

The Waisman Laboratory  
for Brain Imaging and Behavior



University of Wisconsin  
**SCHOOL OF MEDICINE**  
**AND PUBLIC HEALTH**

# Rapid Acceleration of the Permutation Test via Transpositions

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# Abstract

The permutation test is an often used test procedure for determining statistical significance in brain network studies. Unfortunately, generating every possible permutation for large-scale brain imaging datasets such as HCP and ADNI with hundreds of subjects is not practical. Many previous attempts at speeding up the permutation test rely on various approximation strategies such as estimating the tail distribution with known parametric distributions. In this study, we propose the novel transposition test that exploits the underlying algebraic structure of the permutation group. The method is applied to a large number of diffusion tensor images in localizing the regions of the brain network differences.

# Acknowledgement

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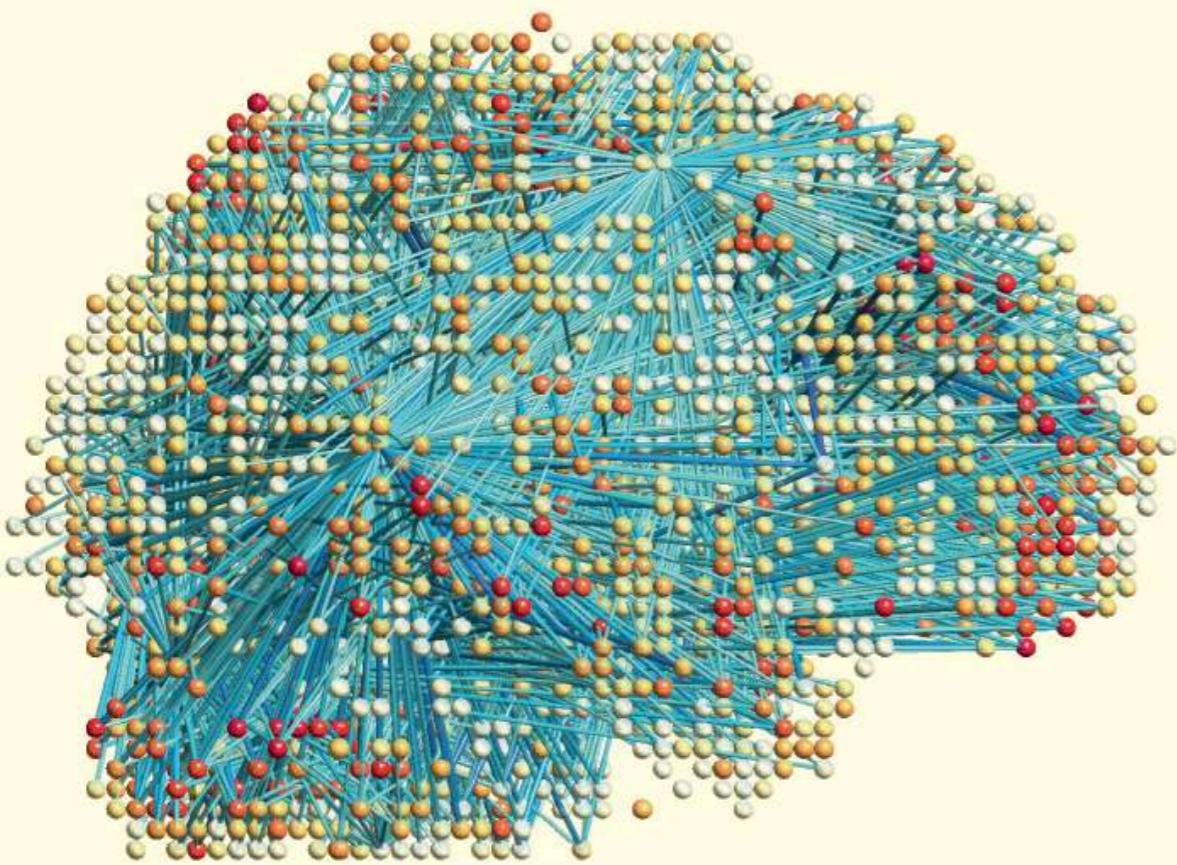
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# BRAIN NETWORK ANALYSIS



Moo K. CHUNG

New book!

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Wisconsin-Madison  
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**Organizers:**

**Joseph Reinhardt**

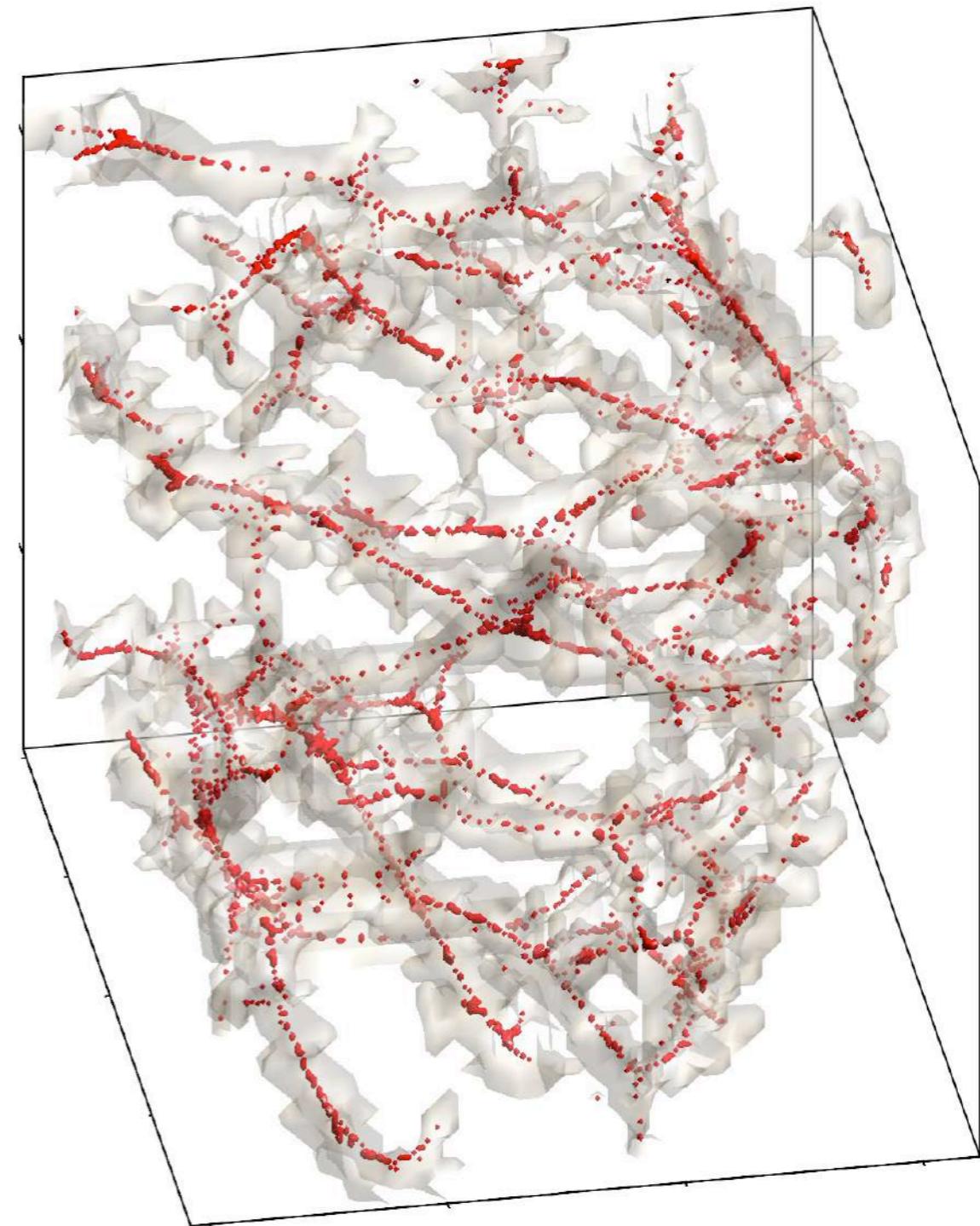
University of Iowa

**Moo K. Chung**

Univ. of Wisconsin-Madison

**Six Invited speakers**

**Best paper awards**



# What is permutation test?

$$\mathbf{x} = (x_1, x_2, \dots, x_m)$$

$$\mathbf{y} = (y_1, y_2, \dots, y_n)$$

$$(\mathbf{x}, \mathbf{y}) = (x_1, \dots, x_m, y_1, \dots, y_n)$$

$$\pi(\mathbf{x}, \mathbf{y}) \in \mathbb{S}_{m+n}$$

Permutation group of order  $m+n$

$$p\text{-value} = \frac{1}{(m+n)!} \sum_{\pi \in \mathbb{S}_{m+n}} \mathcal{I}\left(f(\pi(\mathbf{x}), \pi(\mathbf{y})) \geq f(\mathbf{x}, \mathbf{y})\right)$$



Computational bottleneck

# History of permutation test

Fisher 1935, The Design of Experiment

$$\binom{8}{4} = 70$$

Thompson et al. 2001, Nature Neuroscience

$$\binom{40}{20} = 1.34 \cdot 10^{11}$$

Nichols et al. 2002, Human Brain Mapping

$$\binom{6}{3} = 20$$

Google scholar 434,000 papers.

# Limitation of permutation test

Serious computational bottleneck

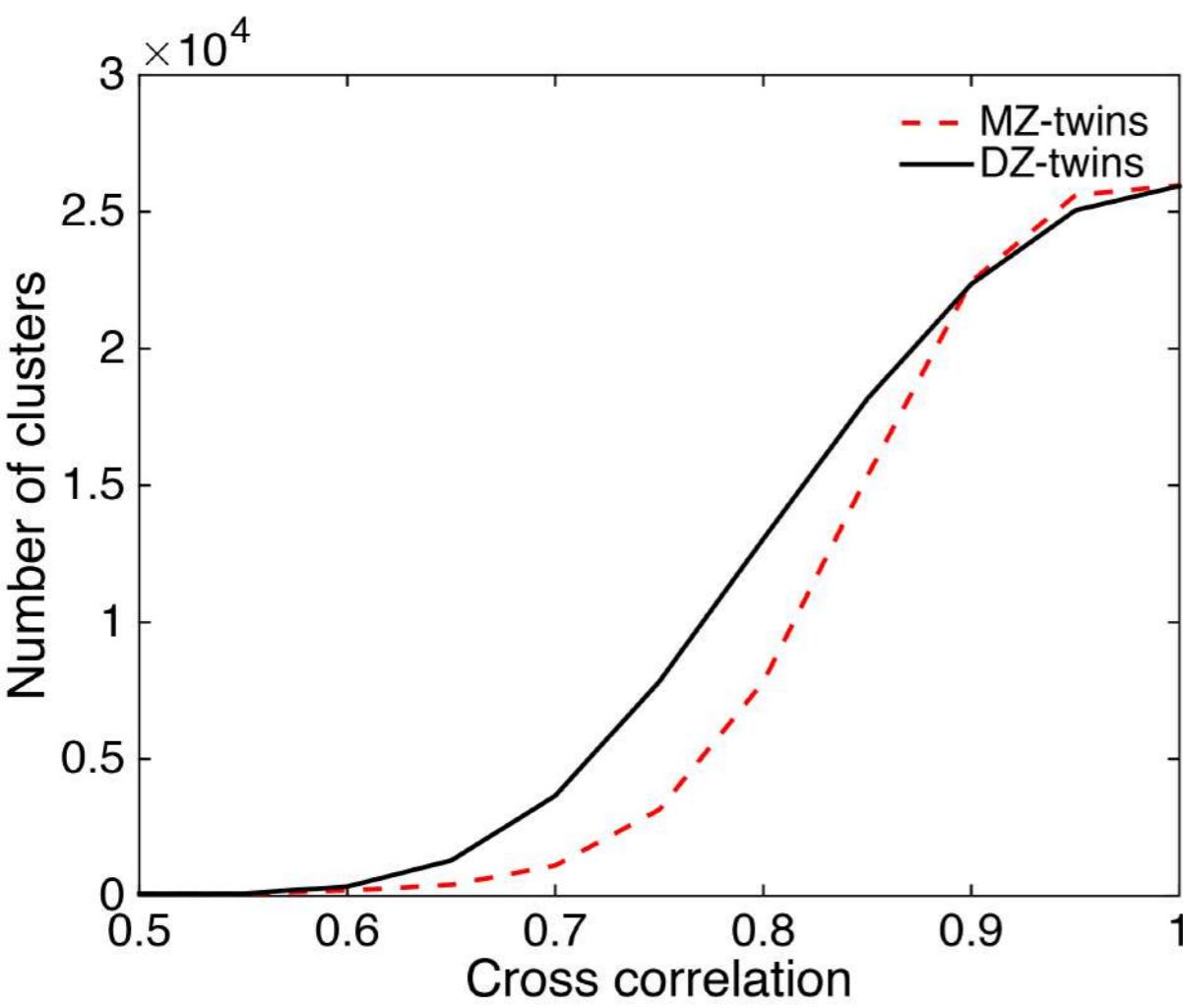
- 1) Need to permute million voxels.
- 2) Very slow: Exponential run time

Thompson *et al.* (2001) used supercomputer:  
1million permutations for

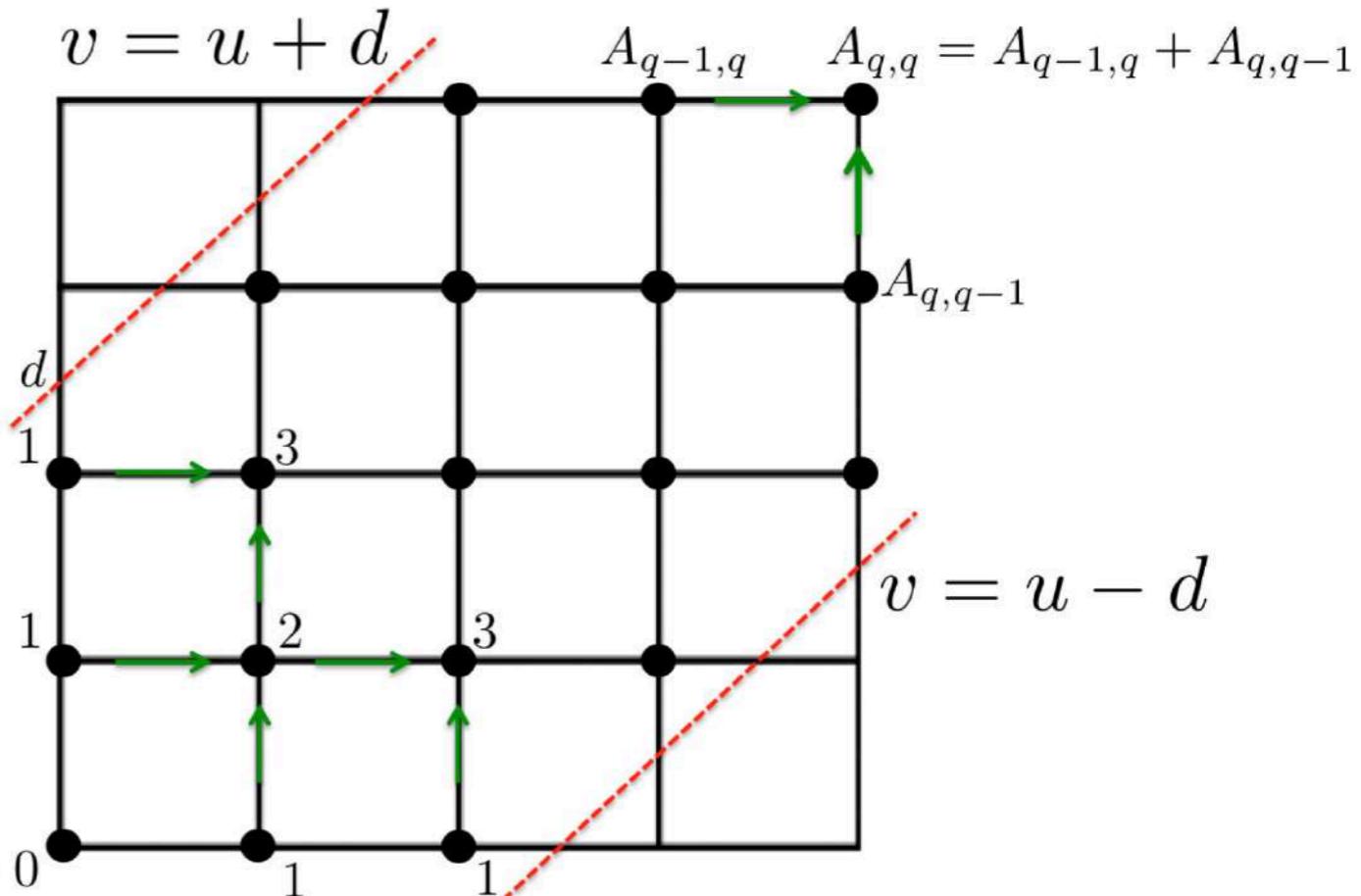
$$\binom{40}{20} = 1.34 \cdot 10^{11} \quad \text{hundred billion}$$

# Pervious method: Exact topological inference (ETI)

$$D_q = \sup_{1 \leq j \leq q} |\beta_i(G_{\lambda_j}^1) - \beta_i(G_{\lambda_j}^2)|$$



$$P(D_q \geq d) = 1 - \frac{A_{q,q}}{\binom{2q}{q}}$$



## Random transposition on the permutation group

$$\mathbf{x} = (x_1, x_2, \dots, x_{i-1}, \textcolor{red}{x_i}, x_{i+1}, \dots, x_m)$$

transpose  $i$ -th and  $j$ -th data

$$\mathbf{y} = (y_1, y_2, \dots, y_{j-1}, \textcolor{red}{y_j}, y_{j+1}, \dots, y_n)$$



$$\pi_{ij}(\mathbf{x}) = (x_1, x_2, \dots, x_{i-1}, \textcolor{red}{y_j}, x_{i+1}, \dots, x_m)$$

$$\pi_{ij}(\mathbf{y}) = (y_1, y_2, \dots, y_{j-1}, \textcolor{red}{x_i}, y_{j+1}, \dots, y_n)$$

Theorem 1: If the test statistic is algebraic, there exists a function  $g$  such that

$$f(\pi_{ij}(\mathbf{x}), \pi_{ij}(\mathbf{y})) = g(f(\mathbf{x}, \mathbf{y}), x_i, y_i)$$

where computational complexity of  $g$  is constant.

Theorem 2: Any permutation in  $S_{m+n}$  can be reachable by a sequence of transpositions.

# Online computation for t-stat.

$$\nu(\mathbf{x}) = \sum_{j=1}^m x_j, \quad \omega(\mathbf{x}) = \sum_{j=1}^m \left( x_j - \frac{\nu(\mathbf{x})}{m} \right)^2$$

$\mathcal{O}(m)$   $\mathcal{O}(3m+2)$

$$\nu(\pi_{ij}(\mathbf{x})) = \nu(\mathbf{x}) - x_i + y_j \quad \mathcal{O}(2)$$

$$\omega(\pi_{ij}(\mathbf{x})) = \omega(\mathbf{x}) - x_i^2 + y_j^2 + \frac{\nu(\mathbf{x})^2 - \nu(\pi_{ij}(\mathbf{x}))^2}{m}$$

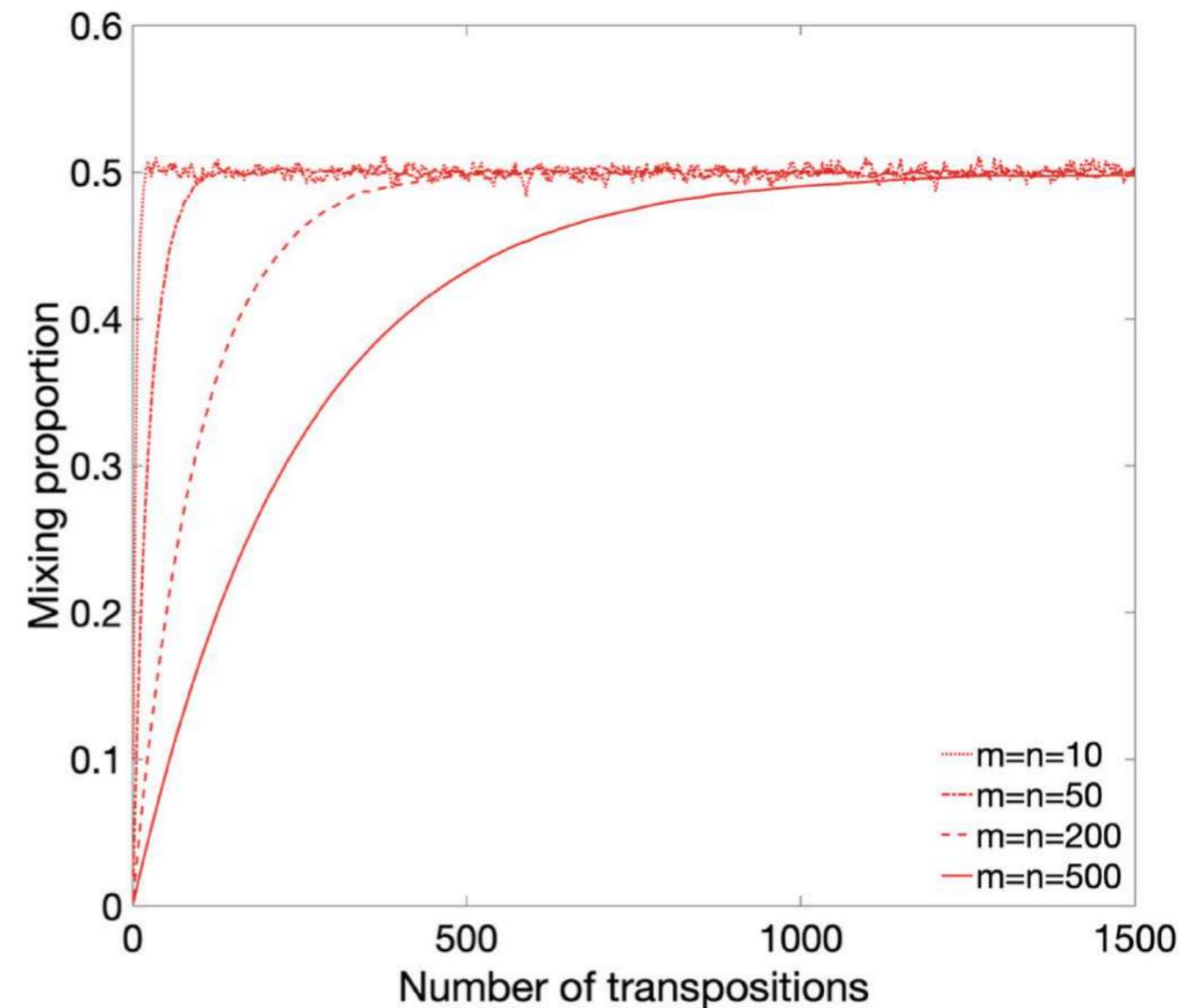
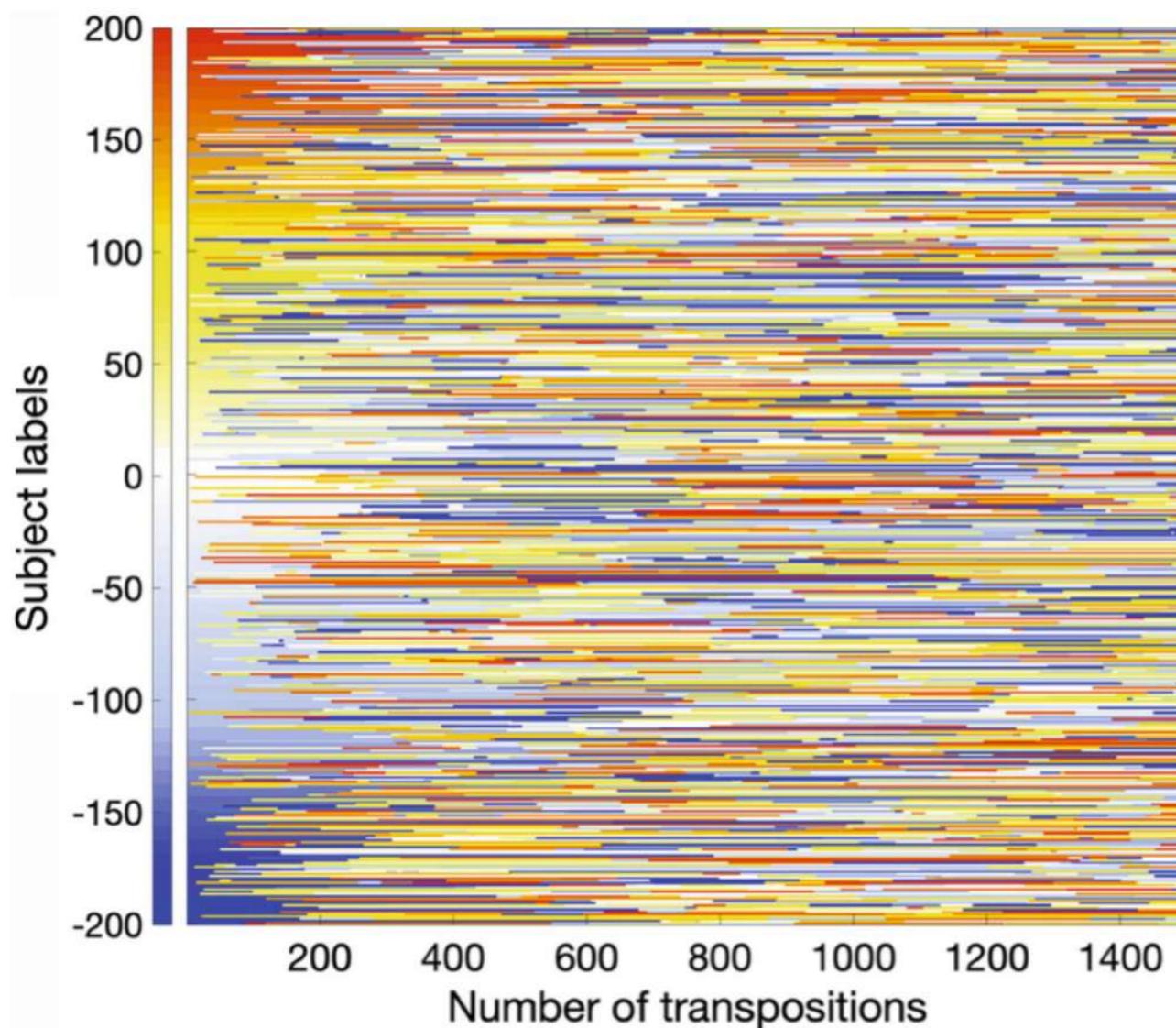
$\mathcal{O}(9)$

T-stat computation per permutation

Permutation test:  $\mathcal{O}(4m+4n+20)$

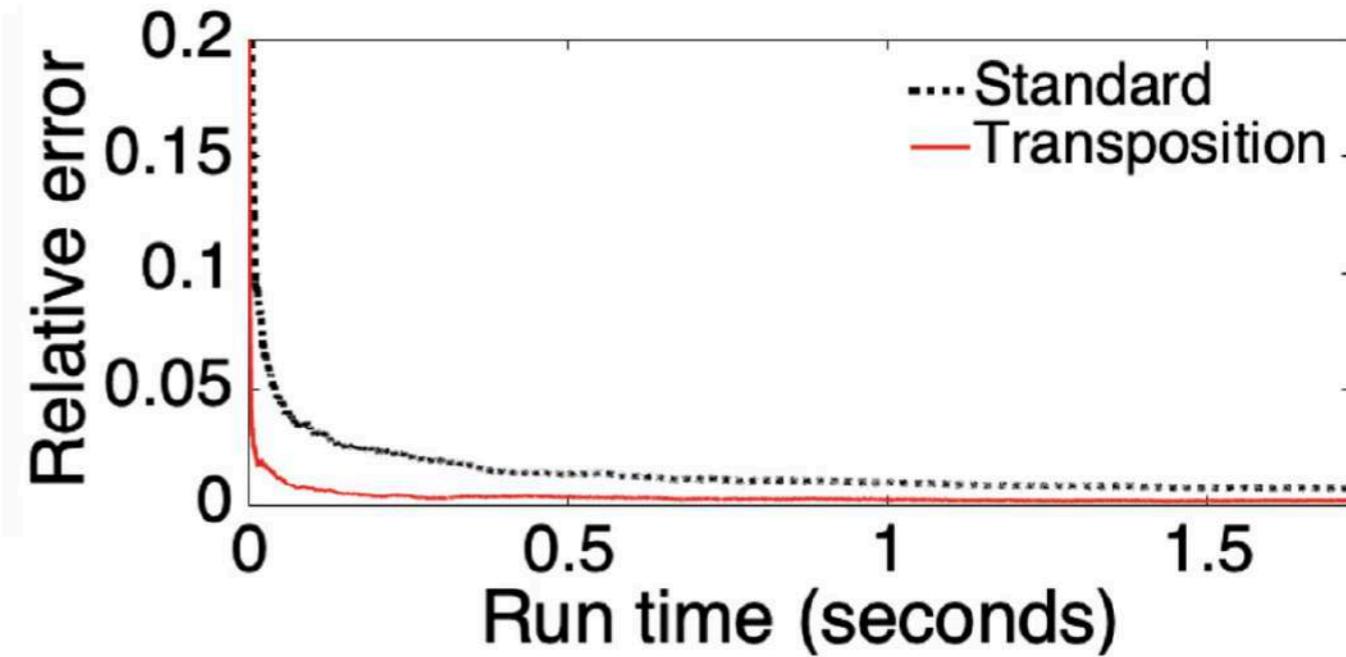
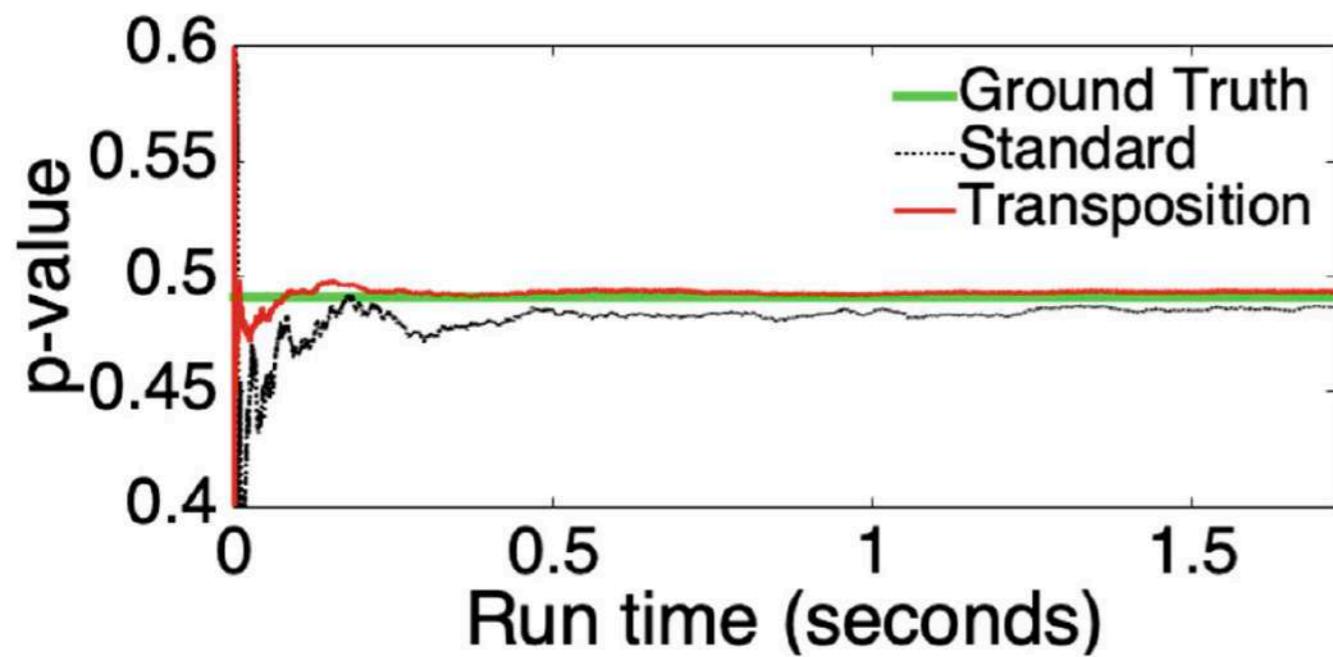
Transposition test:  $\mathcal{O}(35)$

# Simulation: Mixing proportion over transpositions



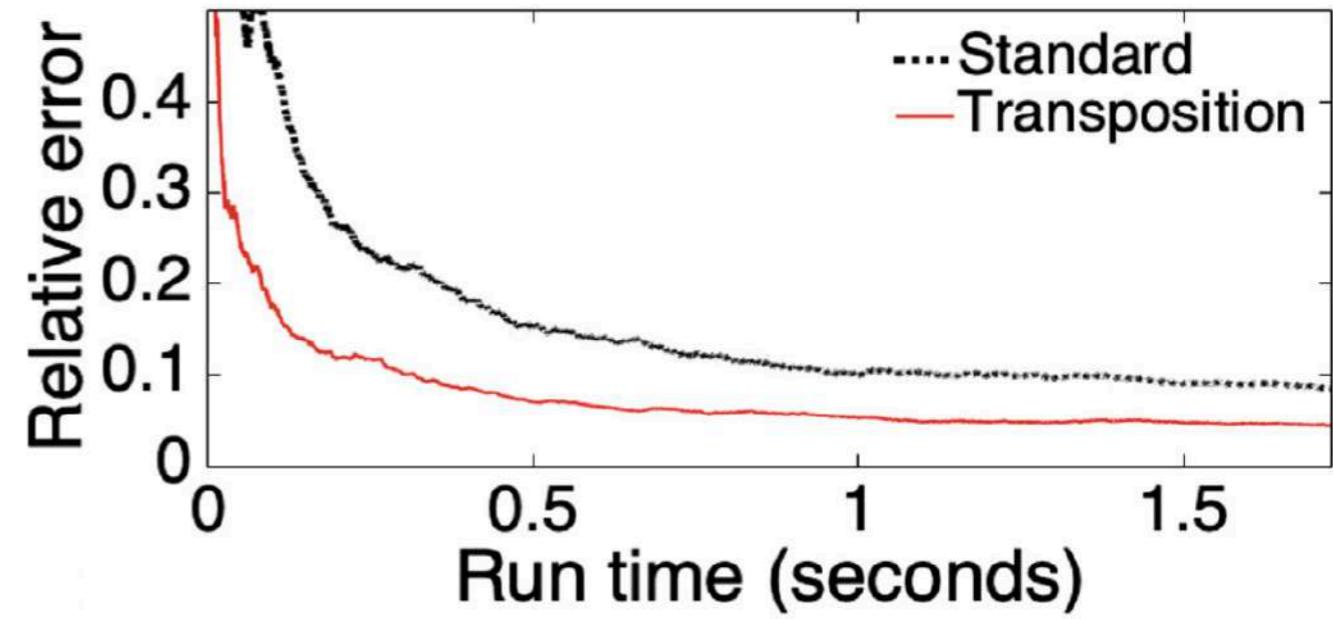
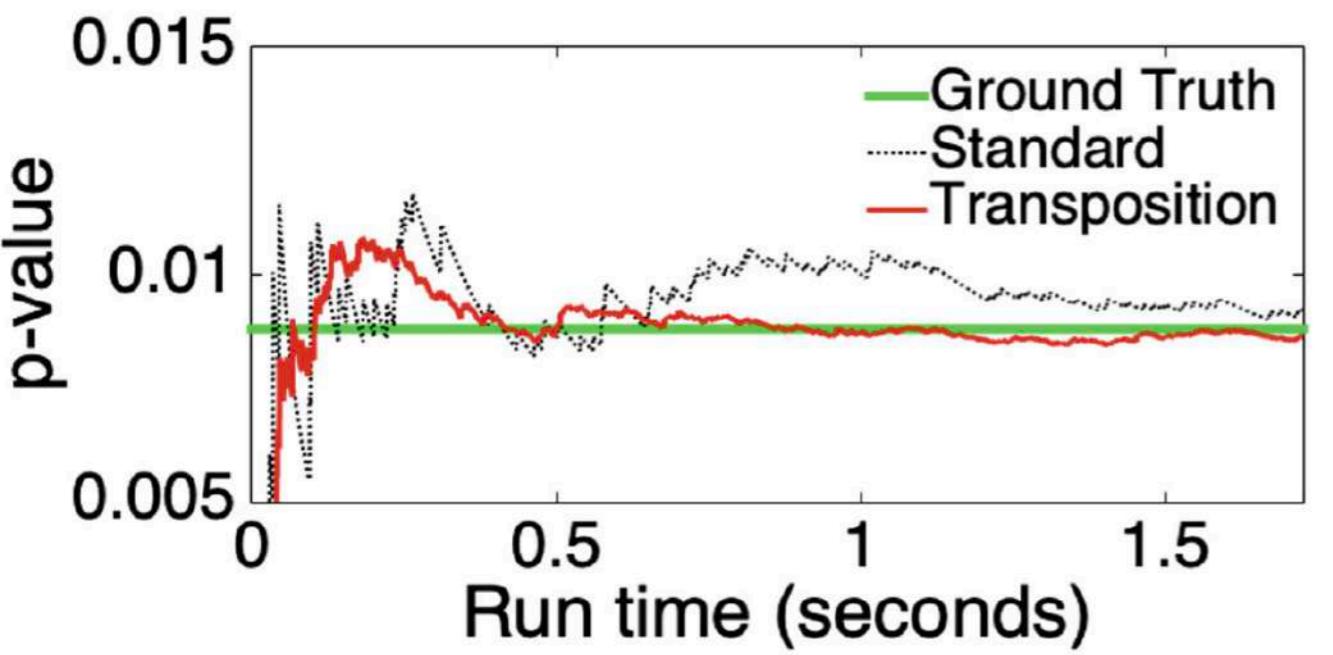
$m=n=10$

$$x_1, \dots, x_m \sim 0.1 + \text{Unif}(0, 1)$$
$$y_1, \dots, y_n \sim \text{Unif}(0, 1)$$



$m=n=100$

$$x_1, \dots, x_m \sim 0.1 + \text{Unif}(0, 1)$$
$$y_1, \dots, y_n \sim \text{Unif}(0, 1)$$



# Matlab code

<http://www.stat.wisc.edu/~mchung/transpositions>

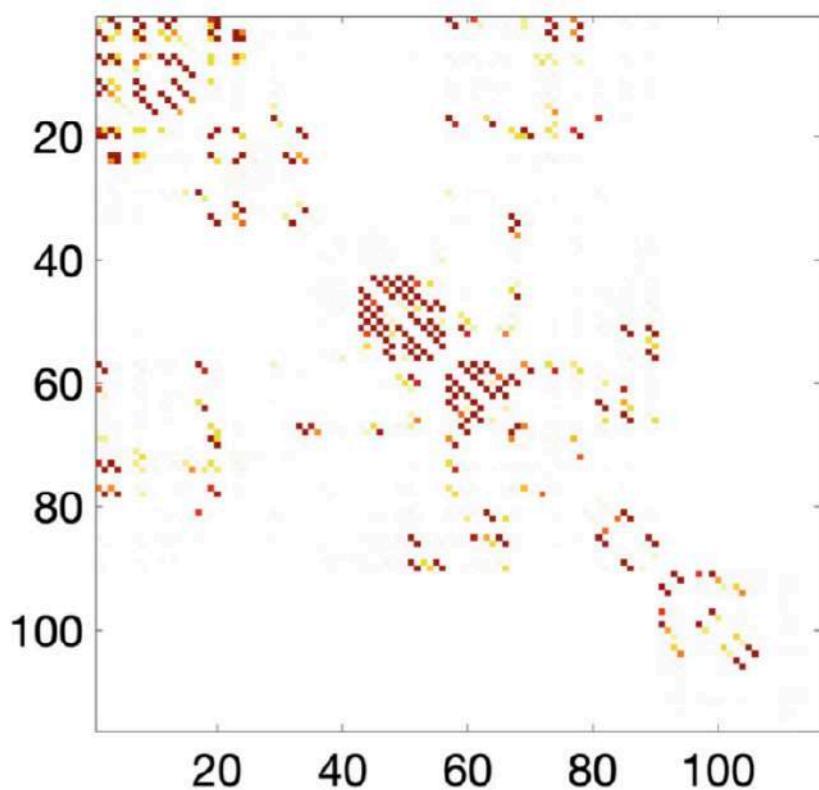
## Permutation test

```
[stat_s, time_s] = test_permute (x ,y, per_s)
```

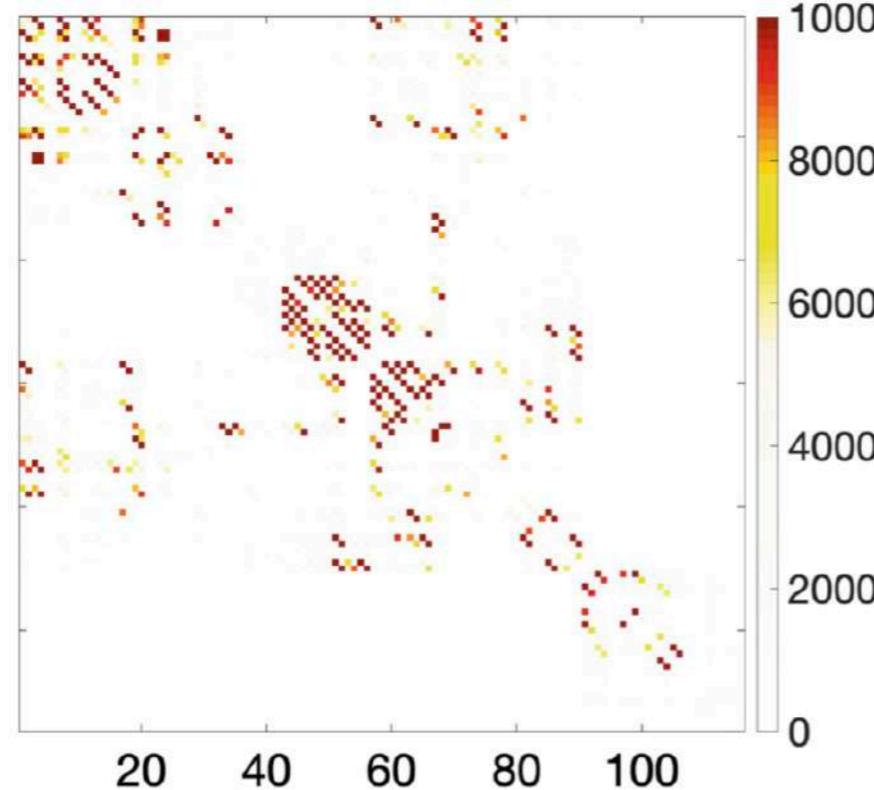
## Transposition test

```
stat_t=[];  
for i=1:10000  
    [stat, time] = test_transpose (x, y, per_t / 10000);  
    stat_t=[stat_t; stat];  
    time_t=time_t + time;  
end
```

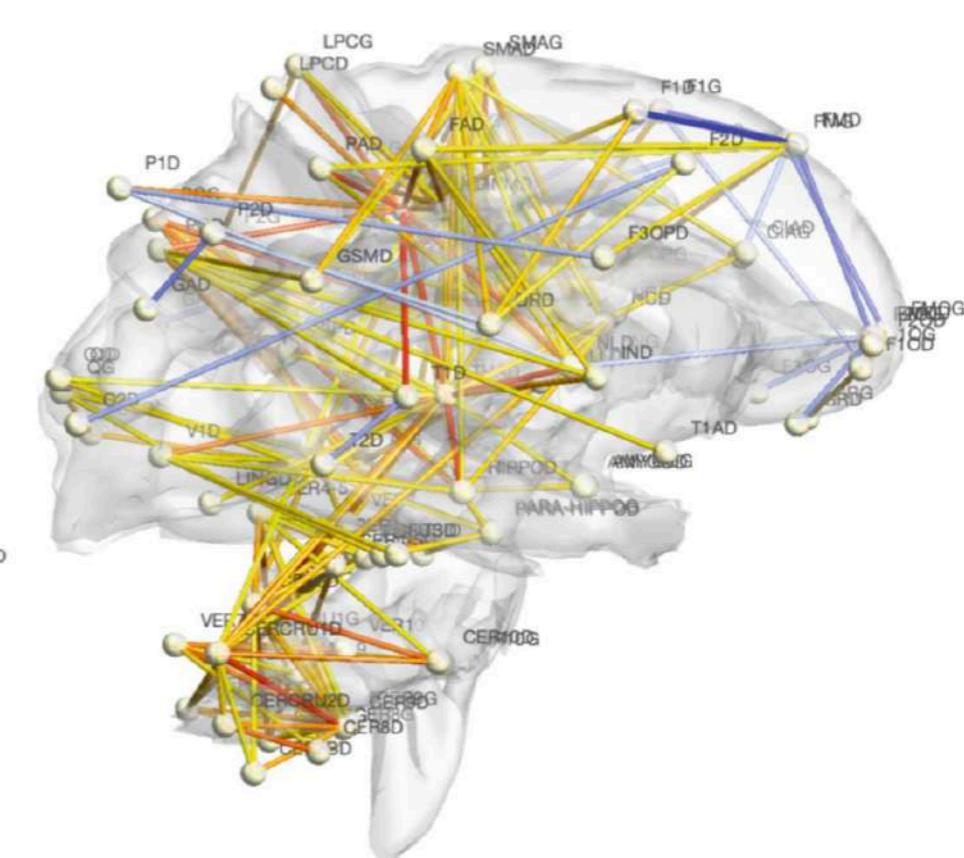
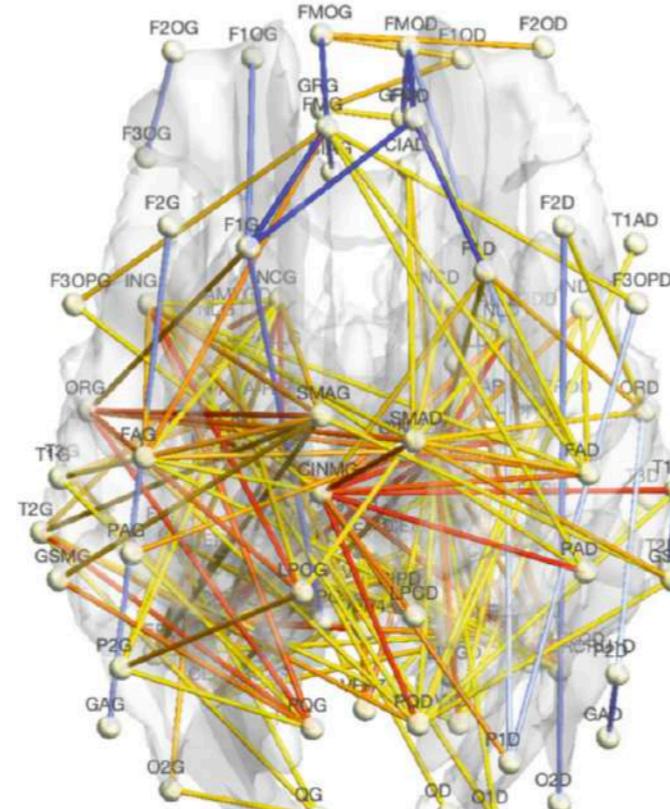
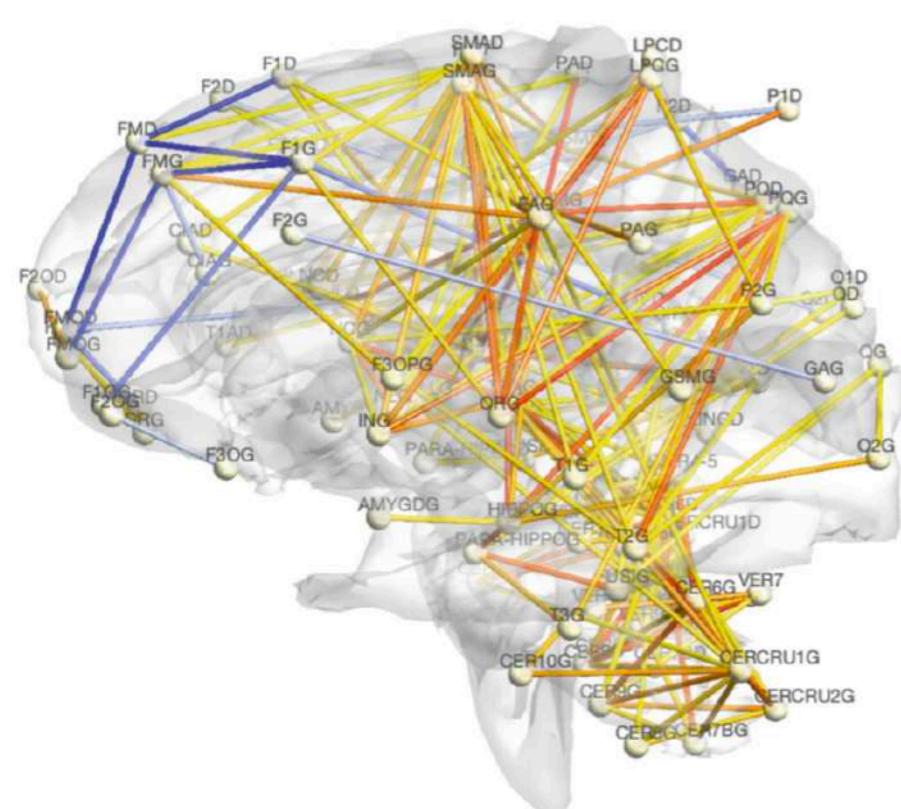
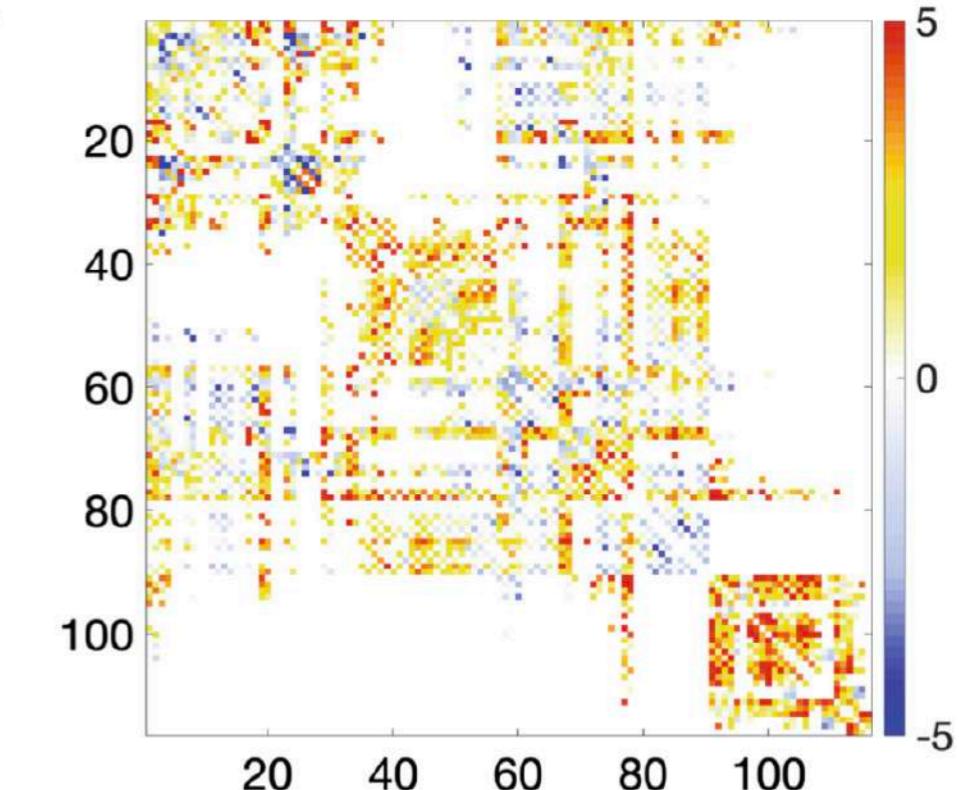
# Female



# Male



# t-stat.



**t-stat (202 females – 154 males)**



Thank you!

Question? [mkchung@wisc.edu](mailto:mkchung@wisc.edu)