# How <u>not</u> to construct functional brain networks

Onerva Korhonen
NIMEG Group Meeting
5.10.2022





**Network:** a model of connections & interactions

Internet, public transport, social networks

Tomás Saraceno: Algo-R(h)i(y)thm

#### Network: a model of connections & interactions

Internet, public transport, social networks



#### Nodes: network's basic elements

Web pages, stops, people

Tomás Saraceno: Algo-R(h)i(y)thm

#### Network: a model of connections & interactions

Internet, public transport, social networks



#### Nodes: network's basic elements

Web pages, stops, people

#### Links: connections between nodes

Web links, transport lines, social relationships

Tomás Saraceno: Algo-R(h)i(y)thm

#### Network: a model of connections & interactions

Internet, public transport, social networks



#### Nodes: network's basic elements

Web pages, stops, people

#### Links: connections between nodes

- Web links, transport lines, social relationships
- Weights?

Tomás Saraceno: Algo-R(h)i(y)thm

#### Network: a model of connections & interactions

Internet, public transport, social networks



#### Nodes: network's basic elements

Web pages, stops, people

#### Links: connections between nodes

- Web links, transport lines, social relationships
- Weights?
- Direction?

Tomás Saraceno: Algo-R(h)i(y)thm

## Why is the brain a network?

Brain: 10<sup>11</sup> neurons, 10<sup>14</sup> synapses

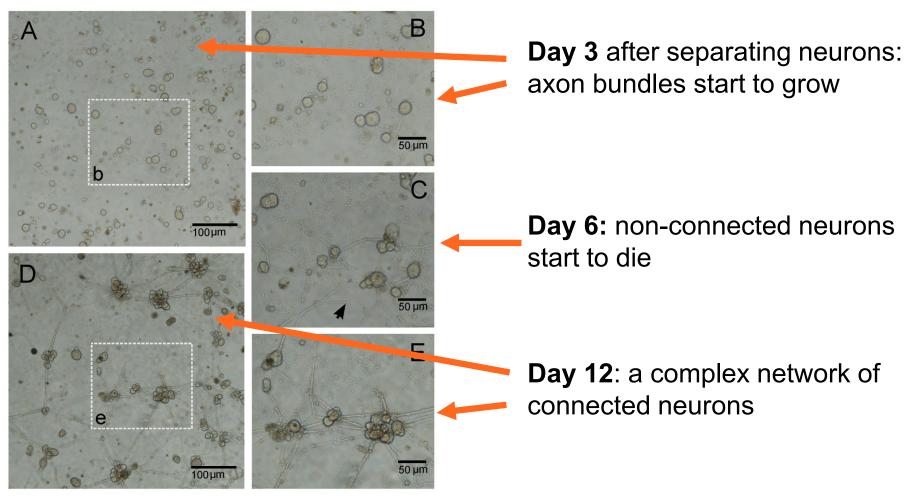
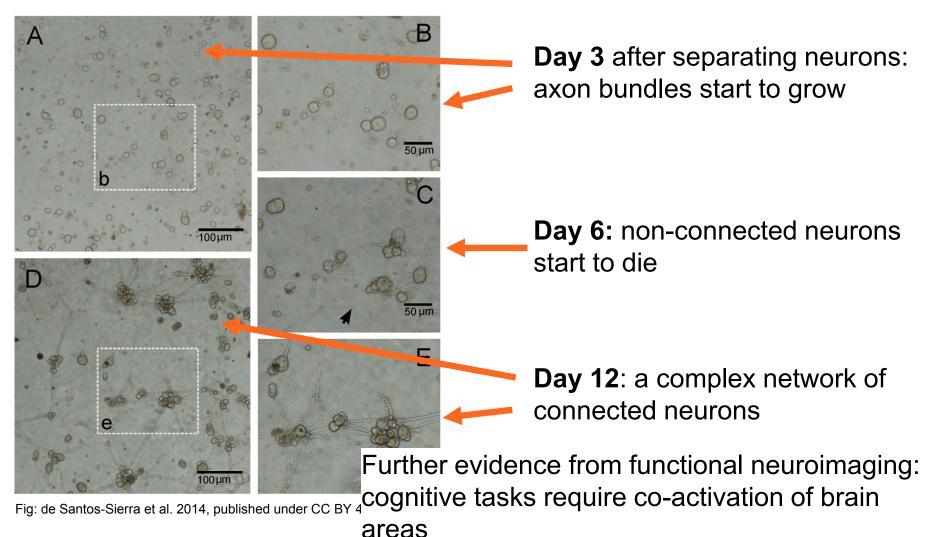


Fig: de Santos-Sierra et al. 2014, published under CC BY 4.0

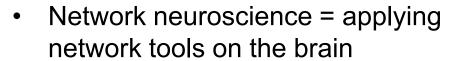
## Why is the brain a network?

Brain: 10<sup>11</sup> neurons, 10<sup>14</sup> synapses



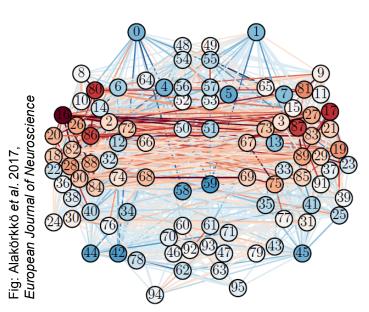
## **Network neuroscience**

(Bassett & Muldoon 2016, Bassett & Sporns 2017)

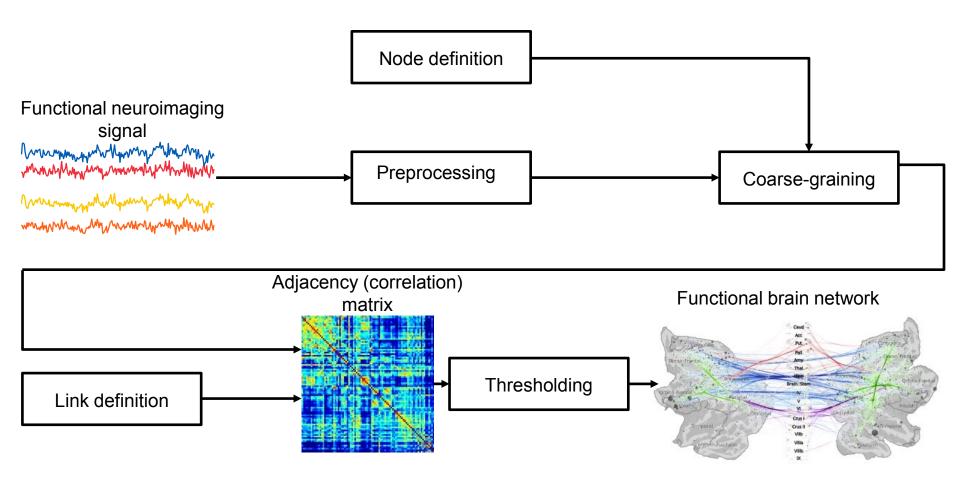




- 1. Understand the healthy brain
- 2. Find causes of diseases
- Broad scales:
  - Molecule neuron brain area human
  - Milliseconds years
- Different brain networks:
  - Structural: anatomic connections
  - Functional: temporal coactivation
  - Effective: causality

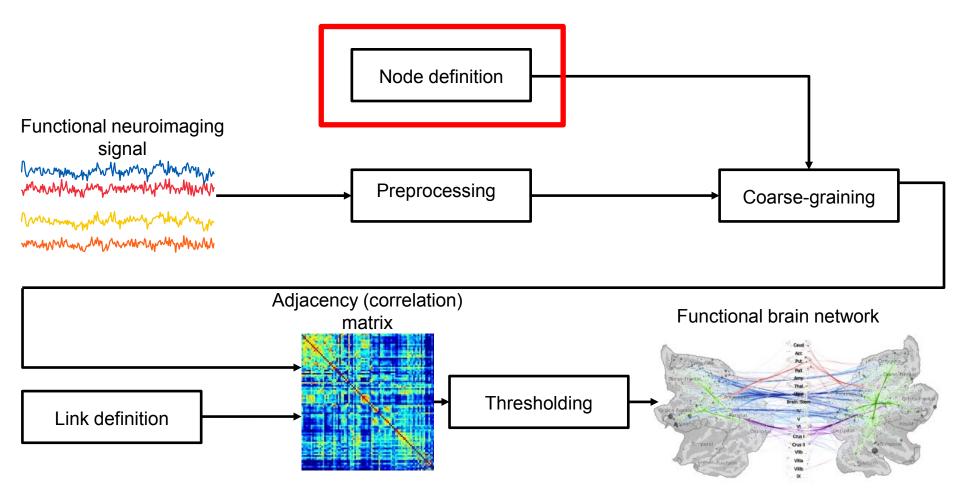


### Functional brain networks: how-to?



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

## Functional brain networks: how-to?



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

## The problem of node definition

#### No natural candidates above the scale of neurons

- => huge variation in node definition
- Number of nodes: from < 100 to 10<sup>5</sup>

**Node definition affects network properties** (e.g. Wang et al. 2009)

**Common strategies** (for a review, Korhonen et al. 2021, section 3.2):

- voxels/vertices
- random clumps of voxels/vertices
- Regions of Interest (ROIs): collections of voxels/vertices

## Voxels vs ROIs

#### Voxels:

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

#### More on this:

- Korhonen et al. 2017
- Ryyppö et al. 2018

#### ROIs:

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

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

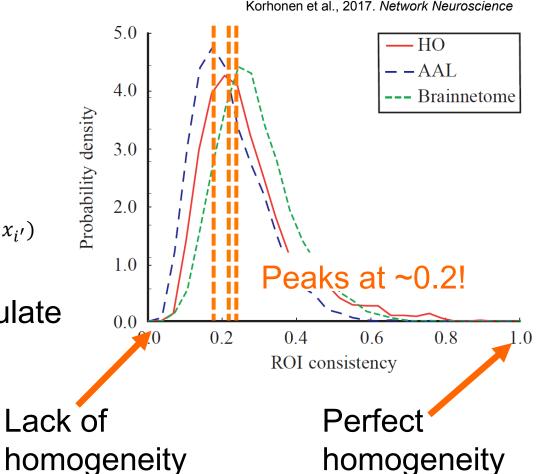
Violent?

## How homogeneous are ROIs?

- Spatial consistency
- = measure of functional homogeneity:

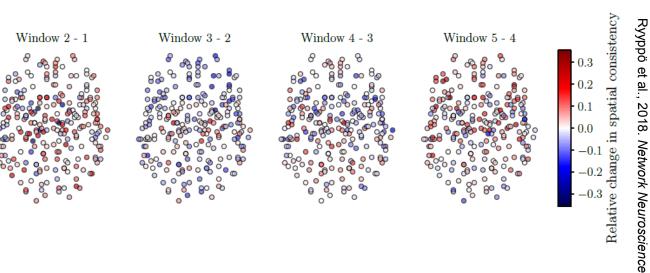
$$\varphi_{spat}(I) = \frac{1}{N_I(N_I - 1)} \sum_{i,i' \in I} C(x_i, x_{i'})$$

- Straightforward to calculate
- Easy to interpret

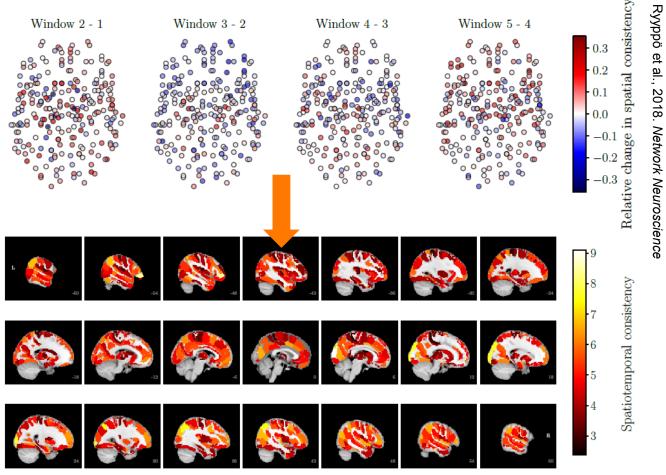


Correlates with ROI size & connectivity

## Spatial consistency changes in time



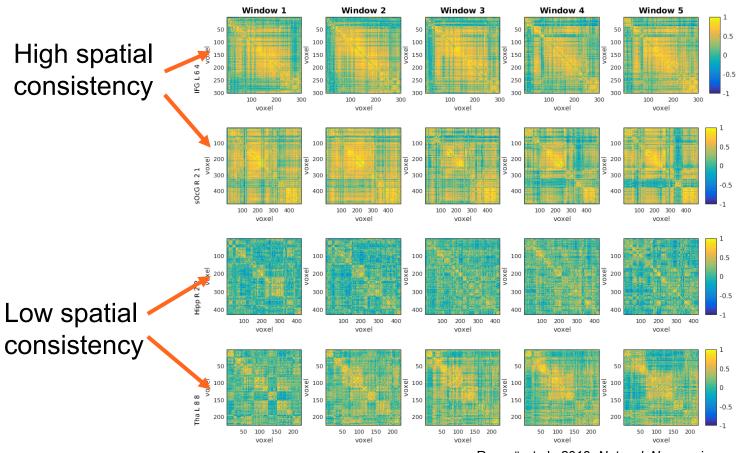
## Spatial consistency changes in time



Spatiotemporal consistency

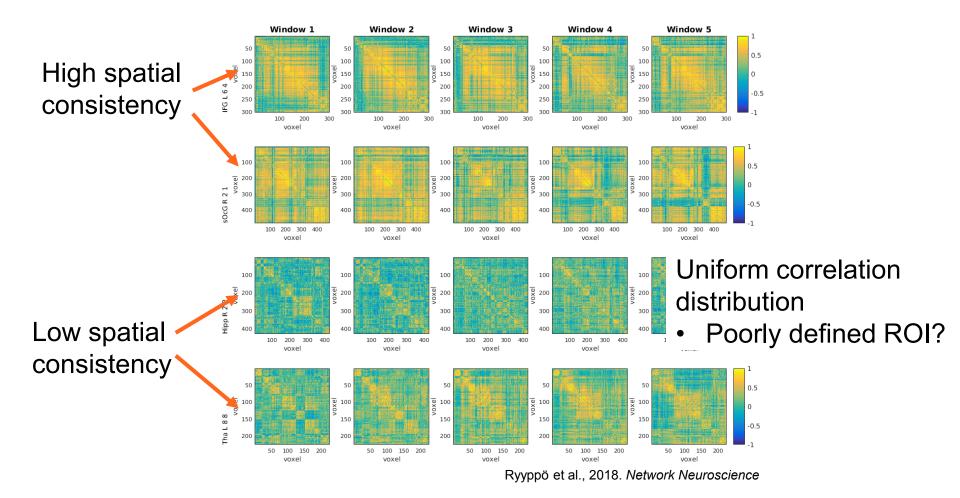
= stability of spat. consistency

# ROIs have rich internal connectivity structure

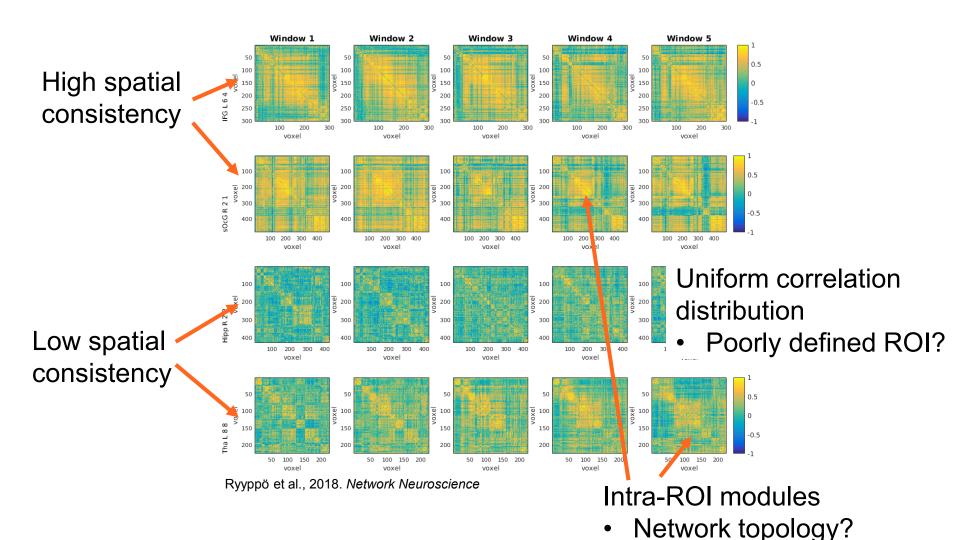


Ryyppö et al., 2018. Network Neuroscience

# ROIs have rich internal connectivity structure



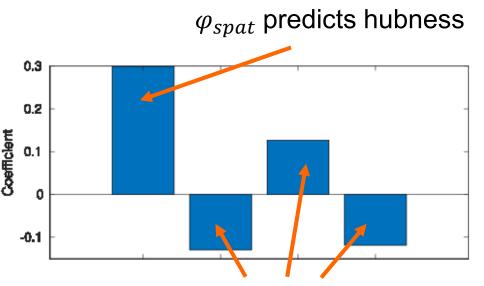
# ROIs have rich internal connectivity structure



## **Consistency predicts topology**

#### **Hub vs non-hub:**

Accuracy:
Training 60.39%
Test 60.23%
(> Random 50.03%)

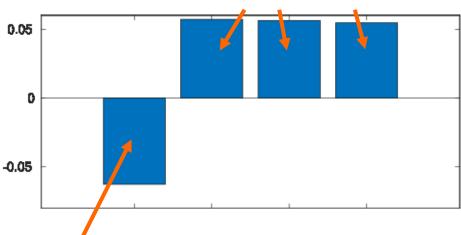


Hubs have lower internal density, high  $\varphi_{st}$ , and uniform in-ROI correlations

#### Provincial vs connector hub

Accuracy: Training 53.23% Test 52.57% (> Random 50.38%)

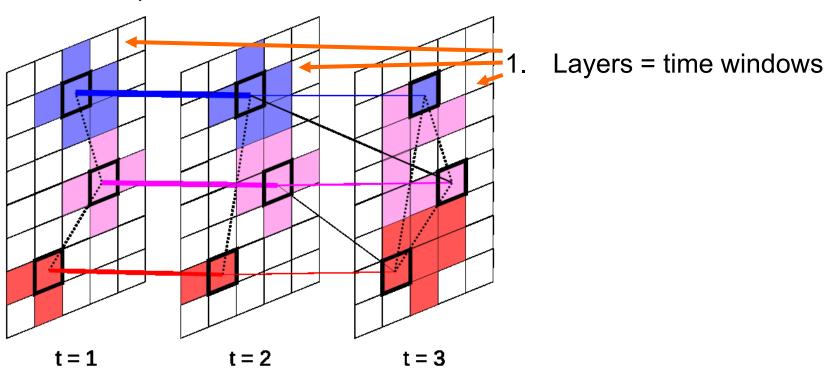
 $\varphi_{spat}$ , varying correlations, and  $\varphi_{st}$  predict provincial role



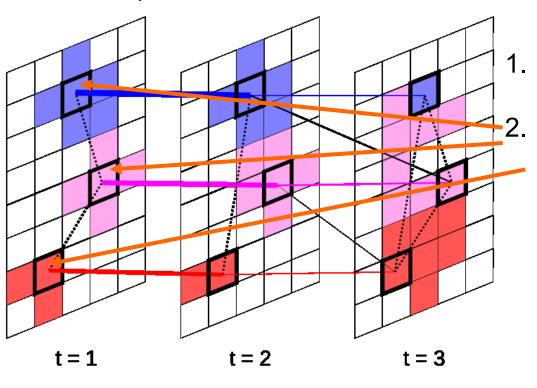
Low internal density = connector hub

(On-going, with T. Nurmi, M. Hakonen, I. Jääskeläinen & M. Kivelä)

Based on multilayer networks (= different connections in the same network), for review: Kivelä et al. 2014



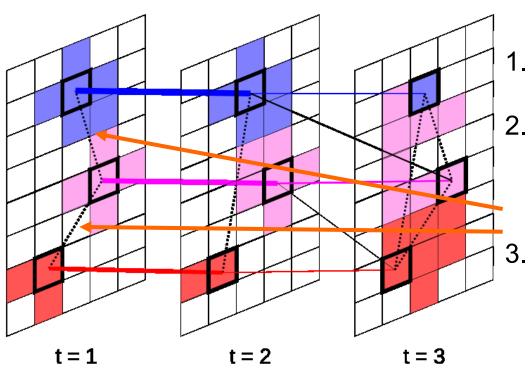
Based on multilayer networks (= different connections in the same network), for review: Kivelä et al. 2014



Layers = time windows

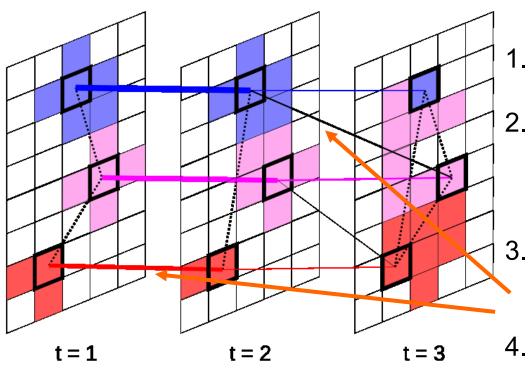
ROIs optimized inside layers for maximal homogeneity (voxel-level clustering)

Based on multilayer networks (= different connections in the same network), for review: Kivelä et al. 2014



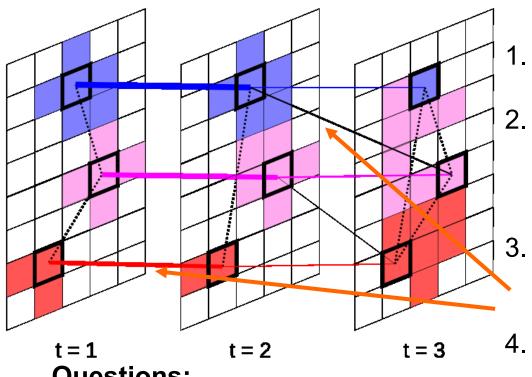
- 1. Layers = time windows
  - ROIs optimized inside layers for maximal homogeneity (voxel-level clustering)
  - Intralayer links = Pearson correlation

Based on multilayer networks (= different connections in the same network), for review: Kivelä et al. 2014



- 1. Layers = time windows
- 2. ROIs optimized inside layers for maximal homogeneity (voxel-level clustering)
  - Intralayer links = Pearson correlation
- 4. Interlayer links = spatial overlap

Based on multilayer networks (= different connections in the same network), for review: Kivelä et al. 2014



- Layers = time windows
  - ROIs optimized inside layers for maximal homogeneity (voxel-level clustering)
    - Intralayer links = Pearson correlation
- Interlayer links = spatial overlap

#### **Questions:**

- How do ROIs change over time? Splitting, merging, disappearing?
- State changes?
- What about Alzheimer's disease?

## **Conclusions**

- It's not trivial to construct a functional brain network
  - Node definition?
  - Not covered today: preprocessing, space, thresholding, link definition, multilayers in multiple-person neuroscience?
  - Know your methods!
- Currently used nodes functionally inhomogeneous
  - Data lost in averaging
  - Can we trust observed 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

- **Alakörkkö, T., Saarimäki, H., Glerean, E., Saramäki, J., & Korhonen, O.** 2017. Effects of spatial smoothing on functional brain networks. *European Journal of Neuroscience* 46(9).
- Bassett, D. S. & Sporns, O. 2017. Network Neuroscience. *Nature Neuroscience* 20(3).
- de Santos-Sierra, D., Sendiña-Nadal, I., Leyva, I., Almendral, J. A., Anava, S., Ayali, A., Papo, D., &
- **Boccaletti, S.** 2014. Emergence of small-world anatomical networks in self-organizing clustered neuronal cultures. *PLoS One* 9(1): e85828.
- Kivelä, M., Arenas, A., Barthelemy, M., Gleeson, J. P., Moreno, Y., & Porter, M. A. 2014. Multilayer networks. *Journal of Complex Networks* 2(3).
- **Korhonen, O., Saarimäki, H., Glerean, E., Sams, M., & Saramäki, J.** 2017. Consistency of Regions of Interest as nodes of fMRI functional brain networks. *Network Neuroscience* 1(3).
- **Korhonen, O., Zanin, M., Papo, D.,** 2021. Principles and open questions in functional brain network reconstruction. *Human Brain Mapping* 42(11).
- **Muldoon, S. F. & Bassett, D. S.** 2016. Network and multilayer network approaches to understanding human brain dynamics. *Philosophy of Science* 82(5).
- 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.
- Ryyppö, E., Glerean, E., Brattico, E., Saramäki, J., & Korhonen, O. 2018. Regions of Interest as nodes of dynamic functional brain networks. *Network Neuroscience* 2(4)
- Wang, J., Wang, L., Zang, Y., Yang, H., Tang, H., Gong, Q., ... He, Y. 2009. Parcellation-dependent small-world brain functional networks: A resting-state fMRI study. *Human Brain Mapping* 30(5).



korhonen/presentations/blob/master/nimeg\_051022.pdf