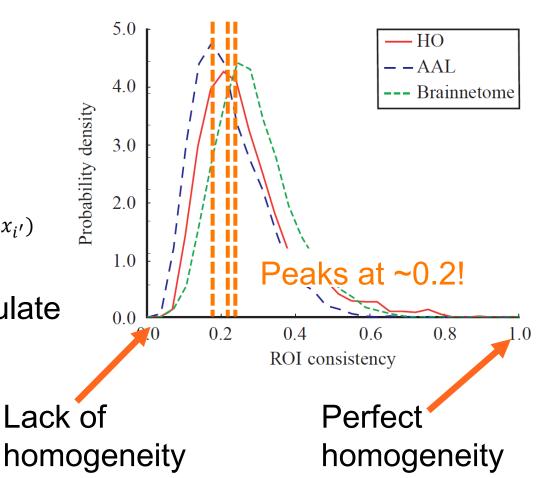


How homogeneous are ROIs?

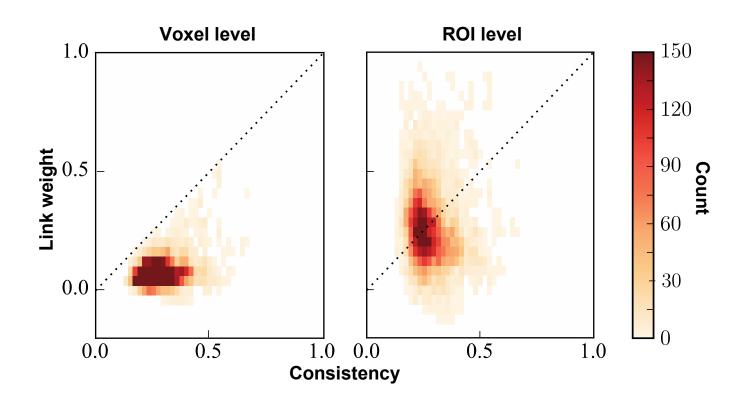
- Spatial consistency
- = measure of functional homogeneity:

$$\varphi_{spat}(I) = \frac{1}{N_I(N_I - 1)} \sum_{i,i' \in I} C(x_i, x_{i'})$$

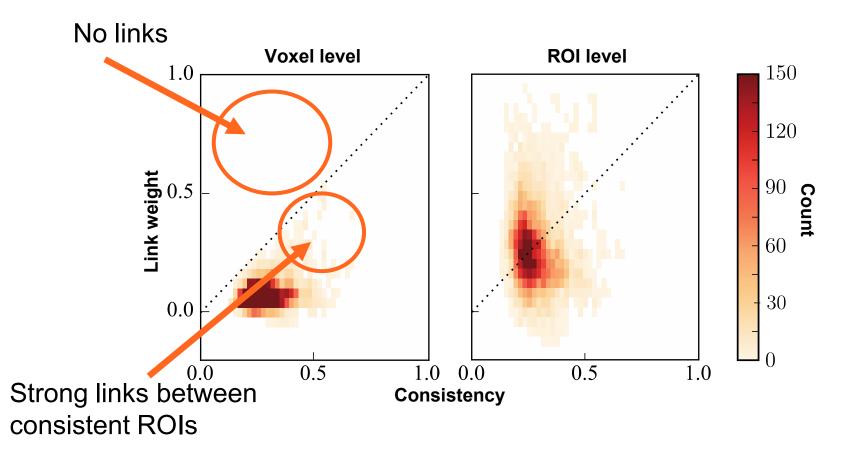
- Straightforward to calculate
- Easy to interpret



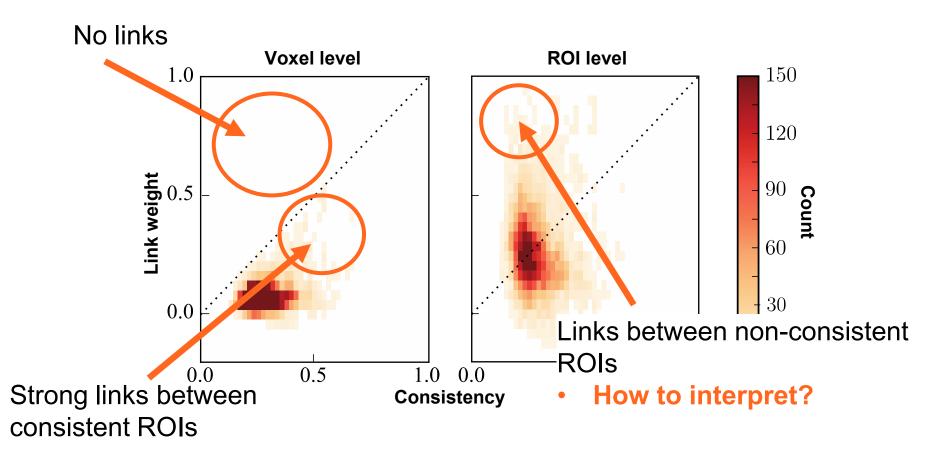
Does consistency predict connectivity?



Does consistency predict connectivity?



Does consistency predict connectivity?



Conclusions

- ROIs are not always functionally homogeneous
- Strong ROI-level correlations between low-consistency ROIs may be spurious
- Does a low spatial consistency tell about
 - a) A bad ROI definition
 - b) High noise level
 - c) Inactivity of the ROI?

Regions of Interest as nodes of dynamic functional brain networks

Ryyppö. E., Glerean, E., Brattico, E., Saramäki, J., & Korhonen, O. 2018, Network Neuroscience

Research questions

- ROIs as nodes of dynamic brain networks?
- Temporal behaviour of spatial consistency?

Methods

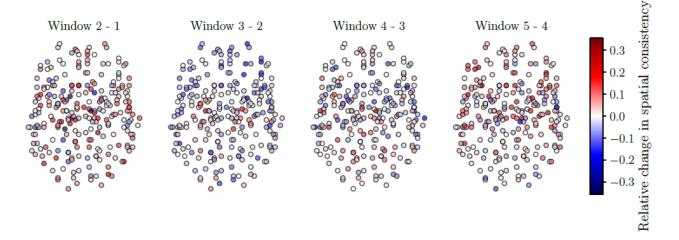
- Two sets of fMRI data:
 - Music listenig (13 subjects)
 - Resting-state (28 subjects)
- ROIs:
 - Brainnetome
 - HO
 - AAL
- Time windows: 80 samples (160s), 50% overlap

Measures

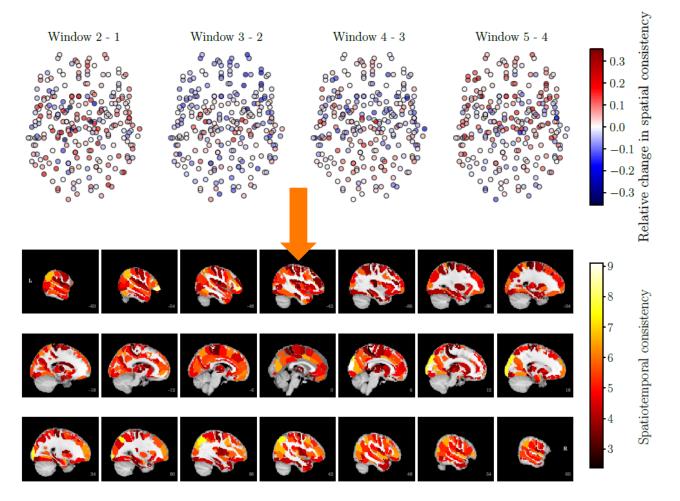
- Spatial consistency φ_{spat} : functional homogeneity of ROI
- Spatiotemporal consistency: time-dependence of φ_{spat}

$$\varphi_{st}(I) = \frac{N_t(N_t - 1)}{2\sum_{t < t'} \frac{\left|\varphi_{spat}(I, t) - \varphi_{spat}(I, t')\right|}{\varphi_{spat}(I, t)}}$$

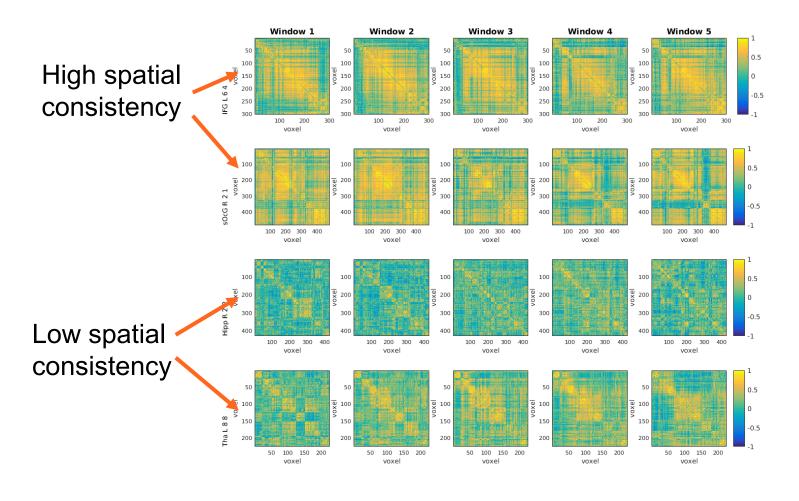
Spatial consistency changes in time



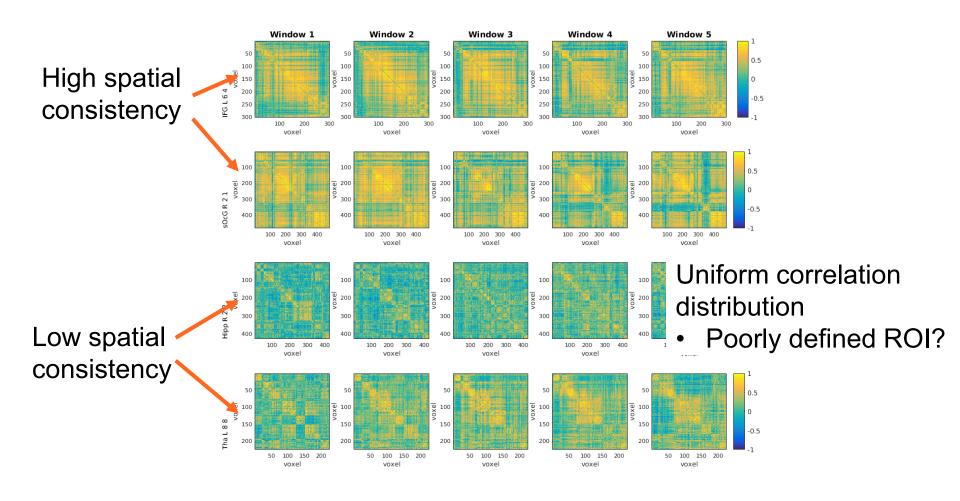
Spatial consistency changes in time



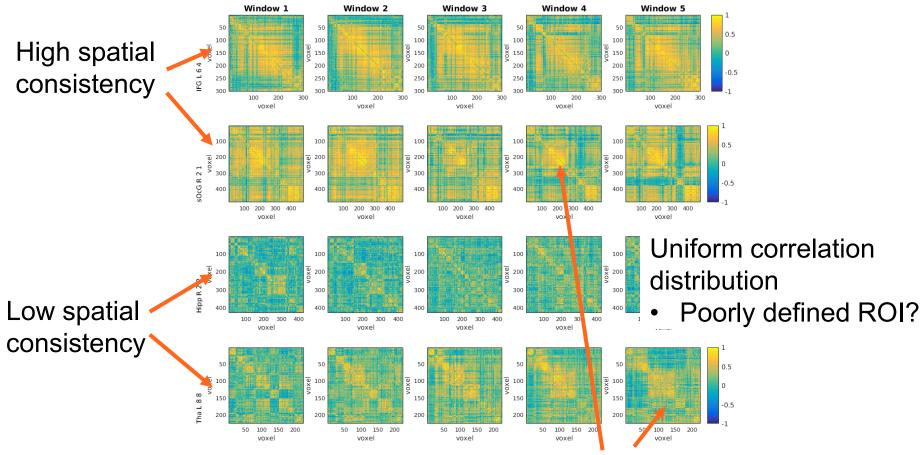
ROIs have rich internal connectivity structure



ROIs have rich internal connectivity structure



ROIs have rich internal connectivity structure



Intra-ROI modules

Network topology?

Conclusions

- Spatial consistency changes in time
 - Reflects activation?
- ROIs have time-dependent internal structure
 - Relates to network topology?
- Do brain networks have stable nodes?

On-going work: Internal connectivity and topological roles of nodes in functional brain networks

with Elisa Ryyppö, Ana Triana & Jari Saramäki

Research questions

- Why does consistency change in time?
- Is consistency connected to network structure?

Methods

- Two sets of fMRI data:
 - Music listenig (13 subjects)
 - Resting-state (28 subjects)
- ROIs: Brainnetome
 - Subject's native space
 - Standard space
- Two link definition approaches
 - Pearson correlation
 - Coarse-graining (Kujala et al. 2016)

Predictors

- ROI size: number of voxels
- 5 measures of internal connectivity:
 - Spatial consistency φ_{spat} : functional homogeneity of ROI
 - Spatiotemporal consistency φ_{st} : time-dependence of φ_{spat}
 - SD of correlations inside ROI: broadness of correlation distribution
 - Self-link weight: number of voxel-level links inside the ROI
 - Internal density: normalized by ROI size

Topological roles

- Global hubness (Klimm et al. 2014):
 - How central the node is?

$$h_i = \frac{s_i - \langle s_R \rangle}{\sigma_R}$$

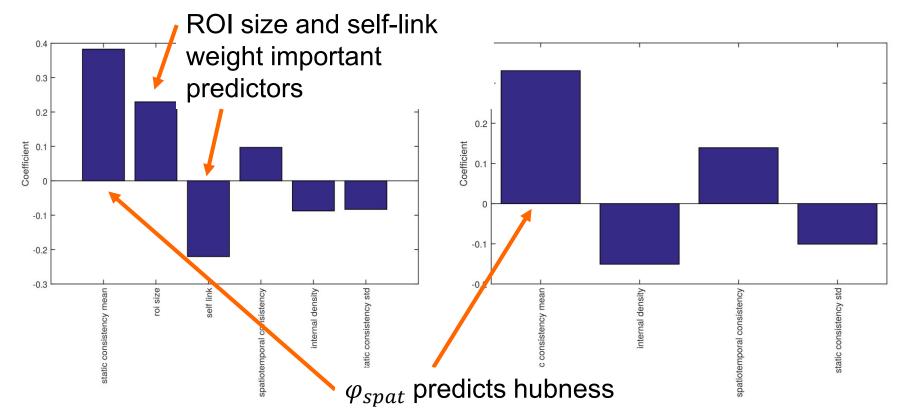
- Connectorness:
 - Does the node build bridges?

$$b_i = \frac{\sum_{l_{ij} > l_t} A_{ij}}{\sum_{l_{ij}} A_{ij}}$$

Prediction with logistic regression

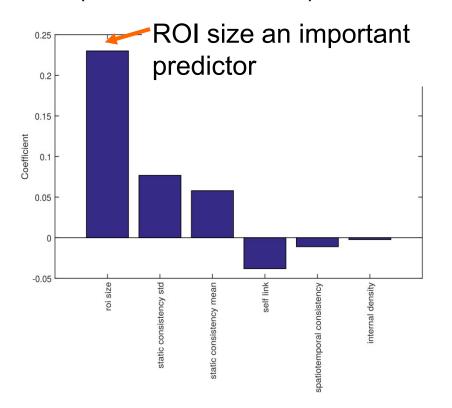
Results: Hub vs non-hub

Accuracy: Training 61.65% Test 60.10% (> Random 50.03%) Accuracy: Training 60.75% Test 59.96% (> Random 50.03%)

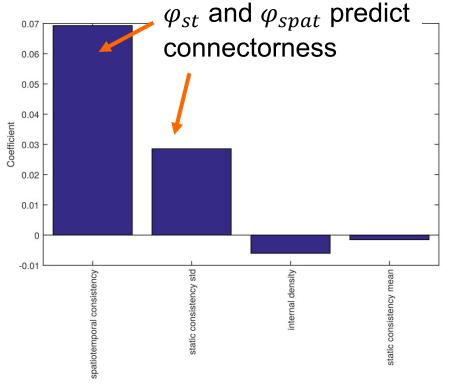


Results: Connector vs non-connector

Accuracy: Training 57.82% Test 55.97% (> Random 50.38%)



Accuracy: Training 52.80% Test 52.17% (> Random 50.38%)



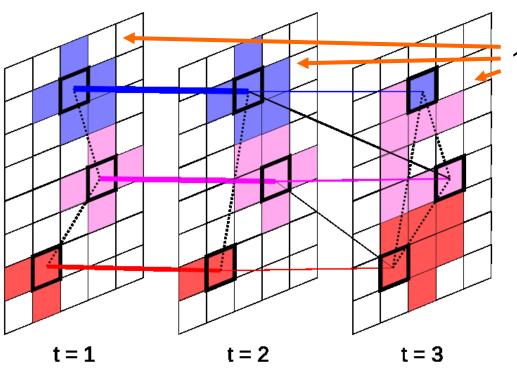
Conclusions

- Internal connectivity measures predict topological roles
 ⇒Varying homogeneity not a technical flaw!
- ROI time series = average of voxel time series
 - \Rightarrow Low homogeneity = lost data
 - ⇒ Flexible nodes needed

On-going work: Multilayer brain networks with flexible nodes

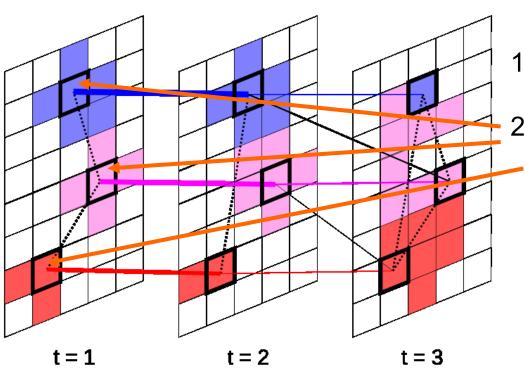
with Tarmo Nurmi, Maria Hakonen, Iiro Jääskeläinen & Mikko Kivelä

 Based on multilayer networks (= different connections in the same network)



1. Layers = time windows

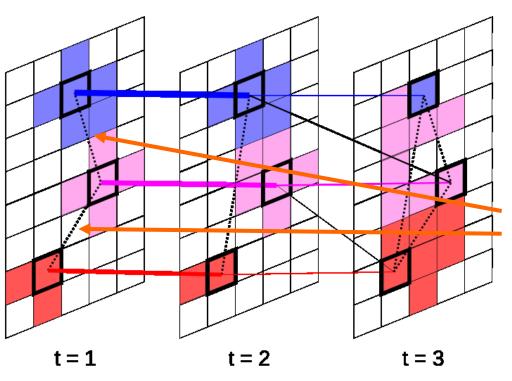
 Based on multilayer networks (= different connections in the same network)



1. Layers = time windows

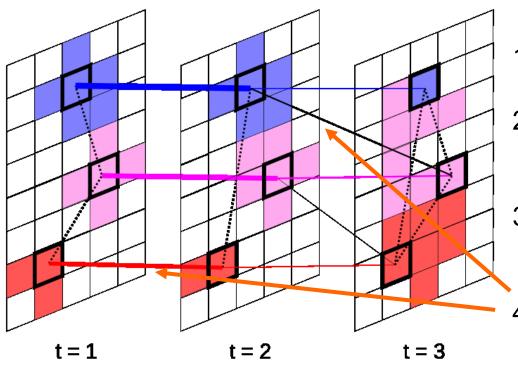
ROIs optimized inside layers for maximal homogeneity

 Based on multilayer networks (= different connections in the same network)



- 1. Layers = time windows
- ROIs optimized inside layers for maximal homogeneity
- 3. Interlayer links = Pearson correlation

 Based on multilayer networks (= different connections in the same network)



- 1. Layers = time windows
- ROIs optimized inside layers for maximal homogeneity
- Interlayer links = Pearson correlation
 - Intralayer links = spatial overlap

General conclusions

- It's not trivial to construct a functional brain network
 - Know your methods!
- Currently used nodes are not functionally homogeneous
 - Data lost in averaging
 - Risk of spurious connectivity?
- Homogeneity changes in time
 - Changes relate to function?
- Low homogeneity isn't a technical flaw
 - ⇒ Can't be fixed by new static nodes
 - ⇒ Flexible nodes needed!

References

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- **Bassett, D. S. & Sporns, O.** 2017. Network Neuroscience. *Nature Neuroscience* 20(3), 353–364.
- Klimm, F., Borge-Holthoefer, J., Wessel, N., Kurths, J., & Zamora-López, G. 2014. Individual node's contribution to the mesoscale of complex networks. *New Journal of Physics* 16: 125006.
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- **Muldoon, S. F. & Bassett, D. S.** 2016. Network and multilayer network approaches to understanding human brain dynamics. *Philosophy of Science* 83(5), 710–720.
- Nummenmaa, L., Saarimäki, H., Glerean, E., Gostopoulos, A., Jääskeläinen, I. P., Hari, R., & Sams, M. 2014. Emotional speech synchronizes brains across listeners and engages large-scale dynamic brain networks. *NeuroImage* 102, 498–509.

