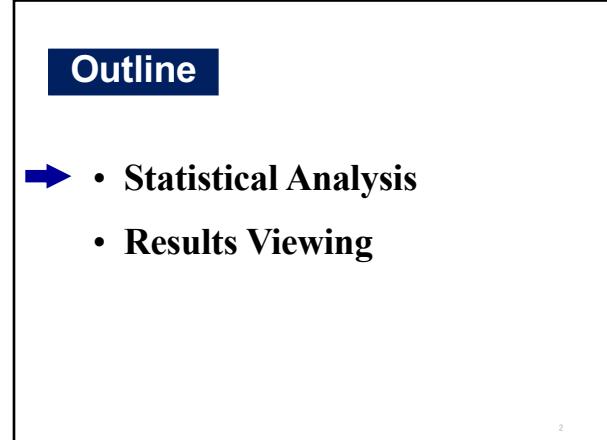
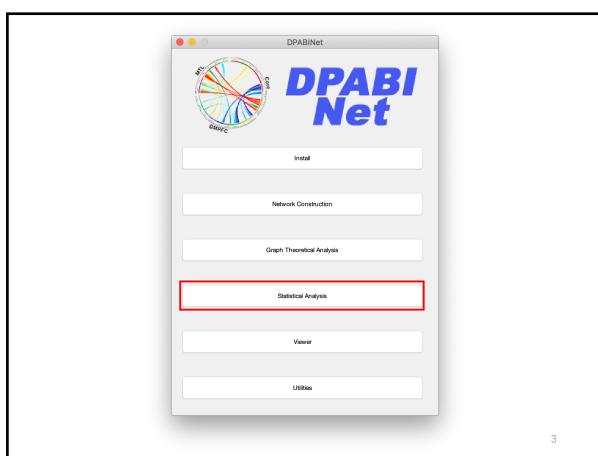


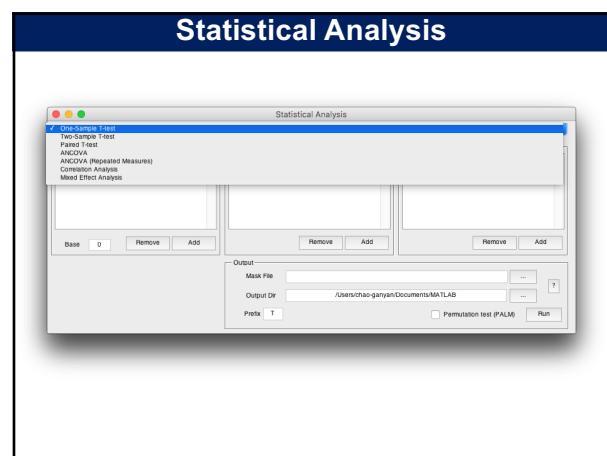
1



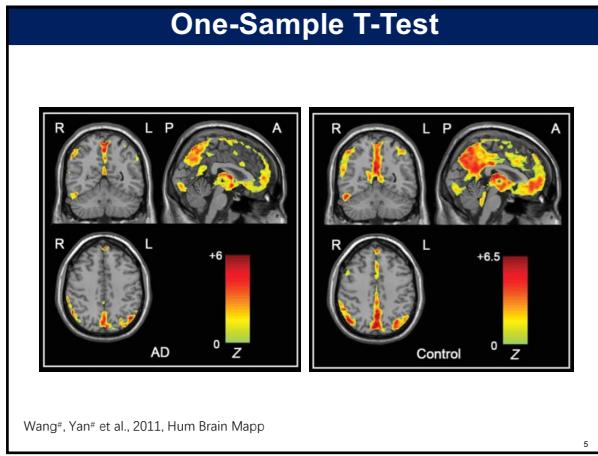
2



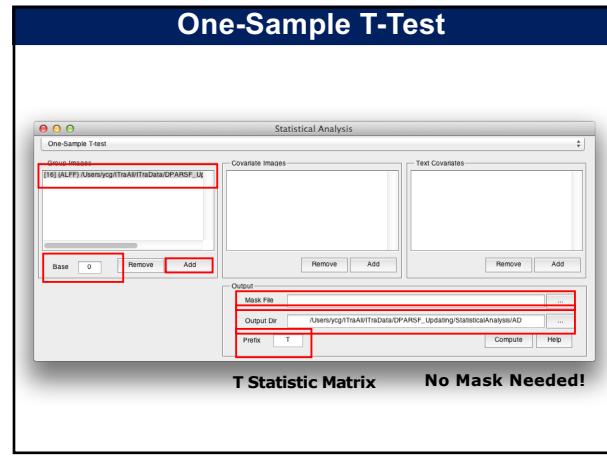
3



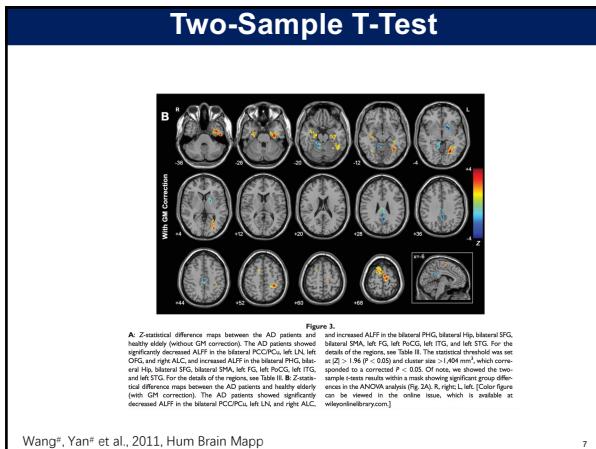
4



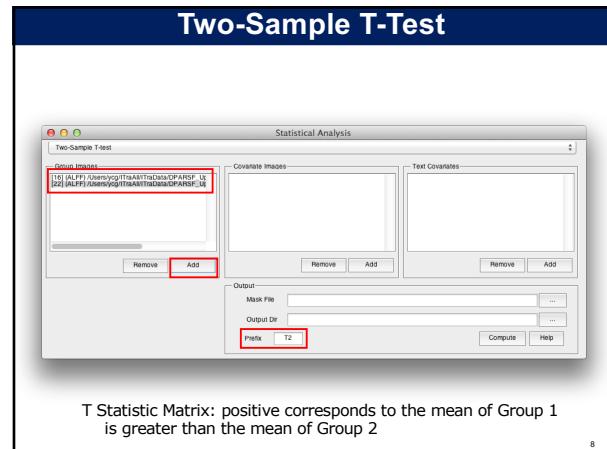
5



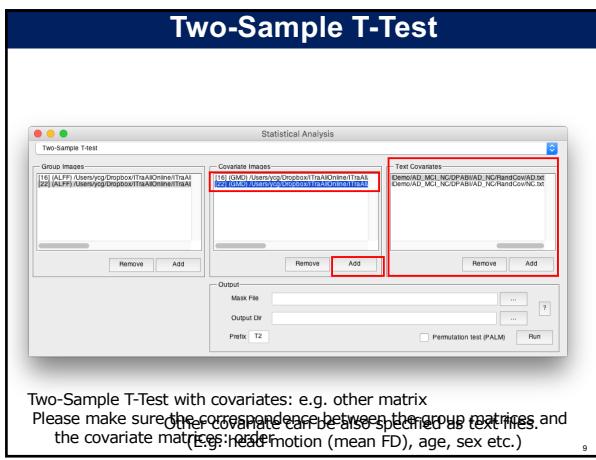
6



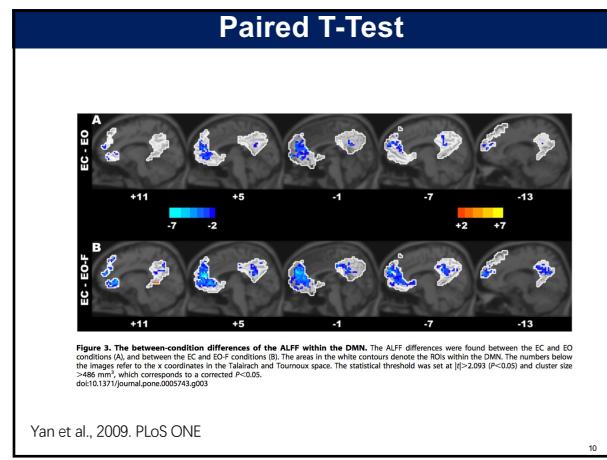
7



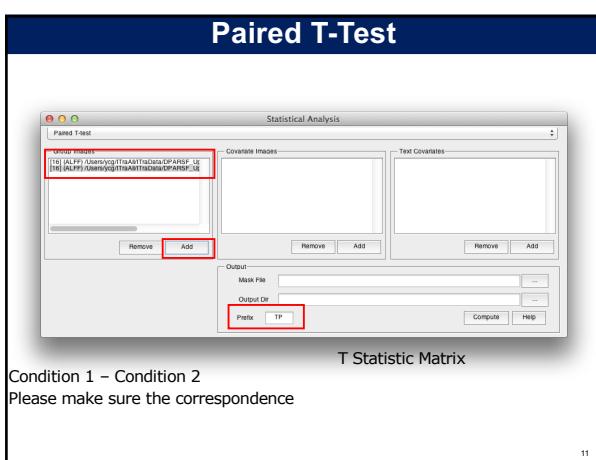
8



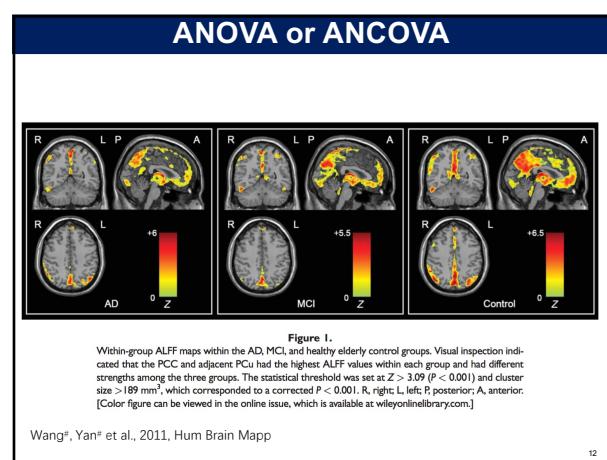
9



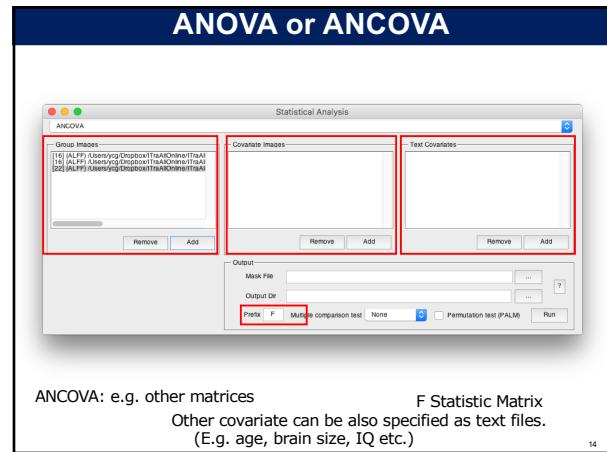
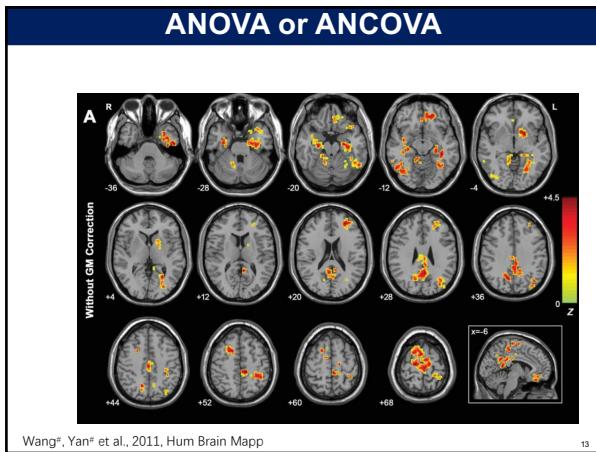
10



11

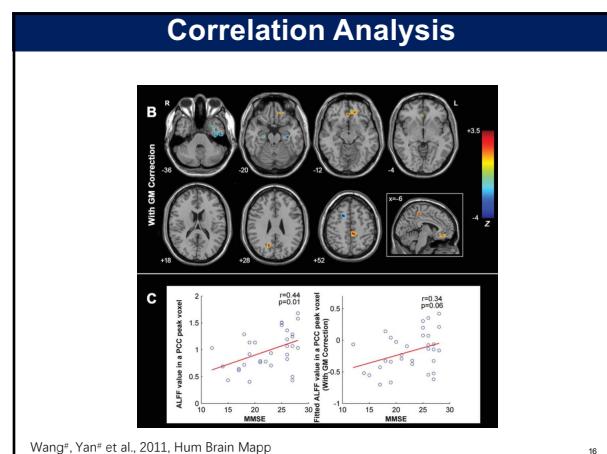
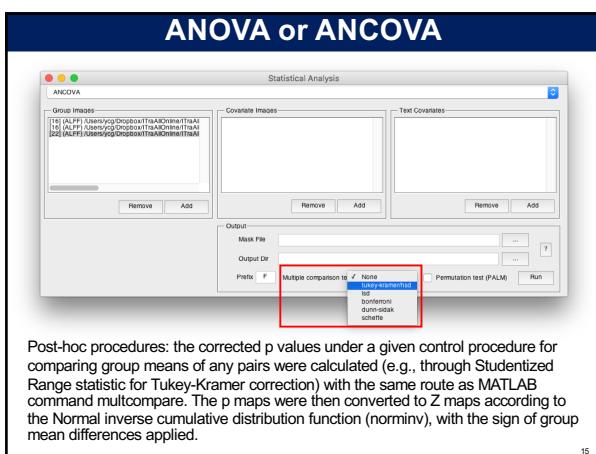


12



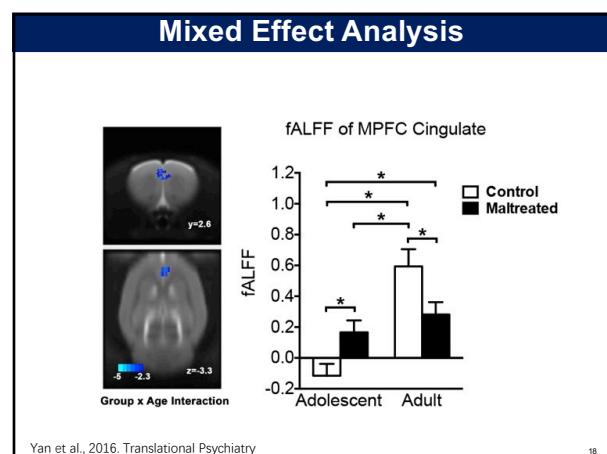
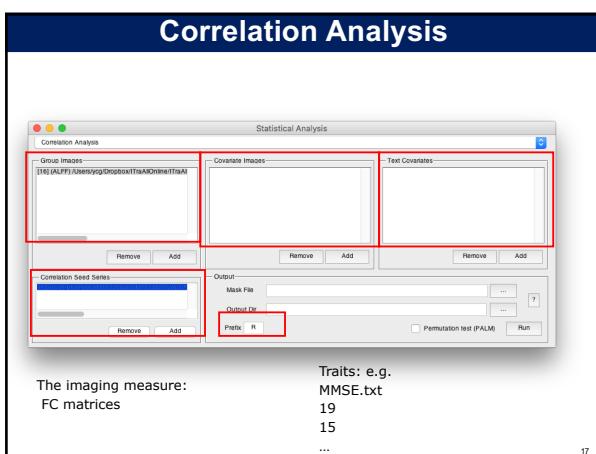
13

14



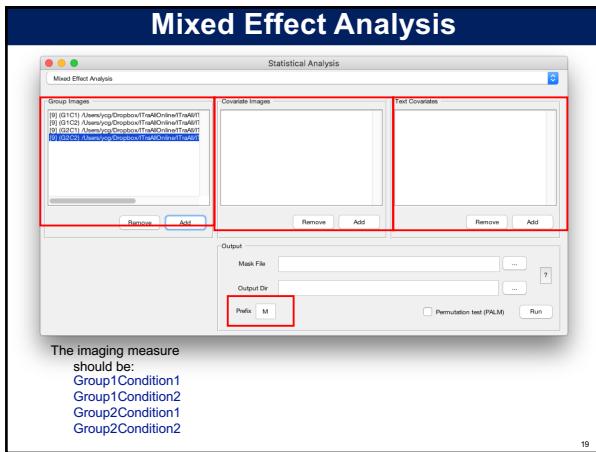
15

16



17

18



19

## Mixed Effect Analysis

- \* ConditionEffect\_T.mat - the T values of condition differences (corresponding to the first condition minus the second condition) (WithinSubjectFactor)
- \* Interaction\_F.mat - the F values of interaction (BetweenSubjectFactor by WithinSubjectFactor)
- \* Group\_TwoT.mat - the T values of group differences (corresponding to the first group minus the second group). Of note: the two conditions will be averaged first for each subject. (BetweenSubjectFactor)

20

## Statistical Analysis

```

Function [b_OLS_brain, t_OLS_brain, TF_PoerContrast_brain, r_OLS_brain, Header] = y_Groupanalysis_lImage(DependentVolume,Predictor,OutputName,
% Input: DependentVolume - 4D data matrix (Dim0xDim1xDim2xDim3) or the directory of 3D image data file or the filename of one 4D image
% Predictor - tple Predictors M (subjects) by N (traits). SHOULD INCLUDE the CONSTANT column if needed. The program will not add constant
% OutputName - output name, (should not have extension such as .img,.nii)
% Maskfile - the mask file
% CovVolume (optional) - 4D data matrix (Dim0xDim1xDim2xDim3) or the directory of image covariates, in which the files should be
% CovVolume = 'CovVolume' - Constant for F-test or 'LevelCov' matrix for T-test
% TF_flag (optional) - 'T' or 'F'. Specify if T-test or F-test need to be performed for the contrast
% IsOutputResiduals (optional) - 1: output the 4D residuals.
% Header (optional) - If DependentVolume is given as a 4D Brain matrix, then Header should be designated.

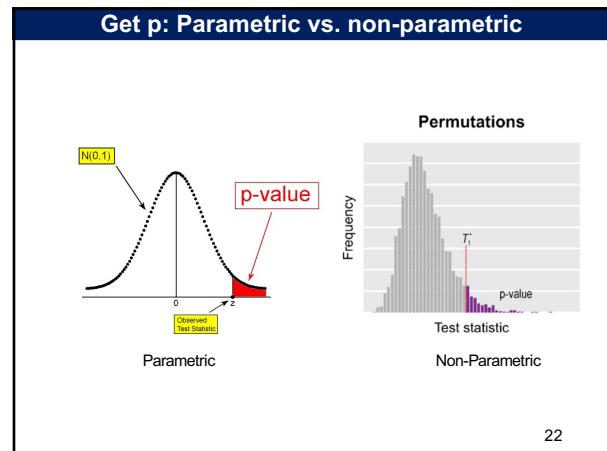
% Output:
% OutputName_b.nii, OutputName_T.nii - beta and t value files results
% OutputName_Residual.nii (optional) - Residual files

% Written by YAN Chao-Gan 12/08/2012
% This work was supported by the Child Mind Institute, 445 Park Avenue, New York, NY 10022, USA
% The Phyllis Green and Randolph Cowen Institute for Pediatric Neuroscience, New York University Child Study Center, New York, NY 10016, USA
% ycg.ycg@gmail.com

{DPABI_Dir}/StatisticalAnalysis/y_GroupAnalysis_lImage.m

```

21



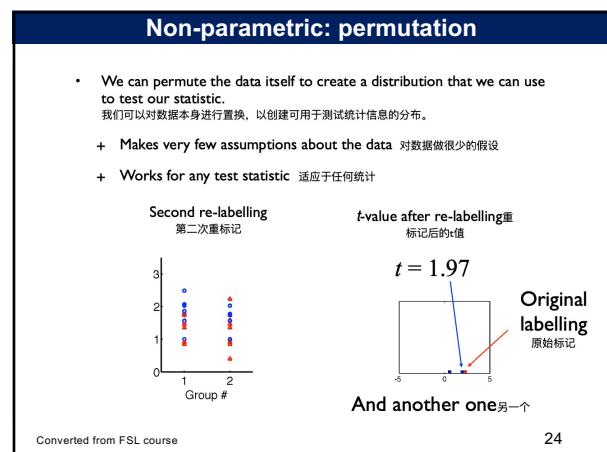
22

## Non-parametric: permutation

- We can permute the data itself to create a distribution that we can use to test our statistic.  
我们可以对数据本身进行置换，以创建可用于测试统计信息的分布。
- Makes very few assumptions about the data 对数据做很少的假设
- Works for any test statistic 适应于任何统计

Winkler et al., 2016. NeuroImage; Converted from FSL course

23



24

## Non-parametric: permutation

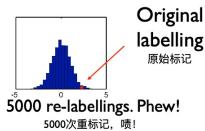
- We can permute the data itself to create a distribution that we can use to test our statistic.
- Makes very few assumptions about the data 对数据做很少的假设
- Works for any test statistic 适应于任何统计

Of the 5000 re-labellings, only 90 had a t-value > 2.27 (the original labelling).

在5000个重新标记中，只有90个的t值>2.27（原始标记）。

i.e. there is only a ~1.8% (90/5000) chance of obtaining a value > 2.27 if there is no difference between the groups

C.f.  $p(x \geq 2.27) = 1.7\%$  for  $t_{18}$   
即如果各组之间没有差异，则只有1.8% (90/5000) 的机会获得>2.27的值。C.F.t18的 $p(x \geq 2.27) = 1.7\%$



Converted from FSL course

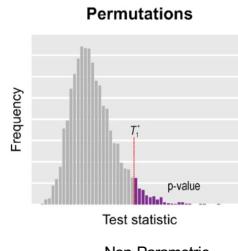
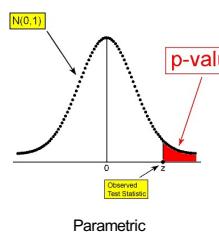
25

## Get p: Parametric vs. non-parametric

	Parametric	Non-parametric
Assumed distribution	Normal	Any
Assumed variance	Homogeneous	Homogenous and Heterogeneous
Typical data	Ratio or Interval	Ordinal or Nominal
Data set relationships	Independent	Any
Usual central measure	Mean	Median
Benefits	Can draw more conclusions	Simplicity; Less affected by outliers

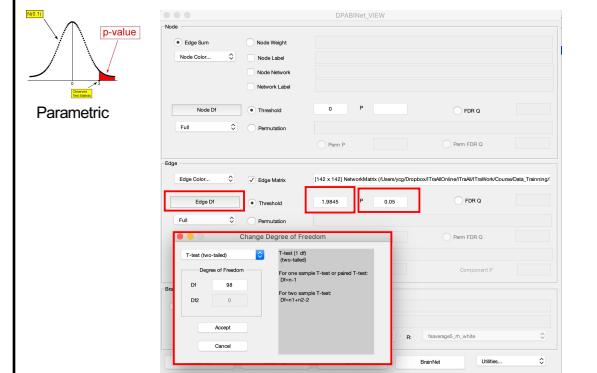
26

## Get p: Parametric vs. non-parametric



27

## Get p: Parametric vs. non-parametric



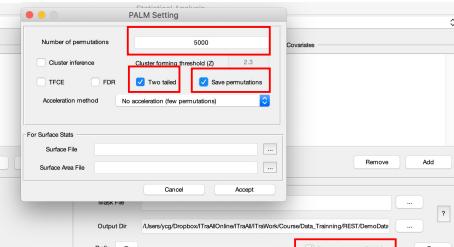
28

27

28

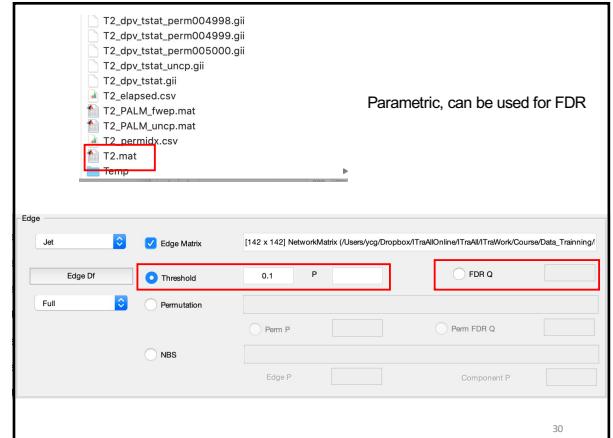
## Non-Parametric

Not for GTA!!!



29

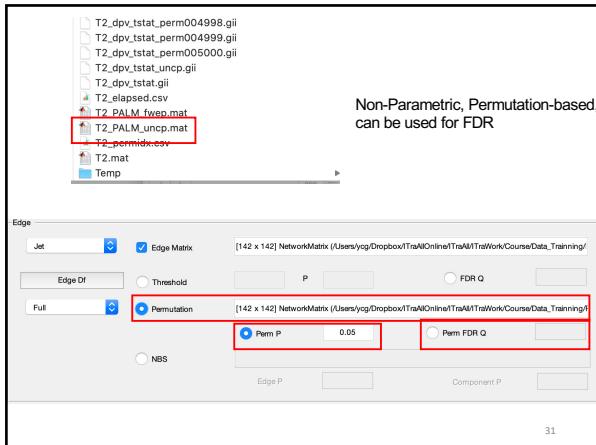
Parametric, can be used for FDR



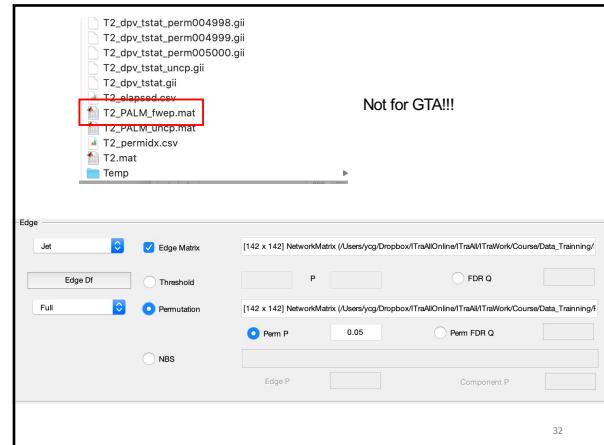
30

29

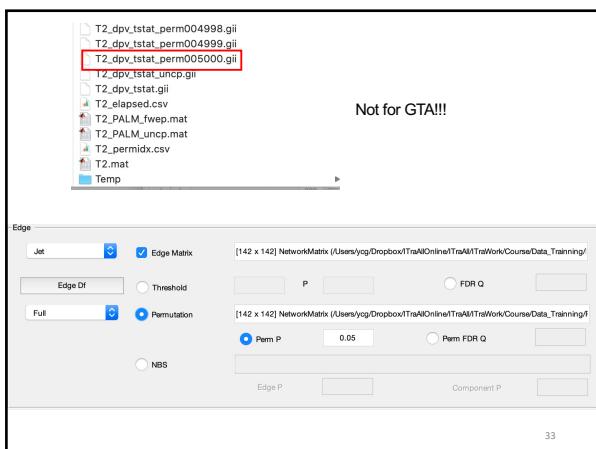
30



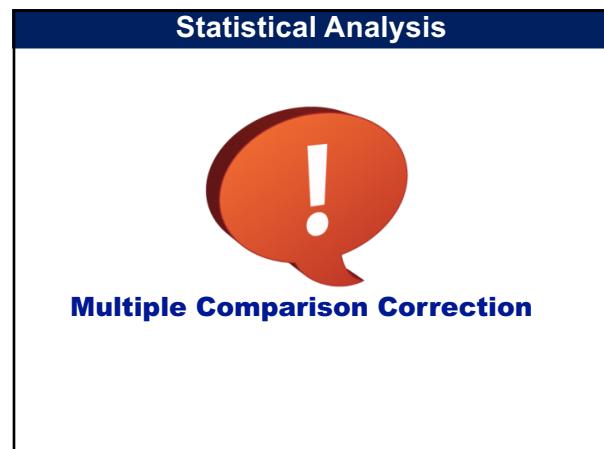
31



32



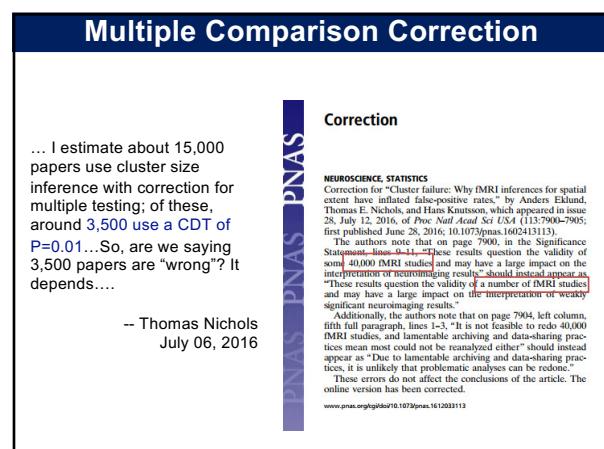
33

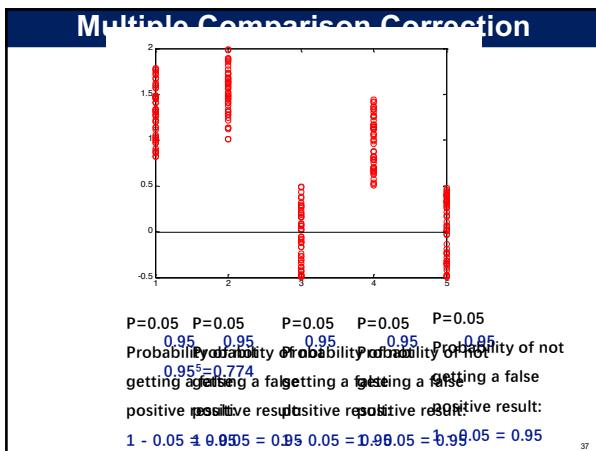


34

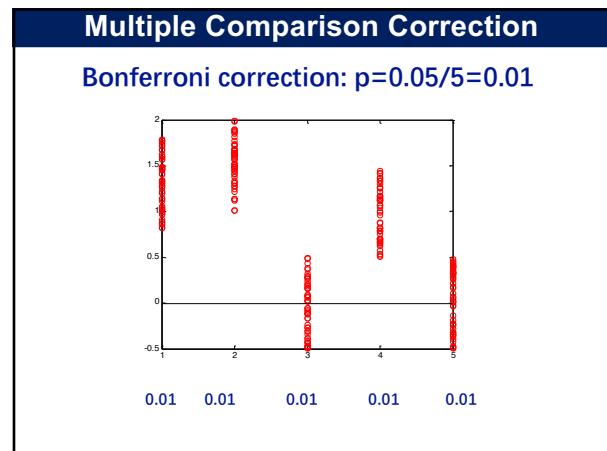


35

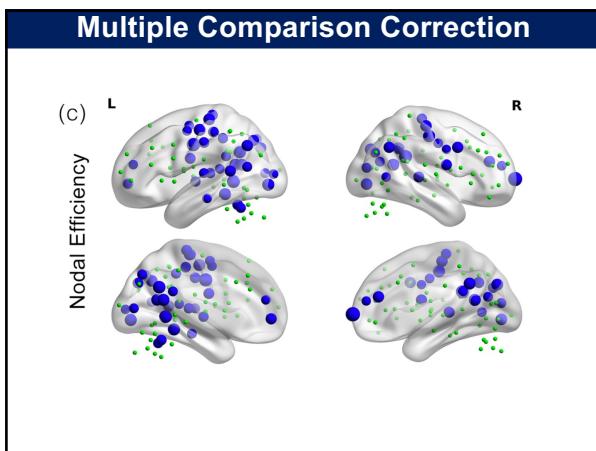




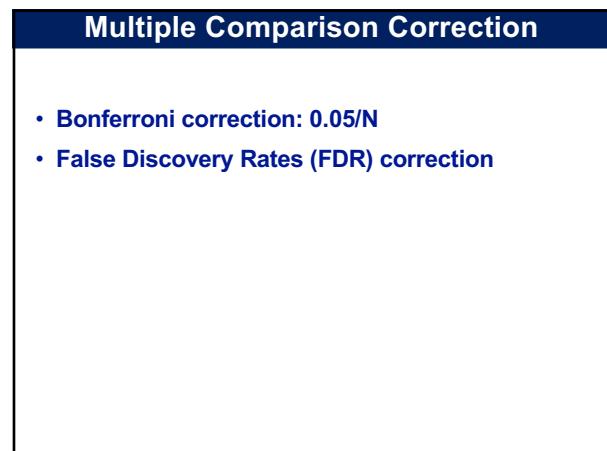
37



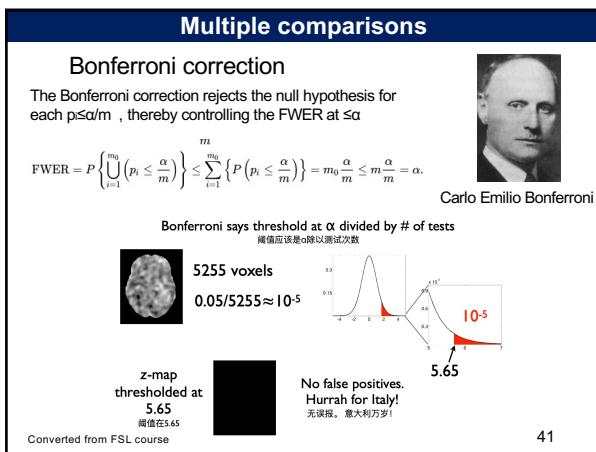
38



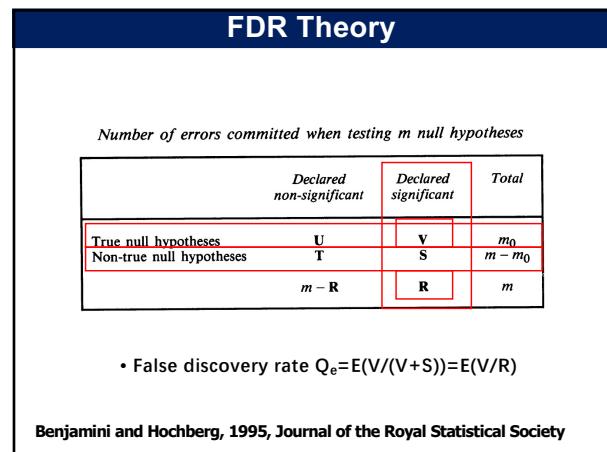
39



40



41

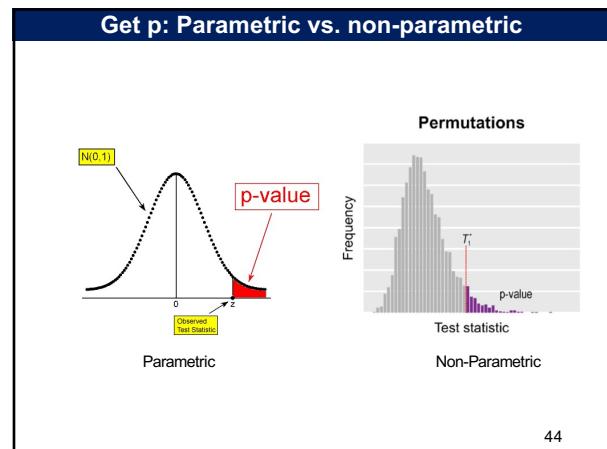


42

## FDR Theory

- Let  $H_1, \dots, H_m$  be the null hypotheses and  $P_1, \dots, P_m$  their corresponding p-values. Order these values in increasing order and denote them by  $P_{(1)}, \dots, P_{(m)}$ . For a given  $q$ , find the largest  $k$  such that  $P_{(k)} \leq q/m$ .
- Then reject (i.e. declare positive) all  $H_{(i)}$  for  $i = 1, \dots, k$ .

43



44

Parametric, can be used for FDR

File list:  
 T2\_dpv\_tstat\_perm004998.gii  
 T2\_dpv\_tstat\_perm004999.gii  
 T2\_dpv\_tstat\_perm005000.gii  
 T2\_dpv\_tstat\_uncp.gii  
 T2\_dpv\_tstat.gii  
 T2\_elapsed.csv  
 T2\_PALM\_fwen.mat  
 T2\_permidx.csv  
 T2.mat  
 Temp

Node settings:  
 Edge Sum  
 Node Weight [142 x 1] NodalEfficiency\_AUC (/Users/jcg/Dropbox/ITraAllOnlineITraAllITraWork/CourseData\_Train)  
 Jet  
 Node Label  
 Node Network  
 Network Label  
 Node DF  
 Threshold 0 P   
 FDR Q   
 Full  Permutation  
 Perm P   
 Perm FDR Q

45

Non-Parametric, Permutation-based, can be used for FDR

File list:  
 T2\_dpv\_tstat\_perm004998.gii  
 T2\_dpv\_tstat\_perm004999.gii  
 T2\_dpv\_tstat\_perm005000.gii  
 T2\_dpv\_tstat\_uncp.gii  
 T2\_dpv\_tstat.gii  
 T2\_elapsed.csv  
 T2\_PALM\_fwen.mat  
 T2\_permidx.csv  
 T2.mat  
 Temp

Node settings:  
 Edge Sum  
 Node Weight [142 x 1] NodalEfficiency\_AUC (/Users/jcg/Dropbox/ITraAllOnlineITraAllITraWork/CourseData\_Train)  
 Jet  
 Node Label  
 Node Network  
 Network Label  
 Node DF  
 Threshold 0 P   
 FDR Q   
 Full  Permutation  
 Perm P   
 Perm FDR Q

46

## Outline

- Statistical Analysis
- Results Viewing

47

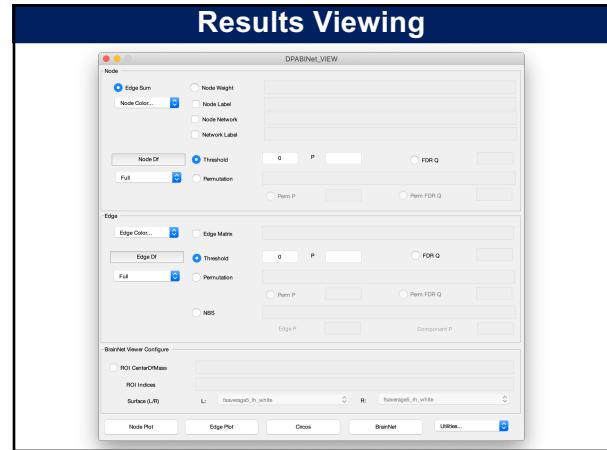


48

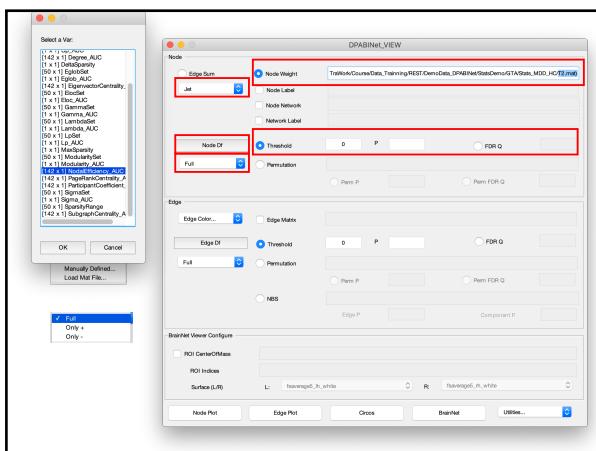
**T2.mat**

Assortativity_AUC	-0.2448
AssortativityPer	1x10 double
Betweenness_AUC	1x142 double
ClusteringCoefficient_AUC	1x50x142 double
ClusteringCoefficientSet	1x5697
Cp_AUC	2.1359
Degree_AUC	1x142 double
DegreeSet	1x50x142 double
DeltaParity	2.9204
Eglib_AUC	-1.7801
EigenvectorCentrality_AUC	1x50x142 double
EigenvectorCentralitySet	1x50x142 double
Elo_AUC	-1.9592
EloSet	1x50 double
Gamma_AUC	0.3687
GammmaSet	1x50 double
Lambda_AUC	1.3723
LambdaSet	1x50 double
Lp_AUC	1.7825
LpSparsity	1x50 double
Modularity_AUC	0.0896
ModularitySet	1x50 double
NodalEfficiency_AUC	1x142 double
NodalEfficiencySet	1x50x142 double
PageRankCentral_AUC	1x142 double
PageRankCentralSet	1x50x142 double
ParticipantCoefficient_AUC	1x142 double
ParticipantCoefficientSet	1x50x142 double
Sigma_AUC	0.2828
SigmaSet	1x50 double
SparsityRange	50x1 double
SpOp	1x1 string
SubgraphCentrality_AUC	1x142 double
SubgraphCentralitySet	1x50x142 double

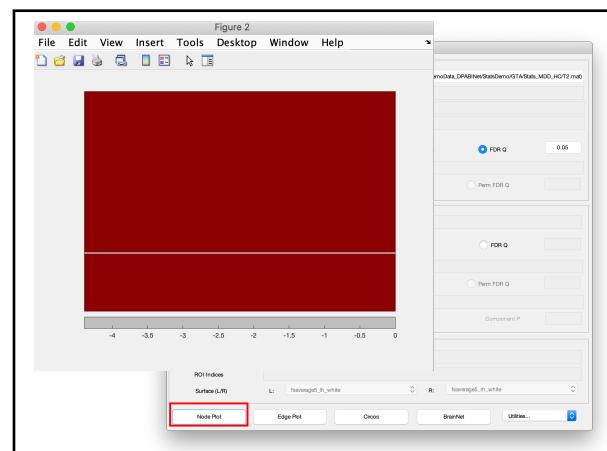
49



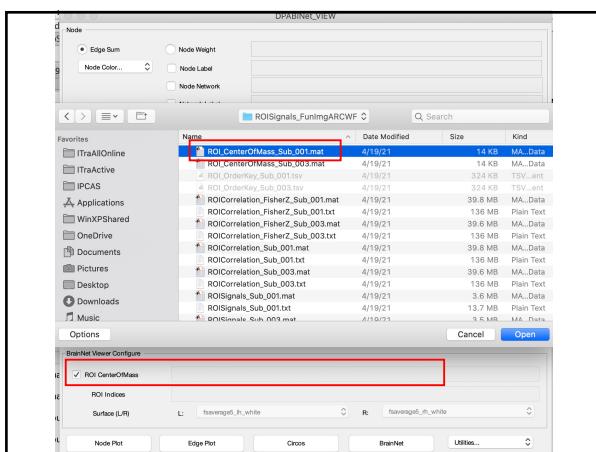
50



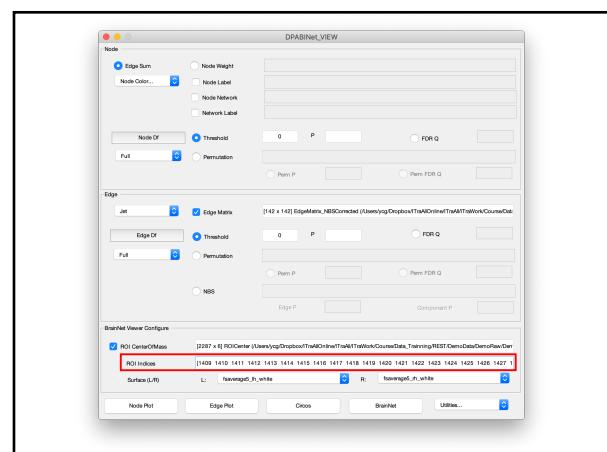
51



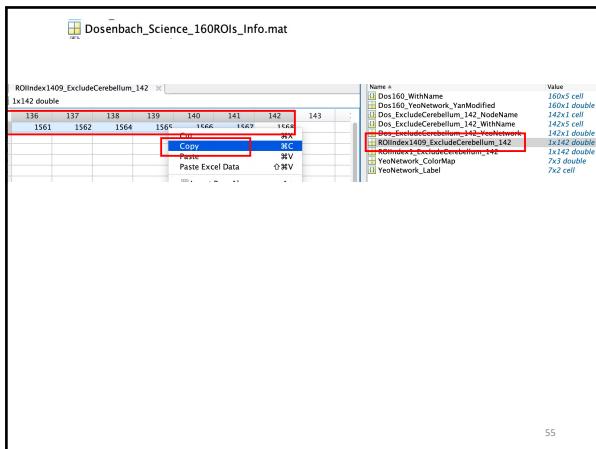
52



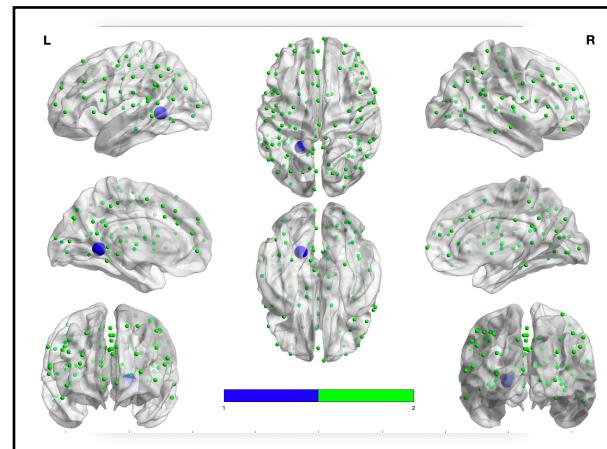
53



54



55



56

### Further Help

**The R-fMRI Course V3.0**

Chao-Gan YAN, Ph.D.  
严超赣  
ygy.yan@gmail.com  
http://rfmri.org  
The R-fMRI Lab  
International Big Data Center for Depression Research  
Institute of Psychology, Chinese Academy of Sciences

<http://rfmri.org/wiki>

**Douyu** The R-fMRI Journal Club

<http://rfmri.org/Course>

Official Account: RFMRILab

57

### DPABI/DPABISurf/DPARSF特训营

**第九届DPABI/DPABISurf/DPARSF  
脑影像基础特训营（云端）通知**  
2021.3.27~3.29

**第一届DPABISurf/DPABINet  
脑网络进阶特训营（北京现场）通知**  
2021.4.24~4.26

定期举办，请关注<http://rfmri.org>

58

### DPABISurf皮层数据预处理业务

DPABISurf V1.4

DPABISurf

特训营学员：99元/例次

59

### Acknowledgments

Chinese Academy of Sciences  
Xi-Nian Zuo  
Wei-Wen Wang  
Fei Luo  
Hangzhou Normal University  
Yu-Feng Zang  
NYU Child Study Center  
F. Xavier Castellanos  
Peking University Sixth Hospital  
Tian-Mei Si  
Jing Liu

- National Natural Science Foundation of China
- National Key R&D Program of China
- Chinese Academy of Sciences

Funding

60

59

10

**Thanks for your attention!**

61

61