Are All Brains Wired Equally?





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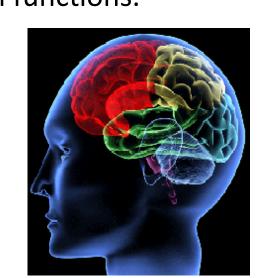
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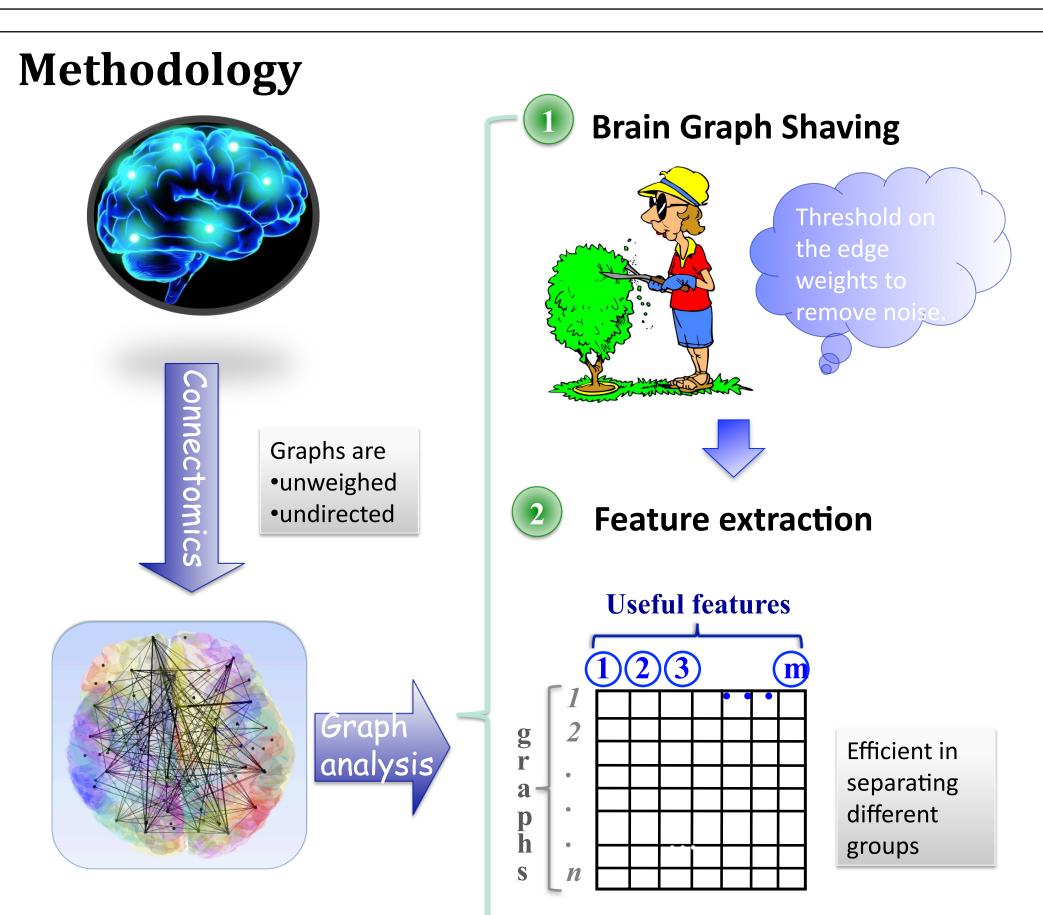
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Motivation

- •Connectomics -- creation of brain connectivity maps.
- •Analyzing the connectomes to understand how the brain functions.
 - Are there differences between different people?
 - Female vs. male
 - High math skills vs. normal
 - Artists vs. scientists...
 - Are all parts of the brain connected similarly?
 - Left hemisphere
 - Right hemisphere





Dataset

- Connectomes of 114 people
- Obtained by Multimodal Magnetic Resonance Imaging (MRI)
- Each connectome is represented as unweighted and undirected graph.
- Voxels/nodes: 492K-916K
- Connections/edges: 9.14M-17.42M
- Attributes per person: age, gender, IQ, creativity index, ...
- Toolkit
- PEGASUS
- Networkx



References

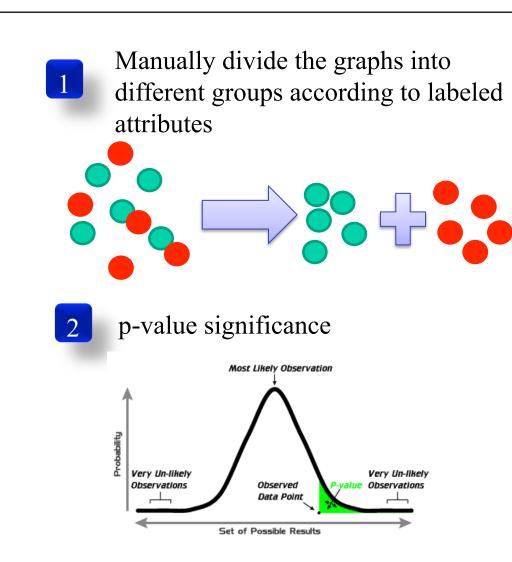
•Gray W, Bogovic JA, Vogelstein JT, Landman BA, PrinceJL, Vogelstein RJ, 'Magnetic resonance connectome automated pipeline:An overview', Pulse, IEEE, vol. 3, no. 2, pp. 4248, 2012. • Kang U, Tsourakakis C, Faloutsos C., 'PEGASUS: A Peta-Scale Graph Mining System -Implementation and Observations.', IEEE International Confer-ence on Data Mining (ICDM), Miami, Florida, USA, 2009.

Our Approach

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Step 1: Brain Difference

- Scalar Features Analysis
- 1 number of nodes
- 2 number of edges
- 3 largest eigenvalue
- 4 number of triangles
- (5) number of connected components
- 6 maximum pagerank
- 7 minimum pagerank



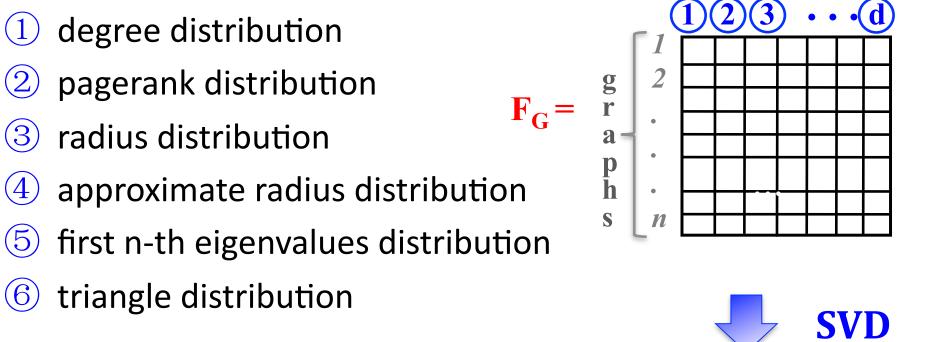
Plot the different invariants and test whether they help distinguish the groups (e.g., female/male).

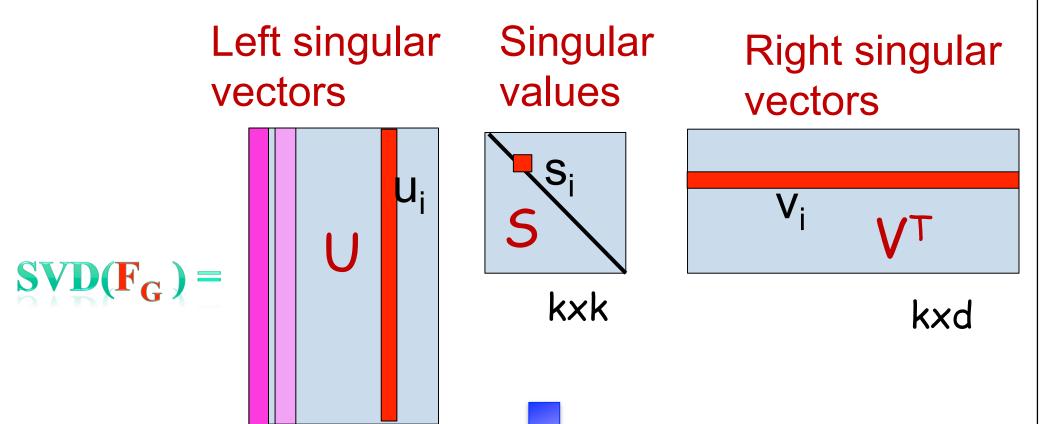
the single vector feature

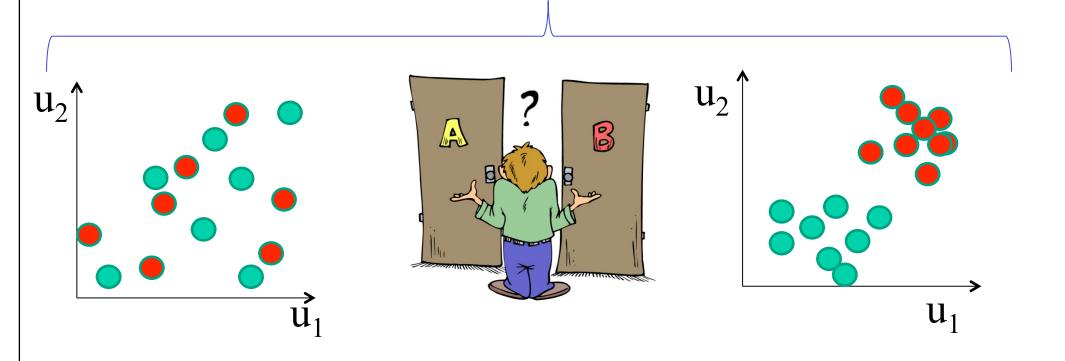
Plot out u₁ VS u₂

Vector Features

- 1 degree distribution
- 2 pagerank distribution
- 3 radius distribution
- 4 approximate radius distribution
- 6 triangle distribution







nxk

Possible conclusion 1 The feature cannot distinguish different groups

 u_1u_2

Possible conclusion 2 The feature contributes in predicting the group memberships.

Experimental Results

Preliminaries

- 114 connectomes in total - Groups divided in two ways:
- 1.Gender
 - 50 females: red dots and 64 males: green dots
- 2.Subject Type: ~ Level of math skills Normal: green dots Low: red dots

High: blue dots

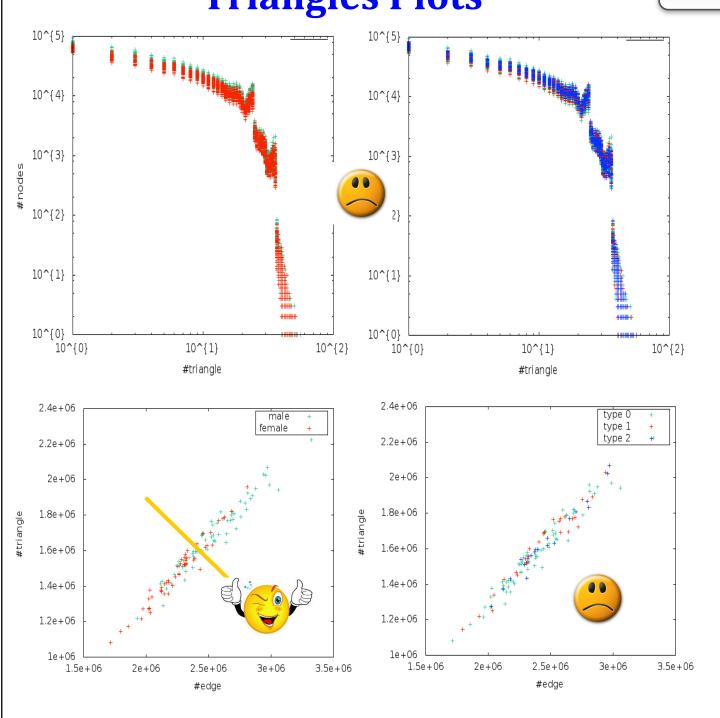
First vs. second principal component of the matrix brain scan X node-degree-distr.

Degree Distribution

Observation:

The extreme cases of female and male connectomes are well separated. However, the degree distribution cannot

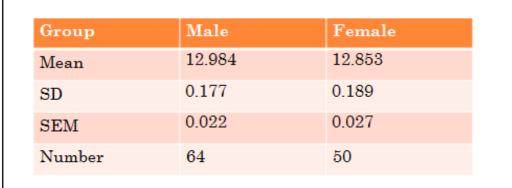
Triangles Plots



Observation:

Triangle distribution cannot separate the groups. Total # of triangles vs. # of edges can help separate females from males, but cannot distinguish the groups w.r.t. the level of math skills.

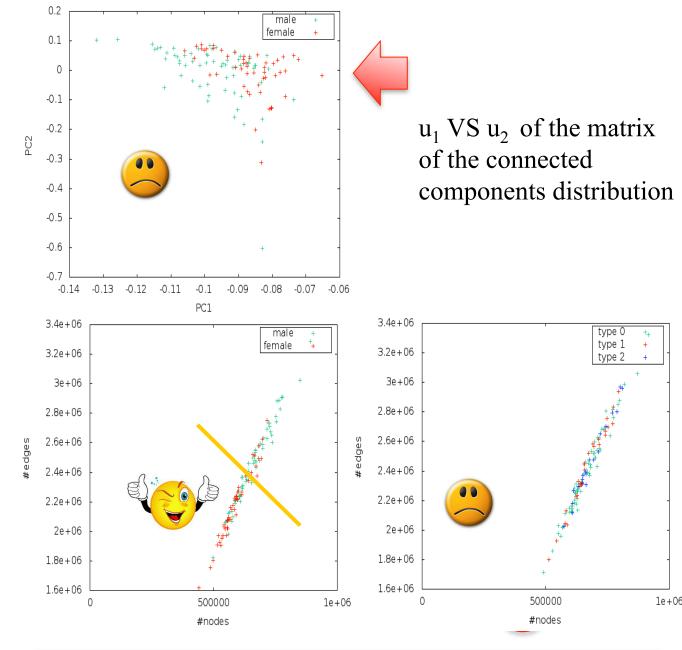
Largest Eigenvalue of Graph Matrix





separate connectomes w.r.t. the mastery of math.

Connected Components



Observations:

- 1) The distribution of connected components does not differ significantly between males and females.
- 2) The #of edges vs. nodes in the giant connected component of each connectome reveals two clusters corresponding to males and females.

The two genders differ significantly in this feature (2-sample t-test: p-value 0.0002).

CONCLUSIONS

- Novel approach: analyze invariants of numerous, big-scale brain graphs (connectomes) in order to do clustering and classification.
- •Obeservations:
- The size (number of edges), as well as the maximum eigenvalues of the brain graphs differ significantly between males and females.
- The degree distribution, and the number of triangles are features that can contribute towards the classification of the scans by gender.

