




认知心理学进阶第十二讲： 功能磁共振数据组分析

严超赣

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http://rfmri.org/yan

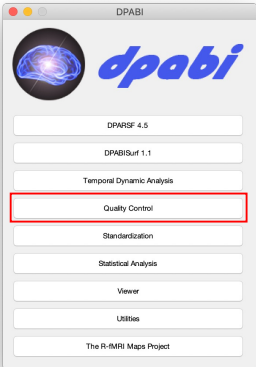
清华大学心理与认知科学系

1

Outline

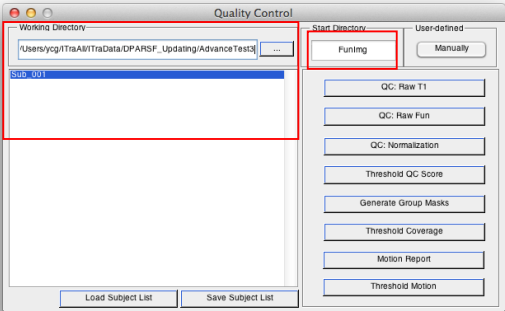
- ➔ • Quality Control
- Statistical Analysis

2



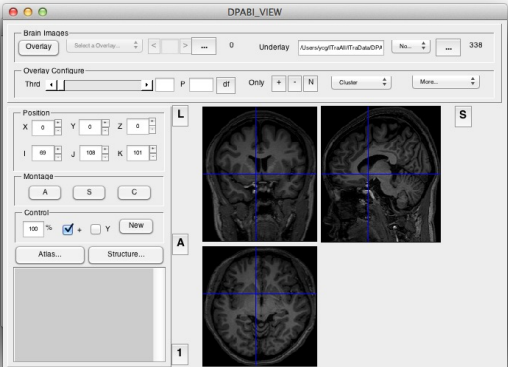
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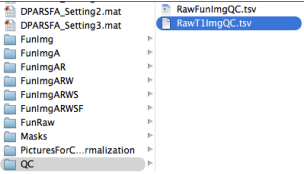
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Quality Control



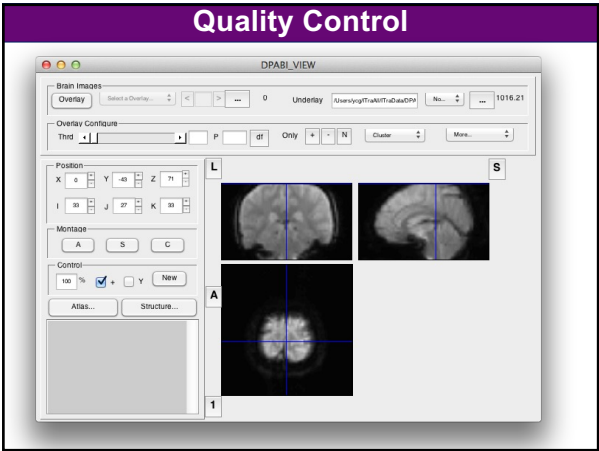
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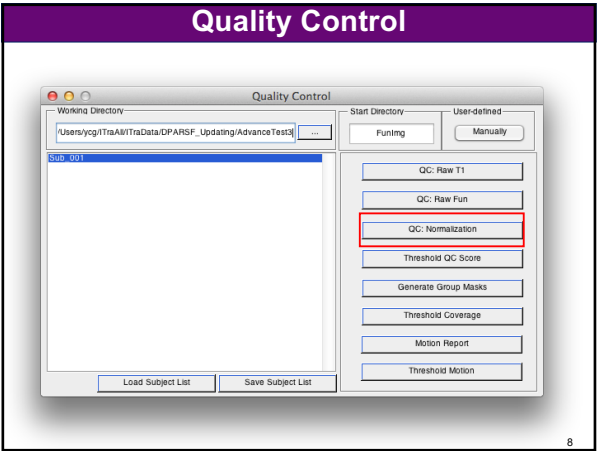


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3				

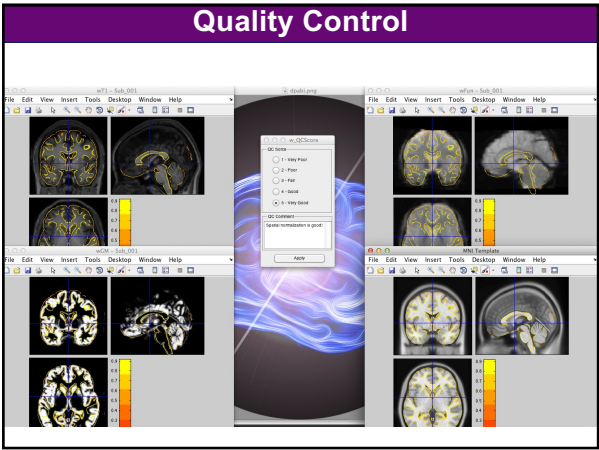
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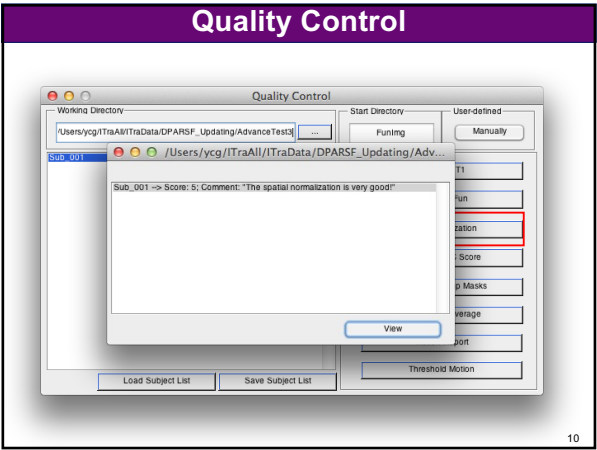
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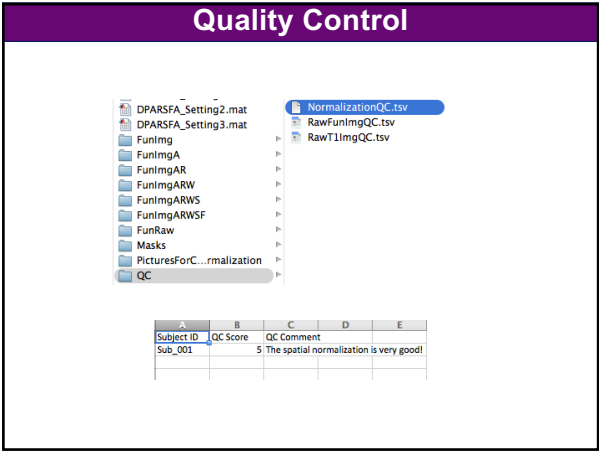
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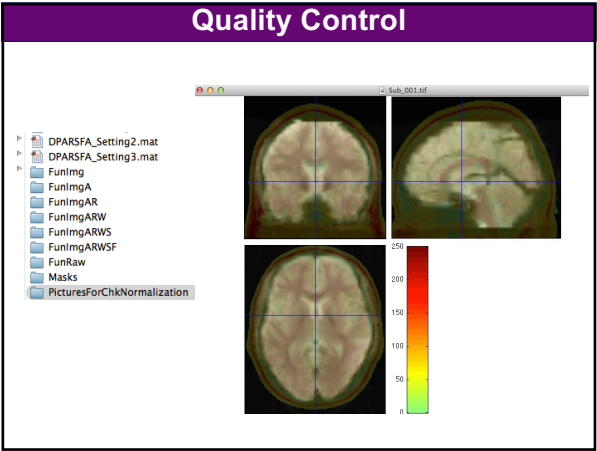
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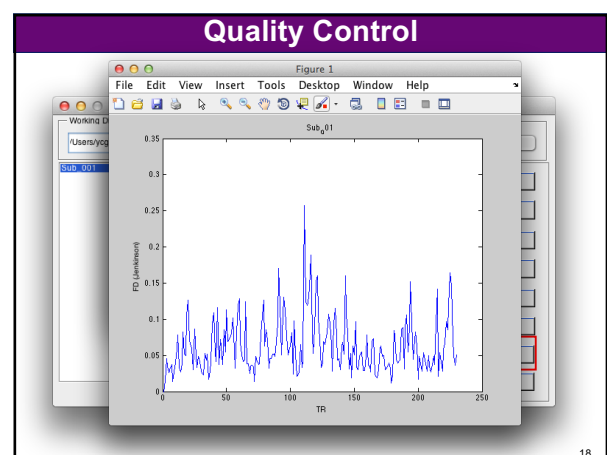
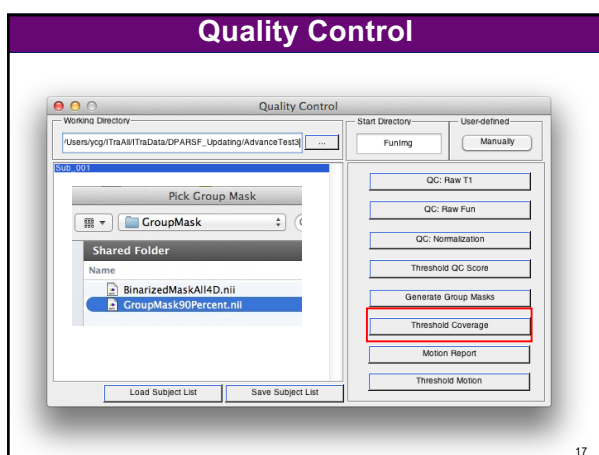
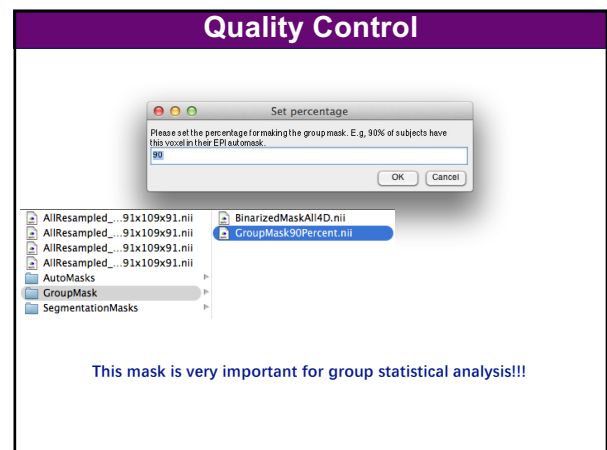
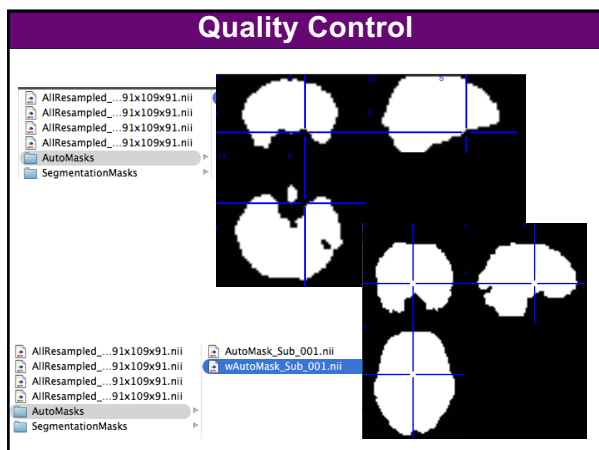
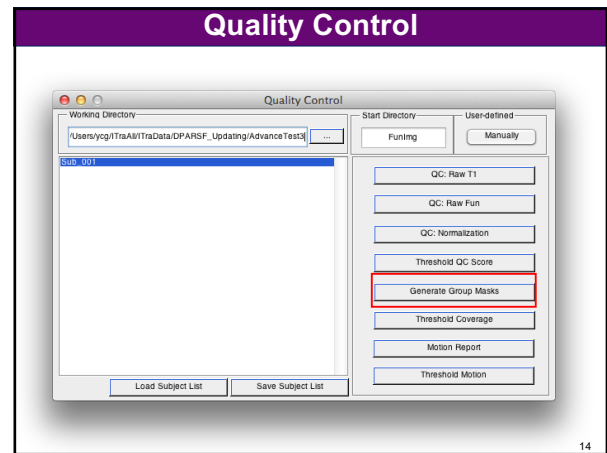
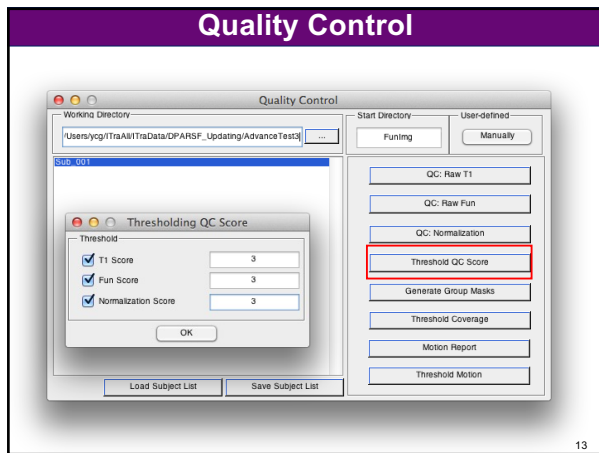
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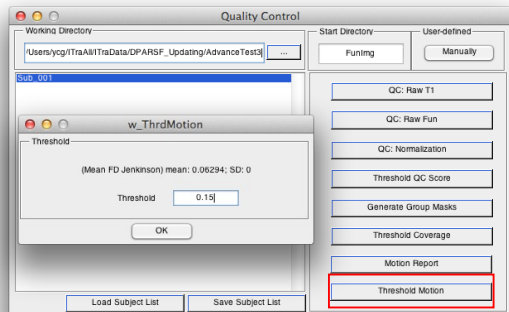
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Quality Control



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Quality Control

- Using the visual inspection step within DPARSF, subjects showing severe head motion in the T1 image and subjects showing extremely poor coverage in the functional images, as well as subjects showing bad registration were excluded
- Subjects with overlap with the group mask (voxels present at least 90% of the participants) less than 2*SD under the group mean overlap (threshold: 92.2%) were excluded
- Subjects with motion (Mean FD Jenkinson greater than 2*SD above the group mean motion (threshold: 0.192) were excluded



Yan et al., 2013, Neuroimage

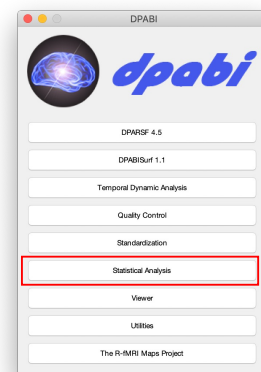
Full Length Article
Standardizing the intrinsic brain: Towards robust measurement of inter-individual variation in 1000 functional connectomes
Chao-Gan Yan^{1,2,3,*}, R. Cameron Craddock^{4,5}, Xi-Nian Zuo⁶, Yu-Feng Zang⁷, Michael P. Milham^{4,5,8}

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Outline

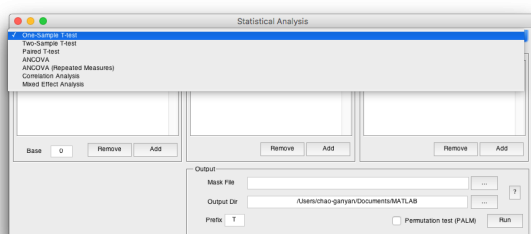
- Quality Control
- ➔ • Statistical Analysis

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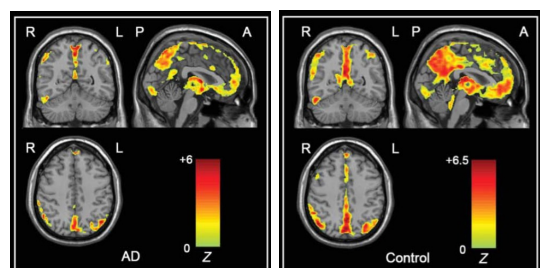
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Statistical Analysis



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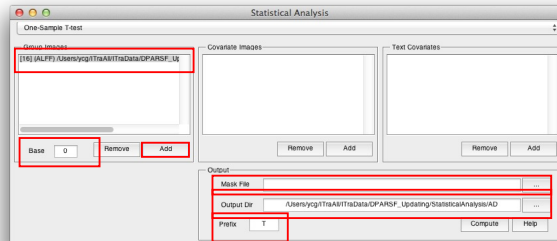
One-Sample T-Test



Wang*, Yan* et al., 2011, Hum Brain Mapp

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One-Sample T-Test



0 for z* images
1 for m* images

T Statistic Image

Group Mask

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Two-Sample T-Test

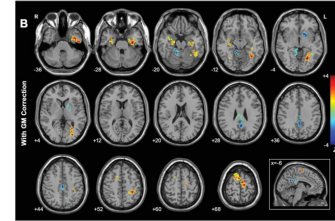
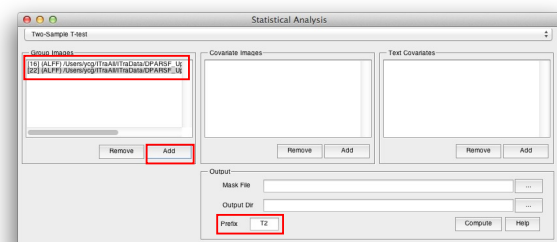


Figure 3. A: Z-statistical difference maps between the AD patients and healthy elderly (without GH correction). The AD patients showed significantly decreased ALFF in the bilateral PCC/PCu, left UN, left OFG, and right ALC, and increased ALFF in the bilateral PHG, bilateral HIp, bilateral SPG, bilateral SPA, left FG, left PCCG, left ITG, and left STG. For the details of the regions, see Table II. The statistical threshold was set at $|Z| > 1.96$ ($P < 0.05$) and cluster size $> 1,404 \text{ mm}^3$, which corresponded to a corrected $P < 0.05$. Of note, we showed the two-sample t-test results with a mask showing significant group difference in the ANCOVA analysis (Fig. 2A). B: Right L, left L. [Color figure can be viewed in the online issue, which is available at www.intellect.com.]

Wang*, Yan* et al., 2011, Hum Brain Mapp

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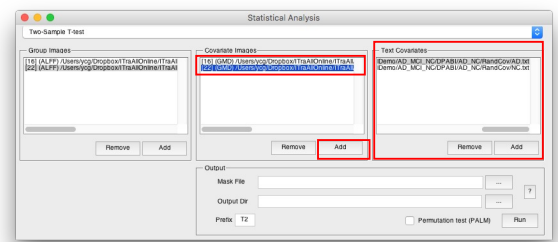
Two-Sample T-Test



T Statistic Image: positive corresponds to the mean of Group 1 is greater than the mean of Group 2

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Two-Sample T-Test



Two-Sample T-Test with covariates: e.g. gray matter proportion images (e.g. head motion (mean FD), age, sex etc.)

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Paired T-Test

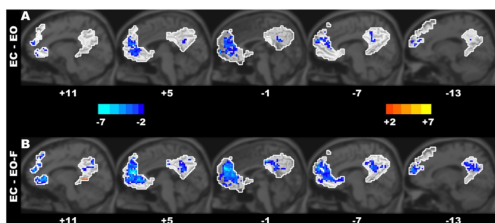
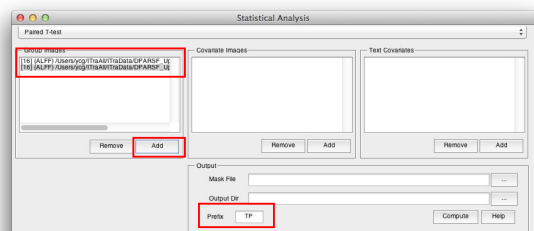


Figure 3. The between-condition differences of the ALFF within the DMN. The ALFF differences were found between the EC and EO conditions (A), and between the EC and EO-F conditions (B). The areas in the white contours denote the ROIs within the DMN. The numbers below the images refer to the x coordinates in the Talairach and Tournoux space. The statistical threshold was set at $|p| > 2.053$ ($P < 0.05$) and cluster size $> 486 \text{ mm}^3$, which corresponds to a corrected $P < 0.05$. doi:10.1371/journal.pone.0005743.g003

Yan et al., 2009, PLoS ONE

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Paired T-Test

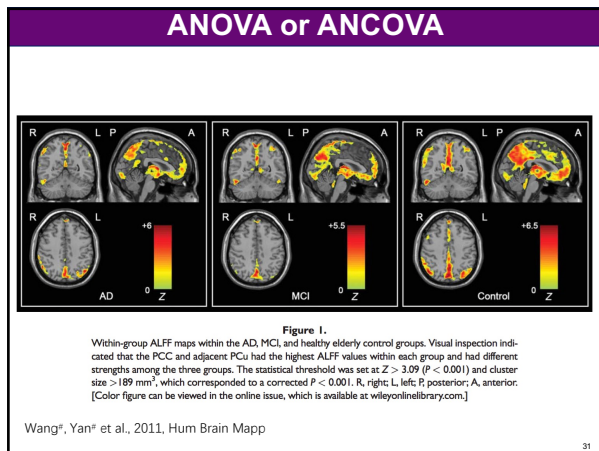


Condition 1 – Condition 2

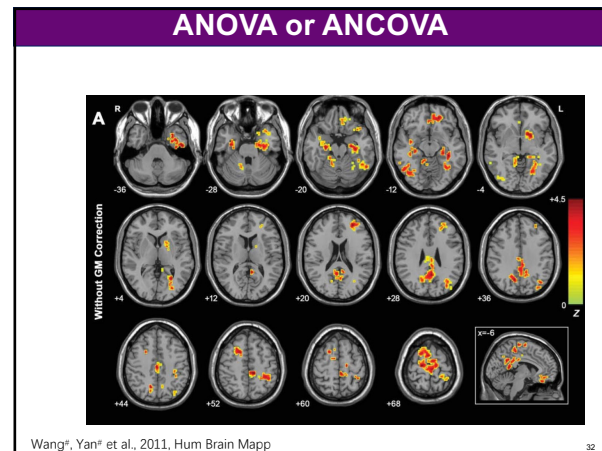
T Statistic Image

Please make sure the correspondence

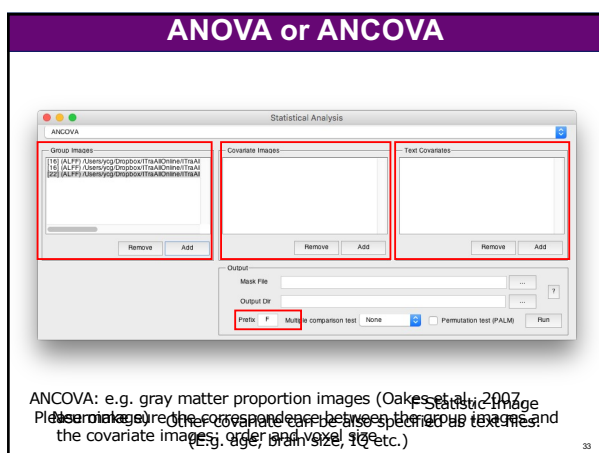
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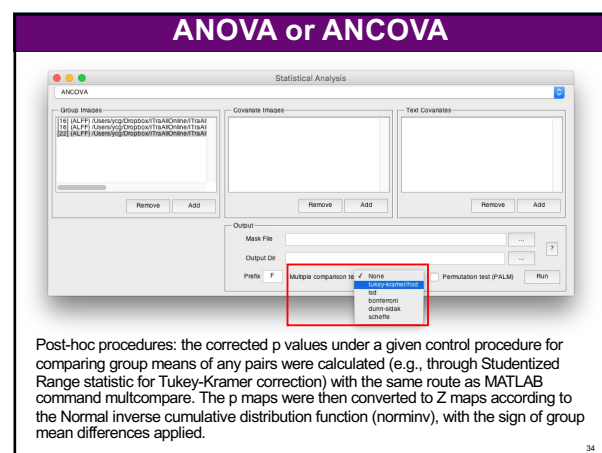
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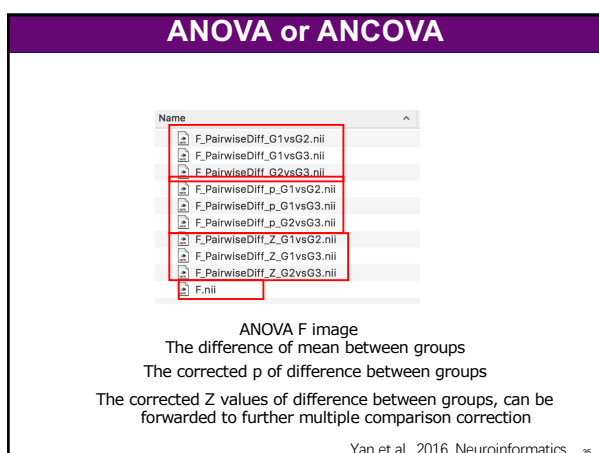
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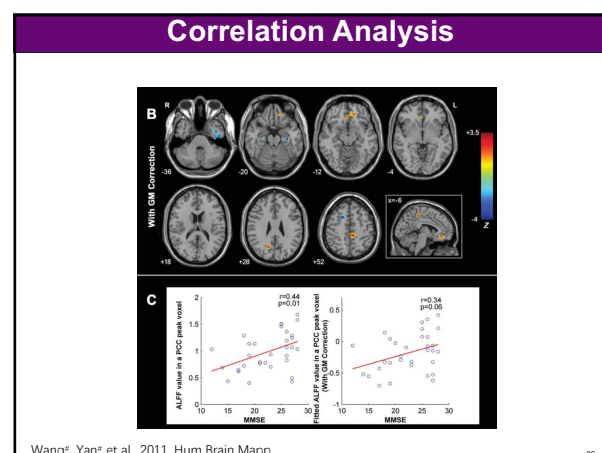
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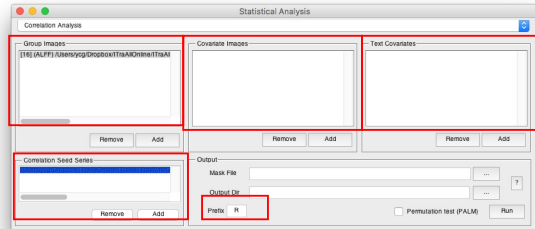


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Correlation Analysis

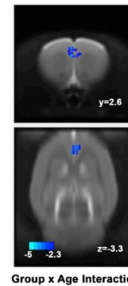


The imaging measure:
ALFF maps

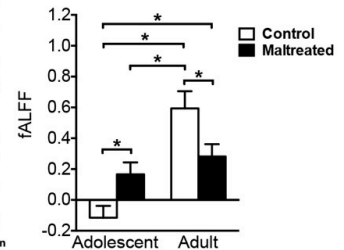
Traits: e.g.
MMSE.txt
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...

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Mixed Effect Analysis



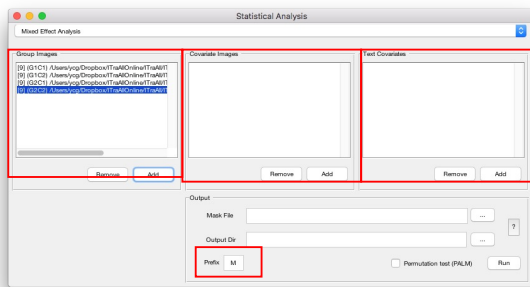
fALFF of MPFC Cingulate



Yan et al., 2016. Translational Psychiatry

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Mixed Effect Analysis



The imaging measure
should be:
Group1Condition1
Group1Condition2
Group2Condition1
Group2Condition2

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Mixed Effect Analysis

- * **_ConditionEffect_T.nii** - the T values of condition differences (corresponding to the first condition minus the second condition) (WithinSubjectFactor)
- * **_Interaction_F.nii** - the F values of interaction (BetweenSubjectFactor by WithinSubjectFactor)
- * **_Group_TwoT.nii** - 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)

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Statistical Analysis

```
function [b_OLS_brain, t_OLS_brain, TF_FactorContrast_brain, r_OLS_brain, Header] = y_GroupAnalysis_Image(DependentVolume, Predictor, OutputName, MaskFile, Covolume, TF_Flag, IsOutputResidual, Header)
% Perform regression analysis
% Input:
% DependentVolume - 4D data matrix (DimXDimYDimZDimTimePoints) or the directory of 3D image data file or the filename of one 4D
% Predictor - 1xN Predictors, N (subjects) by M (traits). SHOULD INCLUDE the CONSTANT column if needed. The program will not add constant
% OutputName - the output name. (should not have extension such as .img, .nii)
% MaskFile - the mask file.
% Covolume (optional) - 4D data matrix (DimXDimYDimZDimTimePoints) or the directory of image covariates, in which the files should be
% Contrast (optional) - Contrast for T-test or F-test. Input as matrix.
% TF_Flag (optional) - 'T' or 'F'. Specify if T-test or F-test need to be performed for the contrast
% IsOutputResidual (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 120823.
% The Nathan Kline Institute for Psychiatric Research, 148 Old Orangeburg Road, Orangeburg, NY 10962, USA
% Child Mind Institute, 465 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.yan@gmail.com

[DPABI_Dir]/StatisticalAnalysis/y_GroupAnalysis_Image.m
```

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Statistical Analysis



Attention!!!

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Statistical Analysis

```
function [b_OLS_brain, t_OLS_brain, TF_Contrast_brain, r_OLS_brain, Header] = y_GroupAnalysis_Image(DependentVolume, Predictor, OutputName,
function [b_OLS_brain, t_OLS_brain, TF_Contrast_brain, r_OLS_brain, Header] = y_GroupAnalysis_Image(DependentVolume, Predictor, OutputName,
% Perform regression analysis
% Input:
% DependentVolume - 4D data matrix (DimXDimYDimZDimTimePoints) or the directory of 3D image data file or the filename of one 4D
% Predictor - the Predictors M (subjects) by N (traits). SHOULD INCLUDE the CONSTANT column if needed. The program will not add constant
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% MaskFile - the mask file.
% Covolume (optional) - 4D data matrix (DimXDimYDimZDimTimePoints) or the directory of image covariates, in which the files should be
% Contrast (optional) - Contrast for T-test for F-test. IncoVX matrix.
% TF_Flag (optional) - 'T' or 'F'. Specify if T-test or F-test need to be performed for the contrast
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% 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 128823.
% The Nathan Kline Institute for Psychiatric Research, 148 Old Orangeburg Road, Orangeburg, NY 10962, USA
% 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.yang@gmail.com
```

{DPABI_Dir}/StatisticalAnalysis/y_GroupAnalysis_Image.m
Smoothness estimation based on the 4D residual is built in this function!!!

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Statistical Analysis

<http://rfmri.org/DemoData>
{Download}/ProcessingDemoData/StatisticalDemo/AD_MCI_NC/

ALFF: AD – NC Two Sample T Test:

- Applied smooth kernel in preprocessing: [4 4 4]
- Smooth kernel estimated on 4D residual: [6.77 6.88 6.71]
- Smooth kernel estimated on statistical image (T to Z, as in easythresh): [6.90 7.33 6.94]

ReHo: AD – NC Two Sample T Test:

- Applied smooth kernel in preprocessing: [4 4 4]
- Smooth kernel estimated on 4D residual: [8.10 8.50 7.93]
- Smooth kernel estimated on statistical image (T to Z, as in easythresh): [8.33 8.94 8.24]

Thus, only using smooth kernel applied in preprocessing is NOT sufficient!!!



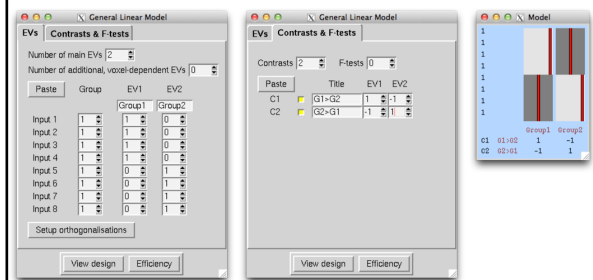
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Statistical Analysis

```
function [b_OLS_brain, t_OLS_brain, TF_Contrast_brain, r_OLS_brain, Header] = y_GroupAnalysis_Image(DependentVolume, Predictor, OutputName,
function [b_OLS_brain, t_OLS_brain, TF_Contrast_brain, r_OLS_brain, Header] = y_GroupAnalysis_Image(DependentVolume, Predictor, OutputName,
% Perform regression analysis
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% IsOutputResidual (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 128823.
% The Nathan Kline Institute for Psychiatric Research, 148 Old Orangeburg Road, Orangeburg, NY 10962, USA
% 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.yang@gmail.com
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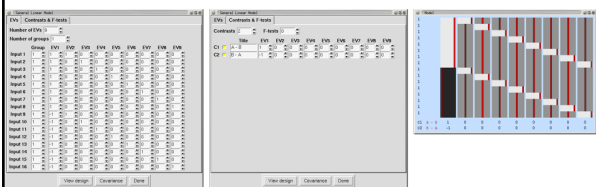
Statistical Analysis



<http://fsl.fmrib.ox.ac.uk/fsl/fslwiki/GLM>

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Statistical Analysis



<http://fsl.fmrib.ox.ac.uk/fsl/fslwiki/GLM>

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Statistical Analysis



Multiple Comparison Correction

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Multiple Comparison Correction



The last 15 years of fMRI research might be totally useless.

Due to the recent discovery of an fMRI bug, about 40,000 papers on brain research may now be invalid.

PNAS, CrossMark, Eklund et al., 2016

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Multiple Comparison Correction

... I estimate about 15,000 papers use cluster size inference with correction for multiple testing; of these, around 3,500 use a CDT of $P=0.01$... So, are we saying 3,500 papers are "wrong"? It depends....

-- Thomas Nichols
July 06, 2016

Correction

NEUROSCIENCE, STATISTICS

Correction for "Cluster failure: Why fMRI inferences for spatial extent have inflated false-positive rates," by Anders Eklund, Thomas E. Nichols, and Hans Knutson, which appeared in issue 28, July 12, 2016, of *Proc. Natl. Acad. Sci. USA* (113:7900-7905; first published June 28, 2016; 10.1073/pnas.1602413113).

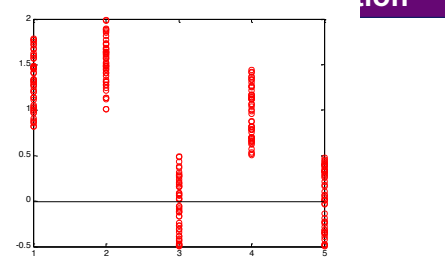
The authors note that on page 7900, in the Significance Statement, lines 4-5, "these results question the validity of some 40,000 fMRI studies" should instead appear as "These results question the validity of a number of fMRI studies and may have a large impact on the interpretation of weakly significant neuroimaging results."

Additionally, the authors note that on page 7904, left column, fifth full paragraph, lines 1-3, "It is not feasible to redo 40,000 fMRI studies, and lamentable archiving and data-sharing practices mean most could not be reanalyzed either" should instead appear as "Due to lamentable archiving and data-sharing practices, it is unlikely that problematic analyses can be redone." These errors do not affect the conclusions of the article. The online version has been corrected.

www.pnas.org/cgi/doi/10.1073/pnas.1612031113

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Multiple Comparison Correction

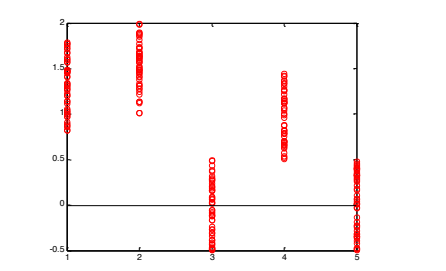


$P=0.05$ $P=0.05$ $P=0.05$ $P=0.05$ $P=0.05$
 Probability of not getting a false positive result: $1 - 0.05 = 0.95$ $1 - 0.05 = 0.95$ $1 - 0.05 = 0.95$ $1 - 0.05 = 0.95$ $1 - 0.05 = 0.95$
 Probability of getting a false positive result: $0.95 \times 0.95 \times 0.95 \times 0.95 \times 0.95 = 0.774$

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Multiple Comparison Correction

Bonferroni correction: $p=0.05/5=0.01$



0.01 0.01 0.01 0.01 0.01

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Multiple Comparison Correction

- False Discovery Rates (FDR) correction
- Family-Wise Error (FWE) correction
 - Bonferroni correction: $0.05/5=0.01$
 - Gaussian Random Field theory correction
 - Monte Carlo simulations (AlphaSim)
 - Threshold-Free Cluster Enhancement
 - Permutation test

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FDR Theory

Number of errors committed when testing m null hypotheses

	Declared non-significant	Declared significant	Total
True null hypotheses	U	V	m_0
Non-true null hypotheses	T	S	$m - m_0$
	$m - R$	R	m

• False discovery rate $Q_e = E(V/(V+S)) = E(V/R)$

Benjamini and Hochberg, 1995, *Journal of the Royal Statistical Society*

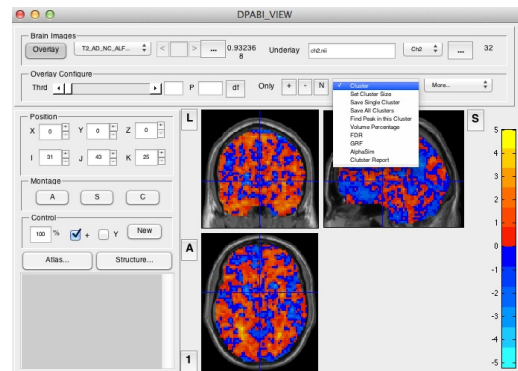
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FDR Theory

- 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 $P_{(k)} \leq kq/m$.
- Then reject (i.e. declare positive) all $H_{(i)}$ for $i = 1, \dots, k$.

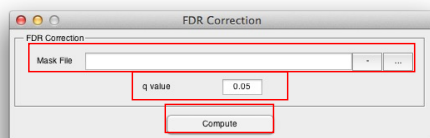
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FDR Theory



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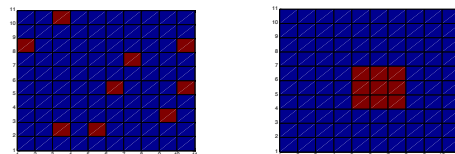
FDR Theory



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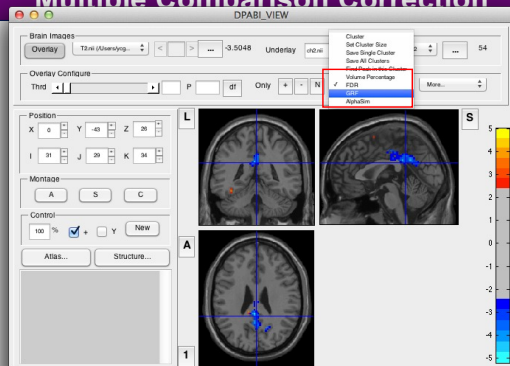
Multiple Comparison Correction

Gaussian Random Field Theory Correction Monte Carlo simulations (AlphaSim)



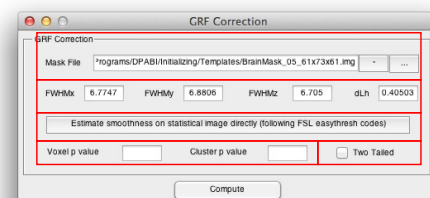
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Multiple Comparison Correction



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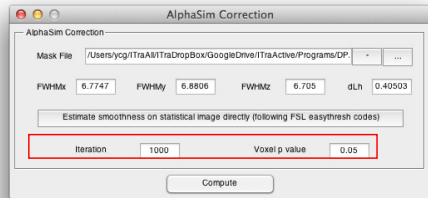
Multiple Comparison Correction



Voxel $Z > 2.3$, Cluster $P < 0.05$, Two One-Tailed Corrections:
equivalent to
Voxel $P < 0.0214$, Cluster $P < 0.1$, Two Tailed.

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Multiple Comparison Correction



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Multiple Comparison Correction

Cl Size	Frequency	Cum Prop p/Voxel	Max Freq	Alpha	
1	235971	0.619898	0.009613	0	1.000000
2	76150	0.819945	0.006282	0	1.000000
3	32297	0.904789	0.004131	0	1.000000
4	15940	0.946664	0.002763	0	1.000000
5	8476	0.968930	0.001863	0	1.000000
6	4786	0.981503	0.001265	1	1.000000
7	2767	0.988772	0.000860	19	0.999000
8	1606	0.992991	0.000586	51	0.980000
9	1011	0.995647	0.000405	127	0.929000
10	585	0.997184	0.000276	132	0.802000
11	391	0.998211	0.000194	172	0.670000
12	236	0.998831	0.000133	146	0.498000
13	164	0.999262	0.000093	107	0.352000
14	98	0.999519	0.000063	78	0.245000
15	69	0.999701	0.000043	61	0.167000
16	37	0.999798	0.000029	30	0.106000
17	22	0.999856	0.000020	22	0.076000
18	22	0.999913	0.000015	21	0.054000
19	11	0.999942	0.000010	11	0.033000
20	7	0.999961	0.000007	7	0.022000
21	5	0.999974	0.000005	5	0.015000
22	5	0.999987	0.000003	5	0.010000
23	4	0.999997	0.000002	4	0.005000
24	1	1.000000	0.000000	1	0.001000

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Threshold-Free Cluster Enhancement (TFCE)

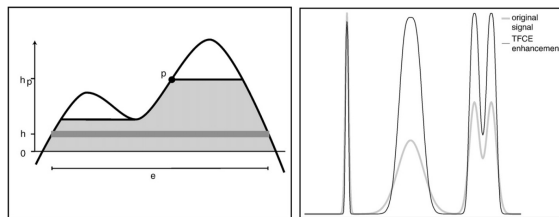
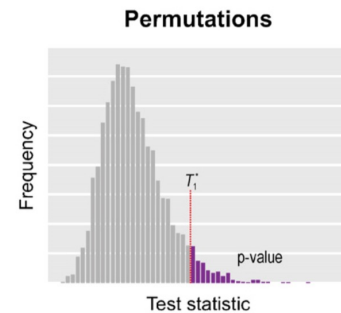


Fig. 1. Illustration of the TFCE approach. Left: the TFCE score at voxel p is given by the sum of the scores of all incremental supporting sections (one such is shown as the dark grey head) within the area of "support" of p (light grey). The score for each section is a simple function of its height h and extent e . Right: example input image and TFCE-enhanced output. The input contains a focal, high signal, a much more spatially extended, lower, signal and a pair of overlapping signals of intermediate extent and height. The TFCE output has the same maximal values for all three cases, and preserves the distinct local maxima in the third case.

Smith et al., 2009. Neuroimage

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Permutation Test

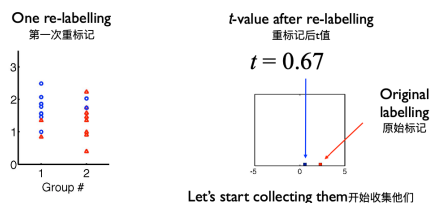


Winkler et al., 2016. Neuroimage

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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



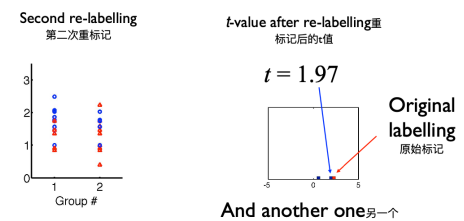
Winkler et al., 2016. Neuroimage; Converted from FSL course

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Non-parametric: permutation

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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



Converted from FSL course

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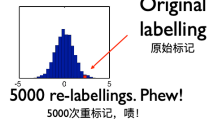
Non-parametric: permutation

- 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

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
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. t8dfp (x 2.27) = 1.79%

C.f. $p(x \geq 2.27) = 1.79\%$ 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. t8dfp (x 2.27) = 1.79%



Converted from FSL course

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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

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Multiple Comparison Correction

Cluster failure: Why fMRI inferences for spatial extent have inflated false-positive rates

Anders Eklund^{a,b,c,1}, Thomas E. Nichols^{a,*}, and Hans Knutsson^{a,c}

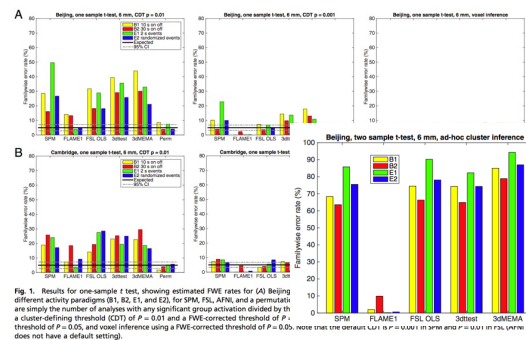
^aDivision of Medical Informatics, Department of Biomedical Engineering, Linköping University, S-581 85 Linköping, Sweden; ^bDivision of Statistics and Machine Learning, Department of Computer and Information Science, Linköping University, S-581 83 Linköping, Sweden; ^cCenter for Medical Image Science and Visualization, Linköping University, S-581 83 Linköping, Sweden; ^dDepartment of Statistics, University of Warwick, Coventry CV4 7AL, United Kingdom; and ^eWMG, University of Warwick, Coventry CV4 7AL, United Kingdom

Edited by Emery N. Brown, Massachusetts General Hospital, Boston, MA, and approved May 17, 2016 (received for review February 12, 2016)

Eklund et al., 2016. PNAS

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Multiple Comparison Correction



Eklund et al., 2016. PNAS

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Multiple Comparison Correction

12 May 2015, RW Cox, 3dClustSim, level 2 (MINOR), type 5 (MODIFY)
Eliminate edge effects of smoothing by padding and unpadding

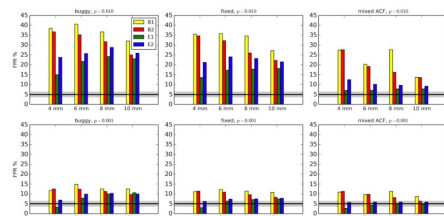
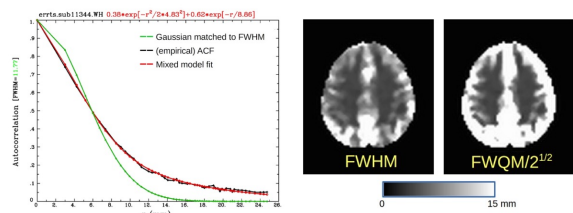


Figure 1. False Positive Rates (FPRs) for various software scenarios, with 1000 2-sample t-tests (as in [1,2]) using 20 subjects' data in each sample. "buggy" and "fixed" means the cluster-size thresholds were selected using the Gaussian shape model with the FWHM being the median of the 40 individual subject's values; "buggy" via 3dClustSim before the bug fix; "fixed" via 3dClustSim after the bug fix. "mixed ACF" means the cluster-size threshold was selected using the (Eq. 1) model for spatial correlation of the noise, with the a.b.c. parameters being the median of the 40 individual subject's values (estimated via program 3dFWHMx). Two different per-voxel p-value thresholds (1-sided tests, as used in [2]) are shown. The black line shows the nominal 5% false positive rate (out of 1000 trials), and the gray band shows its theoretical 95% confidence interval, 3.65-6.35%. As in [2], different smoothing values were tested (4-10 mm). B1 = 10 s block; B2 = 30 s block; E1 = regular event related; E2 = randomized event related.

Cox et al., 2016. bioRxiv

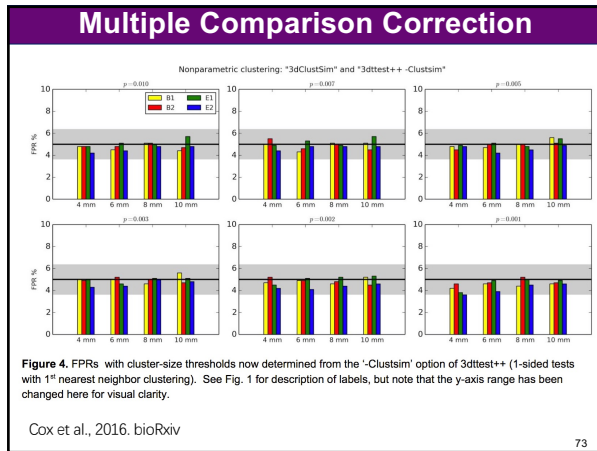
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Multiple Comparison Correction

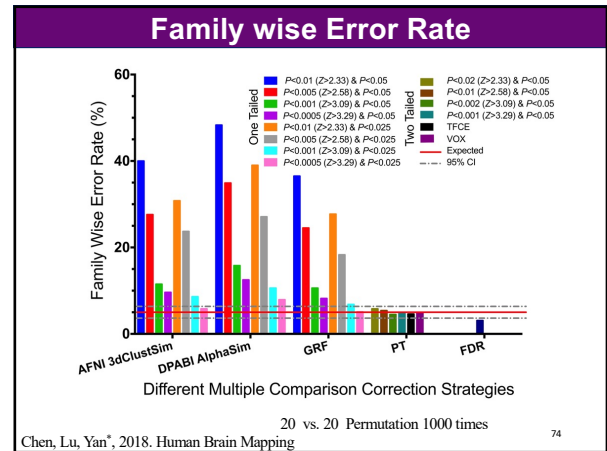


Cox et al., 2016. bioRxiv

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Family wise Error Rate

TABLE I. FWER and cluster size of ALFF (smoothness: 7.94 × 7.31 × 6.86) without GSR under corrections of GRF Theory, AFNI 3dClustSim, and DPABI AlphaSim

(One-tailed twice)		AFNI 3dClustSim		DPABI AlphaSim		GRF	
Voxel threshold	Cluster threshold	FWER	Cluster size	FWER	Cluster size	FWER	Cluster size
P < 0.01 (Z > 2.33)	P < 0.05	40.0%	66.05 ± 0.73	48.3%	60.24 ± 1.68	36.5%	69.35 ± 1.09
P < 0.005 (Z > 2.58)	P < 0.05	27.6%	43.59 ± 0.42	34.9%	39.45 ± 1.13	24.5%	46.70 ± 0.75
P < 0.001 (Z > 3.09)	P < 0.05	11.5%	19.98 ± 0.34	15.8%	18.40 ± 0.61	10.6%	21.29 ± 0.46
P < 0.0005 (Z > 3.29)	P < 0.05	9.6%	14.53 ± 0.25	12.5%	13.93 ± 0.54	8.2%	15.82 ± 0.39
P < 0.01 (Z > 2.33)	P < 0.025	30.8%	74.50 ± 1.14	39.0%	67.72 ± 2.36	27.7%	78.96 ± 1.24
P < 0.005 (Z > 2.58)	P < 0.025	23.7%	47.01 ± 0.59	27.1%	44.48 ± 1.60	18.3%	53.48 ± 0.85
P < 0.001 (Z > 3.09)	P < 0.025	8.6%	22.63 ± 0.75	10.6%	21.00 ± 0.67	6.8%	24.94 ± 0.41
P < 0.0005 (Z > 3.29)	P < 0.025	5.8%	17.33 ± 0.22	7.9%	16.03 ± 0.71	5.1%	18.51 ± 0.50

20 vs. 20 Permutation 1000 times

Chen, Lu, Yan*, 2018. Human Brain Mapping

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Family wise Error Rate

TABLE II. FWER under correction of three kinds of cluster-based correction with the strictest threshold, 6 versions of PT-based correction as well as FDR correction

Voxel threshold	Cluster threshold	FWER									
		ALFF	fALFF	ReHo	DC	VMHC	ALFF with GSR	fALFF with GSR	ReHo with GSR	DC with GSR	VMHC with GSR
Smoothness (mm, x/y/z)		7.94 × 7.34 × 9.56 × 7.86 × 6.83 × 7.99 × 7.32 × 9.24 × 8.06 × 6.11 × 11.88 ×	7.31 × 7.42 × 8.72 × 7.97 × 6.87 × 7.31 × 7.41 × 8.56 × 8.16 × 6.41 × 11.57 ×								
AFNI 3dClustSim (one-tailed)	$P < 0.0005$ (Z > 3.29)	8.8%	8.8%	8.8%	8.8%	8.8%	8.8%	8.8%	8.8%	8.8%	8.8%
DPABI AlphaSim (one-tailed)	$P < 0.0005$ (Z > 3.29)	7.9%	8.3%	8.5%	10.2%	9.0%	7.8%	7.7%	7.8%	8.3%	9.6%
GRF (one-tailed)	$P < 0.0005$ (Z > 3.29)	5.1%	5.5%	4.9%	7.4%	5.2%	4.8%	5.9%	5.3%	5.1%	6.4%
PT cluster extent correction (one-tailed)	$P < 0.001$ (Z > 2.58)	2.8%	3.0%	2.8%	3.6%	3.2%	2.8%	3.9%	3.2%	2.8%	3.3%
PT cluster extent correction (two-tailed)	$P < 0.01$ (Z > 2.33)	4.0%	4.0%	4.0%	4.6%	4.0%	3.8%	4.5%	4.0%	4.0%	4.4%
PT TCCE	$P < 0.002$ (Z > 2.58)	4.5%	4.1%	5.3%	4.8%	4.2%	4.5%	5.0%	5.1%	4.7%	4.3%
PT VOX	$P < 0.002$ (Z > 2.58)	4.8%	4.9%	4.5%	4.9%	3.4%	4.3%	4.8%	5.4%	4.2%	3.9%
FDR correction	$P < 0.001$ (Z > 3.09)	4.6%	3.9%	5.7%	5.0%	4.3%	5.3%	4.2%	5.7%	4.7%	4.8%
PT TCCE	$P < 0.001$ (Z > 3.29)	4.9%	4.9%	5.7%	5.0%	4.7%	4.0%	4.3%	5.4%	4.0%	4.6%
PT VOX	$P < 0.001$ (Z > 3.29)	3.1%	3.4%	4.4%	2.4%	3.9%	4.1%	2.8%	3.6%	3.5%	1.6%
FDR correction		3.1%	3.4%	4.4%	2.4%	3.9%	4.1%	2.8%	3.6%	3.5%	1.6%

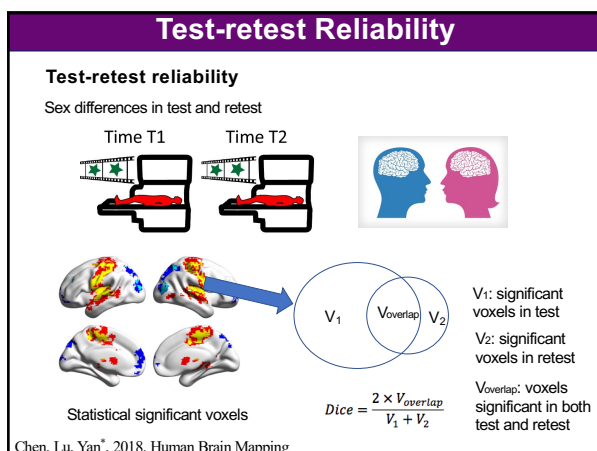
The smoothness in the second row is the estimated effective smoothness of the final metric maps fed to statistical analysis, and was different from the applied smoothness (4 mm FWHM) in pre-processing. The effective smoothness was used in 3 permutations of cluster-based correction (i.e., GRF theory correction, AFNI 3dClustSim and DPABI AlphaSim).

20 vs. 20 Permutation 1000 times

Chen, Lu, Yan*, 2018. Human Brain Mapping

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Test-retest Reliability

TABLE III. Test-retest reliability of sex differences for all R-fMRI metrics with and without GSR under correction of three kinds of cluster-based correction with the strictest threshold, six kinds of PT-based correction and FDR correction, calculated between the first and second sessions in the CORR dataset

	Voxel threshold	Cluster threshold	Test-retest reliability (dice coefficient)									
			ALFF	fALFF	ReHo	DC	VMHC	ALFF with GSR	fALFF with GSR	ReHo with GSR	DC with GSR	VMHC with GSR
AFNI 3dClustSim (one-tailed)	P < 0.0005 (Z > 3.29)	P < 0.025	0.65	0.51	0.50	0.34	0.39	0.64	0.48	0.44	0.28	0.24
DPABI AlphaSim (one-tailed)	P < 0.0005 (Z > 3.29)		0.65	0.51	0.49	0.34	0.39	0.64	0.48	0.45	0.27	0.27
GRF (one-tailed)	P < 0.002 (Z > 2.58)		0.64	0.51	0.50	0.35	0.39	0.65	0.48	0.43	0.28	0.24
PT cluster extent correction (one-tailed)	P < 0.001 (Z > 3.09)	P < 0.05	0.65	0.70	0.56	0.45	0.40	0.62	0.68	0.45	0.30	0.40
PT cluster extent correction (two-tailed)	P < 0.001 (Z > 3.09)	P < 0.05	0.67	0.66	0.52	0.32	0.33	0.60	0.63	0.46	0.27	0.32
PT TCCE	P < 0.001 (Z > 3.09)	P < 0.05	0.63	0.55	0.51	0.36	0.38	0.63	0.52	0.47	0.23	0.32
PT VOX	P < 0.001 (Z > 3.29)	P < 0.05	0.64	0.51	0.48	0.37	0.38	0.64	0.48	0.44	0.28	0.26
FDR correction			0.68	0.75	0.54	0.48	0.44	0.66	0.74	0.44	0.31	0.42
			0.66	0.74	0.46	0.37	0.32	0.62	0.51	0.38	0.24	0.14
			0.64	0.67	0.54	0.39	0.37	0.63	0.64	0.47	0.23	0.29

For test-retest reliability for all the 31 kinds of multiple comparison correction strategies, please see Supporting Information Table S13

➤ Moderate test-retest reliability

➤ ALFF, fALFF, ReHo are better than DC and VMHC

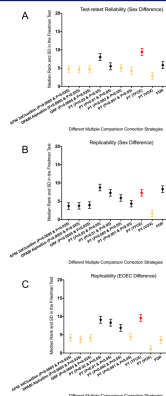
212 M vs. 208 F × 2 times

Chen, Lu, Yan*, 2018. Human Brain Mapping

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PT with TFCE outperforms

Permutation test TFCE, a strict multiple comparison correction strategy, reached the best balance between family-wise error rate (under 5%) and test-retest reliability / replicability

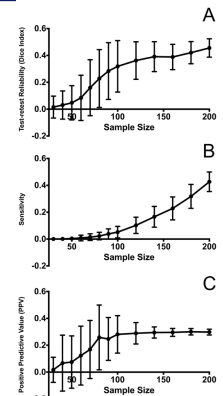


Chen, Lu, Yan*, 2018. Human Brain Mapping

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Sample Size Matters

Randomly draw k subjects from the "SWU 4" site in the CORR dataset, which has two sessions of 116 males and 105 females



Chen, Lu, Yan*, 2018. Human Brain Mapping

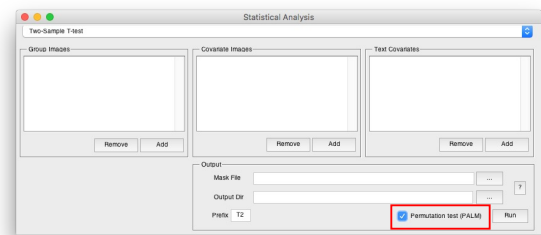
80

Reproducibility of R-fMRI Metrics on the Impact of Different Strategies for Multiple Comparison Correction and Sample Sizes

- Permutation test with TFCE reached the best balance between FWER and reproducibility
- Although R-fMRI indices attained moderate reliabilities, they replicated poorly in distinct datasets (replicability < 0.3 for between-subject sex differences, < 0.5 for within-subject EOEC differences)
- For studies examining effect sizes similar to or even less than those of sex differences, results from a sample size < 80 (40 per group) should be considered preliminary, given their low reliability (< 0.23), sensitivity (< 0.02) and PPV (< 0.26).

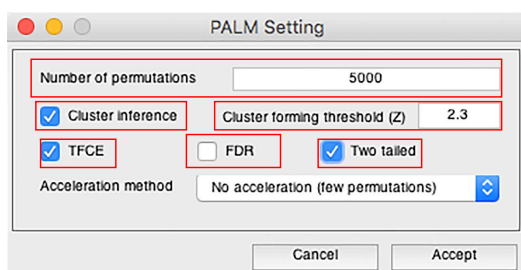
81

Permutation Test



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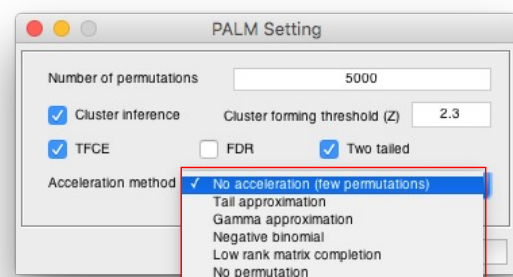
Permutation Test



Based on PALM: Winkler, A.M., Ridgway, G.R., Douaud, G., Nichols, T.E., Smith, S.M., 2016. Faster permutation inference in brain imaging. Neuroimage 141, 502-516.

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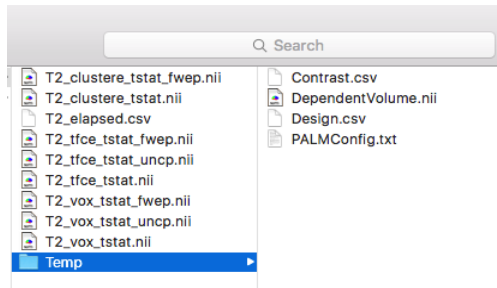
Permutation Test



Based on PALM: Winkler, A.M., Ridgway, G.R., Douaud, G., Nichols, T.E., Smith, S.M., 2016. Faster permutation inference in brain imaging. Neuroimage 141, 502-516.

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Permutation Test



Based on PALM: Winkler, A.M., Ridgway, G.R., Douaud, G., Nichols, T.E., Smith, S.M., 2016. Faster permutation inference in brain imaging. Neuroimage 141, 502-516.

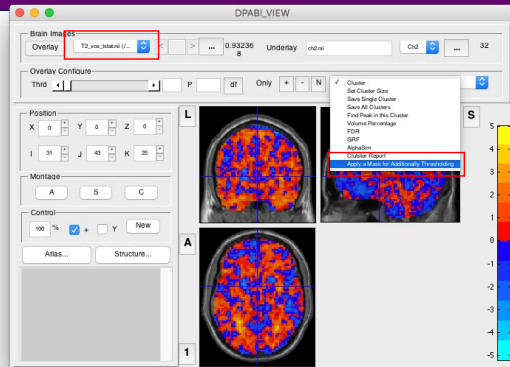
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Permutation Test

1. `_vox_tstat.nii` is the T value of a voxel.
2. `_vox_tstat_uncp.nii` is the p value corresponds to the rank of the observed T value within the permutations FOR A GIVEN VOXEL (the null distribution is the permuted T values of that given voxel). Computing the rank is one of the ways in which the p-value can be obtained (it's then divided by the number of permutations).
3. `_vox_tstat_fwep.nii` is the p value corresponds to the rank of the observed T value within the permutations of maximum T values across all the voxels (the null distribution is composed by the maximum T value across all the voxels for each permutation). For the corrected, the distribution of the maximum is used as reference, and the rank (or quantile) of a given voxel in relation to that distribution is used to obtain p-values.
3. `_clustere_tstat.nii` is simply the size (in voxels) of the cluster. This number acts as the test statistic.
4. `_clustere_tstat_fwep.nii`: p-values computed in the same way as 3, i.e., using the distribution of the maximum cluster size.
5. The TFCE maps are similar to Points 1, 2 and 3.

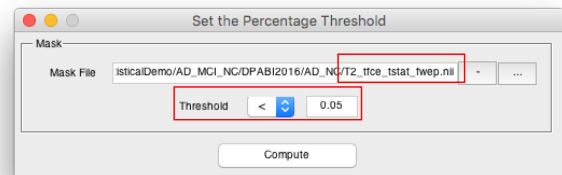
86

Permutation Test



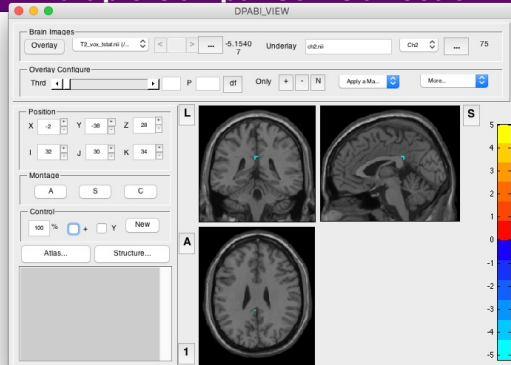
87

Permutation Test



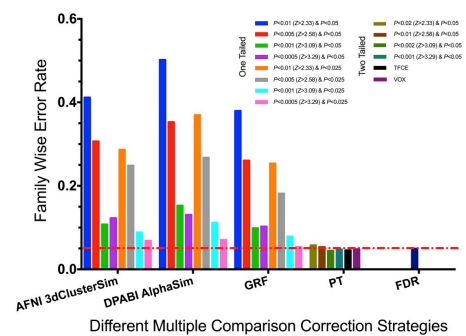
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Multiple Comparison Correction



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Multiple Comparison Correction

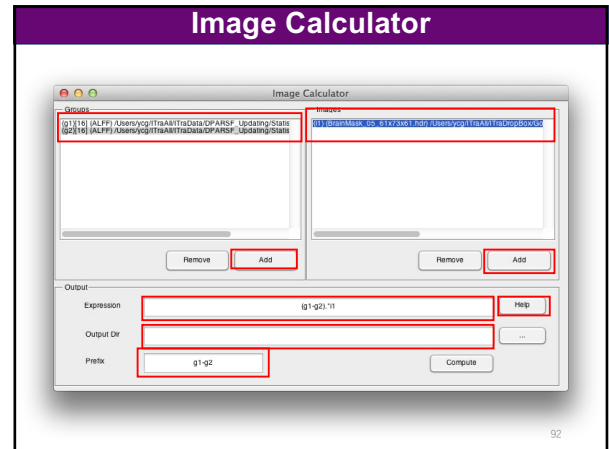


Chen, Lu, Yan, Human brain mapp. 2017. 20 vs. 20 Permutation 1000 times

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Image Calculator

Example expressions:

- (a) **g1-i1** Subtract 1 from each image in group 1
- (b) **g1-g2** Subtract each image in group 2 from each corresponding image in group1
- (c) **i1-i2** Subtract image 2 from image 1
- (d) **i1>100** Make a binary mask image at threshold of 100
- (e) **g1.*To4D((i1>2.3),100)** Make a mask (threshold at 2.3 on i1) and then apply to each image in group 1 (group 1 has 100 images)
- (f) **mean(g1)** Calculate the mean image of group 1
- (g) **(i1-mean(g1))/std(g1)** Calculate the z value of i1 related to group 1
- (h) **corr(g1,g2,"temporal")** Calculate the temporal correlation between two groups, i.e. one correlation coefficient between two "time courses" for each voxel.
- (i) **corr(g1,g2,"spatial")** Calculate the spatial correlation between two groups, i.e. one correlation coefficient between two images for each "time point".

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Reading and Writing functions

Reading:

```
[Data Header] = y_Read('brodmann.nii');
Data = 181*217*181 double
Header = Structure
```

Processing:

```
BA20Data = (Data==20);
```

Writing:

```
y_Write(BA20Data, Head, 'BA20.img');
```

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Further Help

The R-fMRI Course V2.1

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<http://wiki.rfmri.org>

The R-fMRI Journal Club

Official Account: RFMRILab

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Thanks for your attention!

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