

# Evaluate testing data (multi-class) - Lasso

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Note: The two differences between Lasso and Tree-based methods are:

1. Lasso has its own inherent feature selection process.
2. Lasso's vimp will be based on how many times the feature exist in all runs. Regression coefficients may be presented for binary outcomes

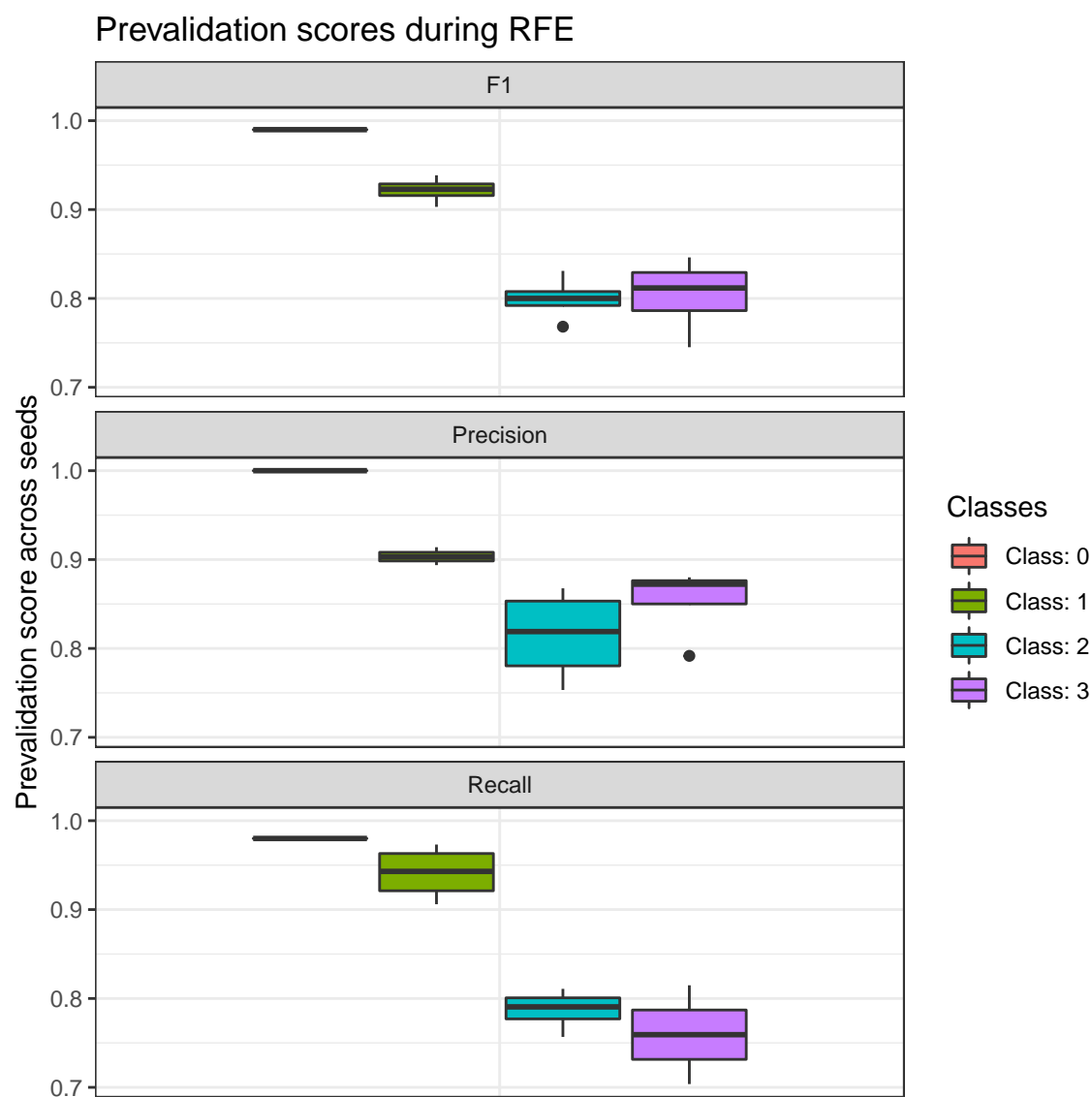
```
## user input
project_home <- "~/EVE/examples"
project_name <- "lasso_multi_outCV_test"
```

## 0. Load Data

```
## 300 of samples were used
## 100 of full features
## 4 runs, each run contains 3 CVs.
## Labels:
##
##    0    1    2    3
## 50 149   74   27
```

# 1. Scores

## 1.1 Scores per Class

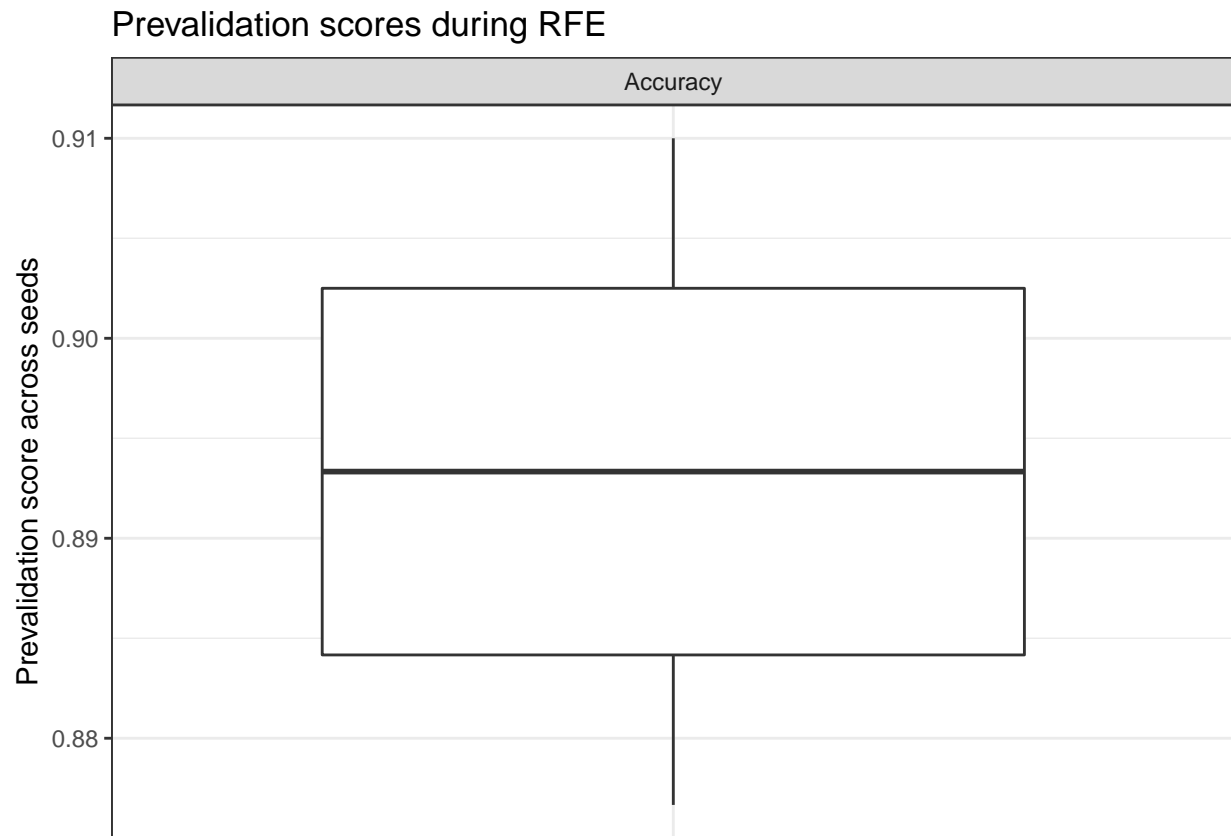


## Confusion Matrix

```
## confusion matrix at feature size = 100
## sum across 4 seeds
```

```
##           Reference
## Prediction  0    1    2    3
##           0 196    0    0    0
##           1   0 561   53    7
##           2   0  35 233   19
##           3   4   0  10   82
```

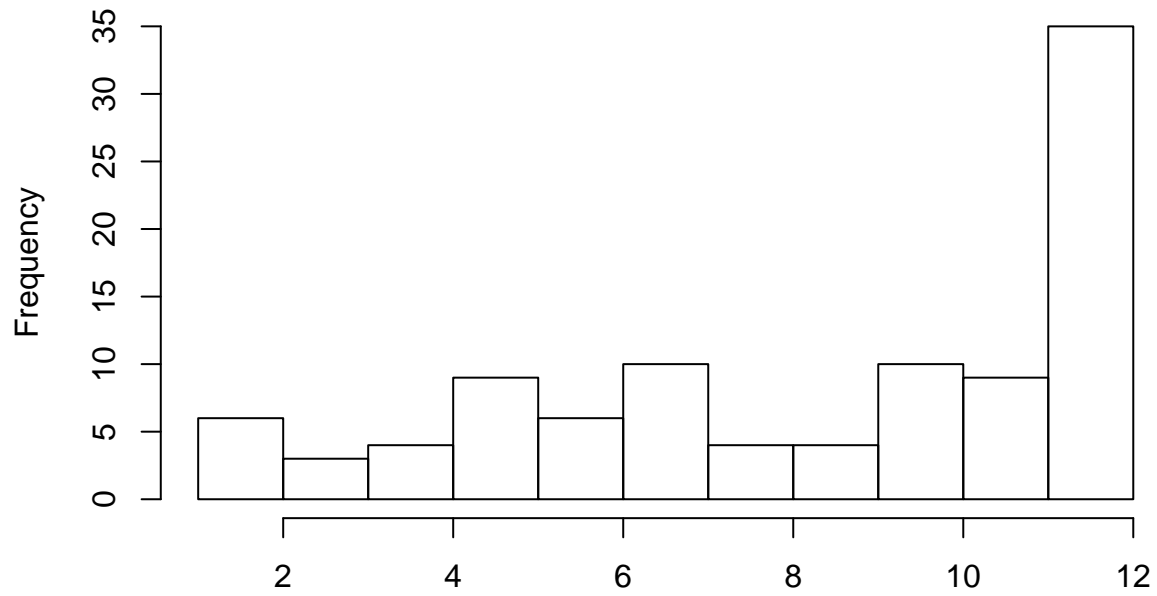
## 1.2 Average score



## 2. Important Features

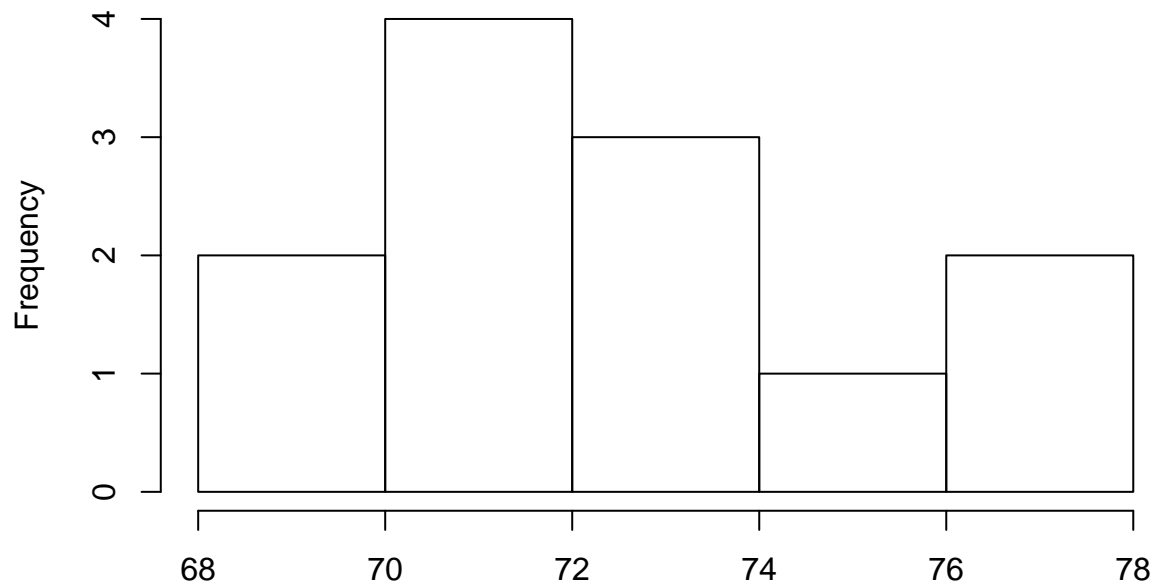
For Lasso, we calculate how many times a given features is being used in all the runs.

**distribution across 4 seed x 3 CV**



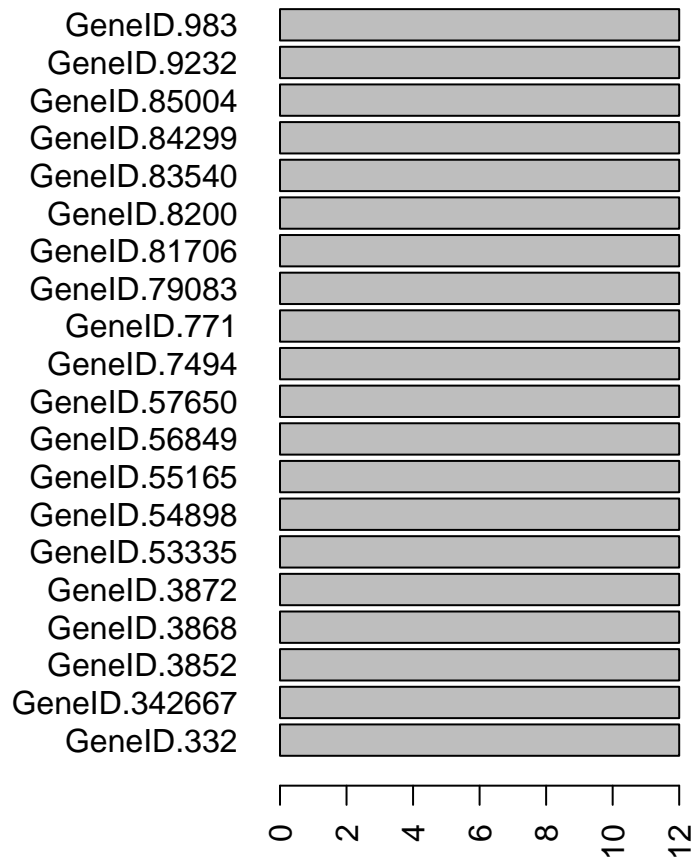
# of times a feature is selected by lasso (alpha= 0.5 )

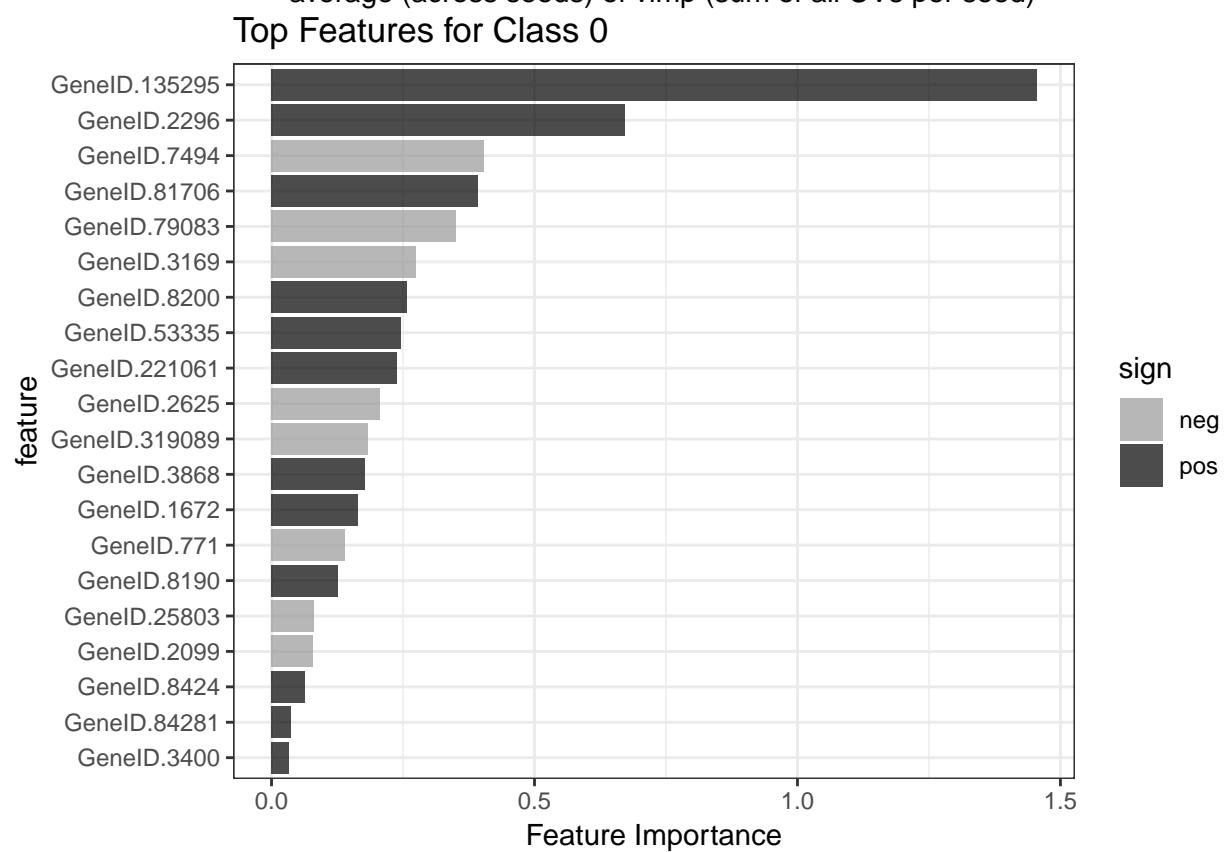
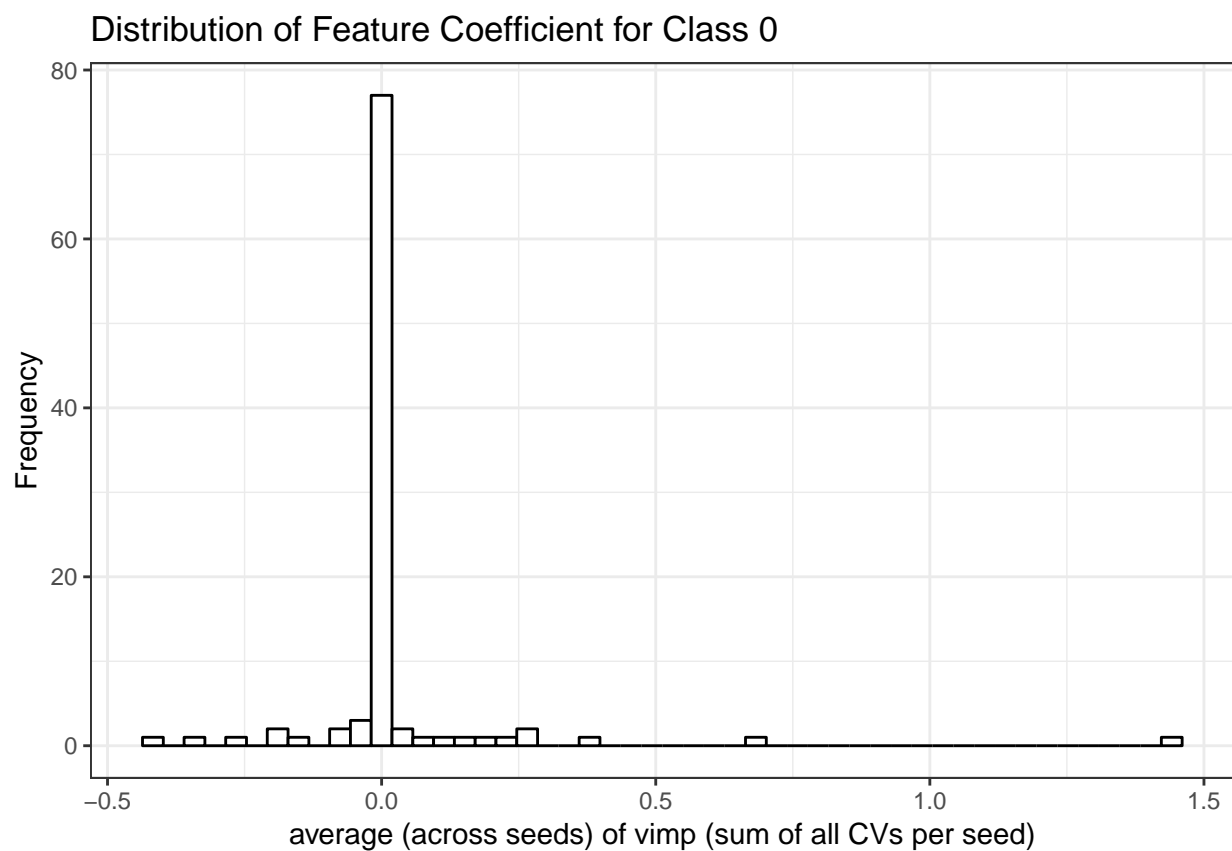
**summary of model size across 12 runs**



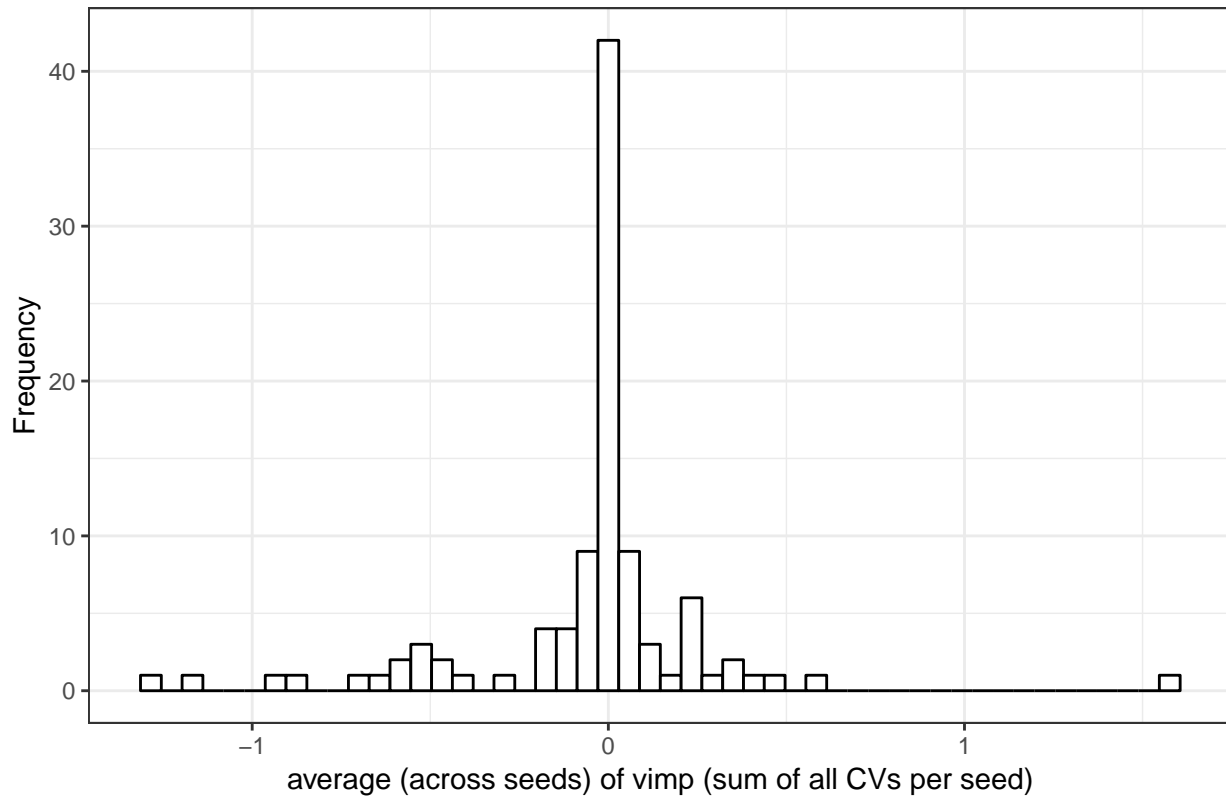
number of retained features in one run

### most used features

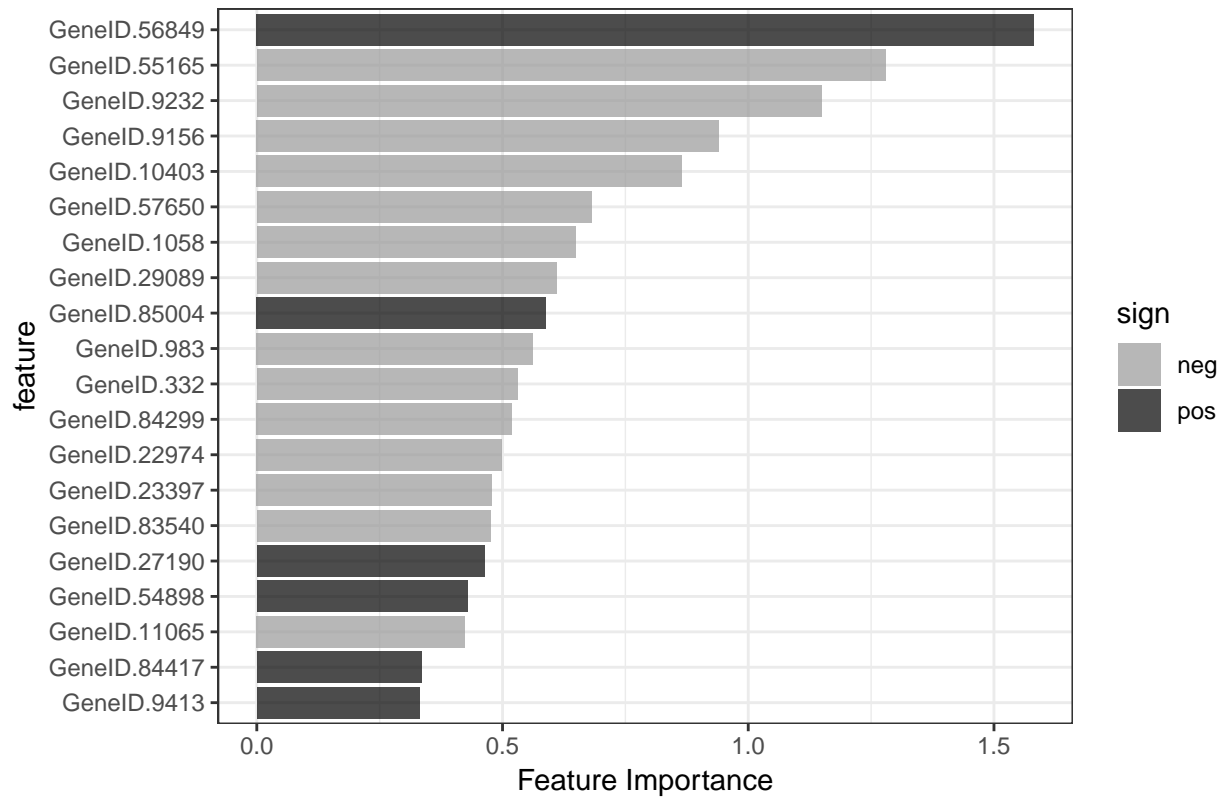




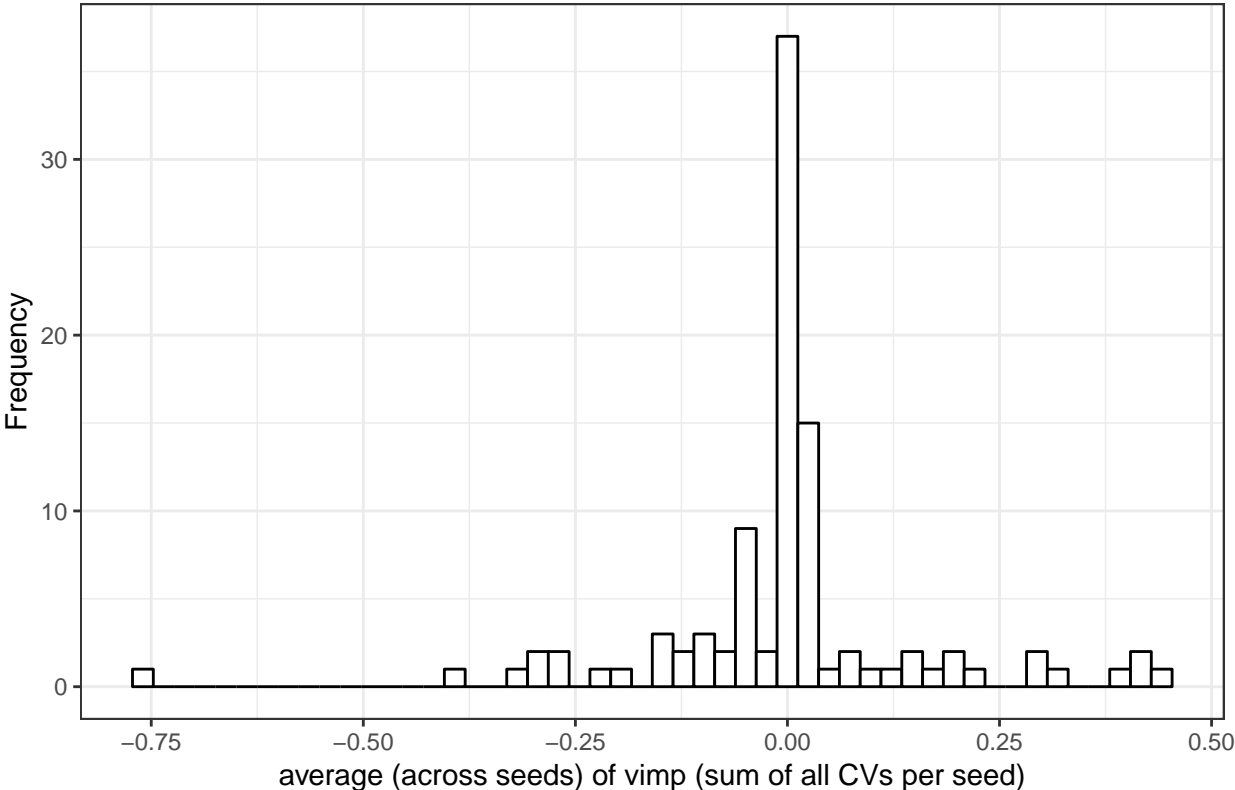
Distribution of Feature Coefficient for Class 1



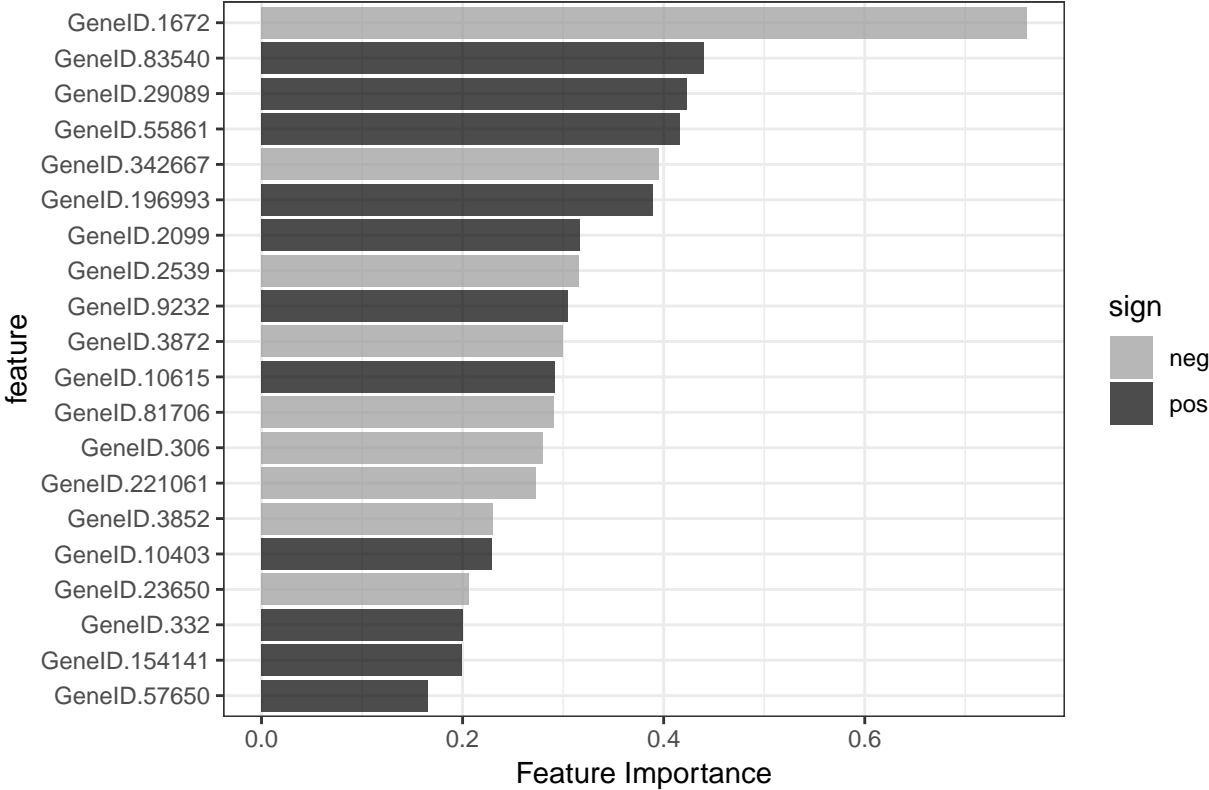
Top Features for Class 1



Distribution of Feature Coefficient for Class 2

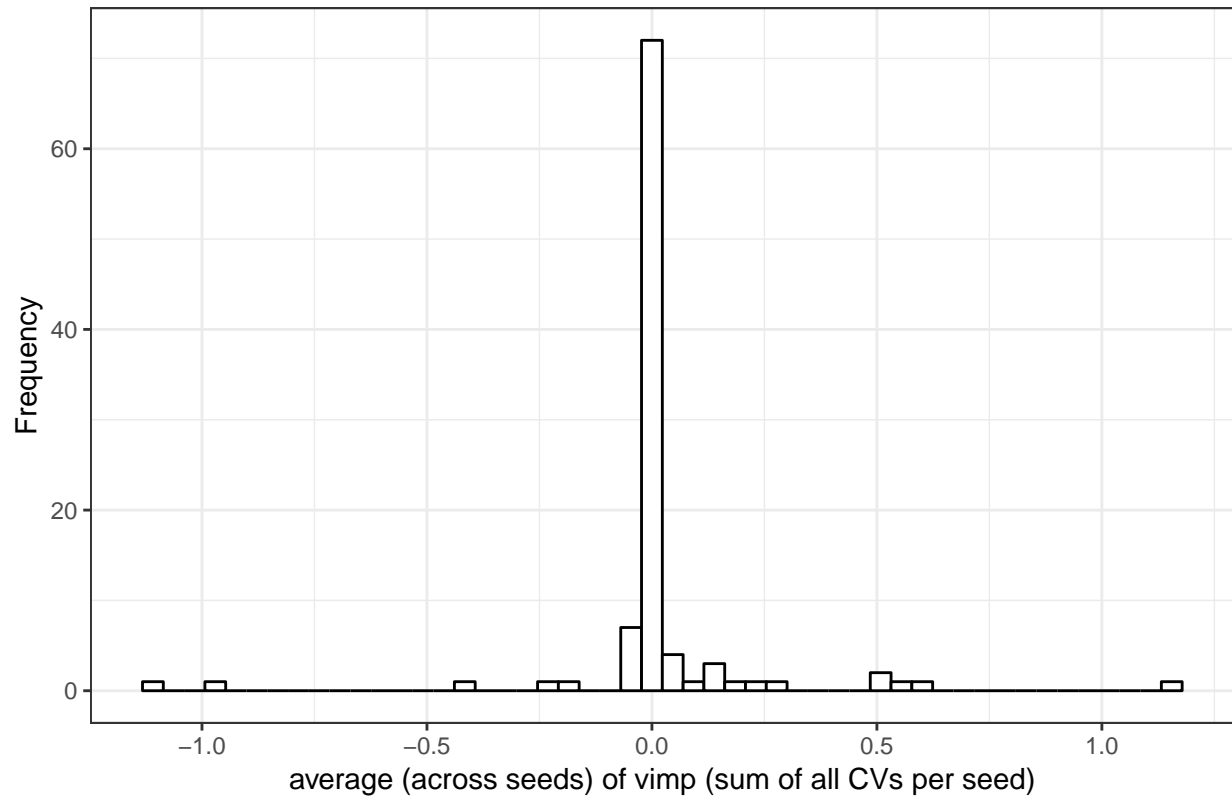


Top Features for Class 2





Distribution of Feature Coefficient for Class 3



Top Features for Class 3

