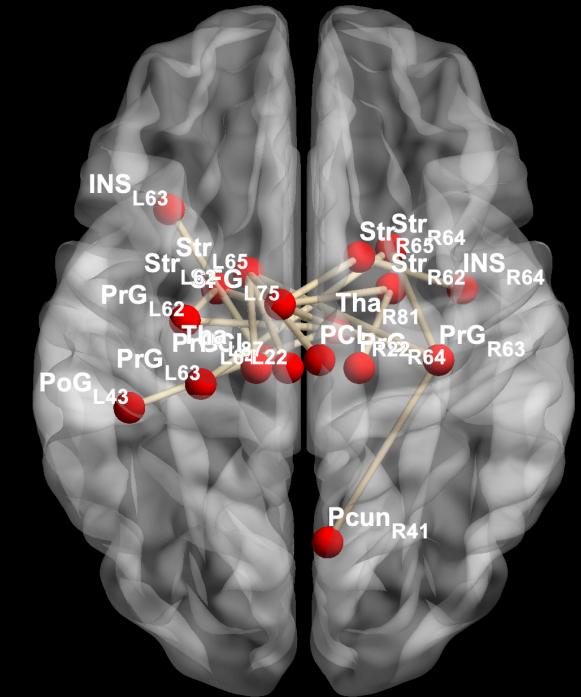
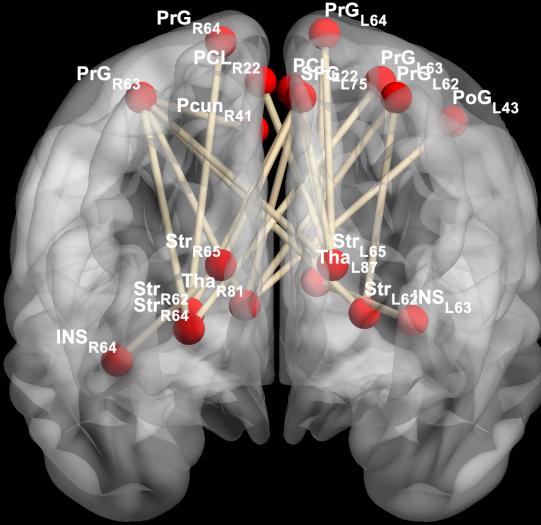


Brain networks.



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CS-E5700
Hands-on Network Analysis

A''

Aalto University

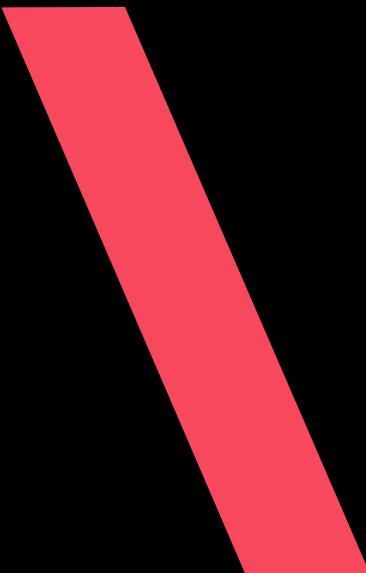
Background.

Our group has a background in bioinformation tech., neuroscience, and medicine.



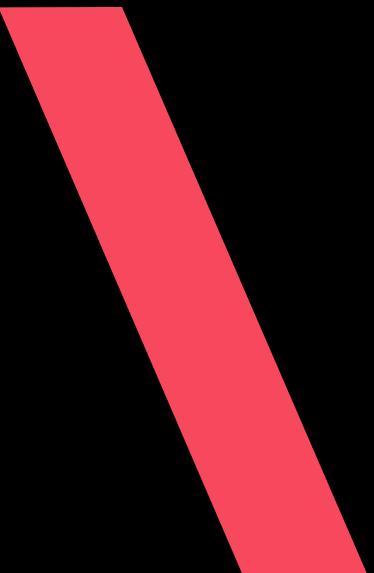
- We want to learn how to model brain data/activity as a network from real-world data.
- How to calculate different network metrics and apply statistical tests?

**Brain network,
connectome,
contains neural
connections in the
brain.**



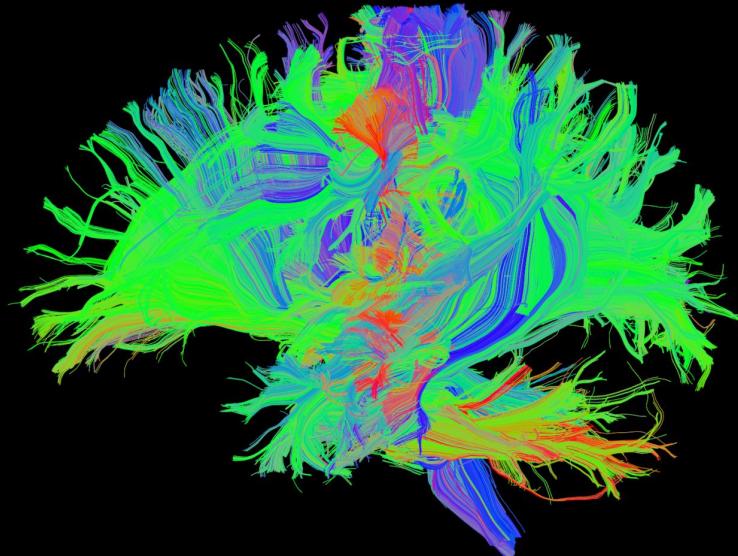
- Scale can vary from the synaptic connections to inter-regional pathways.
- Functional connectome can be mapped with neuroimaging methods (fMRI, MEG etc.).
- Connectivity matrix: brain regions as nodes, time-series correlation as edge weights.

What is the meaning of differences in connectomes?



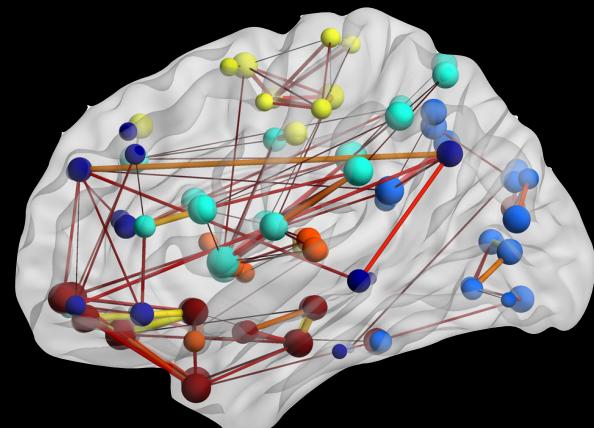
- Differences are associated with experimental effects, medical conditions, and between-group differences.
- E.g. faulty wiring in psychiatric disorders or enchanted connectivity in some talented subpopulation.
- Connectome can be used as a tool for diagnostics.

Fig 1: Human Brain Connectome



From <https://perso.telecom-paristech.fr/comercier/>

Fig 2: Functional brain network



From BrainNet Viewer. https://www.nitrc.org/project/list_screenshots.php?group_id=504&screenshot_id=381

- **Structural networks are based on physical connections (neurons) measured with e.g. diffusion MRI.**
- **Links of the functional networks are not necessarily physical links (i.e. no neuron); they are mathematical models of the data.**

Autism spectrum disorder (ASD) is a developmental disorder.

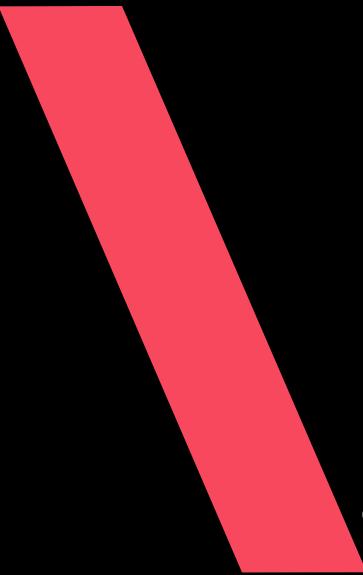


- Neuroanatomically, pathological findings of autism are unspecific.
- ASD isn't a homogeneous disorder, and it has various causes and a vast range of functional symptoms.
- We cannot generalize our results to the whole ASD.

Our data was a
selection from public
Autism Brain Imaging
Data Exchange
(ABIDE) data set.

- We had resting-state fMRI from 47 ASD male patients with age, sex-matched neurotypical (NT) controls.
- One subject pair was excluded from the analysis.

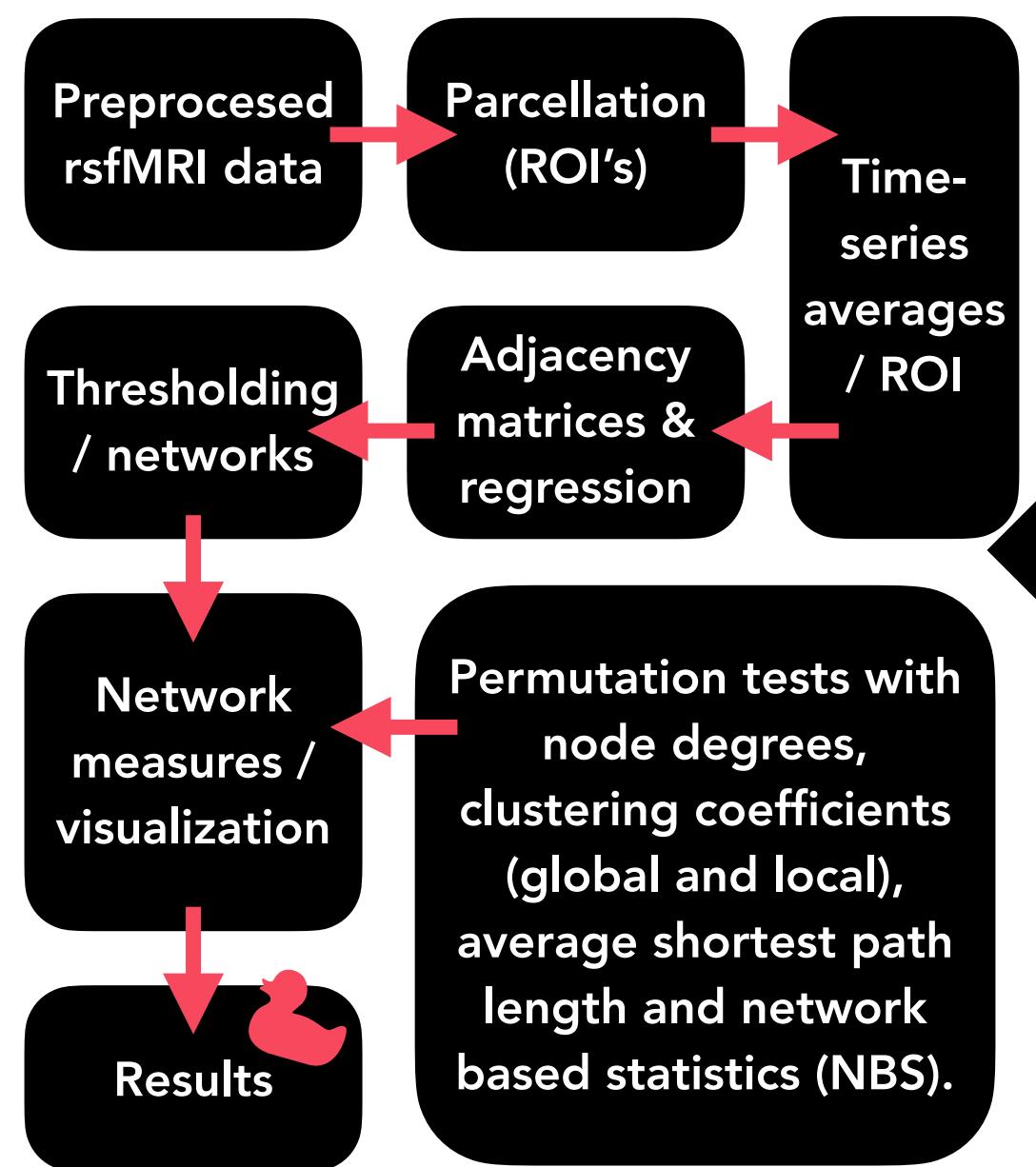
Our goal was to
find hyper- and
hypoconnected
links at ASD
patients.



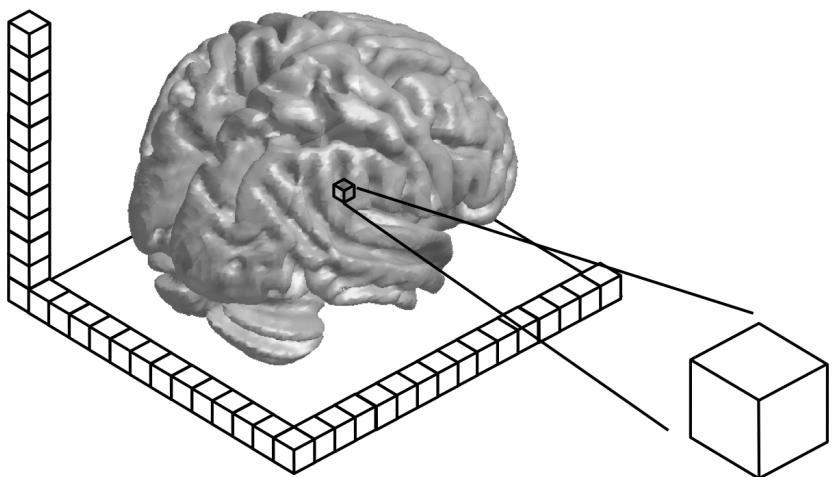
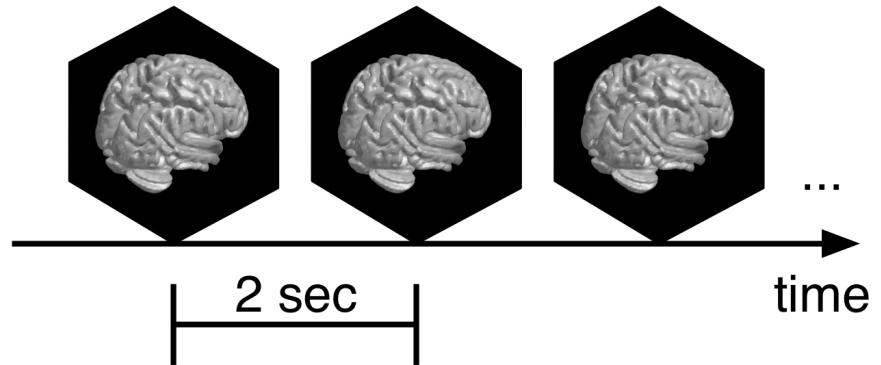
- Previous studies with the same data have found thalamocortical hyperconnectivity and amygdala-cortical.
- Can we reproduce the results?

Networks.

Flowchart of the project.



- We got already preprocessed resting-state fMRI data from the ABIDE data set.
- Preprocessing included the following steps:
 - Slice-timing and motion correction.
 - Nuisance signal regression and temporal filtering (0.009–0.1 Hz)
 - Normalization to MNI152 stereotactic space (2 mm³ isotropic). It was done with linear and non-linear registrations.

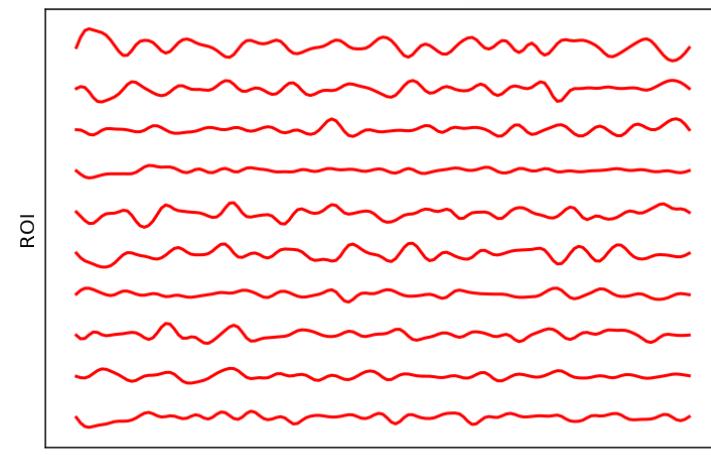
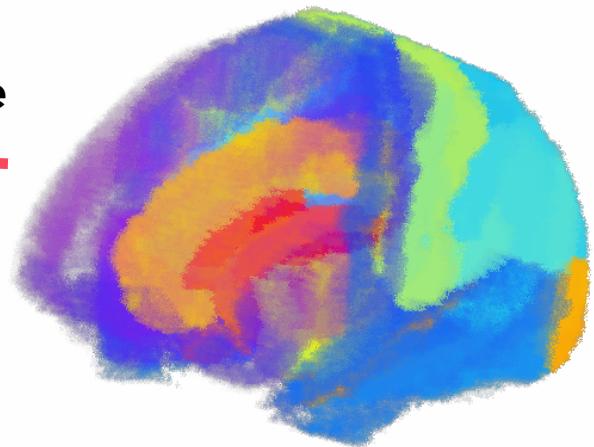


Single voxel Blood Oxygen Level (BOLD) signal

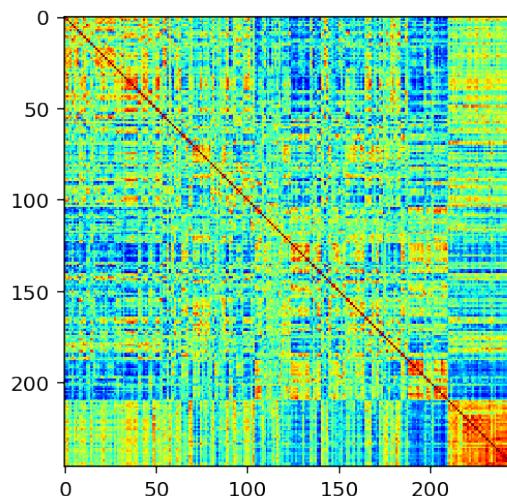
- Multiple ***time series*** are measured from different voxels at once.
- ***Time series*** can be considered independently or try to investigate mutual relationships.

Fig: Enrico Glerean (2017). NBE-E4530
Special Course in Human Neuroscience:
Human brain connectivity

From parcellation we
got ROI time
series

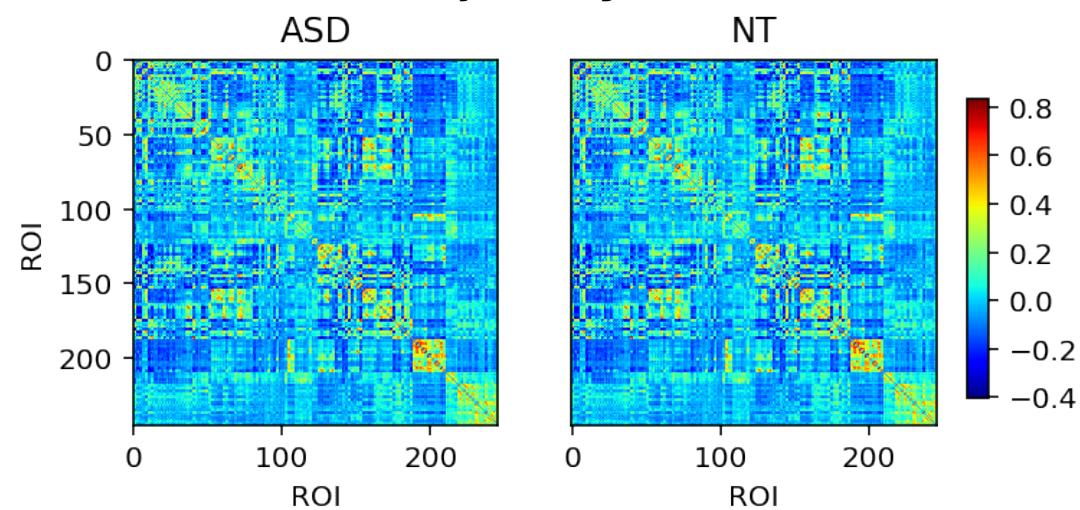


Time-series
correlations
to adjacency matrix

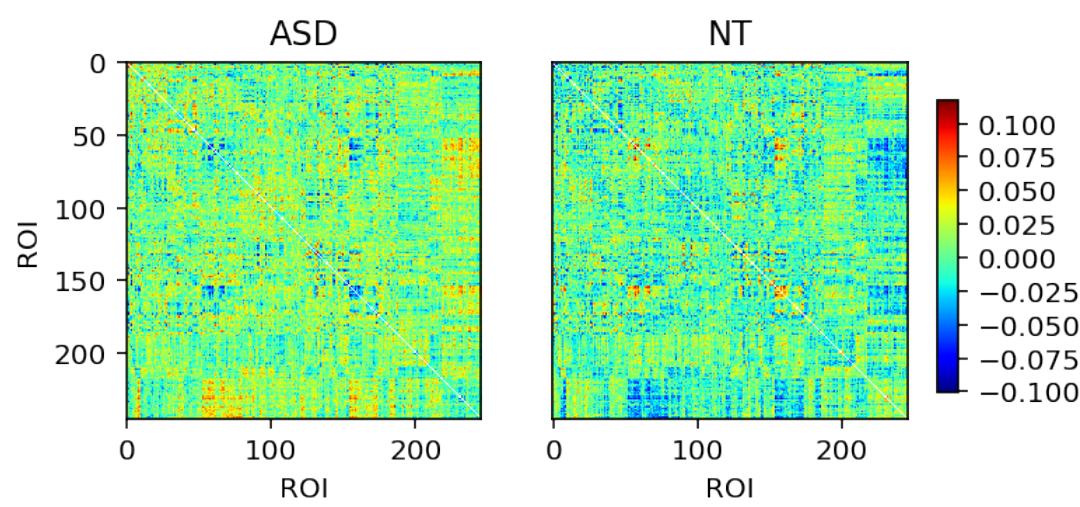


- **Parcellation:** from 117,457 grey matter voxels to 246 ROI's.
- Correlations between **time series (ROI)** were computed.
- **Adjacency matrices** were created from the **correlation coefficients (link weights)** of the time series.
- Adjacency matrices hold the information of link weights between ROI's.

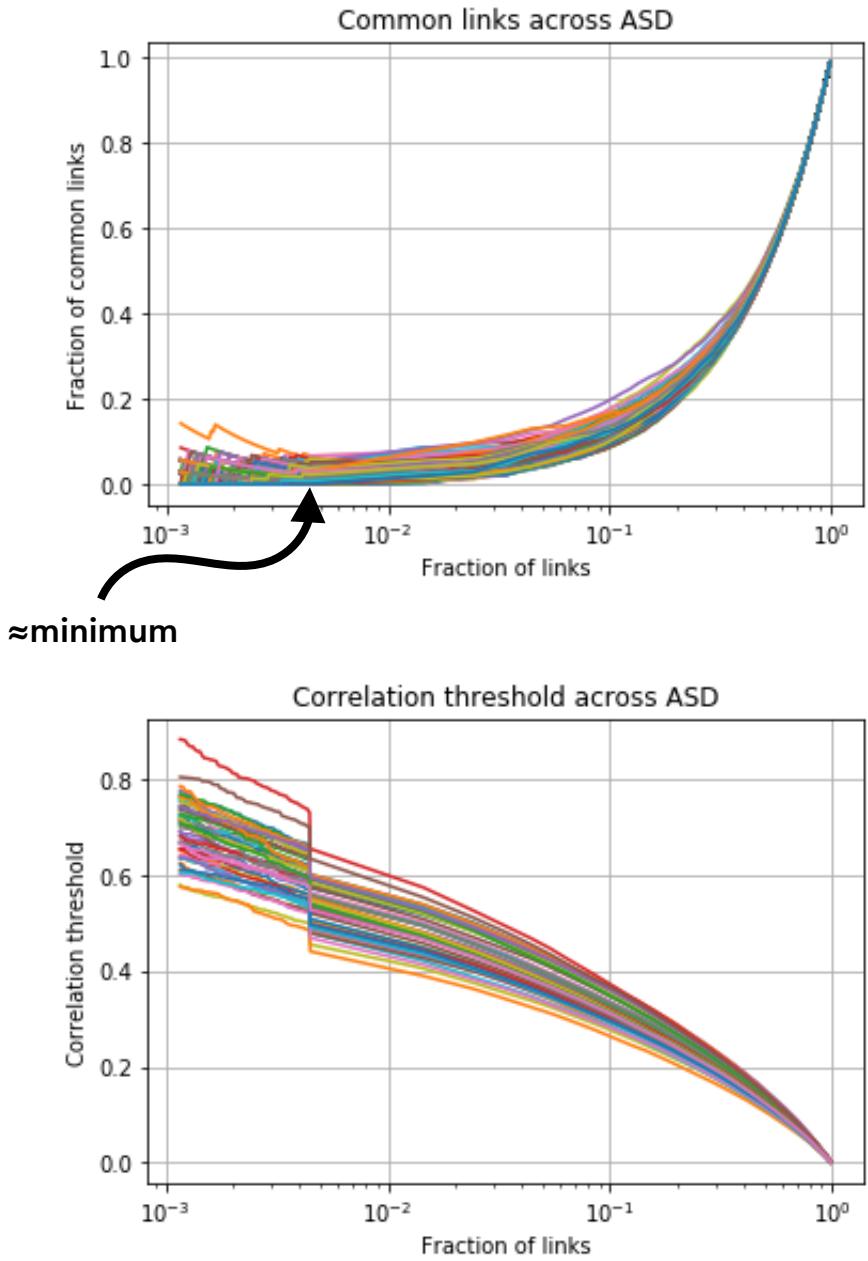
Mean adjacency matrices



Regressed mean adjacency matrices



- Regression was used to tackle issues that arise from different imaging devices used in the data collection and movement of subjects.
- The effect of regression can be seen from the mean adjacency matrices.



- Thresholding lets us focus only on the interesting part of the data and avoid heavy computation.
- There is not a single method to do thresholding.
- We wanted to optimize the threshold value to give the biggest difference between the two groups.
- 100 strongest links from each subject were chosen (\approx minimum overlap).

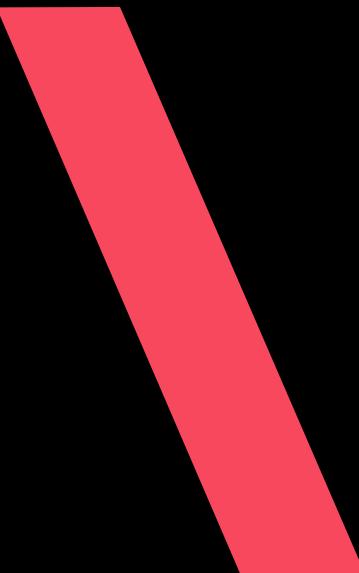
* Similar figures done for neurotypicals

Should the network be fully connected or not?



- The selection of only the 100 strongest links led to a loss of subcortical information.
- Fully connected network, MinST + 100 strongest provides information about subcortical – cortical connections.

Using network properties as a diagnostic marker.



- **Degree distribution (node-wise):**
 - Compatible with the existence of hubs.
- **Clustering coefficient (node-wise):**
 - Measures how many neighbors of a node are connected to each other. Closely related to local efficiency.
- **Shortest path length (global-level):**
 - The average number of edges needed to get from any node in the network to any other node in the network. Inversely related to global efficiency.
- **Network-based statistic (NBS) (edge-wise):**
 - Identify significantly different links that form a connected structure instead of individual links.

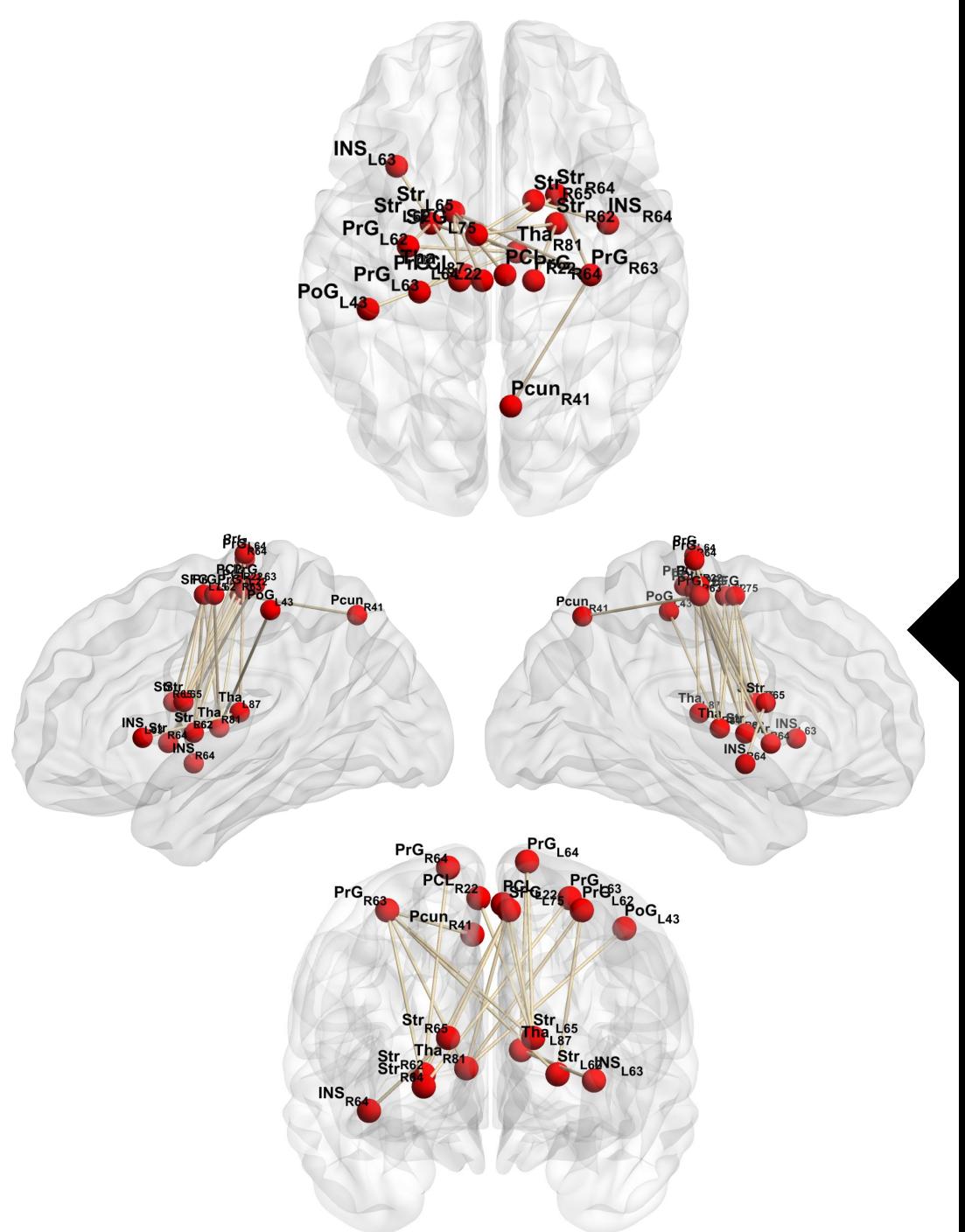
Results.

Permutation test results from 100 strongest + MinST networks.

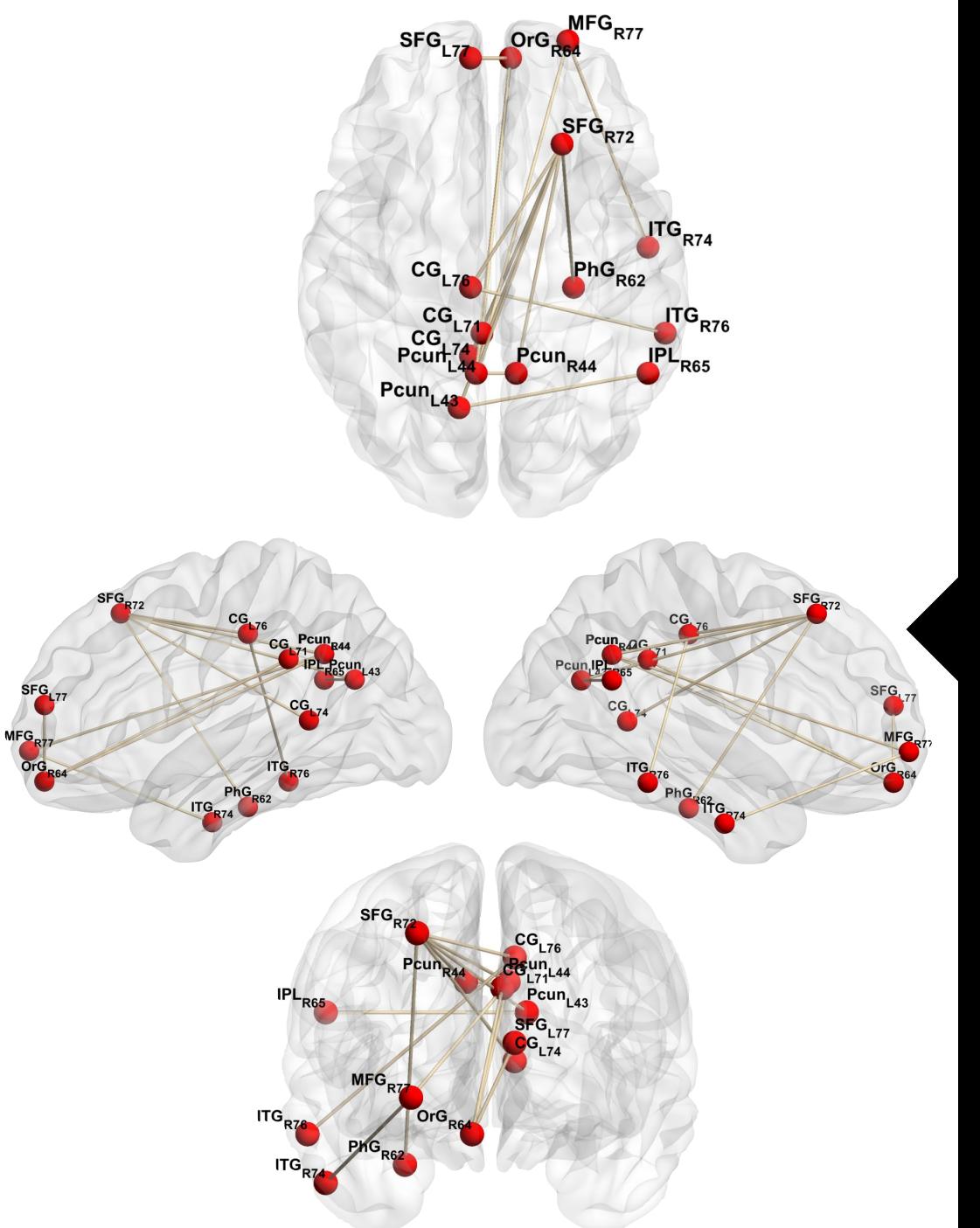
	ASD > NT i.e. hyperconnected	ASD < NT hypoconnected
Degree *	Right superior middle frontal gyrus, left post central gyrus, right striatum	Right inferior temporal gyrus, left amygdala, left striatum
Clustering coefficient (Global) *	No significant difference between groups	No significant difference between groups
Clustering coefficient (Local) *	Precentral gyri (L+R), left paracentral lobule	Left superior and middle frontal gyri, inferior temporal gyri (L+R), right inferior parietal lobe, left precuneus, left cingulate gyrus
Average shortest path length *	No significant difference between groups	No significant difference between groups

* 20 000 permutations, FDR correction (BH) and alpha 0.05.

Table doesn't include all significant nodes. Full list is available at the end of presentation



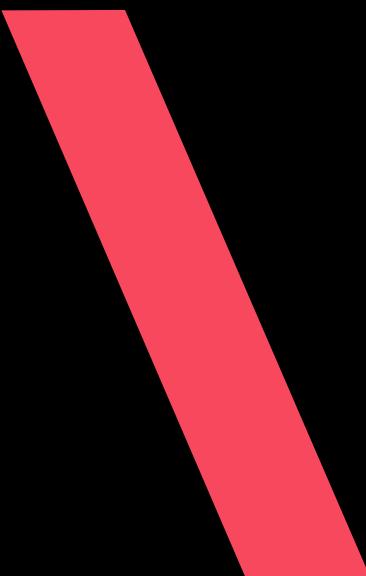
- Network based statistics (NBS): 5000 iterations, threshold = 3.5.
- NBS uses unthresholded adjacency matrices.
- The figure on the left represents ASD>NT, i.e., the hyperconnectivity.
- The test shows thalamocortical hyperconnectivity, similar to the previous study.



- Similar NBS test for ASD<NT, i.e., the hypoconnectivity.
- We see unilateral hypoconnectivity.
- Cingulate gyrus and precuneus nodes are highlighted in the test (edge to frontal lobe).

Conclusion.

We found
thalamocortical
hyperconnectivity
and unilateral
hypoconnectivity.



- Hypoconnectivity result differs from previous studies with the same data set.
 - It could be due to different parcellation and regression.

We learned how
network analysis
is done with real-
life fMRI data.



- ASDs have network differences compared to NTs.
- How permutation tests can be used to assess the statistical significance of network metrics and how brain networks can be visualized.



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A!

Aalto University

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Kiashemshaki Maryam & Tommila Timo**

Hands-on Network Analysis

Appendix.

Appendix 1. Degree permutation results.

	ASD > NT i.e. hyperconnected	ASD < NT hypoconnected
Degree. Node and p-values	SFG_R_7_4 0.02925 MFG_R_7_4 0.01435 OrG_L_6_2 0.0166 MTG_R_4_2 0.00935 IPL_L_6_4 0.04515 PoG_L_4_2 0.03375 INS_R_6_3 0.00485 INS_R_6_6 0.0374 Cun_L_5_3 0.01685 Str_R_6_2 0.00625	MFG_R_7_2 0.02605 MFG_R_7_6 0.01025 STG_L_6_4 0.02505 PhG_L_6_4 0.01345 Pcun_L_4_2 0.008 INS_L_6_3 0.0162 INS_R_6_4 0.0253 CG_L_7_2 0.01725 OcG_L_4_2 0.00825 Str_R_6_6 0.01755

Appendix 2. Clustering coefficient permutation results.

	ASD > NT i.e. hyperconnected	ASD < NT hypoconnected
Clustering coefficient (Local). Node and p-values	SFG_L_7_3 0.0033 IFG_L_6_2 0.0498 PrG_L_6_1 0.01885 PrG_R_6_5 0.03215 PCL_L_2_1 0.0259 Tha_R_8_7 0.0 Tha_L_8_8 0.0 Tha_R_8_8 0.0	SFG_L_7_5 0.0291 OrG_R_6_1 0.04025 MTG_R_4_1 0.049 ITG_L_7_4 0.0344 PhG_L_6_4 0.03335 IPL_R_6_1 0.0174 PoG_L_4_3 0.03095 PoG_R_4_4 0.0446 CG_L_7_3 0.02795 CG_L_7_6 0.00815 Tha_R_8_4 0.0 Tha_R_8_7 0.0 Tha_R_8_8 0.0

Appendix 3. NBS permutation results.

	ASD > NT i.e. hyperconnected	ASD < NT hypoconnected
NBS links and test statistics.	PrG_R_6_3 to Pcun_R_4_1. Test stat: 4.07 PrG_L_6_2 to Str_L_6_2. Test stat: 3.59 PrG_R_6_3 to Str_L_6_2. Test stat: 3.52 SFG_L_7_5 to Str_R_6_2. Test stat: 3.66 PrG_R_6_4 to Str_R_6_2. Test stat: 4.70 PrG_L_6_3 to Str_R_6_4. Test stat: 4.16 PrG_R_6_3 to Str_R_6_4. Test stat: 3.81 SFG_L_7_5 to Str_L_6_5. Test stat: 3.81 PrG_R_6_3 to Str_L_6_5. Test stat: 3.83 PrG_L_6_4 to Str_L_6_5. Test stat: 3.82 PCL_L_2_2 to Str_L_6_5. Test stat: 3.69 PCL_R_2_2 to Str_L_6_5. Test stat: 3.89 SFG_L_7_5 to Str_R_6_5. Test stat: 4.14 INS_R_6_4 to Str_R_6_5. Test stat: 3.72 SFG_L_7_5 to Tha_R_8_1. Test stat: 4.28 PrG_L_6_2 to Tha_R_8_1. Test stat: 3.87 PrG_R_6_3 to Tha_R_8_1. Test stat: 3.51 PoG_L_4_3 to Tha_R_8_1. Test stat: 3.76	SFG_L_7_7 to OrG_R_6_4. Test stat: 3.74 MFG_R_7_7 to ITG_R_7_4. Test stat: 3.58 SFG_R_7_2 to PhG_R_6_2. Test stat: 4.28 SFG_R_7_2 to Pcun_L_4_3. Test stat: 3.97 IPL_R_6_5 to Pcun_L_4_3. Test stat: 3.93 SFG_R_7_2 to Pcun_L_4_4. Test stat: 3.68 MFG_R_7_7 to Pcun_L_4_4. Test stat: 3.52 OrG_R_6_4 to Pcun_L_4_4. Test stat: 4.92 SFG_R_7_2 to Pcun_R_4_4. Test stat: 4.39 Pcun_L_4_4 to Pcun_R_4_4. Test stat: 3.52 OrG_R_6_4 to CG_L_7_1. Test stat: 3.54 SFG_R_7_2 to CG_L_7_4. Test stat: 3.53 SFG_R_7_2 to CG_L_7_6. Test stat: 3.60 ITG_R_7_6 to CG_L_7_6. Test stat: 3.83