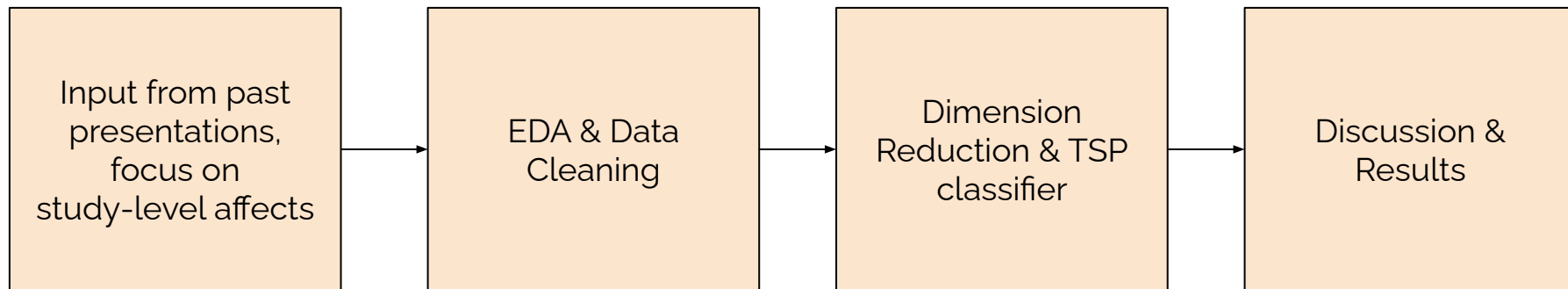

Analysis on pCR Binary Response Data

STAT 117 Final Project, Miela Foster

Roadmap



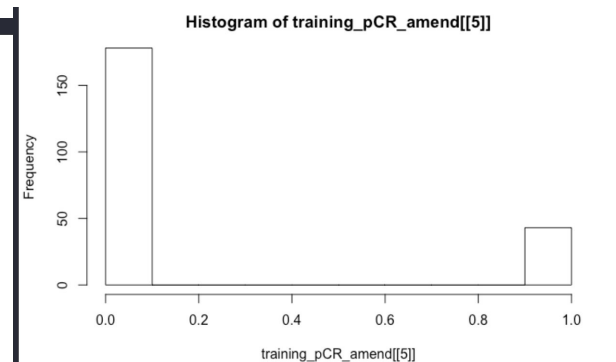
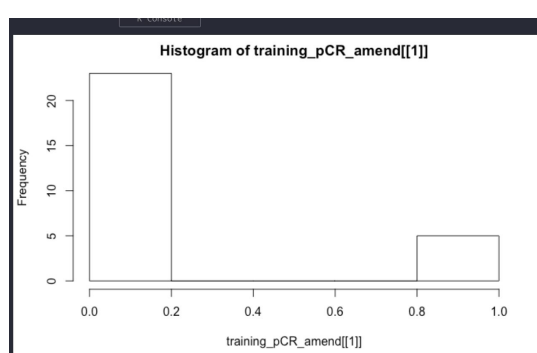
Past Presentations

- High degree of variability between studies (size, number of genes, number of participants, types of genes)
- Feature Extraction Methods
- Training and Test Set Split

My approach: focus on individual studies in order to control ambiguity of results, keep consistent training and test split, focus of decreasing size of studies to reduce potential overfitting (feature extraction), use TSP for familiarity

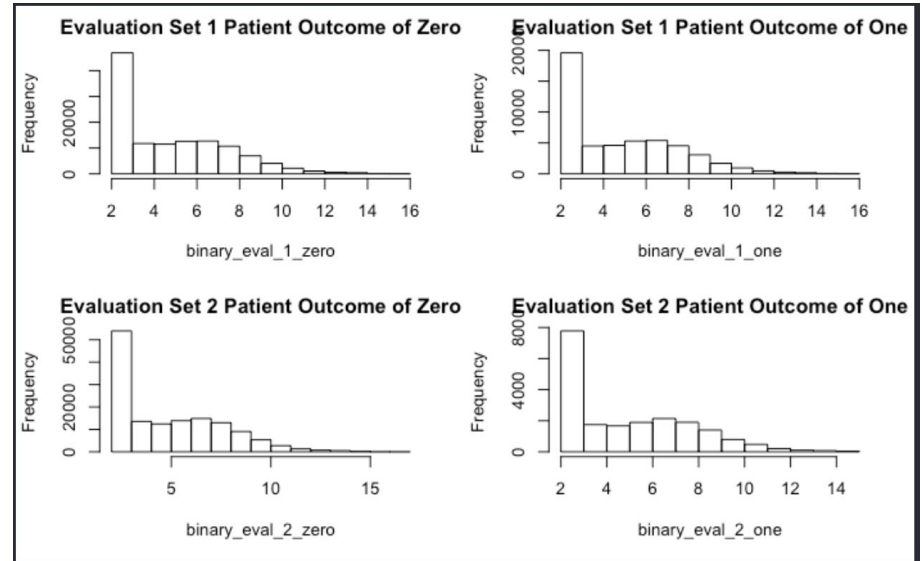
EDA & Train/Test Split

- Remove 3 studies because of missing data
- Random number generator to split the train and test set
 - 10 in training set, 3 in test set
- Evaluate the proportion of each response in each training set
- Interested in looking at the rate of each outcome



Further Investigate the Studies with Variability in Outcome

- What is the distribution of gene expression for each type of outcome
- Focus on studies 7 (eval 1) and 9 (eval 2) because they had most variability in binary outcome
- No clear discernibility based on distribution



Extended Interest in the Correlation between Data

Use **Canonical Correlation** to investigate the highest degree of relatedness between data (note linear relationships)

*use same example sets, study 7 and study 9 *

Association Between 0
Binary Outcome in 1st
Dimension:

0.95300842

Association Between 1
Binary Outcome in 1st
Dimension:

0.9481966

Association Between
Binary Outcomes in 1st
Dimension:

0.9412148

High association in the data, and outcomes within each study even more related. Points to feature extraction and dimension reduction.

K-means & TSP classifier

- K-means will help cluster data, allow for feature extraction
- Used TSP classifier for familiarity (krange = 6)

Paired each training and test set based on the number of participants in the study. Tried to match as close as possible. Interested in studying training sets with high variability (1, 10, 7 , 9)

Study 1 Paired with Testing 3

Study 9 Paired with Testing 1

Study 10 Paired with Testing 2

Study 7 Paired with Testing 1

Method: Perform K Means on 4 training sets, 5 clusters. Extract dataset based on the cluster grouping, run TSP classifier for each cluster number, and extract AUC from testing set assigned to that training dataset

Results from K-Means/TSP

study 1 and study 7 results shown

Study 7
Baseline:
0.5538

Cluster 1
AUC:
0.5115

Cluster 2
AUC:
0.55385

Cluster 3 AUC:
0.756923
**(barely beat
the baseline!!)**

Cluster 4
AUC:
0.50385

Cluster 5
AUC:
0.5038

Study 1:
Baseline:
0.5

Cluster 1
AUC: 0.714

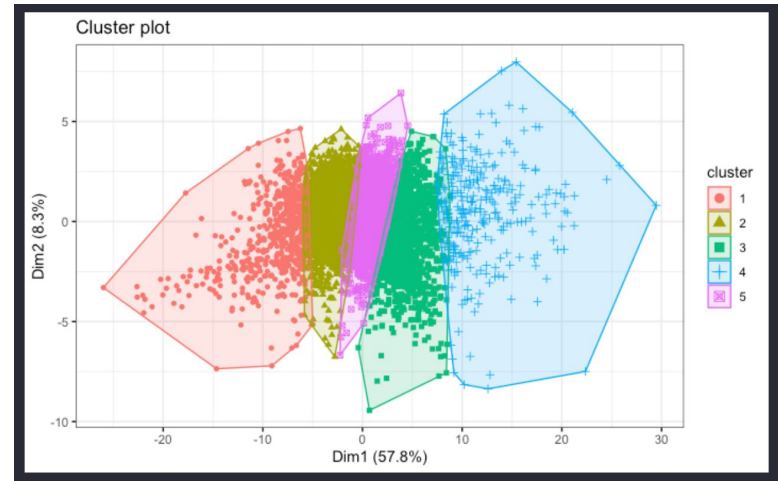
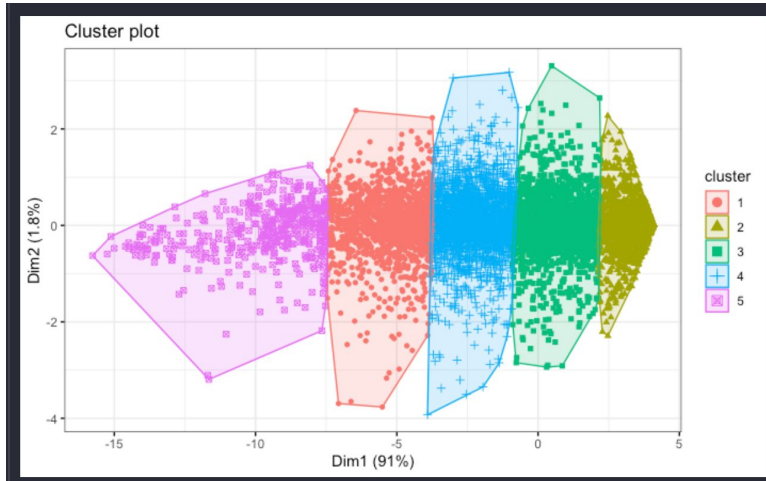
Cluster 2
AUC:
0.5188

Cluster 3
AUC:
0.7321

Cluster 4
AUC: 0.661

Cluster 5
AUC:
0.5357

Why did Cluster 3 Perform well?



- Cluster 3 SD for Study 7 : **1.05 (second lowest)**
- Cluster 3 SD for Study 1 : **1.027 (second lowest)**
- Study 1 Cluster 3 Size: 1682
- Study 7 Cluster 3 Size: 1780

Discussion

- Is k-means robust enough for feature extraction?
 - PCA/SVD, other clustering algorithms, Manifold Learning
 - Did my method help remove study to study issues?
 - What might be the results when I try my algorithm on all the interesting training sets I noted?
 - Why cluster 3?
 - Should I focus on participant size at all?
 - Concerns about the test/train split
 - Concerns about how correlated our data is
-