

## Spatial Modeling: What do we want?

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- We want to find the signal
  - Where is it?
  - How big is it?
- Explicit Model
  - Each blob parameterized
    - x,y,z location
    - Volume
  - Standard errors & confidence intervals on location, volume

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## Spatial Modeling: What do we get?

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- Sorry
  - While naive estimates available...
    - Location of local maximum
    - Volume above a threshold
  - ... no standard errors, no inference
- What is available?
  - Massively Univariate Modeling

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## Spatial Modeling: Massively Univariate Modeling

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- Massively Univariate Models
  - Fit temporal model at every voxel
    - No information shared across space
  - Create statistic at each voxel
    - GLM & contrast
- How to threshold the statistic image?
  - Usual  $\alpha = 0.05$  threshold?
  - Yields 5% false positive rate
  - Yet we're performing 100,000 tests!
  - Is 5,000 false positives OK?

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## Multiple Comparisons: Overview

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- Multiple comparisons problem
  - Usual threshold produces too many false positives
- What to control?
  - Control *per-voxel* false positive rate
    - That is, do nothing
  - Control *FamilyWise false positive Error rate* (FWE)
    - FWE one or more false positives, anywhere
    - FWER is chance of getting *any* false positive
- Adjust threshold to control FWER
- Other false positive measure
  - *False Discovery Rate* (FDR)

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## Multiple Comparisons: FWER-Controlling Methods

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- Bonferroni
  - An old friend, but stingy
- Random Field Theory
  - Hi tech, Hi maintenance
- Nonparametric Permutation Test
  - Lo tech, trustworthy

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## MCP: Bonferroni

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- Usual approach to MCP
- Use  $\alpha/S$  as critical value instead of  $\alpha$ 
  - $S$  = Number of tests
- Controls FWER
  - But is conservative, especially for (spatially) correlated data

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## MCP: FWE & Maximum Statistic

- Important connection between FWE & Maximum

$$\begin{aligned} P(FWE) &= P(\cup_i \{T_i > u\} | \mathcal{H}_0) \\ &= P(\text{One or more voxels} > u | \mathcal{H}_0) \\ &= P(\text{Maximum is above threshold } u | \mathcal{H}_0) \end{aligned}$$

- Distribution of maximum is key
  - Just as 95%ile of null dist<sup>n</sup> gives 5%  $\alpha$  threshold, 95%ile of null max dist<sup>n</sup> gives 5%  $\alpha$  FWE-controlling threshold!

D:Nrm,Mx

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## MCP: FWE & Maximum Statistic

- If we can find the distribution for the max statistic we're done!
- Two approaches
  - Random field theory
  - Nonparametric permutation

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## RFT: That spells Random Field Theory

- Random Field is statistical process
  - Ocean surface
  - Temperature map of USA
  - A fMRI statistic image
- Rich theory available for continuous random fields

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## RFT: Getting at the Max with the Euler Characteristic

- Euler (say "oiler") Characteristic ( $EC$ )
  - Take continuous random field
  - Threshold it at level  $u$
  - $EC = \#blobs - \#holes$
- At high thresholds, no holes
  - $EC$  then just counts blobs
- At very high thresholds
  - $EC$  is just zero or one
  - Either one or no blobs

N:Wor92

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## RFT: Getting at the Max with the Euler Characteristic

- Expected  $EC$ 
  - Mean or average
  - By definition
 
$$E(EC) = \sum_k kP(EC = k)$$
- For high thresholds, expected  $EC$  is just what we want!
 
$$\begin{aligned} E(EC) &= \sum_k kP(EC = k) \\ &= 0P(EC = 0) + 1P(EC = 1) + 2P(EC = 2) + \dots \\ &\approx P(EC = 1) \\ &= P(\text{"Maximum is above threshold } u\text{"}) \end{aligned}$$
- The expected  $EC$  approximates the max dist<sup>n</sup>

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## RFT: Getting at the Max with the Euler Characteristic

- So we can obtain p-values using
  - $P\{\max_{i=1}^S T_i > u\} \approx E(EC_u)$
- Expected  $EC$  formula for Gaussian random field
  - $E(EC_u) = (\text{RESELS})(u^2 - 1) \exp(-u^2/2)/(2\pi)^2(4 \log(2))^{3/2}$
  - Where
    - $u$  is the threshold
    - RESELS is the RESolution ELEment
- RESEL is search volume in units of resolution

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## RFT: RESELS

- Smoothness described in terms of FWHM
  - FWHM = Full Width at Half Maximum

D: fwhm

- RESEL formula

$$\text{RESEL} = \frac{S}{f_x f_y f_z}$$

→ Where

$S$  is search volume, in  $\text{mm}^3$

$f_x, f_y, f_z$  is FWHM smoothness in mm in  $x, y,$  &  $z$

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## RFT: Intuition

- Get a feel for corrected p-values!
- Say I observe a statistic with value  $u$

$$P\{\max_{i=1}^S T_i > u\} \approx E(EC_u) = (\text{RESELS})(u^2 - 1) \exp(-u^2/2) \dots$$

All other things equal...

- ... as  $u$  increases
  - Corrected p-values go \_\_\_\_, i.e. \_\_\_\_ significance
    - Hint: As  $u$  gets higher, can ignore  $u^2$  term
- ... as number of RESELS increases
  - Corrected p-values go \_\_\_\_, i.e. \_\_\_\_ significance

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## RFT: Intuition

$$P\{\max_{i=1}^S T_i > u\} \approx E(EC_u) = (\text{RESELS})(u^2 - 1) \exp(-u^2/2) \dots$$

- ... as search volume  $S$  increases
  - Corrected p-values go \_\_\_\_, i.e. \_\_\_\_ significance
  - Remember  $\text{RESEL} = \frac{S}{f_x f_y f_z}$
- ... as resolution  $f$  increases
  - Corrected p-values go \_\_\_\_, i.e. \_\_\_\_ significance
  - Remember  $\text{RESEL} = \frac{S}{f_x f_y f_z}$
  - Remember resolution is measured in FWHM, so bigger  $f$  means [ sharper | fuzzier ] statistic images

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## RFT: Assumptions

- Raw image data are multivariate Gaussian
  - Can check Normality at each voxel
  - Impossible to check multivariate Normality for every pair of voxels
- Smoothness
  - Images are sufficiently smooth to approximate continuous field
    - Rule of thumb: FWHM 2-3 times voxel size
  - Insufficient smoothness results in
    - Conservative voxel-wise p-values
    - Liberal cluster-size p-values

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## RFT: Assumptions

- Stationarity
  - Smoothness the same *everywhere*
  - Assumption needed for cluster size p-values
  - *Not* needed for voxel size p-values

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## Spatial Inference: SPM

- The following slides are addendum to the SPM intro slides

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## Spatial Inference: “Scopes of Inference”

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- SPM’s framework on inference
- Define Clusters
  - Use intensity *and* cluster size threshold
- Consider 4 “levels”
  - Average, Set, Cluster, & Voxel
- Each has successively greater spatial specificity
  - Though successively less sensitivity

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## Spatial Inference: “Scopes of Inference”

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- Average
  - Assesses  $\bar{F}$ , average of  $F$  image
  - Small p-value rejects  $\mathcal{H}_0$ : No activation anywhere
    - No localization power
- Set
  - Assesses  $c$ , number of clusters
  - Given intensity & cluster size threshold
  - Small p-value rejects  $\mathcal{H}_0$ : No activation anywhere
    - No localization power

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## Spatial Inference: “Scopes of Inference”

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- Cluster
  - Assesses  $k$ , number of voxels in a cluster
    - Given intensity threshold
  - Small p-value rejects  $\mathcal{H}_0$ : No activation in cluster
    - Localization to blob
    - However, can’t point to particular voxel
- Voxel
  - Assesses  $z$  or  $t$  or  $F$  statistic at a voxel
  - Small p-value rejects  $\mathcal{H}_0$ : No activation here
    - Localization to precise voxel

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## Spatial Inference: SPM

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- Important to understand each component of SPM p-value results page
- Top
  - Interactive MIP
    - Drag & Drop to move
    - Right click to get menu
  - Interactive Design Matrix
    - Click to get values
- Middle
  - Interactive table of corrected & uncorrected p-values
    - Click on x,y,z to move cursor
    - Click on number to show it in Matlab window
    - Right click in margin to get menu

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## Spatial Inference: SPM

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- Bottom
  - Details on results
  - Key key key!
    - Smoothness must be  $>\approx$  2-3 voxel size
    - This is needed for  
\_\_\_\_\_ to be valid

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## Spatial Inference: SPM etc

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- SPM.mat
  - Contains essentials of fitted model results
  - Design matrix, number of voxels, etc
- SPMcfg.mat
  - “SPM Configuration file”
  - Input to ‘Estimate’ button
  - Specifies everything necessary to fit a model

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## Spatial Inference: SPM etc

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- `SPM_fmRIDesMtx.mat`
  - Specification of fMRI design
  - Does *not* contain any filenames
  - If identical design is used, can copy these into other analysis directories
  - For example
    - I could have created one `SPM_fmRIDesMtx.mat` that the whole class could use, since all of our subject's had identical paradigms
- “Explore Design button works with any of these.