Connectivity-based Parcellation of Human Inferior Parietal Lobule using Diffusion MRI and Probabilistic Tractography

> Joe Xie May 26, 2011

## Outline

- Background
  - Diffusion MRI
  - Human inferior parietal lobule
- Materials & Methods
  - Data Collection
  - Connectivity Map Preparation via preprocessing
  - Unsupervised Classification Approaches (Spectral clustering)
- Results
  - Pseudo truth from Jülich Atlas
  - K means, Mixture Gaussian, and Spectral Clustering
  - Correspondence accuracy metric for parcellation evaluation

#### BACKGROUND



## **Inferior Parietal Lobule**

- Brain region with marked functional heterogeneity involved in visuospatial attention, memory, and mathematical cognition
- Availability of ECoG electrodes to verify and make testable predications in our study
- Consisted of seven cytoarchitectonic regions (PGp, PGa, PF, PFcm, PFm, PFt, Pfop)

#### Prior Knowledge of IPL Connectivity



Rostral IPL areas: targets in the prefrontal, motor, somatosensory, and anterior superior parietal cortex

Caudal IPL areas: targets in the posterior superior parietal and temporal areas

#### **MATERIALS & METHODS**

## Data

- One subject
  - Diffusion weighted data (128x128x70)
    - B value 1000
    - Acquired in 63 gradient directions
  - T1 coronal data (256x256x208)
    - Manually extracted brain data
- T1 MNI 152 1mm standard data (182x218x182)
  - Juelich atlas

## **Tools for Brain Analysis**

- FreeSurfer: automated tools for reconstruction of the brain's cortical surface from structural MRI data, and overlay of functional MRI data onto the reconstructed surface.
- FSL: a comprehensive library of analysis tools for FMRI, MRI and DTI brain imaging data.
   FSL runs on Apples, Linux, and Windows. Most of the tools can be run both from the command line and as GUIs.
- SPM: a statistical package for processing brain data including fMRI, SPECT, PET, EEG, MEG.

## **Juelich Atlas**

Juelich histological (cyto- and myelo-architectonic) atlas A probabilistic atlas created by averaging multi-subject post-mortem cyto- and myelo-architectonic segmentations. The atlas contains 52 grey matter structures and 10 white matter structures. This is an update to the data used in Eickhoff's <u>Anatomy Toolbox</u> v1.5.

The atlas is based on the miscroscopic and quantitative histological examination of ten human post-mortem brains. The histological volumes of these brains were 3D reconstructed and spatially normalized into the space of the MNI single subject template to create a probabilistic map of each area. For the FSL version of this atlas, these probabilistic maps were then linearly transformed into MNI152 space.

## Flowchart



#### Estimation of Distribution of Diffusion using FSL BEDPOSTX

- Bayesian Estimation of Diffusion Parameters Obtained using Sampling Techniques (BEDPOSTX) to build up distribution of diffusion parameters at each voxel
  - Partial model allowing for fiber direction mixed with an isotropic ally diffusion model
  - A parameterized model of the transfer function between a distribution of fiber orientations in a voxel and the measured diffusion weighted signal
  - Use of Markov Chain Monte Carlo (MCMC) sampling to estimate the posterior distribution on parameters of interest

## WGMI Partition using Freesurfer

- White gray matter interface (WGMI) Partition
  - Gray matter does not have enough connectivity information for parcellation
  - Atlas based cortical registration (a2009 atlas)
  - Seed regions: inferior parietal lobule (IPC) including angus and super marginal
  - Target regions: all cortical regions except IPC

Connectivity Matrix Calculation using Probabilistic Tractography (FSL PROBTRACKX)

• Each value in the connectivity matrix indicates the probability that the seed particle can reach the target region through probabilistic tractography

#### ID # of target regions



connectivity probability = (number of particles that reached the target region) / (total number of particles issued from the seed voxel)

#### Juelich Atlas for Verification



Post-process group averaged probability map to obtain the function field labels with highest probability  $_{14}$ 

## Labeling Approaches

• K–Means Clustering

• Mixture of Gaussians (EM Clustering)

• Spectral Clustering (Graph – cut)

## Spectral Clustering

- Spectral Clustering
  - Build the similarity graph through pair-voxel correlation of connectivity similarity and spatial affinity
  - Solve the normalized graph-cut problem through
    Eigen decomposition of similarity matrix

## Build the Similarity Graph



#### Normalized cut of the Similarity Graph

Normalized cut

 $Ncut(A,B) = \frac{cut(A, B)}{asso(A,V)} + \frac{cut(A,B)}{asso(B,V)}$ 



Shi & Malik, 2000

#### RESULTS

## Data Summary

- Left Hemisphere IPL Parcellation (LH-IPL)
  - 667 voxels selected as seed for probabilistic tractography
  - 148 targets are selected for probabilistic tractography, 3 targets are discarded due to lack of enough connectivity
- Right hemisphere IPL Parcellation (RH-IPL)
  - 617 voxels selected as seed for probabilistic tractography
  - 148 targets are selected for probabilistic tractography, 2 targets are discarded due to lack of enough connectivity

#### Lh-IPL: 3D Sagittal View





Kmeans (N=5)



EM (N=5)



Spectral clustering (N=5)

## Lh-IPL – 2D Views (Kmeans, N=5)



Grey clusters are the atlas, while the colored ones are clustered by kmeans

## Lh-IPL – 2D Views (EM, N=5)



Grey clusters are the atlas, while the colored ones are clustered by EM

## Lh-IPL – 2D Views (SC, N=5)



Grey clusters are the atlas, while the colored ones are clustered by Spectral Clustering

#### **Normalized Connectivity Matrix**

3211-lh-ipc-Connectivity matrix before spectral clustering



Target regions

Before spectral clustering

#### 100 200 seed voxels 400 A00 500 600 20 40 60 80 100 120 140 Target regions

#### After spectral clustering

3211-lh-ipc-Connectivity matrix after spectral clustering

### Connectivity Similarity Matrix of Spectral Clustering



Before spectral clustering

After spectral clustering

#### Affinity Matrix of Spectral Clustering



300

Seed voxels

100

200

Spatial affinity after spectral clustering (3211-lh-ipc)

Before spectral clustering

After spectral clustering

400

500

600

0.9

0.8

0.7

0.6

0.5

0.4

0.3

0.2

0,1

#### Interpretation of the Clusters (LH-IPL)

	РGр	PGa	PFm	PF	PFt	РҒор	PFcm	Total	Top 3 connected targets
Cluster#1	88 (96.7%)	2	0	1	0	0	0	91	wm_lh_S_temporal_sup wm_lh_S_oc_sup_and_transversal wm_lh_S_intrapariet_and_P_trans
Cluster #2	0	0	4	69 (53.1%)	26	29	0	135	wm_lh_S_postcentral wm_lh_G_and_S_subcentral wm_lh_G_front_inf-Opercular
Cluster #3	48 (40.1%)	25	31	14	0	0	0	119	wm_lh_S_intrapariet_and_P_trans wm_lh_S_interm_prim-Jensen wm_lh_G_parietal_sup
Cluster #4	0	0	0	33	0	66 (46.5%)	43	163	wm_lh_Lat_Fis-post wm_lh_G_and_S_subcentral wm_lh_S_circular_insula_sup
Cluster #5	4	34	53	67 (42.1%)	0	0	1	159	wm_lh_S_interm_prim-Jensen wm_lh_S_temporal_sup wm_lh_G_temporal_middle

#### Additional Study

Bilge Soran Quals Project November 2011

## Outline of Work

- Tried several variants of normalized graph cuts
- Used both connectivity and spatial distance information
- Tried several different connectivity similarity functions
- Tried several different spatial distance functions
- Developed a spatial affinity function
- Tried out a feature selection approach
- Developed a new metric for evaluation

## Similarity matrix computation

• Build a normalized connectivity matrix using probabilistic tractography. The values are normalized by dividing by the largest value of the matrix.

• Build a symmetric spatial distance matrix

## **Connectivity Similarity Function**

$$W_{conn}^{i,j} = exp(-\alpha * f_{conn}(p_i, p_j) / \sigma_{conn}^2)$$

(where  $\sigma$  is a weighting factor and set to 2.)

**Distance Functions:** 

- Euclidean
- Standardized Euclidean
- Mahalanobis
- City Block
- Minkowski

- Cheybchev
- Jaccard
- Cosine
- Correlation
- Hamming

#### **Spatial Affinity Functions**

$$W_{spatial}^{i,j} = exp(-(1-\alpha) * dist(i,j) / \sigma_{spatial}^2)$$

$$W_{spatial}^{i,j} = 2 * (1 - \alpha) * (dist(i,j) < median(dist))$$

(where  $\sigma$  is a weighting factor and set to 0.5.)

$$dist(i, j) = max(|i_x - j_x|, |i_y - j_y|, |i_z - j_z|)$$

$$dist(i,j) = \sqrt{(i_x - j_x)^2 + (i_y - j_y)^2 + (i_z - j_z)^2}$$

 $dist(i,j) = dist(i,j) = (i_x - j_x) + (i_y - j_y) + (i_z - j_z)$ 

## Similarity matrix computation

• Compute the composite similarity matrix with one of the equations below:

$$\begin{split} W^{i,j}_{similarity} &= W^{i,j}_{conn} * W^{i,j}_{spatial} \\ W^{i,j}_{similarity} &= W^{i,j}_{conn} + W^{i,j}_{spatial} \end{split}$$

#### **Graph-Cuts Variants**

- 1. Standard Normalized Graph Cuts
- 2. Normalized Graph Cuts with Feature Selection
- 3. Normalized Graph Cuts with K-means

#### Similarity matrix computation



#### Feature Selection by Target Elimination

- Not all voxels have connections to all target regions.
- The variance of a target region is computed by using the connectivity values in its column of the connectivity matrix with the standard formula:

$$\frac{1}{n} \sum_{i=1}^{n} (x_i - \mu)^2$$

• After computing the variance for each target region, a threshold is applied to select targets with high variances since they are expected to carry discriminative information.

#### Evaluation

• An example table used in evaluation:

	Atlas Cluster 1	Atlas Cluster 2	Atlas Cluster 3	Atlas Cluster 4	Atlas Cluster 5	Atlas Cluster 6	Atlas Cluster 7
NCut Cluster 1	0	0	0	91	15	3	0
NCut Cluster 2	0	0	0	24	0	79	8
NCut Cluster 3	0	53	67	9	0	0	0
NCut Cluster 4	6	78	14	4	0	0	2
NCut Cluster 5	91	31	0	0	0	0	0

#### **Evaluation Metric**

$$A = \left(\sum_{i} max(row_{i})\right) / \left(\sum_{ij} cell_{ij}\right)$$
$$B = \left(\sum_{j} max(column_{i})\right) / \left(\sum_{ij} cell_{ij}\right)$$
$$If(A - B) < 0.2 \qquad R = (A + B)/2$$

Else the resulting parcellation is wrong.

#### RESULTS

#### Parcellation of Subject 3211: Connectivity Matrix Before Parcellation



Target regions

#### Parcellation of Subject 3211: Connectivity Matrix After Parcellation



Target regions

#### Parcellation of Subject 3211: Connectivity Variances Before Parcellation



#### Parcellation of Subject 3211: Variance of Each Cluster After Parcellation



44

#### Parcellation of Subject 3211: Performances of Different Clustering Methods



#### Parcellation of Subject 3211: **Performances of Different Clustering Methods**





**Mean-Shift** 



**Sparse K-means** 



#### Work in Progress

## Parcellation of 19 Subjects: Based on the selected parameters from the parcellation of Subject 3211

	LEFT HEMISPHERE							
	Normalized Cuts with Target Elimination							
	Alpha =0.5							
	Similarity Matrix construction = SUM							
	Distance Function = Jaccard							
SUBJECT	BEST RESULT							
3211	0.650273							
3402	0.603636							
3407	0.520629							
3414	0.697107							
3422	0.607826							
3424	0.621622							
3425	0.537037							
3484	0.635490							
3485	0.638649							
3486	0.607445							
3487	0.604724							
3488	0.686678							
3492	0.615960							
3496	0.621302							
3497	0.567294							
3498	0.588542							
3503	0.648148							
3504	0.685022							
3505	0.644431							

## Parcellation of 19 Subjects: Based on the selected parameters from the parcellation of Subject 3211

RIGHT HEMISPHERE								
	Normalized Cuts with Target Elimination							
	Alpha =0.5							
	Similarity Matrix construction = SUM							
	Distance Function =" Jaccard"							
SUBJECT	BEST RESULT							
3211	0.619906							
3402	0.671701							
3407	0.597689							
3414	0.620827							
3422	0.638840							
3424	0.564067							
3425	0.673877							
3484	0.575038							
3485	0.566434							
3486	0.551825							
3487	0.633803							
3488	0.561350							
3492	0.671875							
3496	0.669528							
3497	0.608730							
3498	0.660506							
3503	0.575816							
3504	0.645977							
3505	0.617366							

# Parcellation of 19 Subjects: Best Parameters (Left Hemisphere)

Method	# of Cluster	α	Distance Function	Spatial Coherence	Similarity Matrix
Normalized Graph Cuts	5	0.9	Hamming	Threshold	Multiply
Normalized Graph Cuts with target elimination	5	0.8	City block	Weight	Sum

## Parcellation of 19 Subjects: Parcellation Evaluation based on the training parameters (Left Hemisphere)

Subject	Normalized Graph Cuts	Normalized Graph Cuts
		With Target Elimination
3211	0.589481	0.661202
3402	0.660000	0.629091
3407	0.586444	0.605108
3414	0.706148	0.696203
3422	0.660870	0.726087
3424	0.689189	0.664619
3425	0.659612	0.567019
3484	0.597028	0.648601
3485	0.638649	0.630747
3486	0.586294	0.621827
3487	0.568504	0.609449
3488	0.696546	0.696546
3492	0.608479	0.569825
3496	0.640039	0.666667
3497	0.594067	0.569465
3498	0.618750	0.632292
3503	0.652778	0.577160
3504	0.615639	0.621145
3505	0.643819	0.605263
AVERAGE	0.632228	0.631490

#### Parcellation with Target Elimination Results: Subjects 3414, 3422, 3488 (Left Hemisphere)



Atlas in 3422's FA space



3414





# Parcellation of 19 Subjects: Best Parameters (Right Hemisphere)

Method	# of Cluster	α	Distance Function	Spatial Coherence	Similarity Matrix
Normalized Graph Cuts	5	0.6	Jaccard	Weight	Sum
Normalized Graph Cuts with target elimination	5	0.6	Jaccard	Weight	Sum

## Parcellation of 19 Subjects: Parcellation Evaluation based on the training parameters (Right Hemisphere)

Subject	Normalized Graph Cuts	Normalized Graph Cuts
		With Target Elimination
3211	0.625392	0.626959
3402	0.753577	0.751987
3407	0.718487	0.710084
3414	0.569952	0.570747
3422	0.652021	0.655536
3424	0.637883	0.646240
3425	0.640599	0.640599
3484	0.593415	0.601072
3485	0.627622	0.628497
3486	0.683212	0.682482
3487	0.590669	0.590669
3488	0.611452	0.615542
3492	0.747917	0.746875
3496	0.657725	0.653433
3497	0.666667	0.666667
3498	0.657588	0.701362
3503	0.600768	0.599808
3504	0.611494	0.611494
3505	0.670802	0.669847
AVERAGE	0.648276	0.651047

#### Parcellation with Target Elimination Results: Subjects 3402, 3407, 3492 (Right Hemisphere)



Atlas in 3402's FA space



3407





#### Comparison of Normalized Graph Cuts

- Standard NGC with feature selection produced best results in most of the tests.
- Standard NGC without feature selection produced results very close to those with feature selection.
- NGC with k-means produced incorrect parcellations according to the metric.

## Conclusion

 Different clustering methods were applied to an anatomical connectivity map, which is obtained by DTI-based tractography of the IPL of a living subject to parcellate it into component regions with different connectivity patterns.

 Among the different methods investigated, normalized graph cuts showed the best performance.

## Conclusion

- The main difficulty of the evaluation was having no ground truth data by which to measure the quality of our parcellation.
- How many different regions exist in the IPL of a human being is still an unknown. Therefore in this work, different numbers of clusters were tried and the evaluation metric was designed to measure the quality of the overlap with different numbers of clusters.