

# Project Theta Progress Report

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# Background

# The Paper

- ▶ Neural Basis of Loss Aversion in Decision-Making Under Risk
- ▶ Decisions influenced more by potential loss than by potential gain?

# Paper Purpose

- ▶ Investigate whether potential loss/gain coded the same neurally
- ▶ Brain activation results in different regions points towards yes

# The Data

- ▶ 16 subjects, 3 runs, 256 unique combinations total
- ▶ Behavioral data, BOLD data

# Initial Analysis

-Basic preprocessing so far: identify dvars (RMS of signal derivative) and fd

## Initial work

# EDA

- ▶ downloaded data
- ▶ simple plots, summary statistics



# Plan

# Models and analysis

- Behavioral analysis

Logistic regression model on the behavioral data:

$$\text{logit}(Y_{resp}) = \beta_0 + \beta_{loss} * X_{loss} + \beta_{gain} * X_{gain} + \epsilon \quad (1)$$

Calculate the the behavioral loss aversion:

$$\lambda = -\beta_{loss} / \beta_{gain} \quad (2)$$

# Models and analysis (Continued)

- ▶ Linear Regression on BOLD data

For each voxel  $i$ , we fit a multiple linear model:

$$Y_i = \beta_{i,0} + \beta_{i,loss} * X_{loss} + \beta_{i,gain} * X_{gain} + \epsilon_i \quad (3)$$

For each voxel, we calculate the neural loss aversion  $\eta_i$ :

$$\eta_i = (-\beta_{loss}) - \beta_{gain} \quad (4)$$

## Models and analysis (Continued)

- ▶ Whole brain analysis of correlation between neural activity and behavioral response across participants

Examine the relationship between neural activity and behavioral using the following regression model:

$$\lambda = \alpha_0 + \alpha_1 * \eta + \epsilon \quad (5)$$

# Models and analysis (Continued)

## ► Inferences on Data

1. Test the normal assumption of residuals in linear models
2. Test of significance of coefficients
3. Calculate the (adjusted) R squares.

# Explanation on model simplification

- ▶ Use of Data

Leave out the regressor euclidean distance to indifference

- ▶ Simplification of regression on BOLD data

Perform simple linear models rather than mixed effect models

# Issues with analyses and potential solutions

- ▶ Selecting specific regions to further explore correlation between neural and behavioral activity
- 1. Further research on brain
- 2. Look for regions with most significant neural loss aversion and regression coefficients
- ▶ Producing heat map

Need to map bold data onto standard brain

## Our Process



# Hardest part of process?

- ▶ Working with the FMRI data and trying to understand our paper
- ▶ Keeping up with documentation

## Success in overcoming these obstacles?

- ▶ Using git workflows to raise issue and problems for the group.
- ▶ Limited success in the FMRI part, still figuring things out.

# Issues facing the team?

- ▶ Debugging each other's code when travis CI fails in a pull request
- ▶ Addressing this by meeting up for teamwork or ask for help
- ▶ Most helpful?
  - ▶ Python/numpy, lab sections with git workflows
- ▶ Most confusing?
  - ▶ FMRI analysis

# What do you need to do to successfully complete the project?

- ▶ Have a clear idea of what we can get done
- ▶ Make work as reproducible as possible

## Diffuculty in reproducibility?

- ▶ Very frustrating if Travis fails on a pull request
- ▶ Remembering to write documentation in the scripts
- ▶ Test functions for plotting functions are hard to write/assert
- ▶ With a lot of work we may be able to get most of it reproducible

## Remaining weeks:

- ▶ Could cover:
  - ▶ Software tools like statmodels, make and toolkits that are frequently used in fMRI research
  - ▶ More regression: linear and logistic