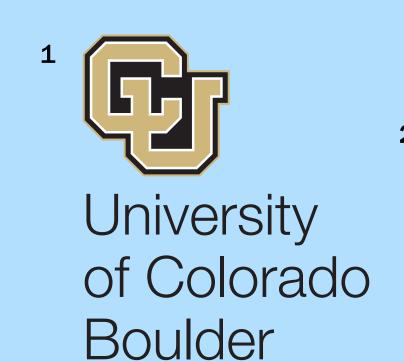
# Characterizing functional discriminability across the brain using large-scale classification





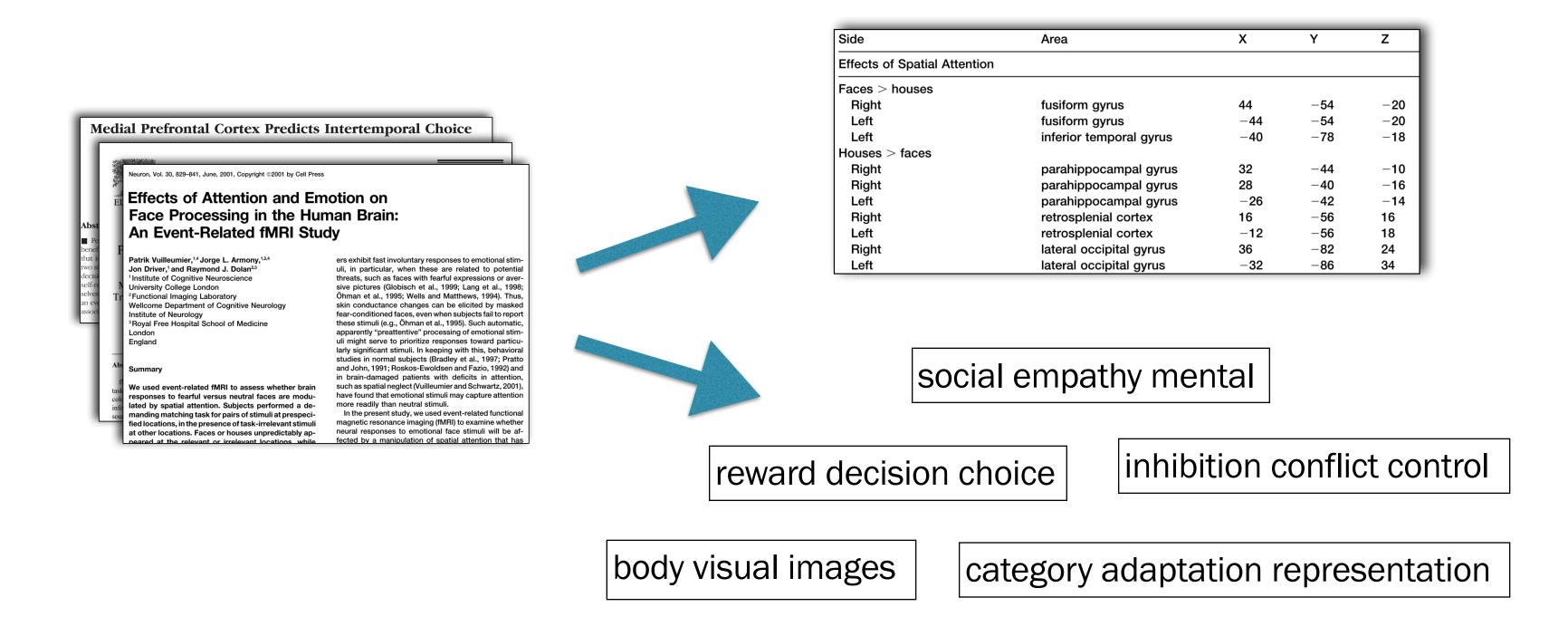
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#### Introduction

- Large-scale fMRI databases are useful for robustly characterizing brain regions associated with cognitive functions
  - E.g. meta-analysis of 'reward' in Neurosynth
- However, it is not clear how useful these functions are for discriminating brain regions from each other
- Knowing what a study was about, how well can you predict a region was activated?
- How well can we discriminate which regions of the brain a study activates, based on the semantics?
  - How do regions differ?

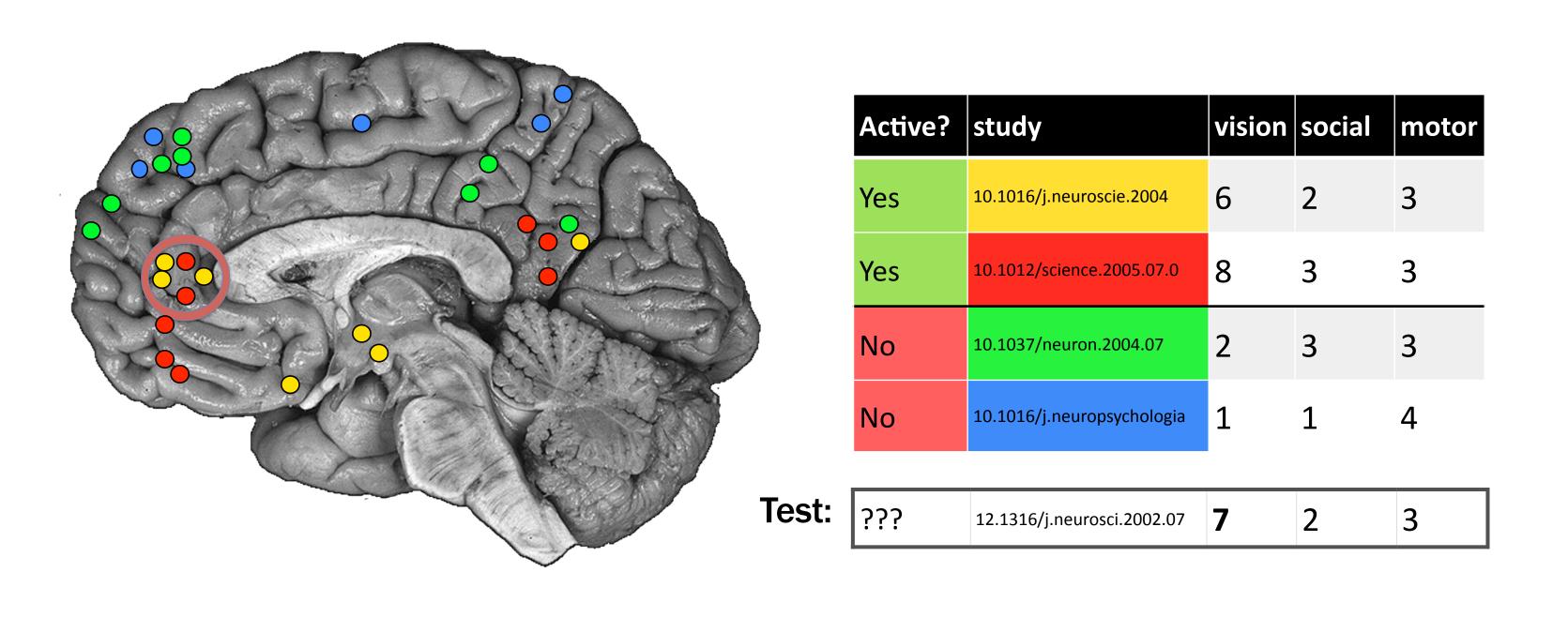
#### Methods

- Neurosynth database (neurosynth.org): 8000+ fMRI studies<sup>1</sup>
- Topic model of word frequencies in papers semantics<sup>2</sup>
- Parcellated brain using k-means on coactivation of voxels (11, 20 and 60 regions)

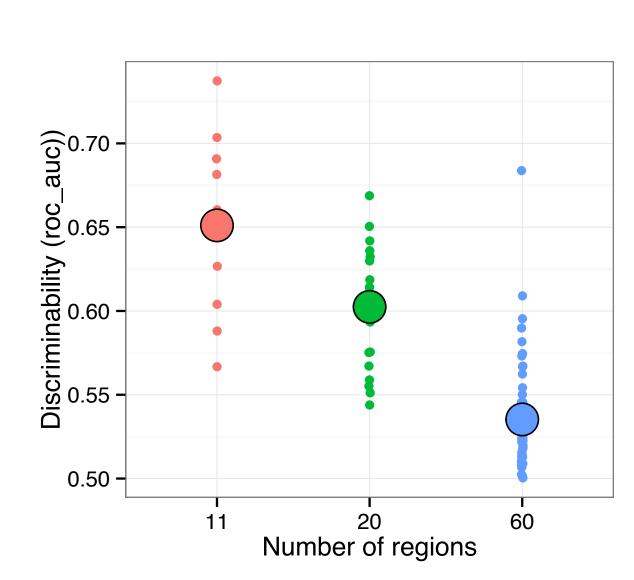


# Classification analysis

- 1. Selected studies that activate and do not activate regions
  - Active 5% of voxels in ROI active
  - Not active zero voxels in ROI active
- 2. For each region individually, trained Ride classifier (python sklearn) to discriminate 'active' vs 'not active' studies using paper semantics (topics) as features
  - Ridge regularized regression classifier produces linearly interpretable weights for each topic
- 3. Evaluated using 4-fold cross-validation (tested models on unseen data) and scored using ROC area under curve

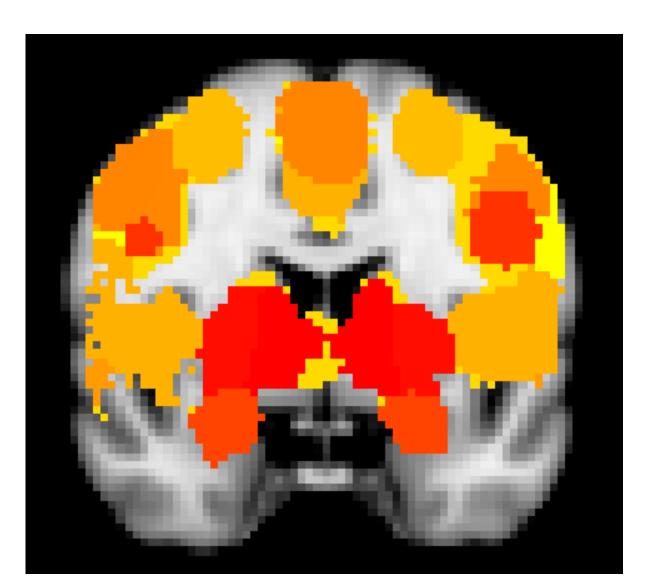


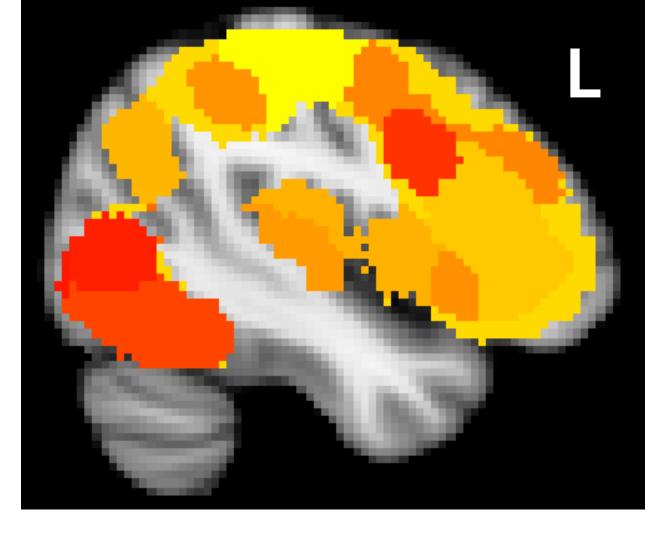
### Classification results

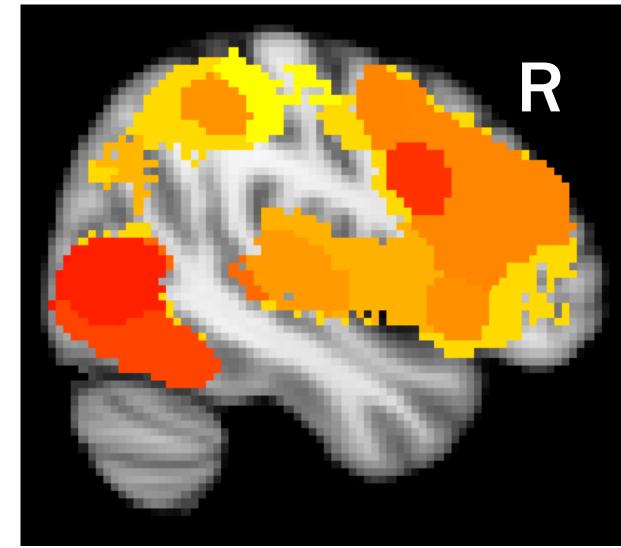


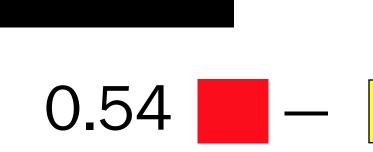
- Above chance classification
- Chance is 0.5
- Decreased accuracy with increased resolution
- Wide variability across brain

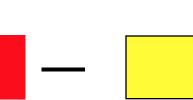


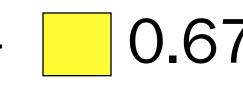




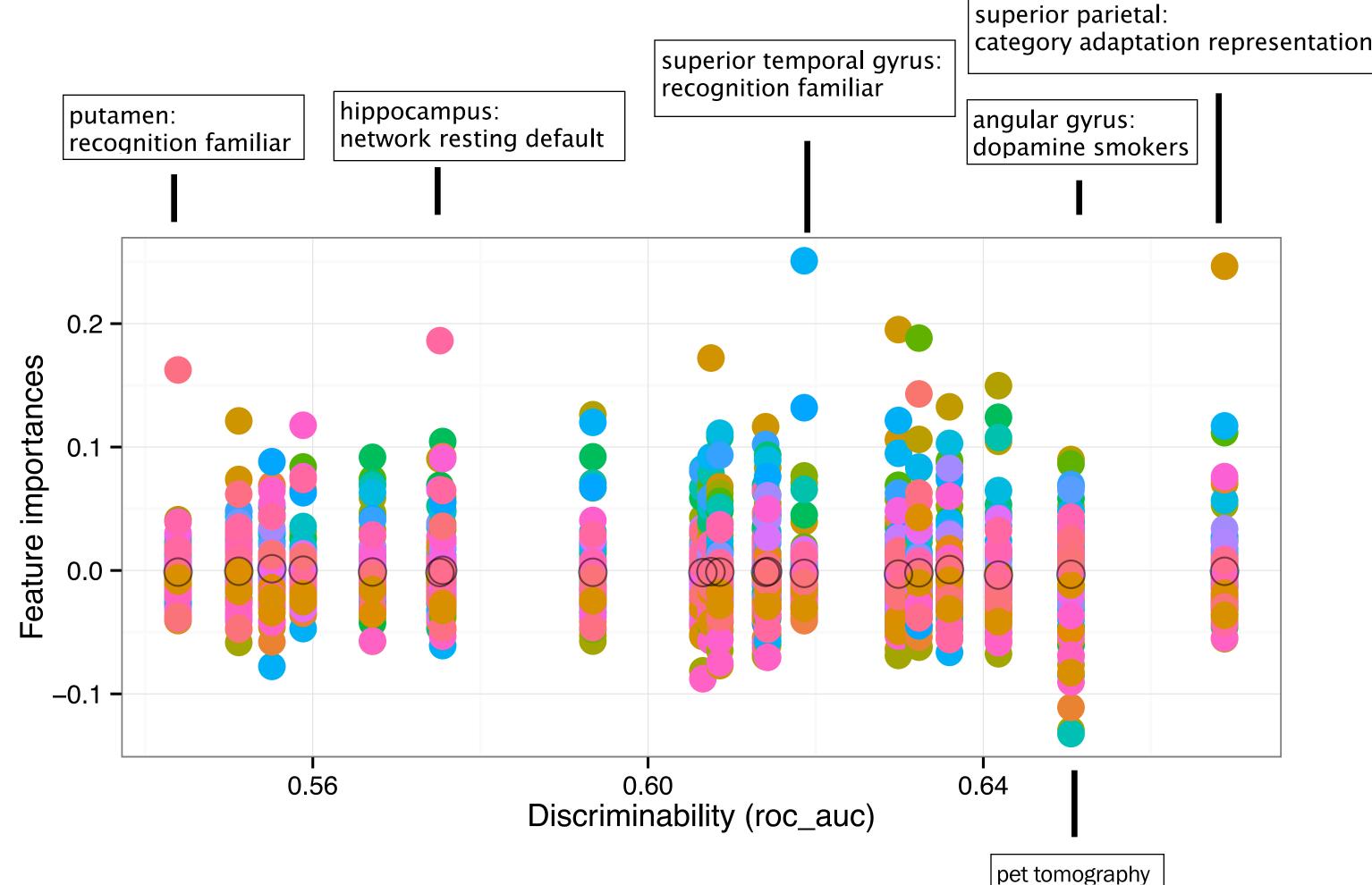








## Which topics are important for classification?



#### Conclusions

- We developed a measure of how discriminable brain regions are using cognitive function in Neurosynth
- Regions vary on discriminability
- Less sparse combination of features yields better discriminability
- Features that support classification mostly agree with univariate approaches, but yield some surprising results

#### References

1. Yarkoni, T., Poldrack, R. A., Nichols, T. E., Van Essen, D. C., & Wager, T. D. (2011). Large-scale automated synthesis of human functional neuroimaging data. Nature Methods 2. Poldrack, R. A., Mumford, J. A., Schonberg, T., Kalar, D., Barman, B., & Yarkoni, T. (2012). Discovering Relations Between Mind, Brain, and Mental Disorders Using Topic Mapping. PLoS Computational Biology,