#### HD4630 Workshop II

First- & Second-Level Analysis

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## Workshop I Recap

- Preprocessing
  - Discard pre-steady state TRs
  - Slice-timing correction
  - Rigid-body motion correction
  - Coregistration of functional & anatomical
  - Normalization to standard template
  - Spatial filtering (Smoothing)
- Quality Control
  - Visual inspection
  - Censoring or "scrubbing" motion

#### Plan for Today

- Discuss first- and second-level analyses in a traditional general linear model (GLM) framework
- Conduct a first-level, fixed-effects analysis in AFNI using uber\_subject.py
- Conduct a second-level, random-effects analysis in AFNI using uber\_ttest.py

## First-level Analysis

First-level analysis involves estimating the β-matrix in

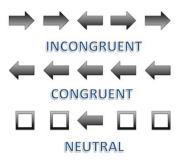
$$Y = X\beta + \epsilon$$

by constructing the contrast matrix X

A great review of the math underlying this is available from Mumford Brain Stats

#### First-Level Analysis, cont.

- We need to input the stimulus timing for each of the conditions in our task
  - We'll collapse the two Flanker 'congruent' and 'incongruent' conditions, ignoring participant accuracy



#### A: Standard model (stimulus effects ignored) Participant 1 Neural activation (arbitrary scale) Stimulus presentations 0 20 40 Participant 2 80 100 120 Neural activation (arbitrary scale) spoon pay climb road take find house desk speak. read see ask 20 40 Participant 3 80 100 120 Neural activation (arbitrary scale) 20 0 40 60 80 100 120 Time (in seconds)

Westfall, Nichols, & Yarkoni bioRxiv

# Second-Level Analysis

 Second-level analysis as implemented in fMRI takes a summary-statistics approach

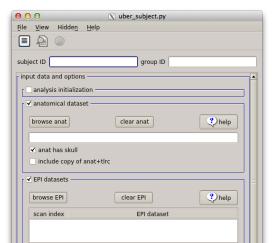
$$\hat{\beta} = X_g \beta_g + \eta^*$$

where beta-hats for each contrast, for each subject are carried forward from the first-level

A great review of the math underlying this is available from Mumford Brain Stats

# To Do: Set Preprocessing Options

- These were discussed in the last workshop
  - For a review, see last week's slides on the course website



## Stimulus Timing Information

- All stimulus timing information is included in the \*events.tsv files for each run
  - However, we need to convert this information into text files that AFNI will accept
- You'll be provided with a folder containing
  - A Python script to convert these files
  - The created text files themselves

## To Do: Enter Stimulus Timing

- In uber\_subject.py, select the provided text files in the section 'Stimulus Timing Files'
- You can also specify the basis function and the file 'type' (see 3dDeconvolve -help for relevant options)

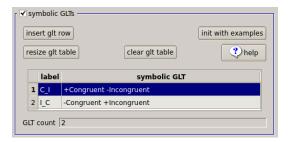


## General Linear Tests (GLTs)

- AFNI refers to contrasts between conditions as General Linear Tests (GLTs)
  - This is because we're working in the General Linear Model (GLM)
- We'll need to specify what tests we would like to conduct
  - Congruent > Incongruent
  - Incongruent > Congruent

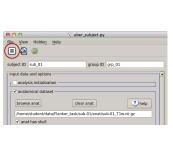
#### To Do: Specify GLTs

- When inputting stimulus timing files, AFNI will automatically associate each condition with your provided label
- You can use those labels to enter your desired GLTs
  - Select 'init with examples' to see example GLTs with your conditions



#### To Do: Review Processing Script

- Once you've entered all relevant parameters for preprocessing and first-level analysis, we can review and run the associated script
- Estimated run time is 40 minutes per subject



```
V file: cmd.an.sub 01
#!/usr/bin/env tcsh
# created by uber subject.pv: version 0.39 (March 21, 2016)
# creation date: Sun Mar 12 18:85:86 2017
# set data directories
set top dir = /home/student/data/Flanker task/sub-01
set anat dir = $top dir/anat
set epi dir = $top dir/func
set stim dir = $top dir/func
# set subject and group identifiers
            = sub 01
set group id = grp 01
# run afni_proc.py to create a single subject processing script
afni proc.py -subj id $subj
        -script proc.$subj -scr_overwrite
        -blocks tshift align tirc volreg blur mask scale regress
        -copy anat $anat dir/sub-01 Tlw.nii.gz
        -tcat remove first trs 0
            $epi dir/sub-01 task-flanker run-1 bold.nii.gz
            Sepi dir/sub-01 task-flanker run-2 bold.nii.gz
        -align opts aea -giant move
        -tlrc base MNI avg152T1+tlrc
        -volreg align to first
        -volreg align e2a
        -volreg_tlrc_warp
        -blur size 6.0
        -regress stim times
```

#### The Design Matrix

- All of the information we have entered thus far will form the X or 'design' matrix
- This is entered into the model

$$Y = X\beta + \epsilon$$

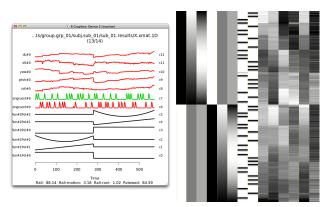
and will allow us to estimate the  $\beta\mbox{-matrix}$  for each participant

#### The Design Matrix, cont.

- After uber\_subject.py, the design matrix is output as a \*.xmat.1D file
- Multiple versions of this file exist, including
  - A list of input stimulus timings (X.stim.xmat.1D)
  - A full design matrix with no motion censoring (X.nocensor.xmat.1D)
  - A full design matrix with motion censoring (X.xmat.1D)
- ▶ In this class, we'll be working with the X.xmat.1D file

#### To Do: Review the Design Matrix

- ExamineXmat reads in the generated \*.xmat.1D file to visualize the time series of the design matrix
- Idgrayplot reads in the generated \*.xmat.1D file to visualize a graph of the design matrix



#### Viewing First-Level results

- We need to carry forward the beta coefficients from the first- to second-level
  - We'll need their index in the created stats file to do that
- To get the correct index, we can view our first-level results in AFNI

## Second-Level Analysis

- Once we're confident at the first-level, we can move forward to second-level analysis
- AFNI also provides a GUI to do this!
  - uber\_ttest.py

