HD4630 Workshop I

Preprocessing & First-Level Analysis

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Why Preprocess?

- fMRI data is acquired with physical, biological constraints
 - MRI machines aren't magic!
 - Sampling neural activity in living, breathing human beings
- We need to compare data across participants to draw (statistically) meaningful conclusions

Basic Preprocessing

- Discard pre-steady state TRs
- Slice-timing correction
- Rigid-body motion correction
- Coregistration of functional & anatomical
- Normalization to standard template
- Spatial filtering (Smoothing)

Preprocessing in AFNI

- AFNI allows you to see as little (or as much) detail as you want when preprocessing
- In this class, we'll be using uber_subject.py, a preprocessing GUI that hides many technical details
 - To see what's happening, we'll have to look at the generated tcsh script

First-level Analysis

- uber_subject.py also includes 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

Plan for Today

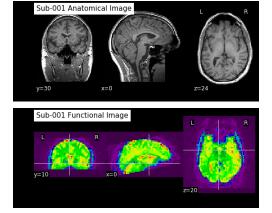
- Review, with examples, basic preprocessing
- Discuss first-level analysis in a traditional general linear model (GLM) framework
- Work along in AFNI using uber_subject.py
 - Preprocess a participant from OpenfMRI ds000102 (Flanker task)
 - Perform a first-level analysis for incongruent vs. congruent Flanker trials

Launching the Docker Container

- We want you to run this locally, so you'll need to launch the docker image you downloaded
- The full instructions to do so are on the course website, but as a quick reminder:
 - For Mac: execute mac_launch.sh in a terminal with Docker running
 - For Windows: execute windows_ip.cmd in a command prompt. Then, supply that IP to windows_launch.sh in a Docker Quickstart Terminal

First Steps: Looking at the Data

- No matter what you're doing, this is the first step!
- Here, we'll need to look at both our anatomical and functional images

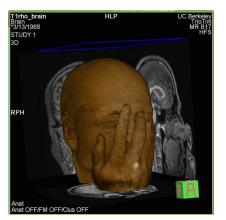


To Do: Open Data in AFNI

- Type afni into the terminal window to open a new session
- For sub-01 of the Flanker task:
 - Read in the anat directory
 - Read in the func directory and scroll through the time series of one run
 - Close afni

What Happened to My Anatomical Image?

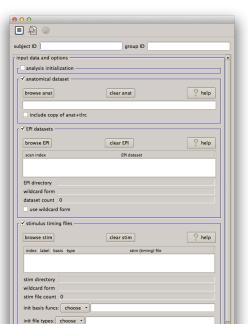
- Defacing is a common step in publicly available MRI data
- Intended to reduce the risk of participant identification



Discard Pre-Steady State TRs

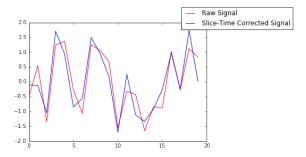
- Pre-steady state or unsaturated volumes show vastly different properties from the rest of an EPI series
- Because of this, some scanners do not reconstruct these "dummy scans" by default
 - Both of our OpenfMRI datasets exclude pre-steady state volumes, so we will not discard any TRs

To Do: Open uber_subject.py



Slice Timing Correction

- In single-shot EPI, each slice in a volume is acquired at a different time
- ▶ In our Flanker task, slices were acquired over a 2000ms TR in an interleaved fashion

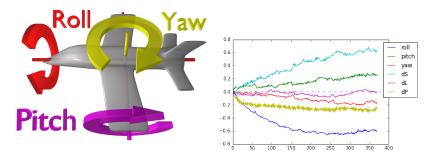


To Do: Slice Timing

- By default, AFNI will check the image header for slice timing information
- You can change the default interpolation used, but we won't address that in this workshop

Rigid-Body Motion Correction

 AFNI will estimate and correct for movement across the entire scan



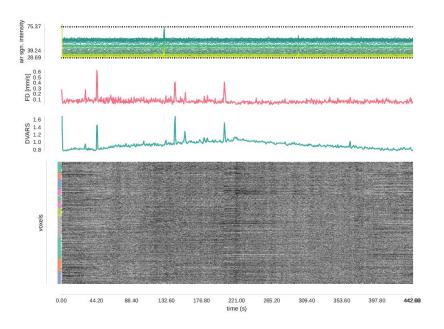
This is a well-behaved participant!

To Do: Set Volume Registration Base

- To perform rigid body motion correction, we need to select a volume to which to register all other volumes
- Since we did not remove any pre-steady state TRs, we can set this option to 'first'

The Limits of Motion Correction

- Movement is a major concern in fMRI analysis
- Participants who show high levels of movement may not be usable
 - Even when correcting for motion, subtle differences may remain



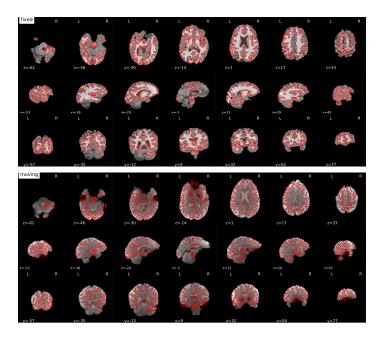
Visualization from MRIQC

To Do: Set Motion Censoring

- 'Motion censor limit' sets the maximum amount of motion allowed in any one TR
- TRs showing motion higher than this limit are 'scrubbed' from the time series

Coregistration

- We need to get the functional and anatomical images into the same space
- By default, these are likely to be out of alignment as participants move between scans
 - After motion correction, all of our functional images should be aligned to one another
 - Aligning these to the anatomical image should therefore be a one step correction



Visualization from fMRIPrep

To Do: Extra Alignment Options

- 'Local Pearson Correlation' is the default coregistration option
 - This is appropriate for T1 to EPI registration, but may need to be changed for other modalities
- Select 'use giant_move' to allow AFNI to align anatomical and functional images where participants have moved substantially

Normalization

- Standard templates currently in use include Talaraich and MNI
 - MNI based on many living subjects rather than one post-mortem
 - Several different versions of the MNI are in use!

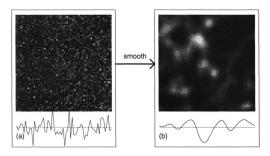


To Do: Extra TLRC Options

- 'TLRC' is short for Talaraich, the default normalization space in AFNI
- Change this to MNI_avg152T1 to align our subject to MNI space
 - This corresponds to the MNI 2009c nonlinear template

Spatial Filtering (Smoothing)

- 1. Increases signal-to-noise ratio
- 2. Helps to ameliorate residual anatomical variability
- Allows data to meet assumptions of Gaussian Random Field theory
 - A fun introduction to this is provided in Worsley, 1996

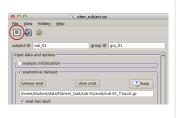


To Do: Set Blur Size

- We need to specify the FWHM (Full-Width at Half-Maximum) of our Gaussian kernel
- ▶ 4mm is the default, but we'll increase this to 6mm (FWHM ≥ 2x voxel size)

Inspecting our Script

Now that we have preprocessing set up, we can look at the parameters for our example subject



```
X file: cmd.ap.sub_01
File
#!/usr/bin/env tcsh
# created by uber subject.py: version 0.39 (March 21, 2016)
# creation date: Mon Mar 6 03:38:06 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 subject and group identifiers
set subi
            = sub 01
set group id = grp 01
# run afni proc.py to create a single subject processing script
afni proc.py -subj id $subi
        -script proc.$subj -scr overwrite
        -blocks tshift align tlrc volreg blur mask scale regress
        -copy anat $anat dir/sub-01 Tlw.nii.gz
        -tcat remove first trs 0
        -dsets
            $epi dir/sub-01 task-flanker run-1 bold.nii.gz
            $epi 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 censor motion 0.3
        -regress est blur epits
        -regress est blur errts
```

First-Level Analysis

- uber_subject.py also incorporates first-level analysis
- For this, we'll need to know the stimulus timing for each of the conditions in the Flanker task
 - We'll collapse the two 'congruent' and 'incongruent' conditions, ignoring participant accuracy

