

# General Linear Model for random-effects fMRI

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#### MRI





3T MR scanner @ Otaniemi (ani.aalto.fi)

#### **f**MRI



#### time

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#### MRI



#### -

#### time

**→** 

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- fMRI is a technique for measuring and mapping brain activity
- fMRI data are time series of image volumes
- Typical dimensions:

#### **122 880** [• 64 x 64 voxels within a slice

- voxels! | about 30 slices
  - 150 300 volume images (time voxel points)

#### One slice of a volume image



(volume element)

**f**MRI



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- fMRI is a technique for measuring and mapping brain activity
- fMRI data are time series of image volumes
- Voxel-by-voxel time-series analysis







#### **f**MRI





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- fMRI is an **indirect** measure of neural activity.
- Absolute magnitude of the fMRI response depends on
  - voxel size, proportion of gray matter in a voxel, local vascular density, number of voxels in a cluster, physiological variability in signal strength, field strength,...
- fMRI is a relative measure of activity: typically you need a control/baseline condition to compare with.



## Typical questions for an fMRI experiment

- Which brain regions are activated during a specific task? (functional localization)
- Is the response to task A larger than the response to task B in a particular brain region?
- Spatial maps

#### "Where" questions



## Preprocessing of fMRI data

- Typical preprocessing steps:
  - Data format conversion
  - Slice timing correction
  - Movement correction
  - Distortion correction
  - Spatial smoothing
  - Spatial normalization

#### **Example voxel time-series**





## Preprocessing of fMRI data

- Typical preprocessing steps:
  - Data format conversion
  - Slice timing correction
  - Movement correction
  - (Distortion correction)
  - Spatial smoothing to increase SNR
  - Spatial normalization to common brain atlas
- Main goal of preprocessing is to reduce non-task-related (uninteresting) variability in the data

#### **Example voxel time-series**





## Statistical analysis of fMRI data

- Typical question: Which brain regions (voxels) are activated by the stimulus or task?
- Standard approach\*:
  - 1. Construct a model of predicted brain activity
  - 2. Fit the model to data
  - 3. Perform statistical tests

\*repeat for each voxel



## General linear model (GLM)





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#### a better model

 Take into account the shape of the BOLD response: convolve the stimulus timing vector with a model of the hemodynamic response function (hrf)





 $y = X\beta + \varepsilon$ 



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$$y = X\beta + \varepsilon$$



$$y = X\beta + \varepsilon$$



\*this is a basic design matrix should model (add columns for) all known effects-of-interest and nuisance variables

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#### $\rightarrow$ Find parameters $\beta$ that best explain the data



#### 4 stimulus categories



Blocked fMRI design (20s 'stimulus on' periods)



#### 96 stimulus types





# Experimental design? Design efficiency?

- Blocked designs
  - Multiple repetitions of stimuli of the same category (or task) shown in "blocks"
  - Good detection power
- Event-related designs
  - More stimulus types
  - Transient activity
  - Good estimation power
- For details on design efficiency, see <u>http://imaging.mrc-</u> <u>cbu.cam.ac.uk/imaging/DesignEfficiency</u>



## General linear model

- $y = X\beta + \varepsilon$
- GLM aims to explain the variation in the measured fMRI time-course in terms of a linear combination of predictors (columns in the design matrix).
- Find parameters  $\beta$  that best explain the data by minimizing the sum of the squared error values ( $\sum \varepsilon^2$ ).



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### Statistical inference

- Voxel-by-voxel hypothesis testing: Does my model explain variance in the data?
- Specify contrast (hypothesis), c, a linear combination of the estimated parameters (e.g., [10])
- Calculate, e.g., the T-statistic for the contrast separately for each voxel

$$T = \frac{c^T \hat{\beta}}{\operatorname{std}(c^T \hat{\beta})}$$



 $\gamma = X\beta + \varepsilon$  $T = \frac{c^T \hat{\beta}}{\operatorname{std}(c^T \hat{\beta})}$ 

fMRI data (one timepoint)







beta map

$$y = X\beta + \varepsilon$$
$$T = \frac{c(\hat{\beta})}{\operatorname{std}(c^T\hat{\beta})}$$



15

10

5

0

-5

spmT map

$$y = X\beta + \varepsilon$$

$$T = \frac{c^T \hat{\beta}}{\operatorname{std}(c^T \hat{\beta})}$$

- 6 - 4 - 2 - 0 -2

8





# Statistical parametric map (T map) thresholded and overlaid on an anatomical MR image



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#### Constructing contrasts

- Are there distinct regions that are specialized for a particular function (*e.g.*, perception of faces)?
- Before neuroimaging: focal lesions → specific perceptual problems
- Why? e.g., behavioral relevance of specific stimulus categories
- Always a network of brain regions
- In practice: Where in the brain stimulus X evokes a larger response than stimulus Y?



#### Faces > scenes





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#### Scenes > faces







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#### Scenes > faces







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## 22 individuals (c: scenes > faces)



Contrast

Design matrix

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## Group analysis?

- How to generalize findings from a sample of subjects to the population (from which the subjects have been drawn)?
- Whole-brain group analysis
  - Spatial normalization: match brains across individuals
    - Warp each individual brain data to a common space (Talairach; MNI)
    - Same voxel ≈ same location in the brain across individuals
  - Power of statistical analysis depends on the quality of normalization
    - Smoothing of functional data increases SNR and overlap of active brain regions across individuals (but aso spreads activations across sulci, increses partial-volume effects and reduces spatial resolution)



## Spatial normalization in volume space

- Spatial normalization: match brains across individuals
  - Warp each individual brain data to a common space (Talairach; MNI)
  - Same voxel ≈ same location in the brain across individuals





#### + deformation (warp) field



#### Spatial normalization in volume space



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#### Spatial normalization in volume space

 Deformation (warp) field can be applied to any other data coregistered with the original MRI image (*e.g.*, functional data)



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# Concatenate all data (fixed-effects analysis)

- Data analyzed as originated from a single subject of a very long experiment
- High statistical power



Statistical analysis: Design

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ect\*bf(1)

ce\*bf(1)



# Concatenate all data (fixed-effects analysis)

- Data analyzed as originated from a single subject of a very long experiment
- High statistical power
- Results cannot be generalized to population!
- →Need to consider that subjects constitute a randomly drawn sample from a large population
- →Random-effects analysis







# Account for between-subject variance (random-effects analysis)





#### Spatial normalization in surface space



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## Surface-based group analysis



## Multiple comparisons correction

- GLM is applied independently to a huge number of voxels (>100 000) = "massively univariate" approach
- At 5% chance level, we might label **5000 voxels** "significant" due to chance
- Different approaches to correct for the multiple tests: Bonferroni correction, Gaussian random field theory, false discovery rate approach, permutation methods,...



## Multiple comparisons correction

- fMRI data has significant spatial correlations (neighboring voxels exhibit similar behavior); for example, Bonferroni is typically too conservative for single subject data.
- You can also control for (reduce the number of) the multiple tests by masking or region-of-interest analysis.
- Importance of meta-analysis and replication studies to identify consistent results across studies.
- For more details, see, for example, Lindquist et al 2015: Zen and the Art of Multiple Comparisons.



## Whole-brain group analysis

- Correspondence problems:
  - Individuals differ both in the global and in the more fine-grained cortical folding patterns
  - Relationship between brain functions and anatomical structures across subjects?





## Whole-brain group analysis

- Correspondence problems:
  - Individuals differ both in the global and in the more fine-grained cortical folding patterns
  - Relationship between brain functions and anatomical structures across subjects?
- → **Region-of-interest** analysis
  - Region(s)-of-interest can be defined based on anatomical landmarks or functional criteria (separate functional localizer scan)



### Region-of-interest analysis

 Region of Interest (ROI) analysis = select a cluster of voxels or brain region *a priori* when investigating a region for effects





## Why region-of-interest analysis?

- To keep the data in the individuals' space (no need to normalize all data to a common space)
- To explore your data
  - Average response time-course within a ROI
- To limit the number of statistical tests (>100 000 vs 1 500)
  - Average activation-level within a ROI
  - Control for multiple tests within the ROI only
- To investigate the function of a region in detail



## Risks of region-of-interest analysis

- You are looking at the "wrong" region
- The effect is not specific to the region
- Avoid "double dipping", "circular analysis"
  - Always use independent data to select the ROIs and ask the research question
  - For more details, see:
    - Kriegeskorte et al 2009: Circular analysis in systems neuroscience the dangers of double dipping.
    - Kriegeskorte et al 2010: Everything You Never Wanted to Know about Circular Analysis, but Were Afraid to Ask



#### Summary



- fMRI data are time-series of image volumes
- Voxel-by-voxel time-series analysis
- **Preprocessing** to reduce non-task/stimulus related variance in the data
- Spatial normalization to a common space necessary when applying whole-brain group analysis



#### Summary



- Statistical analysis...
  - 1) ...aims at localizing the brain regions that show an increased (decreased) response in response to the stimulus or task, and
  - **2)** quantifies the likelihood that an observer effect can be explained by noise fluctuations.
- General linear model: explain the variation in the measured fMRI time-course using a linear combination of predictors.
  - **1)** Find optimal beta weights (parameter estimates) for each predictor.
  - 2) Test the significance of the beta weights (or difference between betas) using, for example, t statistics.
- Massively univariate analysis = statistical tests performed independently for each voxel → statistical (parametric) map → multiple comparison correction







- Random-effects analysis asses the variability of observed effects between subjects (cf. fixed-effect analysis).
- Volume-based vs. surface-based group analysis
- **Spatial correspondence problem** in group-analysis: differences in anatomy, validity of structure-function relationship.
- Region-of-interest analysis (a priori defined, based on independent data, anatomical and/or functional criteria)



## References



Springer



SPM12: <a href="http://www.fil.ion.ucl.ac.uk/spm/software/spm12/">http://www.fil.ion.ucl.ac.uk/spm/software/spm12/</a>

Freesurfer: <u>https://surfer.nmr.mgh.harvard.edu</u>



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## Thank you for your attention! Questions?

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PS. Pay attention to careful planning of your experiment!

