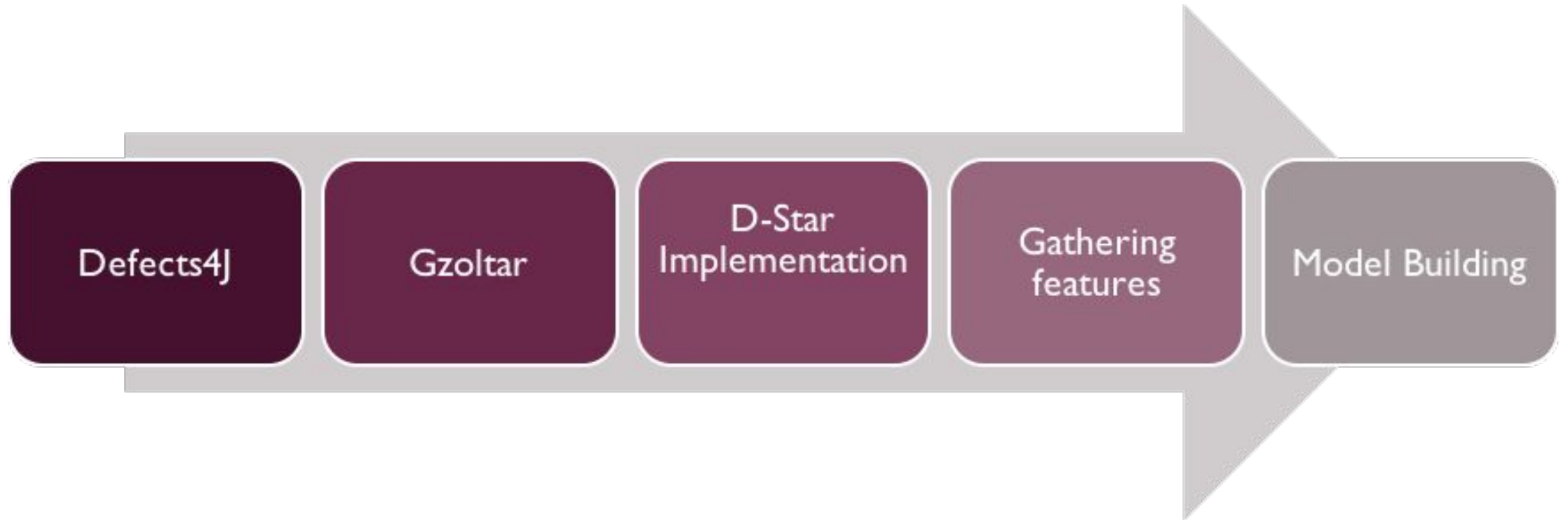

PRAKTIKUM: EXPLORING THE RELATIONSHIP BETWEEN DESIGN METRICS AND SOFTWARE DIAGNOSABILITY USING MACHINE LEARNING

HANDE KARATAY & FURKAN MERT ALGAN

JULY 13, 2018

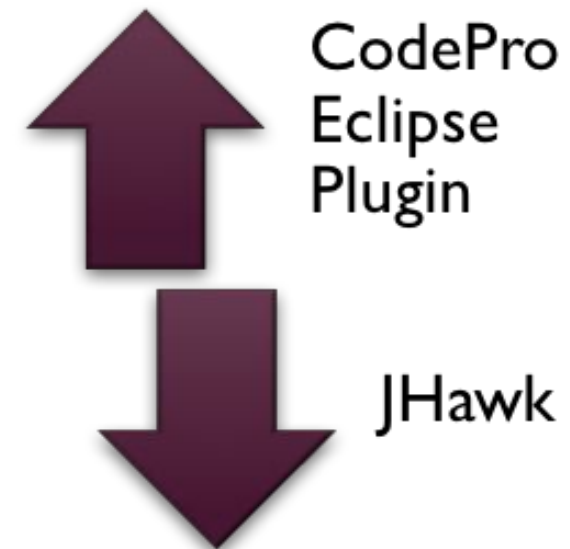
What have we done?



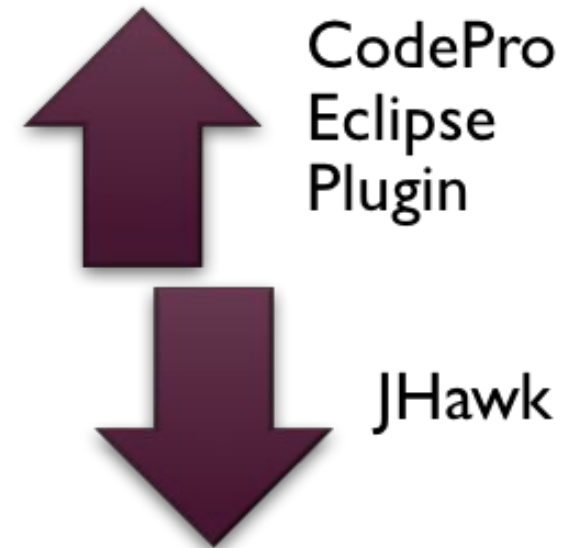
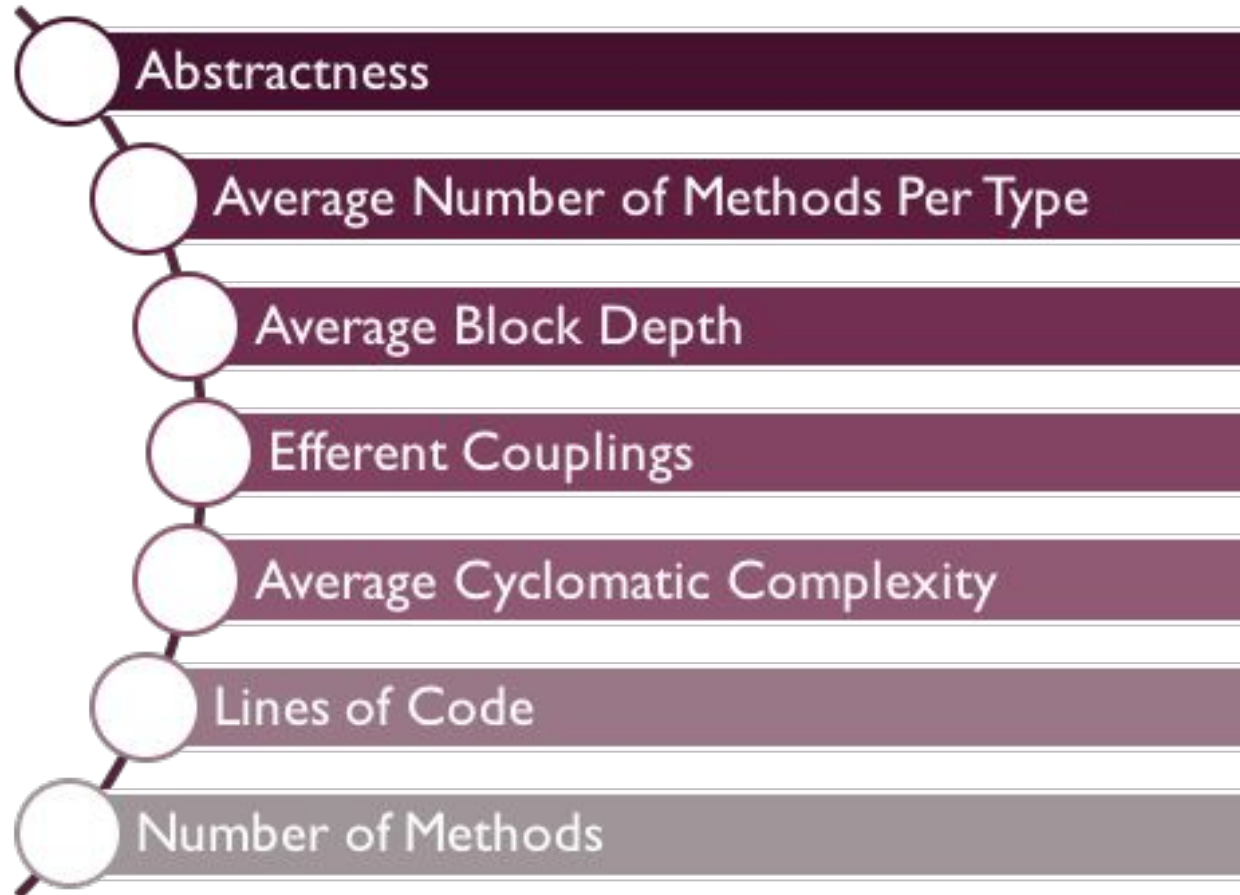
Gathering Features: Static Features

At first...

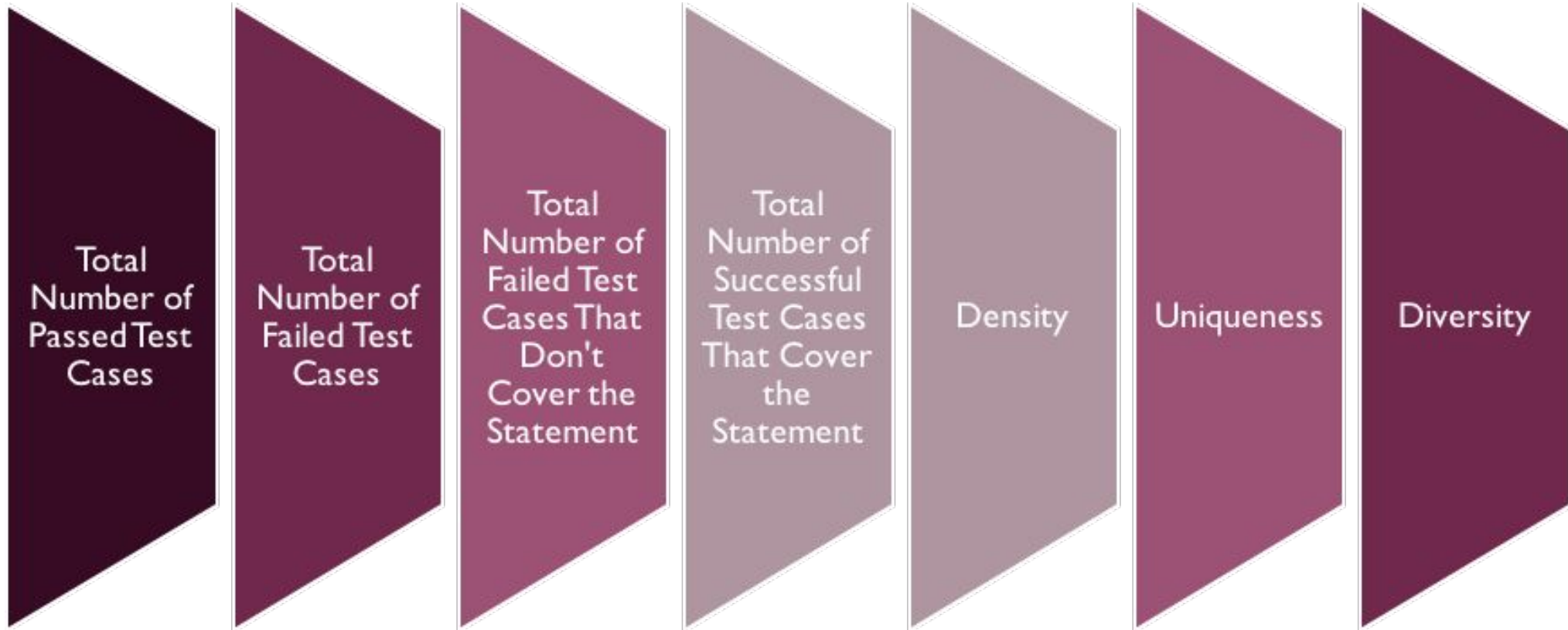
Abstractness	Lines of Code
Average Block Depth	Number of Characters
Average Cyclomatic Complexity	Number of Comments
Average Lines of Code Per Method	Number of Constructors
Average Number of Constructors Per Type	Number of Fields
Average Number of Fields Per Type	Number of Lines
Average Number of Methods Per Type	Number of Methods
Average Number of Parameters	Number of Packages
Comments Ratio	Number of Semicolons
Efferent Couplings	Number of Types



Gathering Features: Static Features



Gathering Features: Test Suite Characteristics



Gathering Features: Bug Characteristics

**Extracted
from the ARP
Platform**

File Count

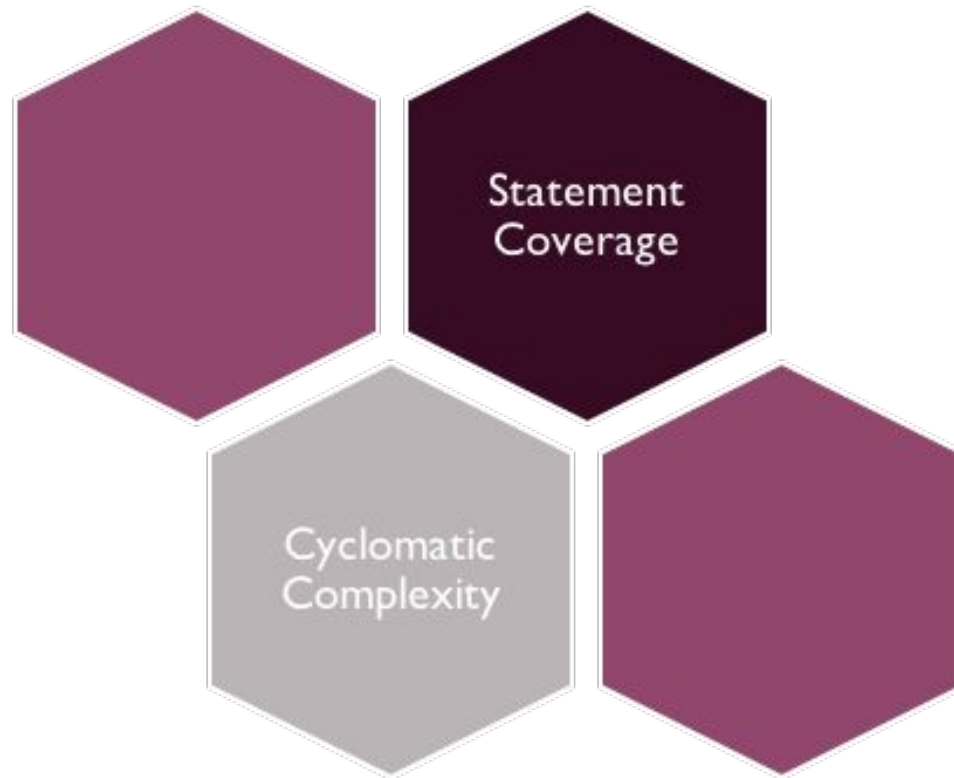
Line Count

Statement Coverage/Dynamic

Relevant Test Count

Triggering Test Count

Gathering Features: Dynamic Features



Our Target Value

- Use **D-Star** algorithm to rank methods

$$DStar(s) = \frac{(N_{cf})^*}{N_{uf} + N_{cs}}$$

- 1: Rank 5 between 200 methods

N_{cf} : Number of failed test cases that cover the statement

- 2: Rank 5 between 100 methods

N_{uf} : Number of failed test cases that do not cover the statement

N_{cs} : Number of successful test cases that cover the statement

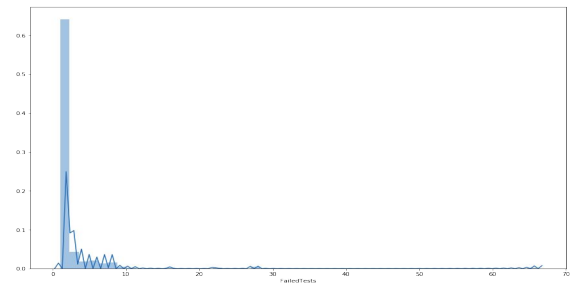
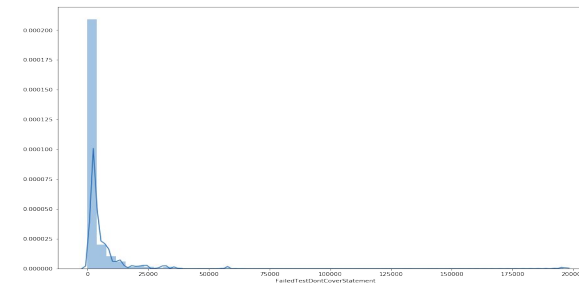
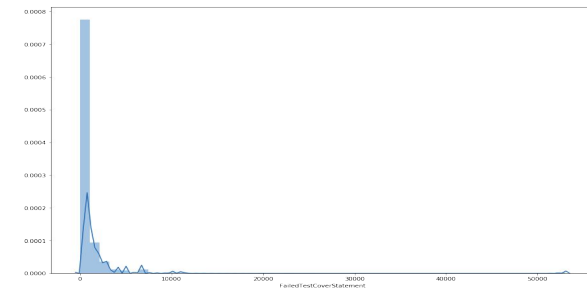
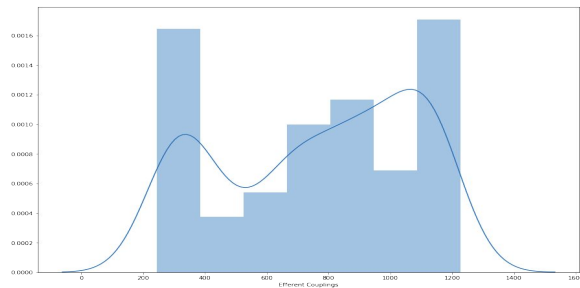
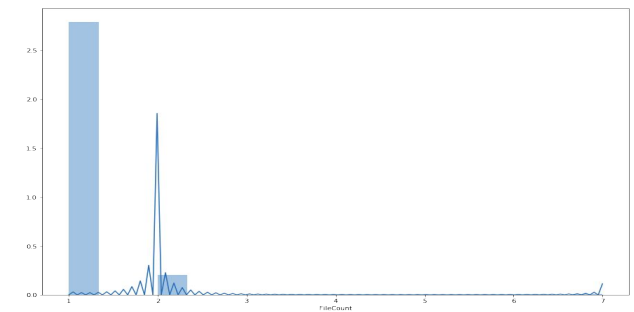
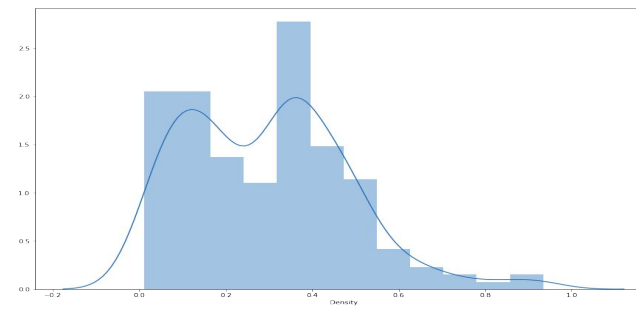
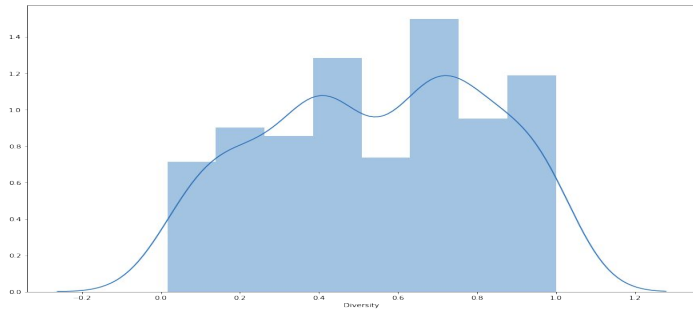
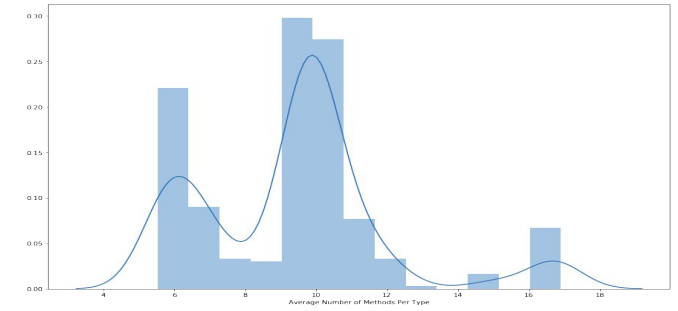
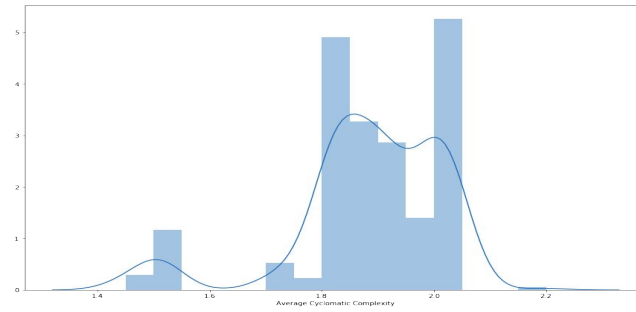
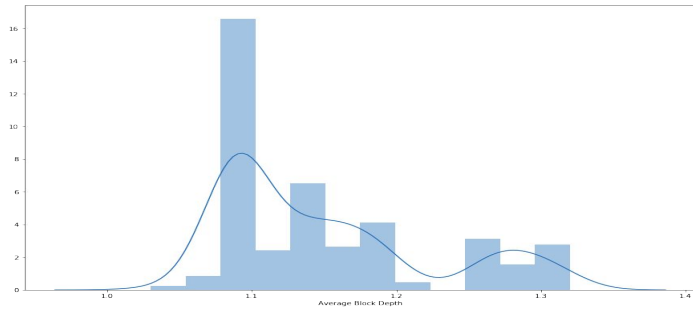
- Proposed Solution: Normalize

$$target\ value = \frac{Dstar\ rank}{number\ of\ methods\ in\ buggy\ version}$$

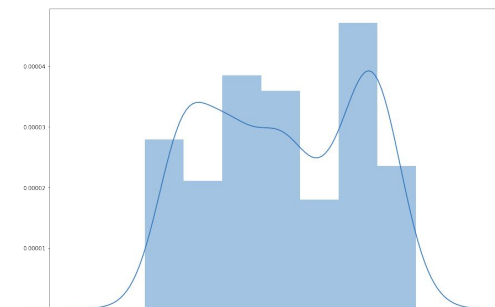
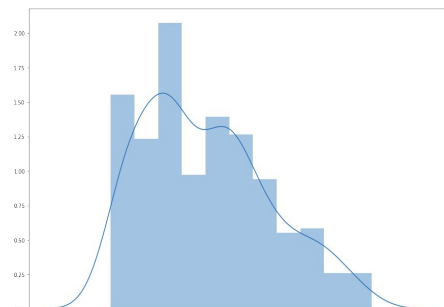
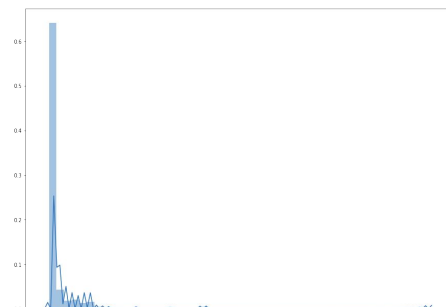
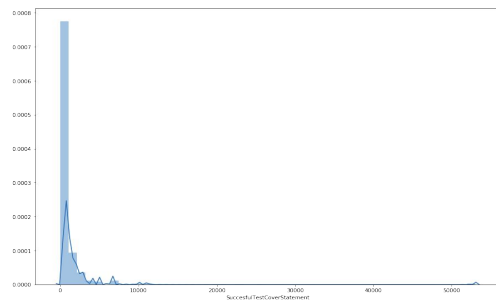
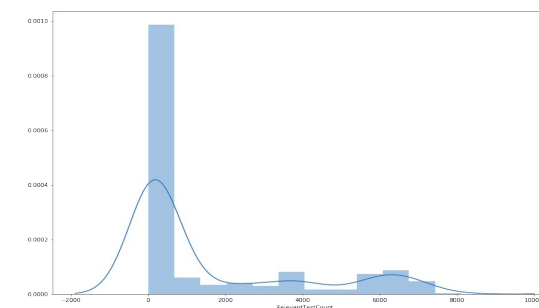
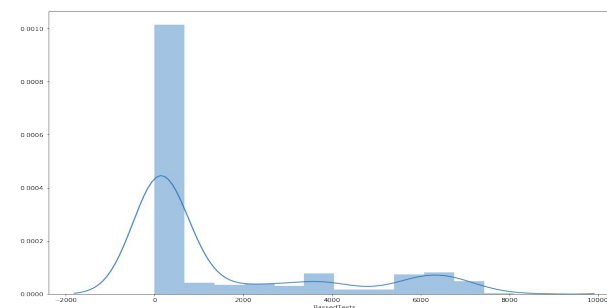
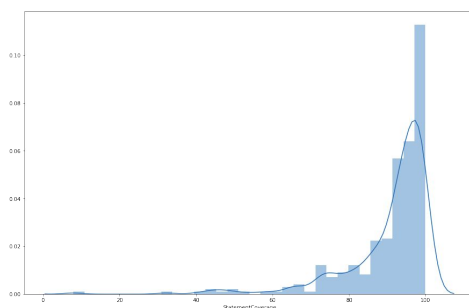
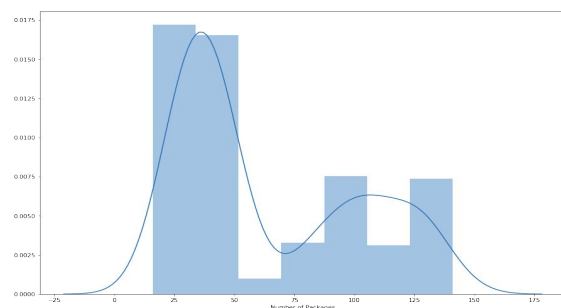
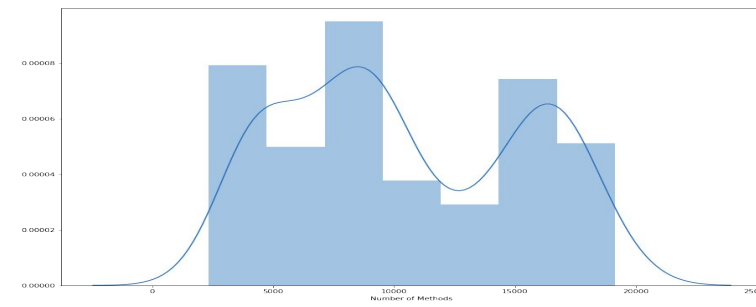
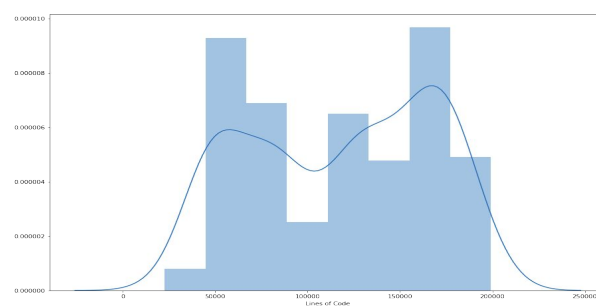
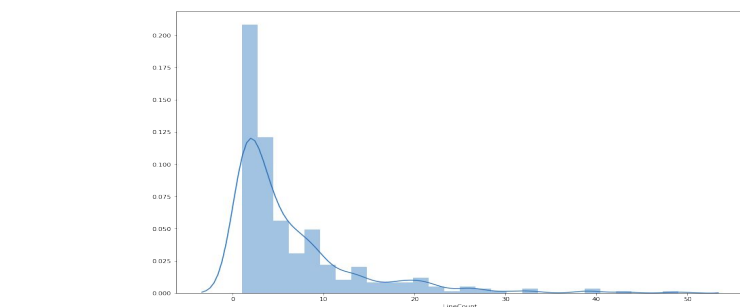
Model Building



Metric Statistics

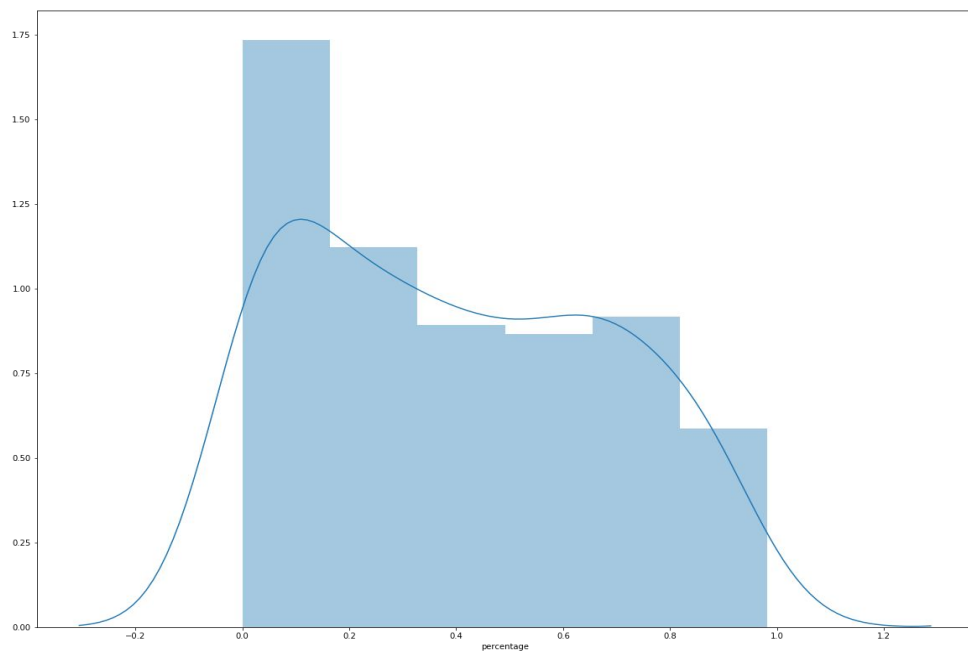


Metric Statistics

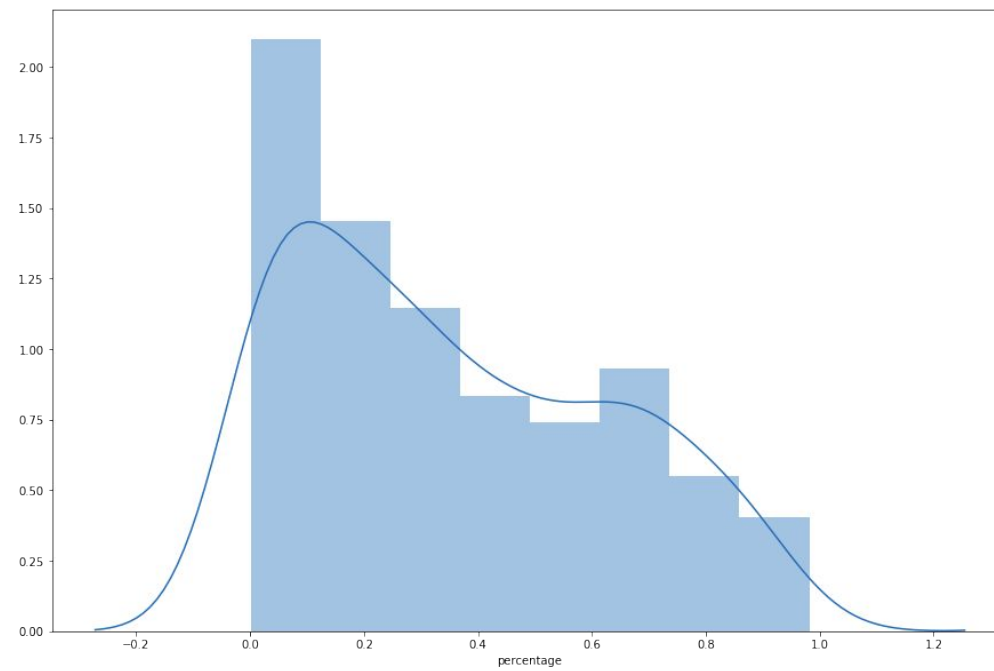


Target Statistics

With More Features

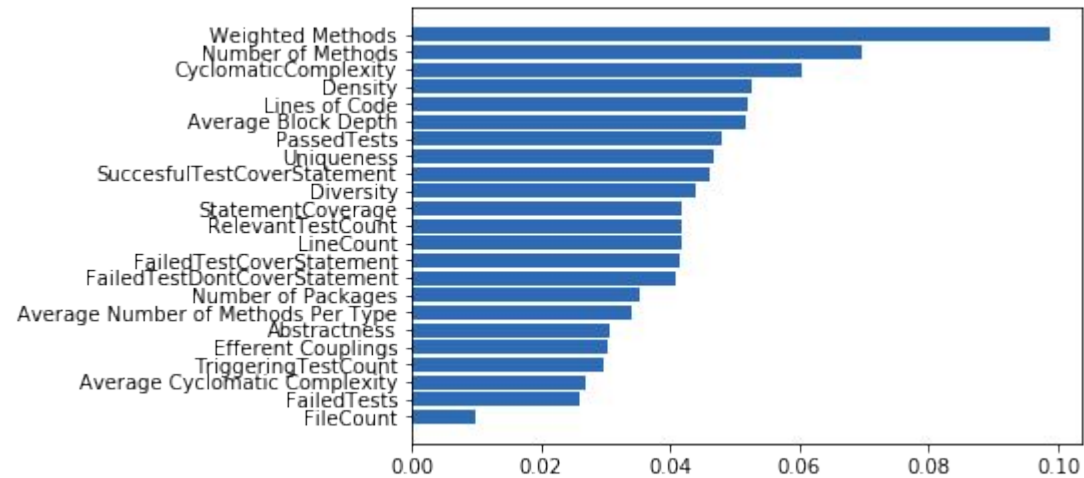


With More Data

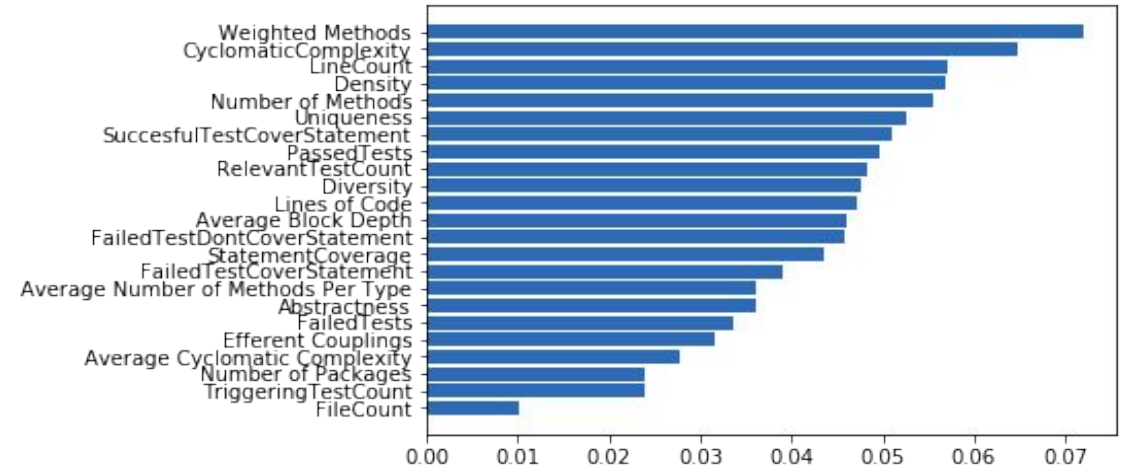


Feature Selection

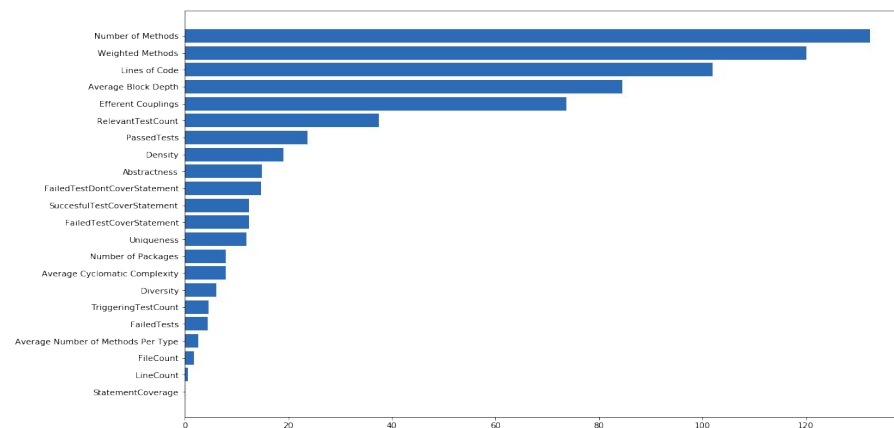
4 classes



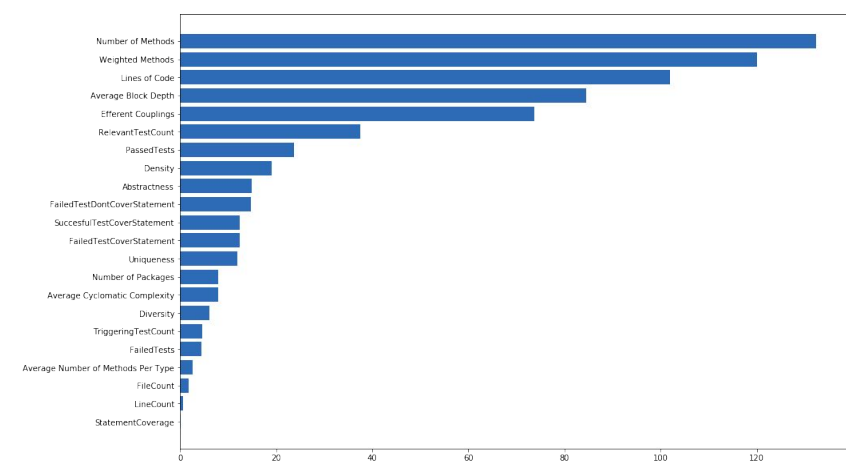
6 Classes



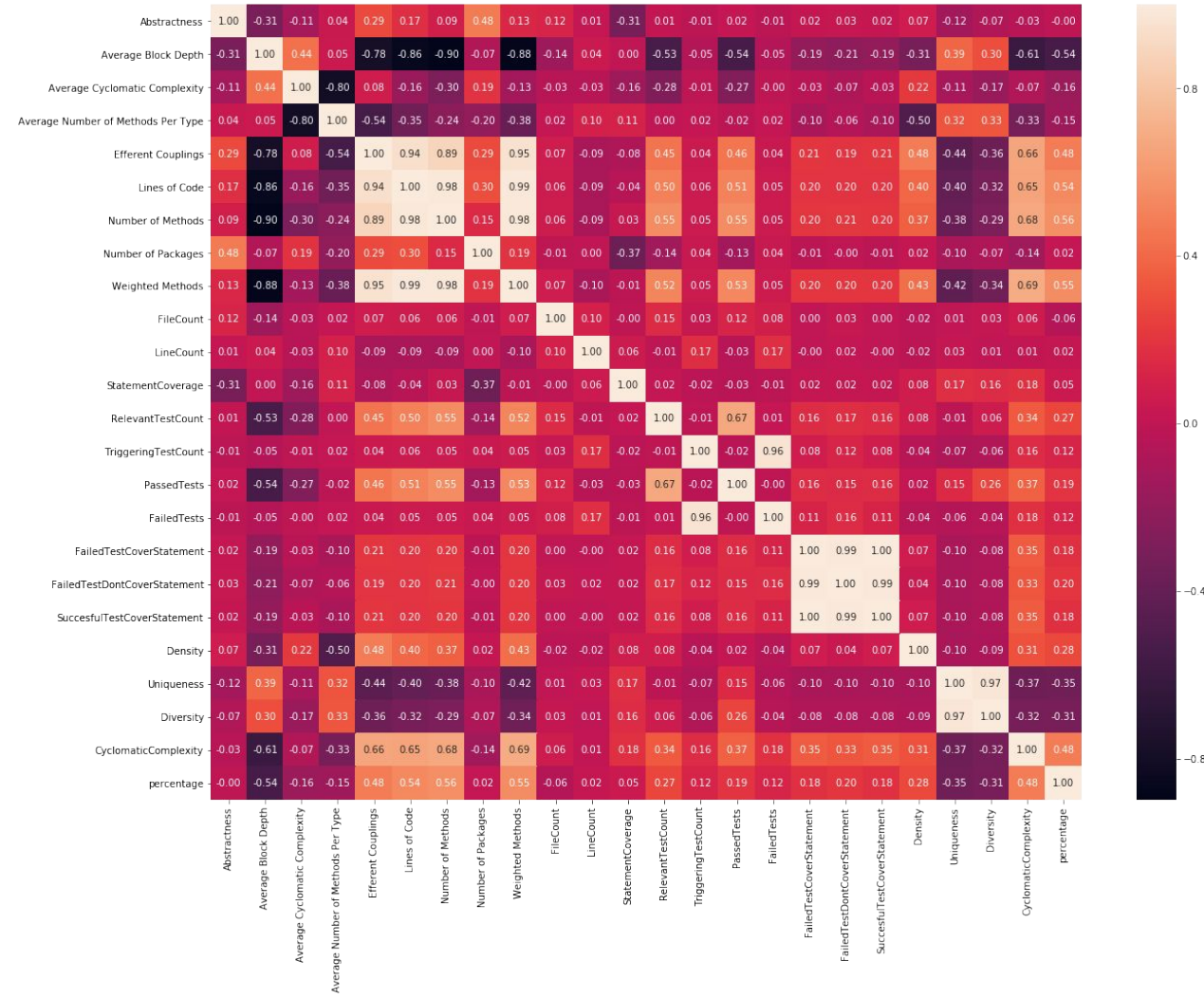
F Test



P Test

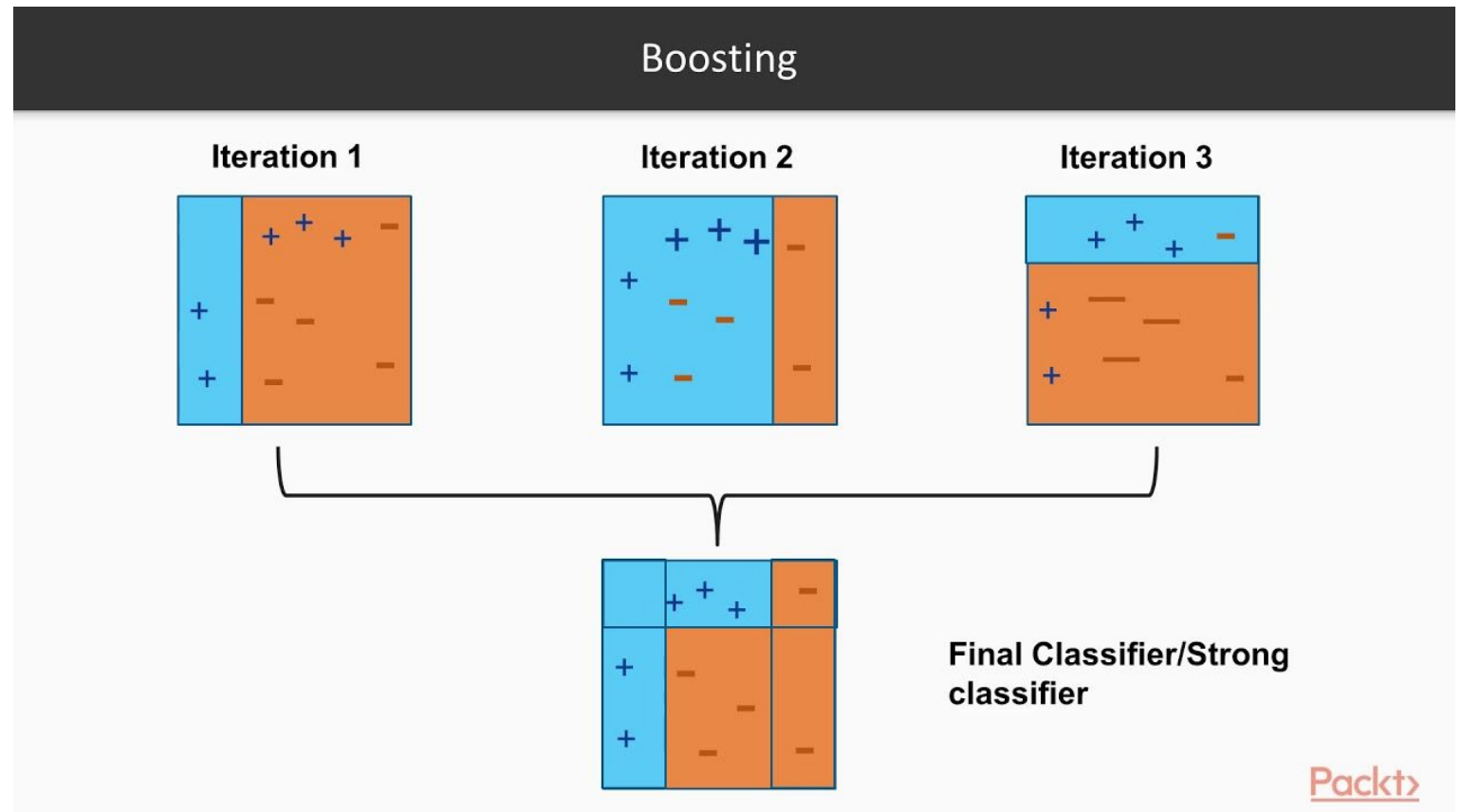


Correlations of Metrics

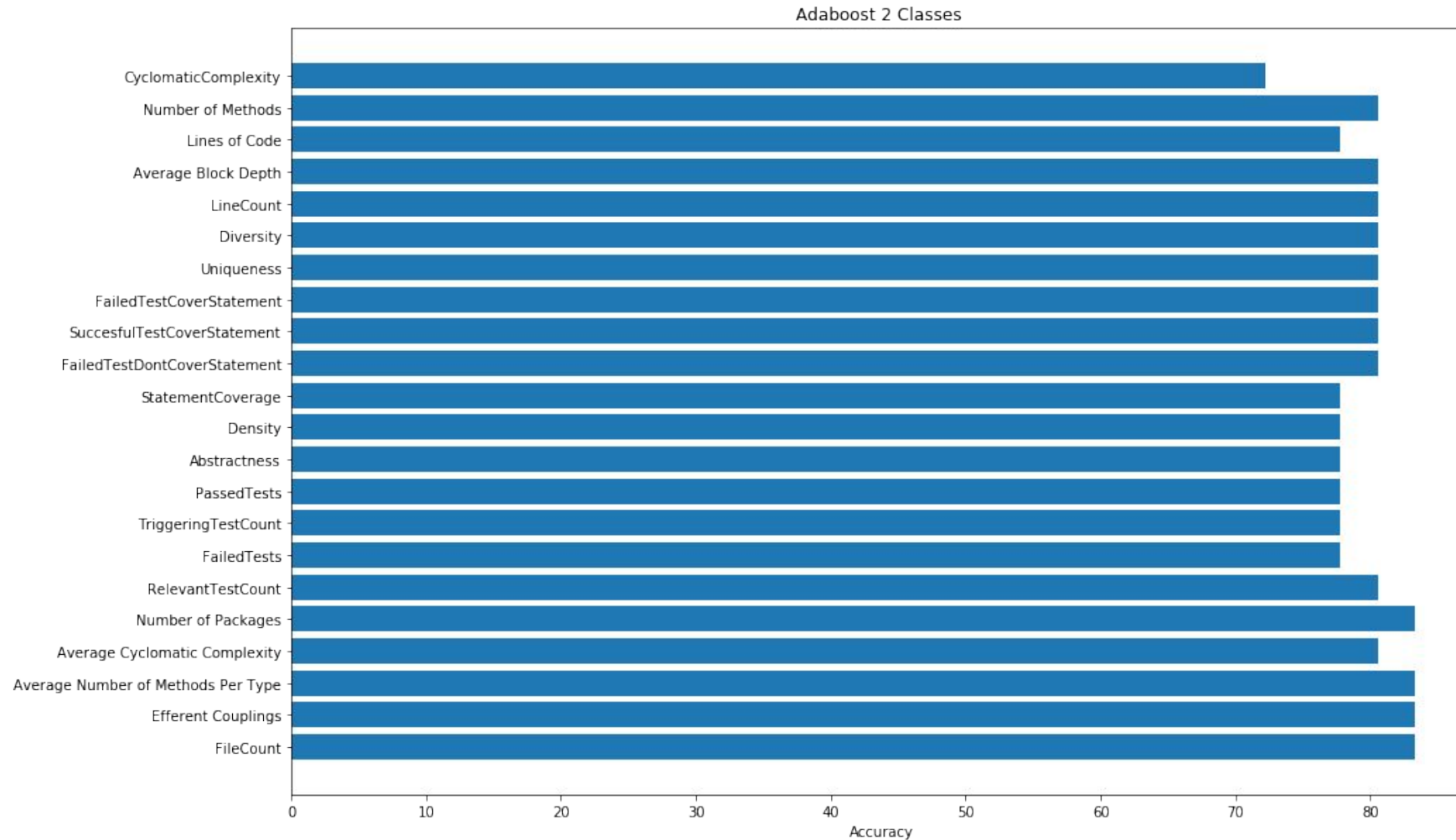


Adaboost

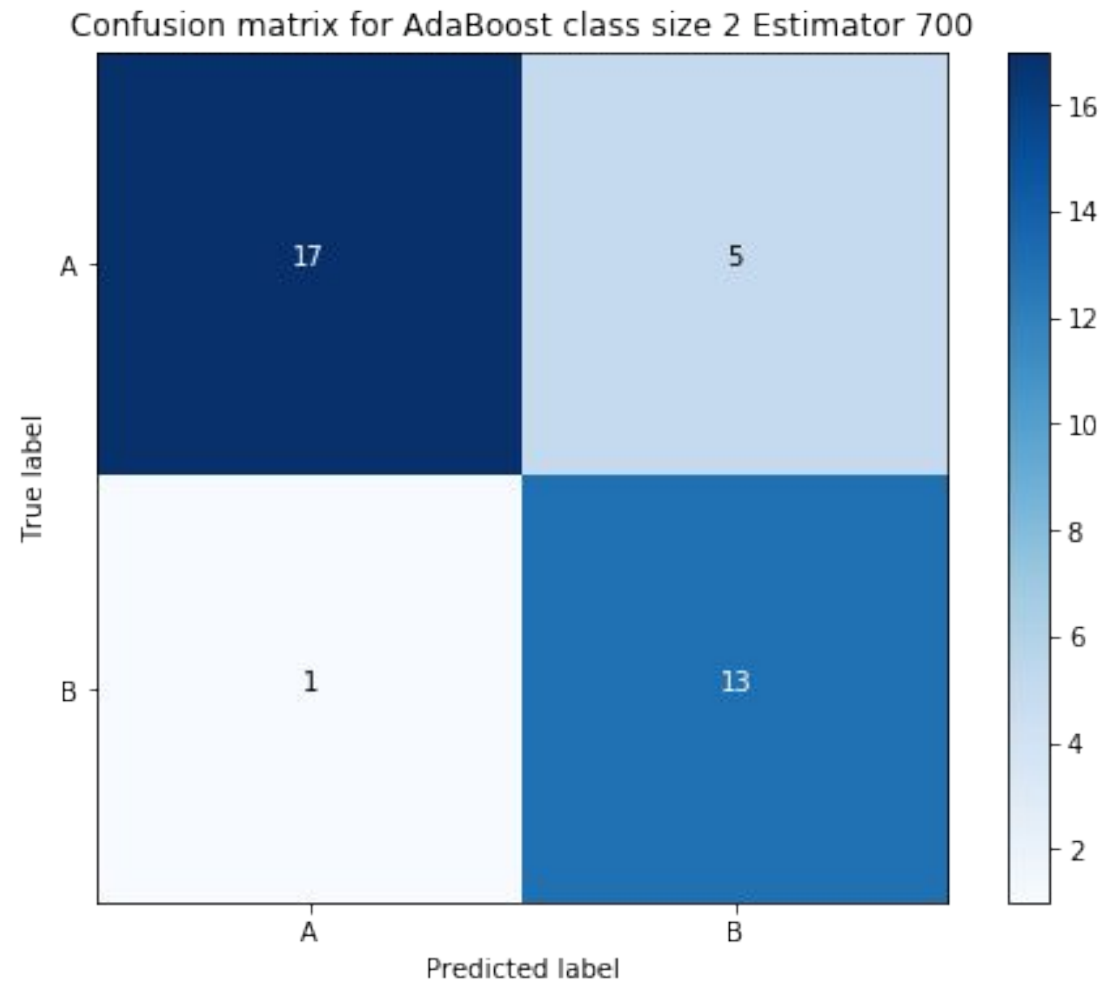
- Augmentation
- Fast



Adaboost(2 Classes)

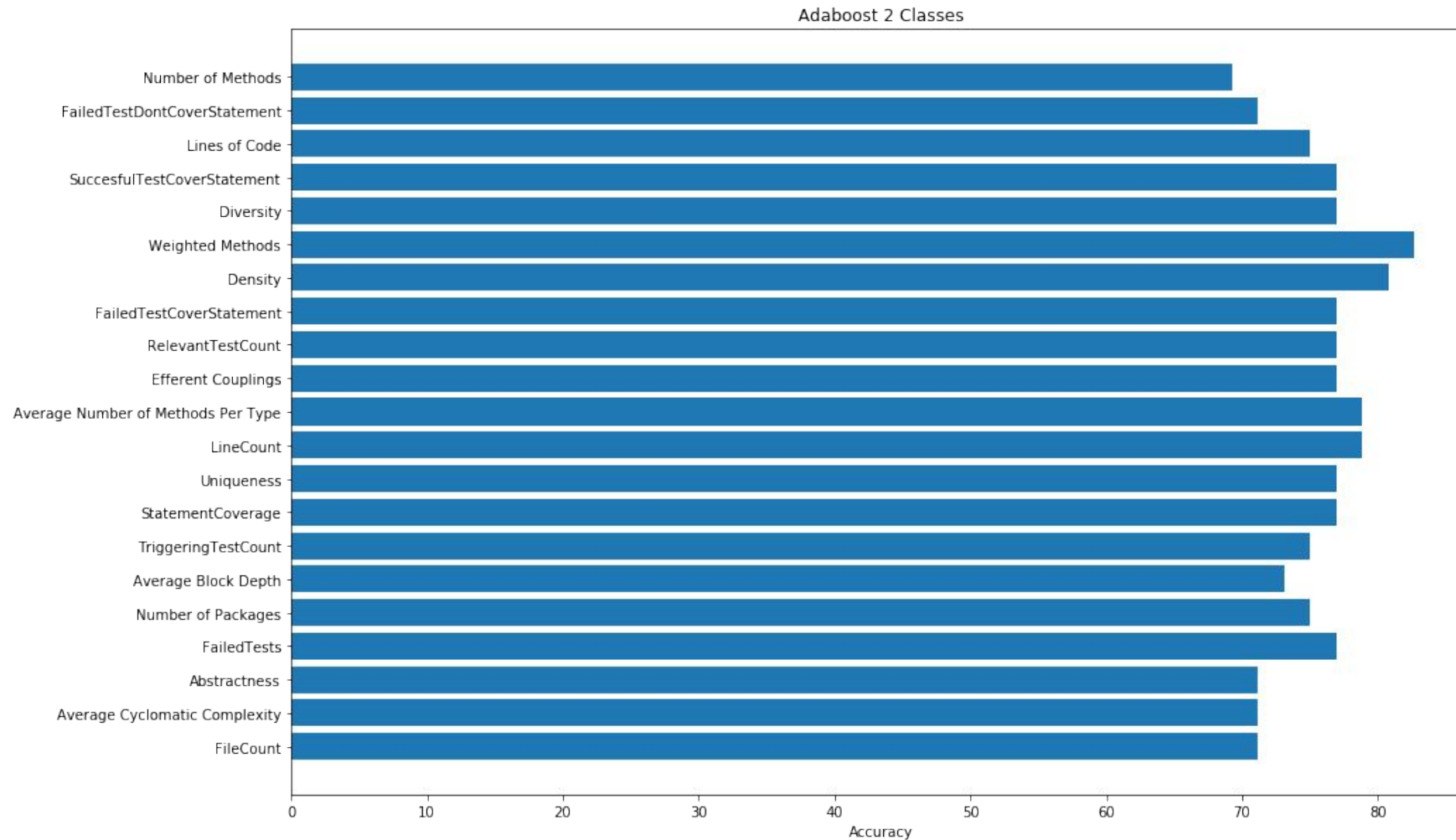


Adaboost

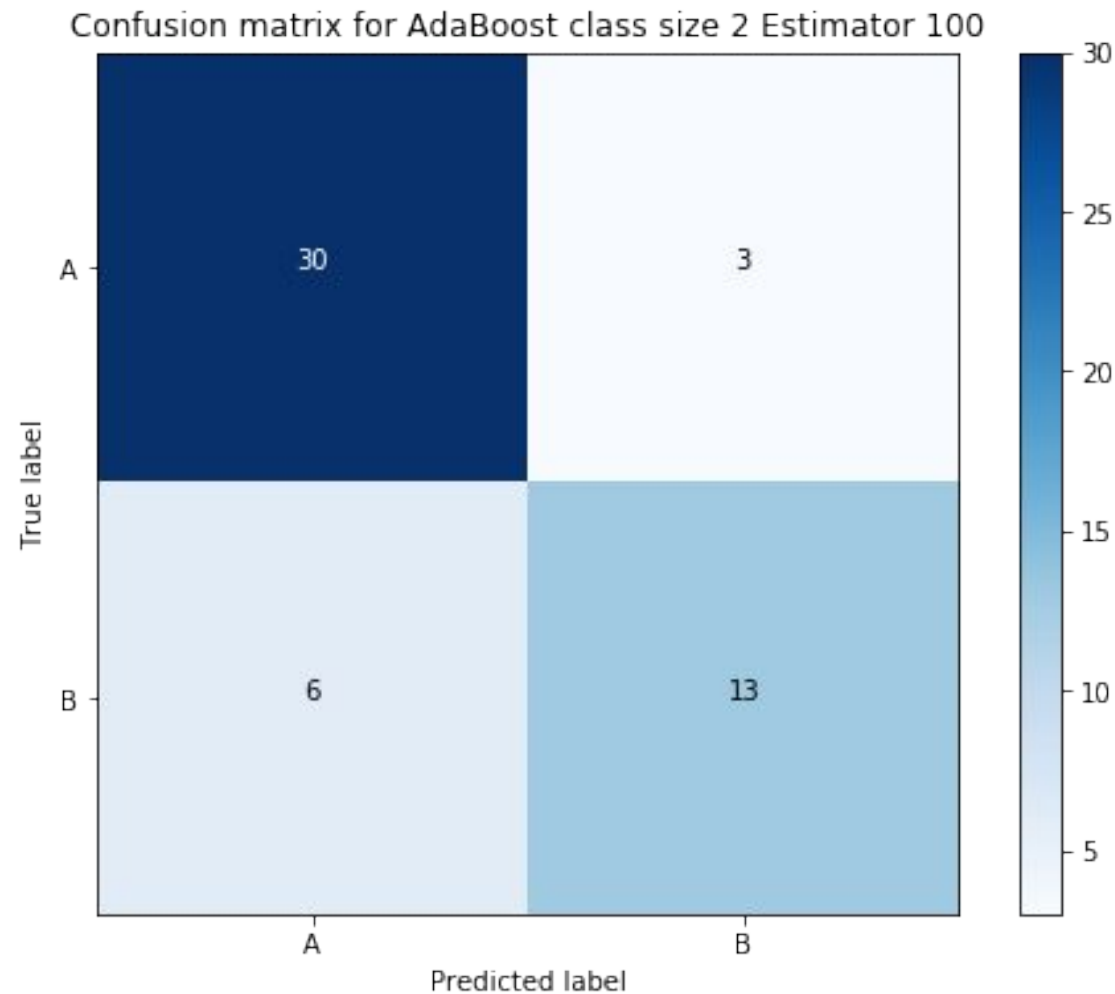


Acc. = 83%

Adaboost with Math



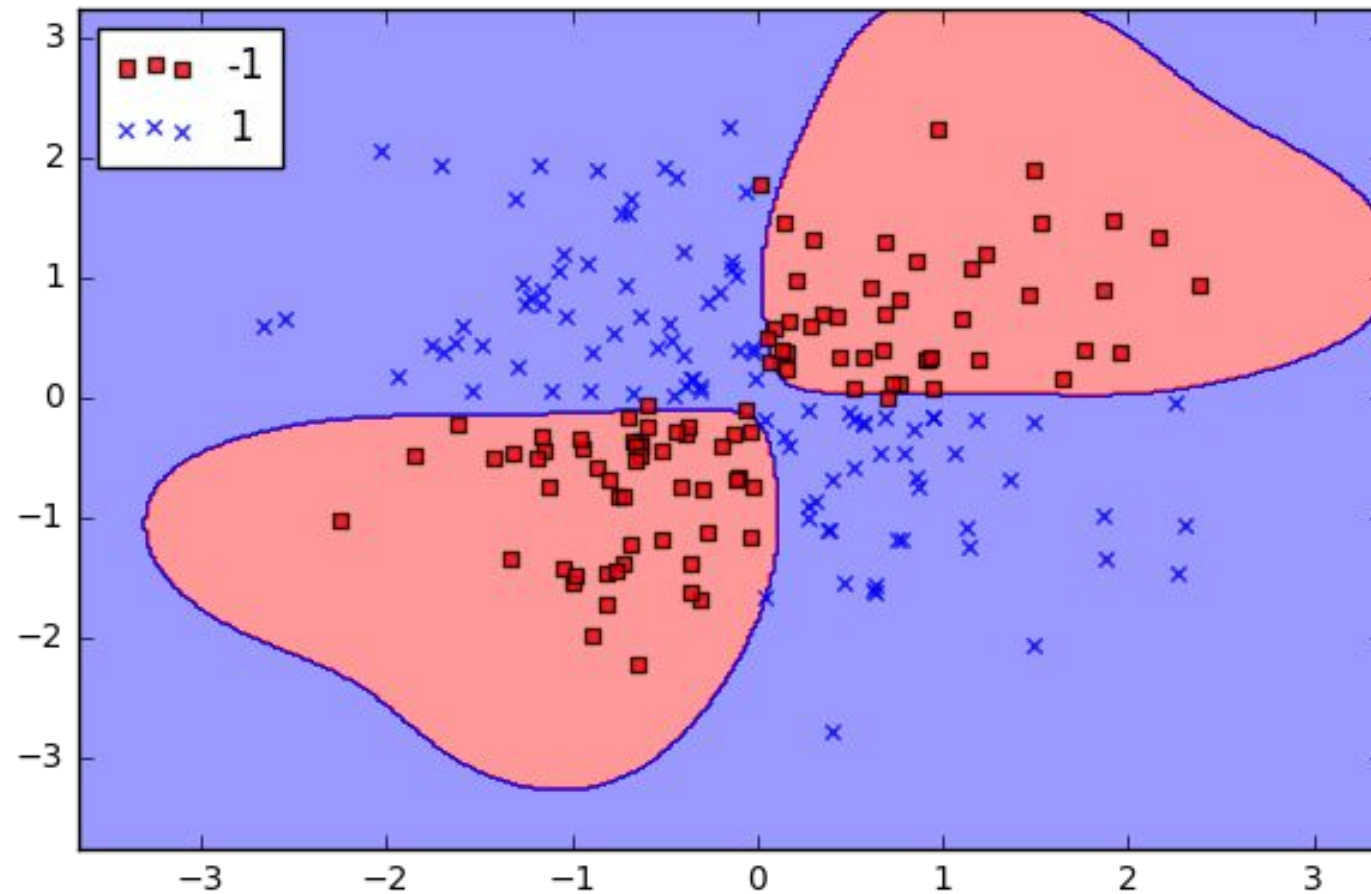
Adaboost with Math



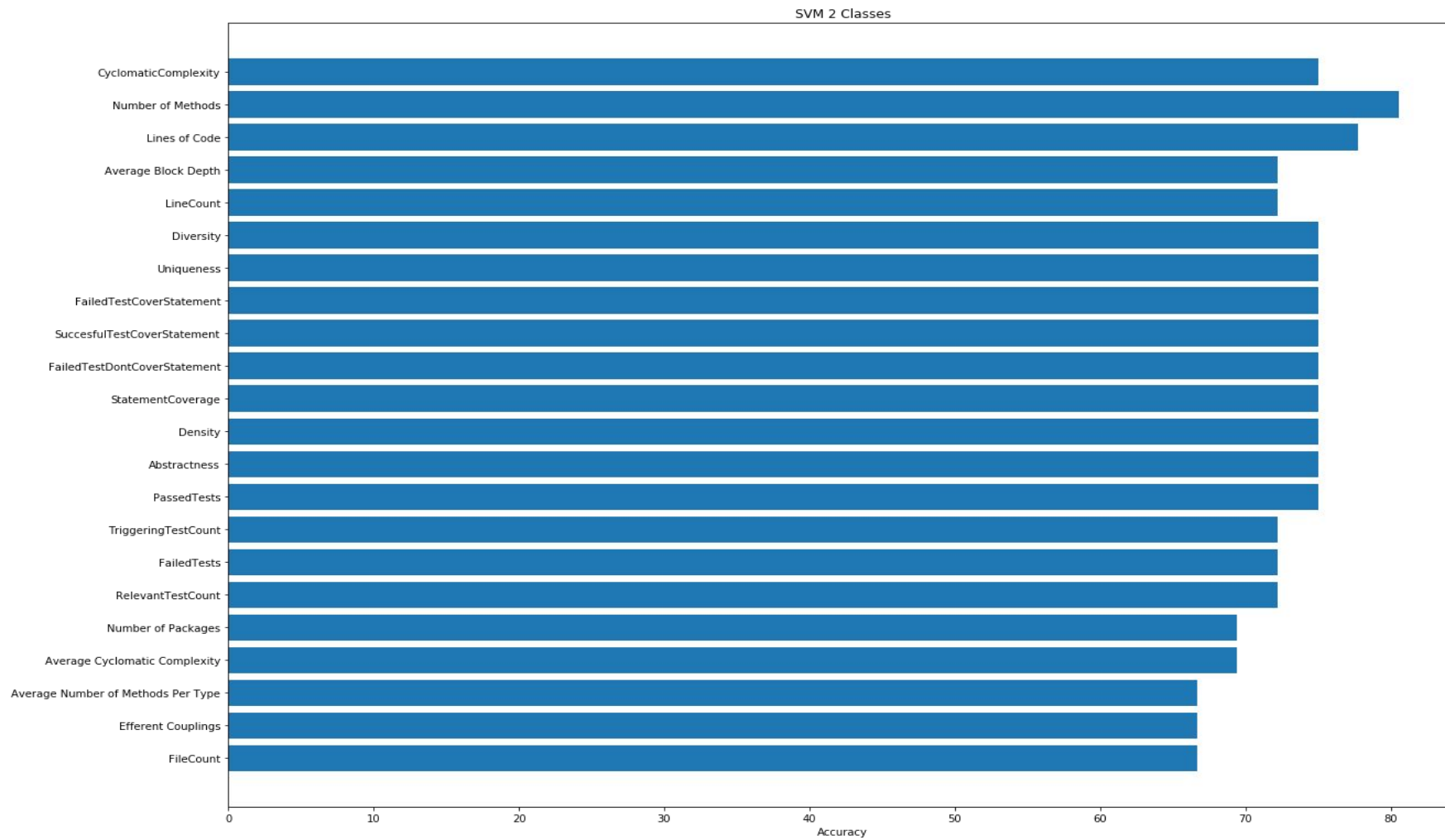
Acc. = 82.6%

SVM

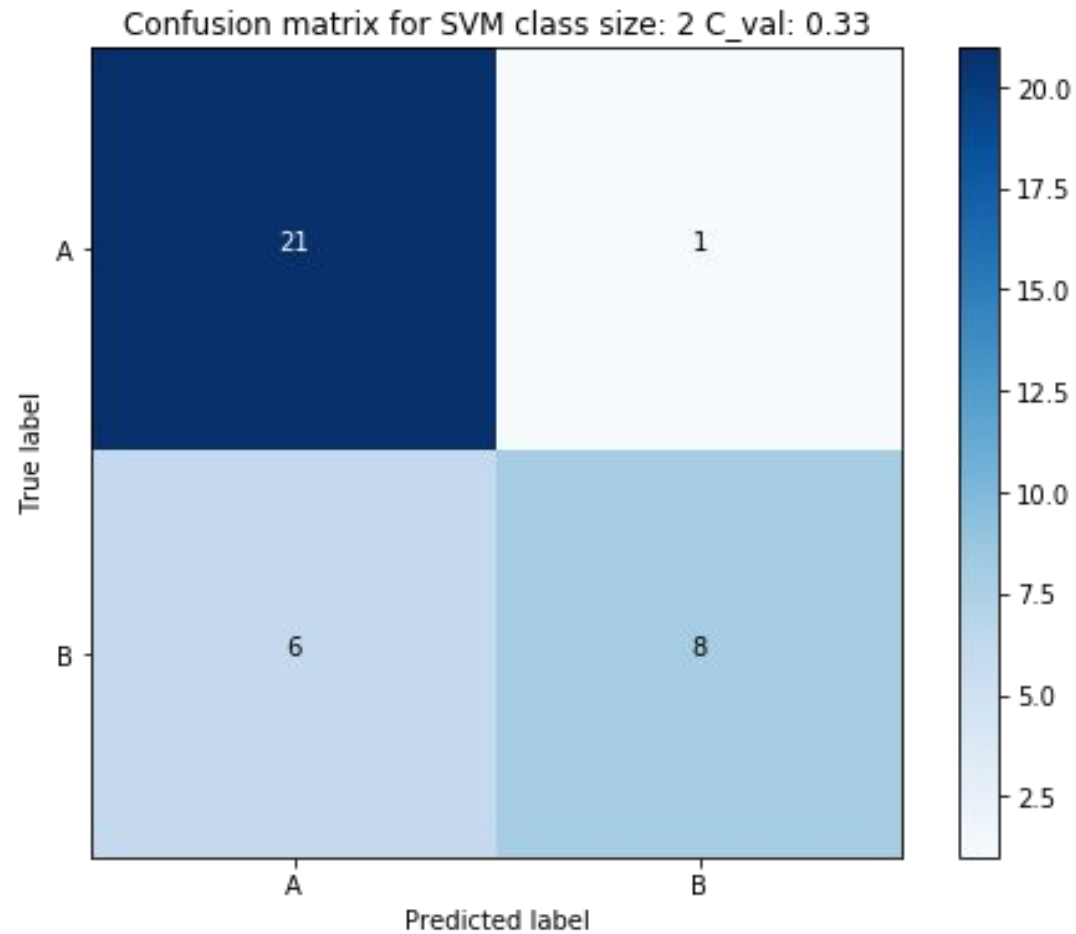
- Kernel
- Robust



SVM(2 classes)

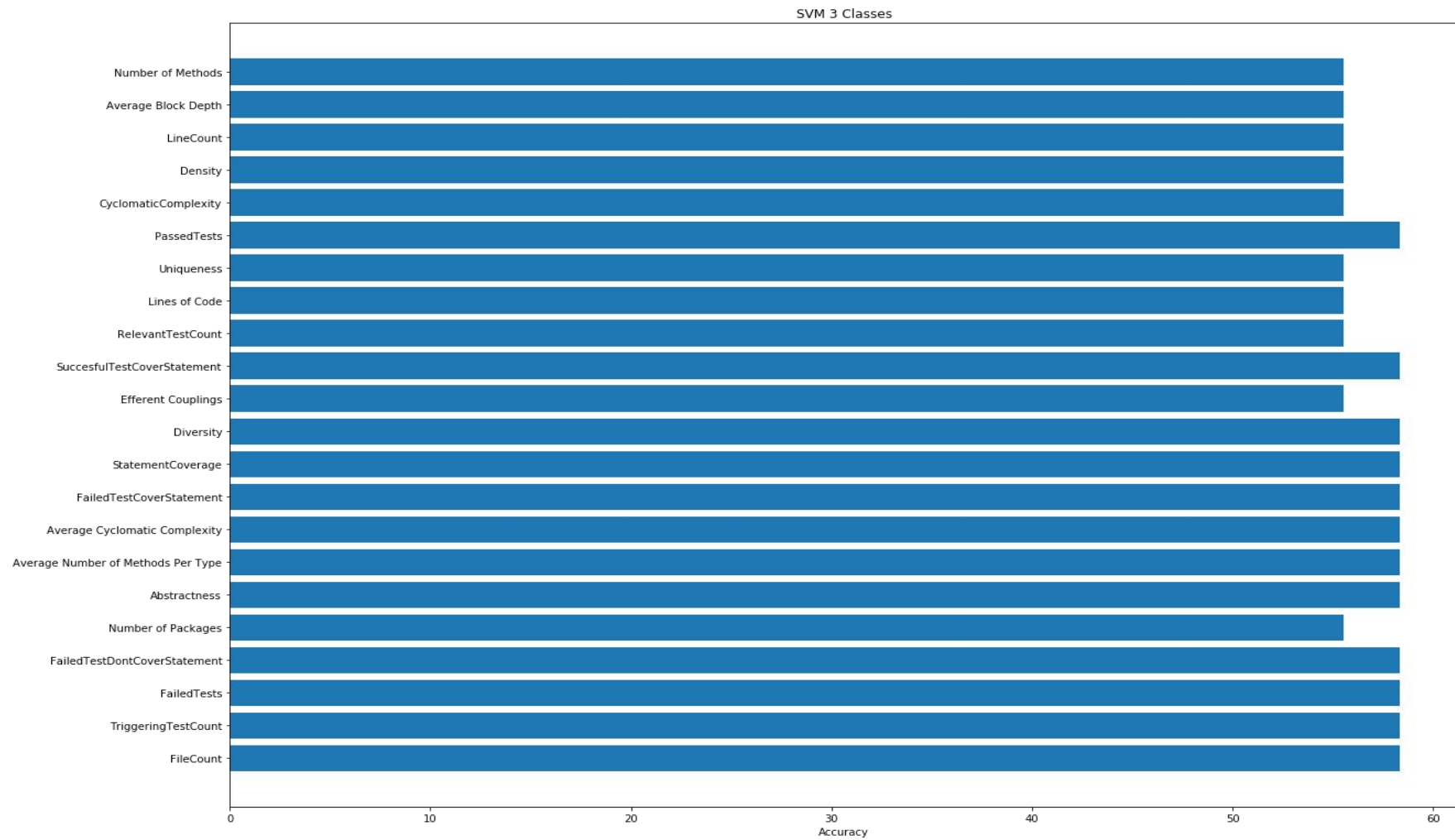


SVM(2 classes)

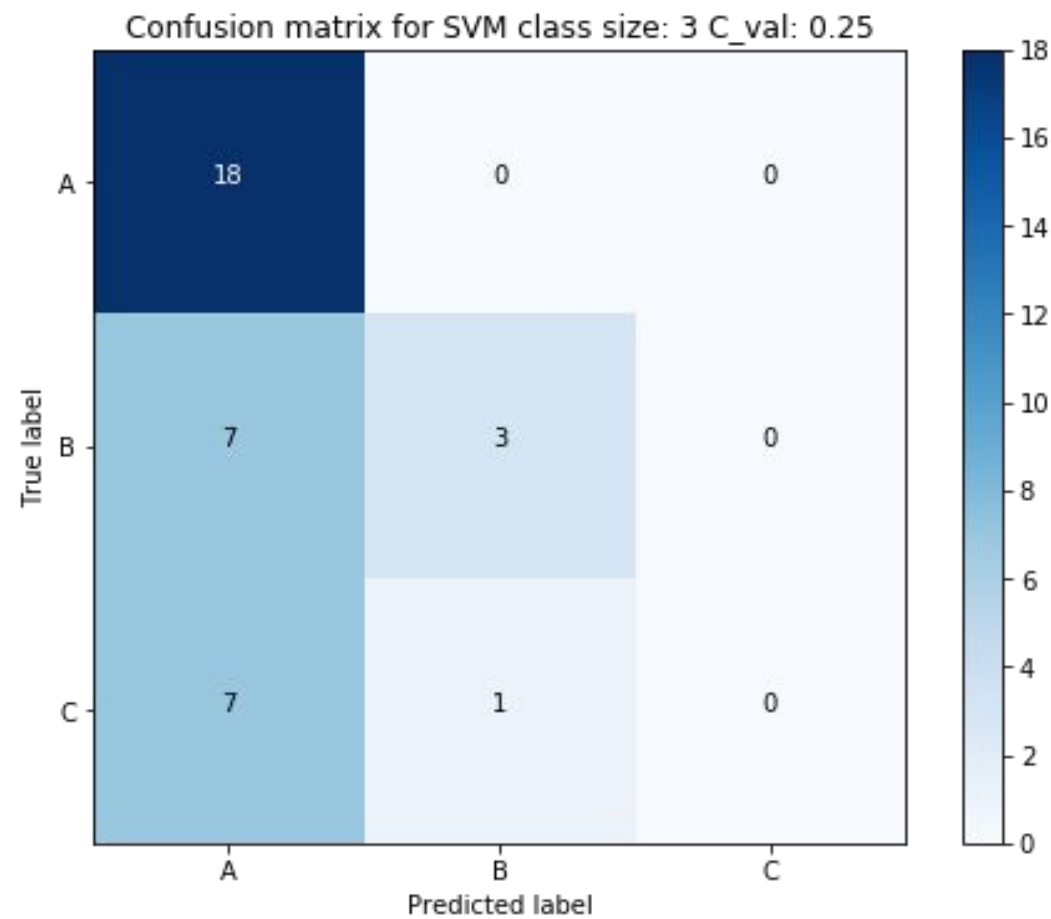


Acc. = 80.5%

SVM (3 classes)

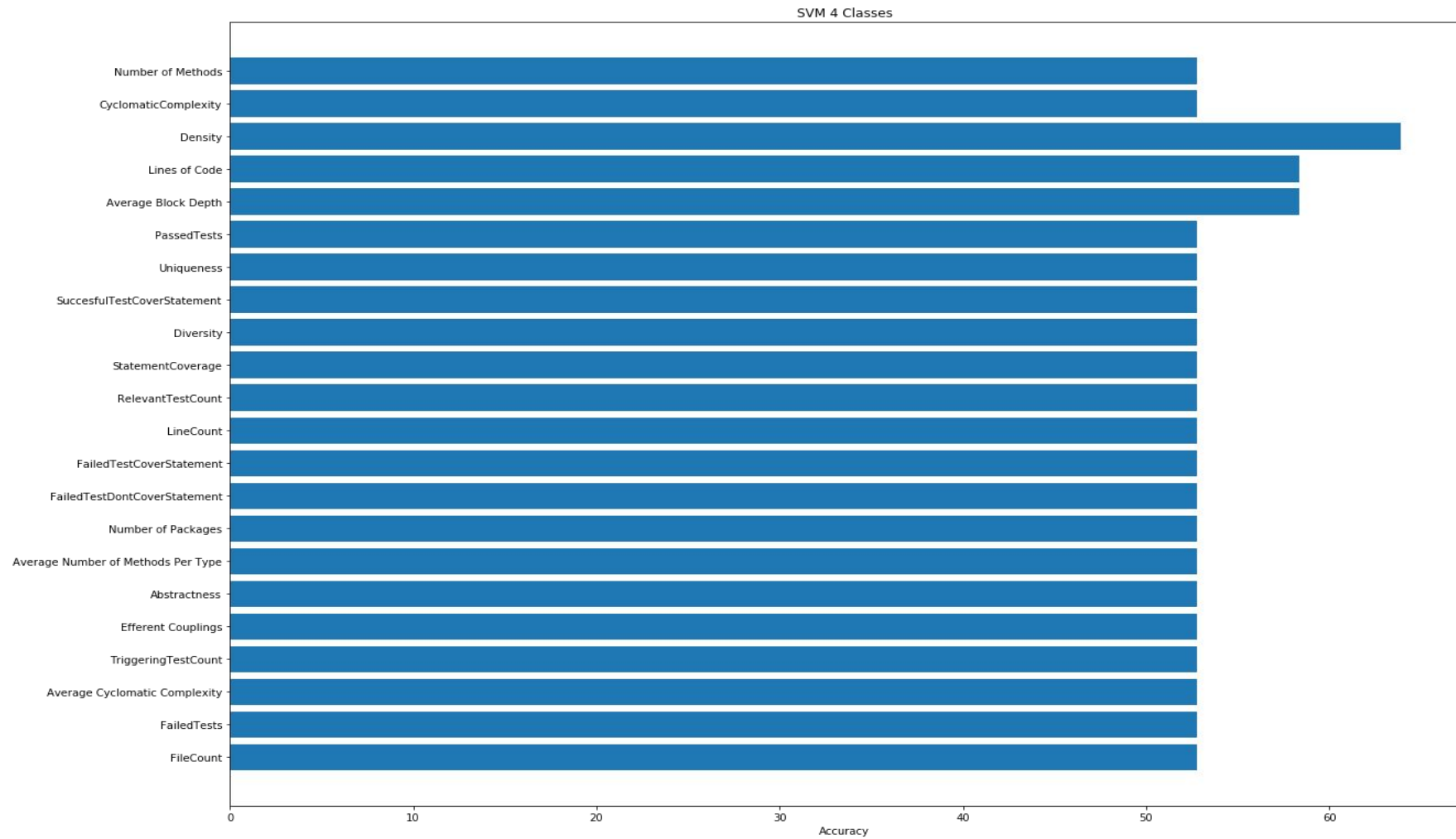


SVM (3 classes)

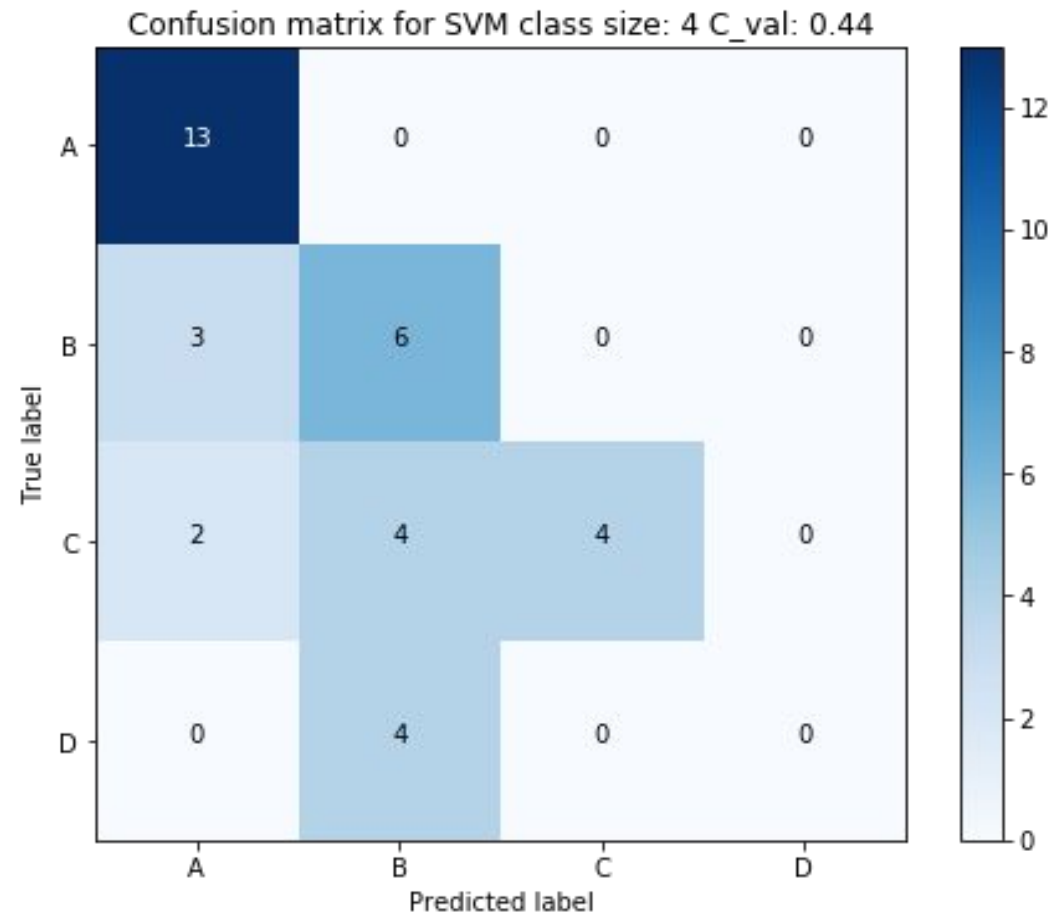


Acc. = 58.3%

SVM (4 classes)



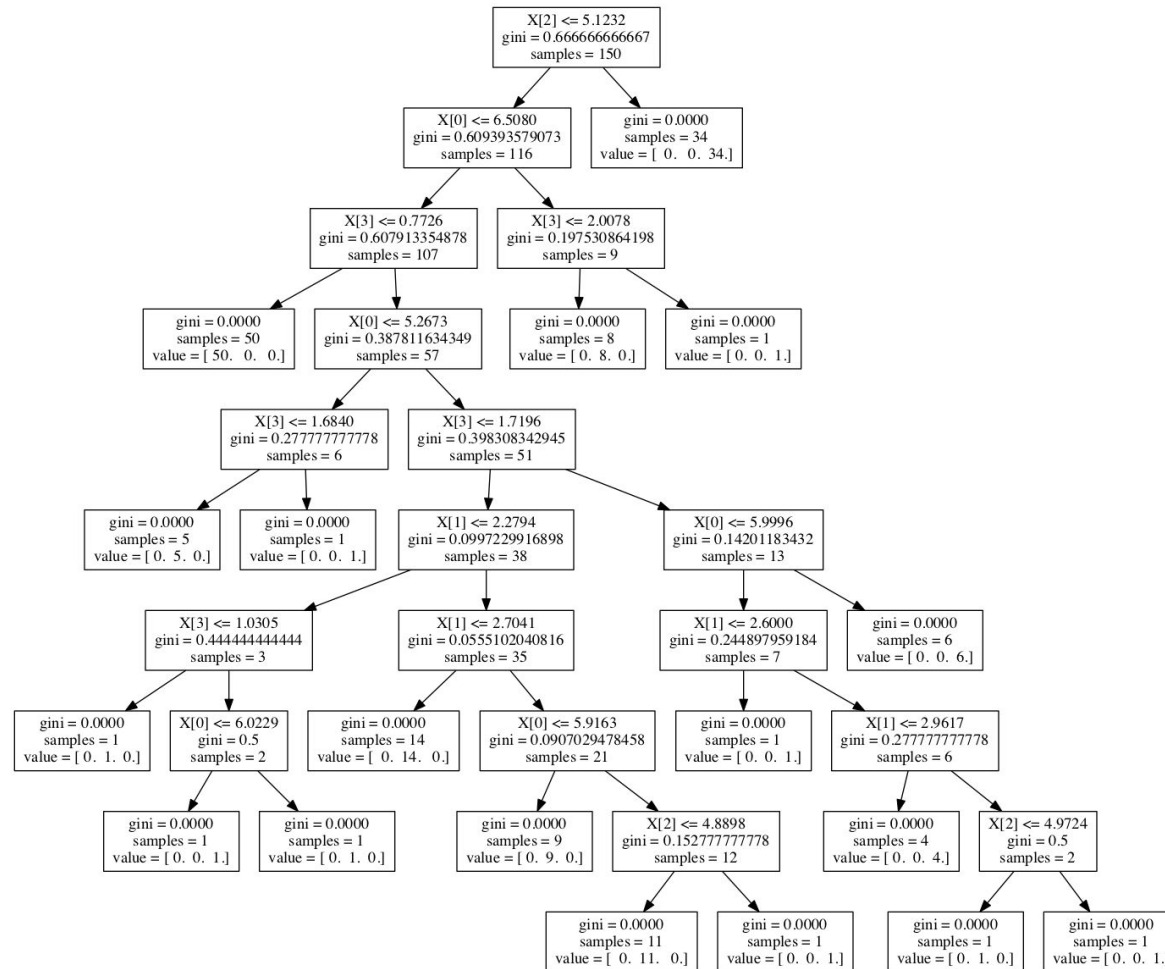
SVM(4 classes)



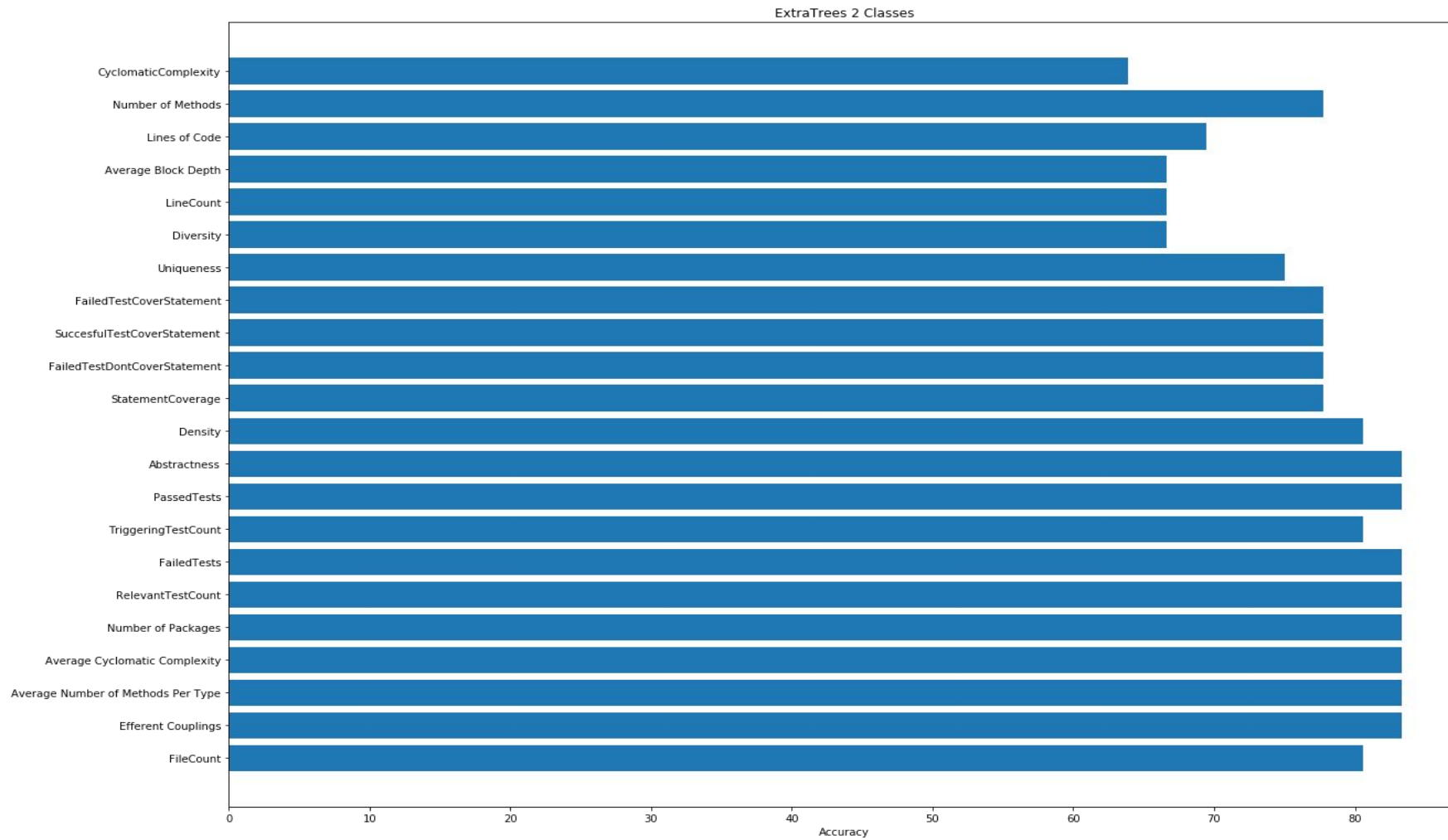
Acc. = 63.88%

ExtraTrees

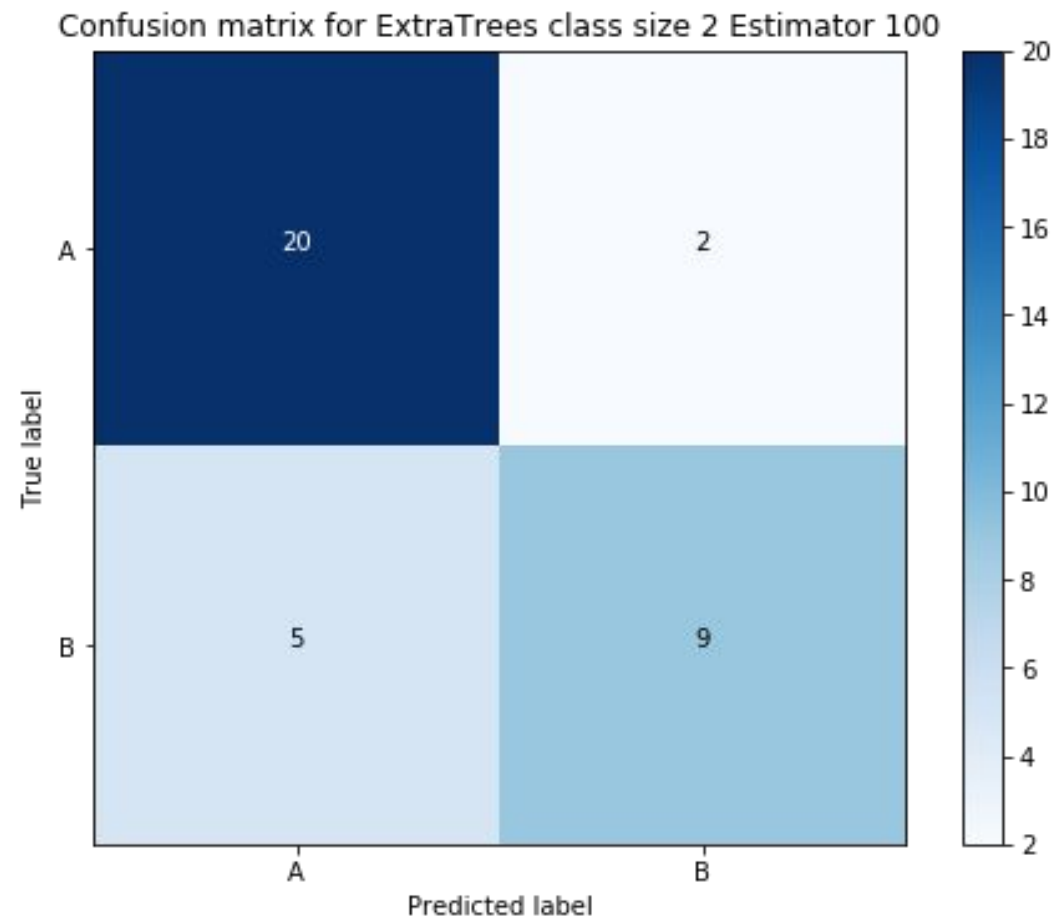
- ExtRa



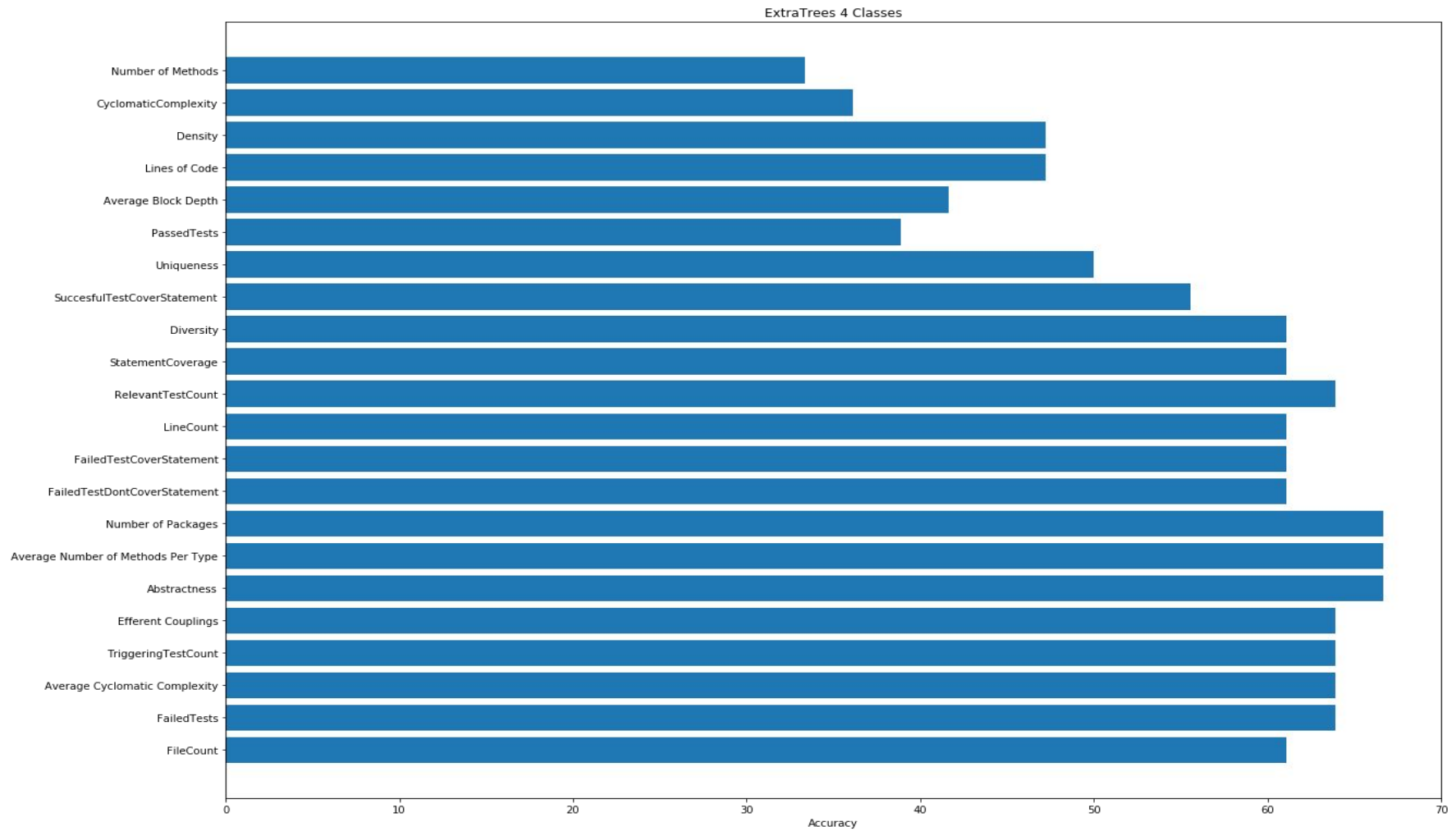
ExtraTrees



ExtraTrees

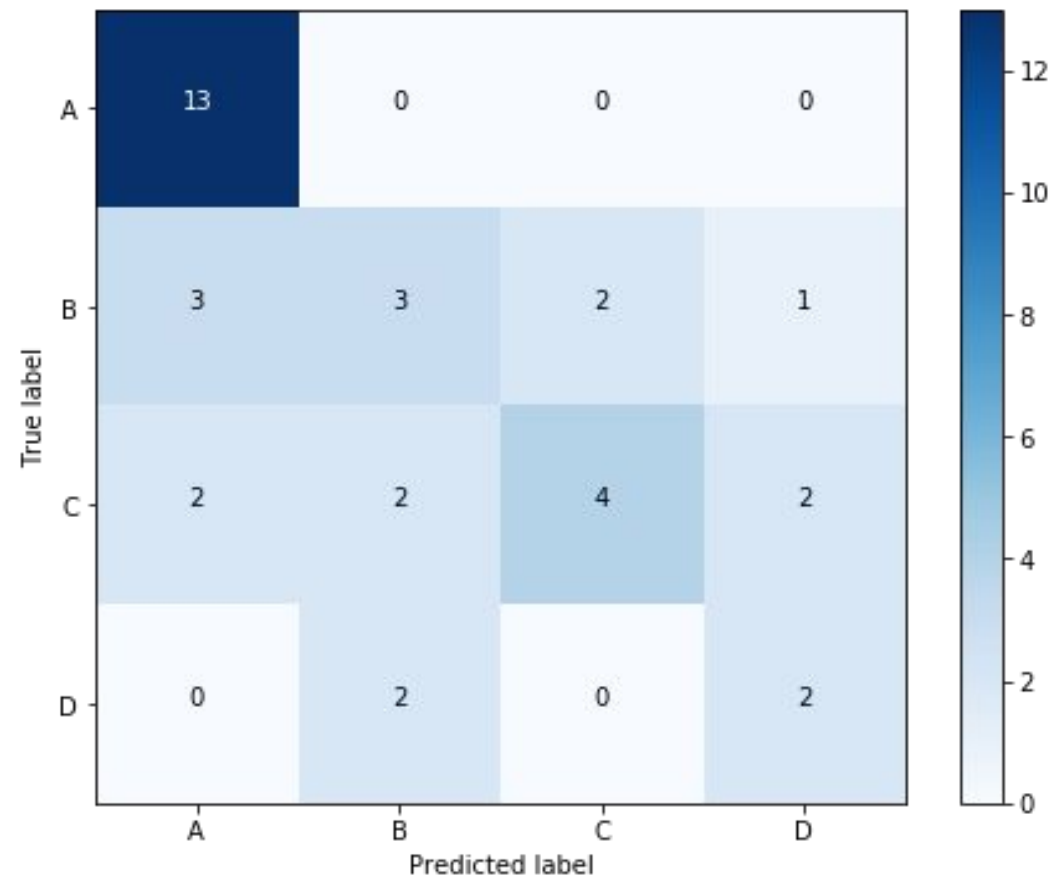


ExtraTrees



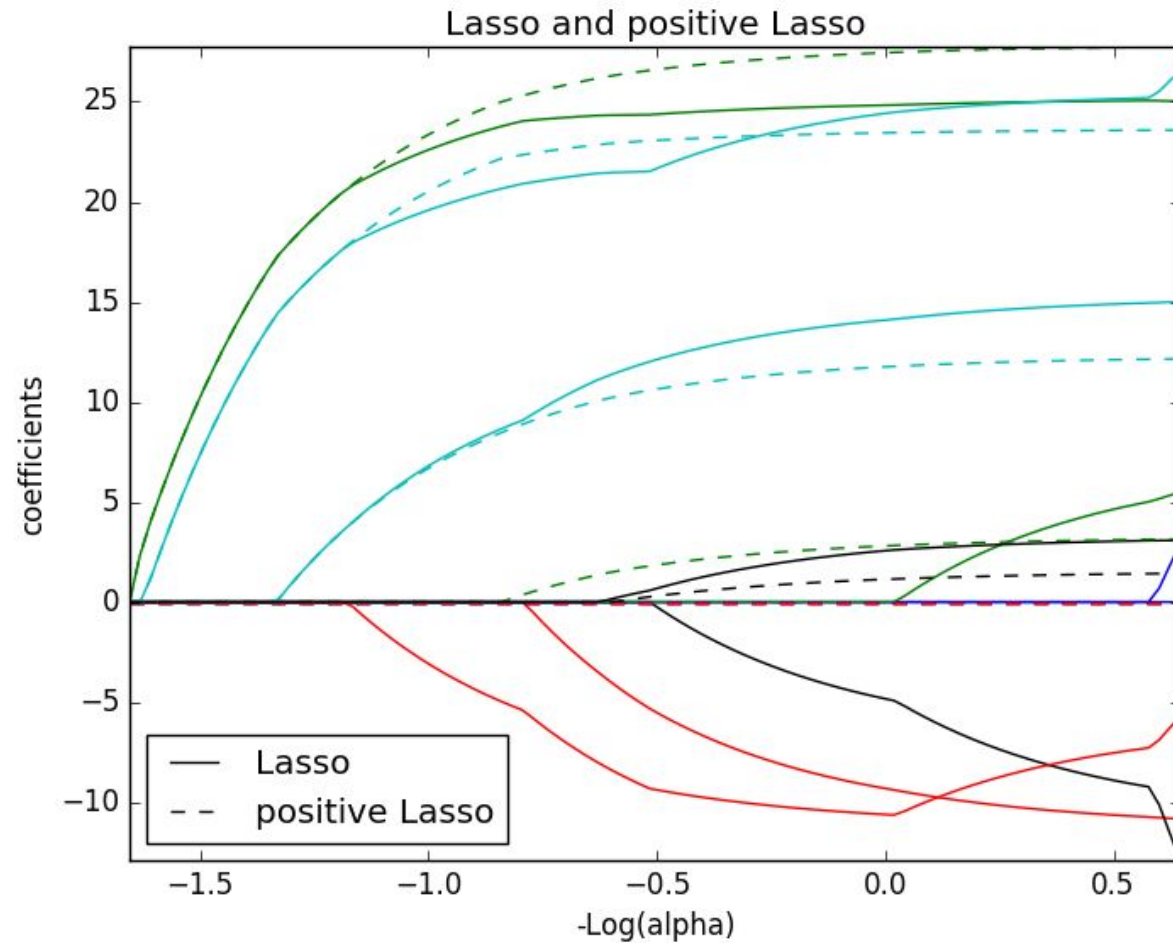
ExtraTrees

Confusion matrix for ExtraTrees class size 4 Estimator 1700

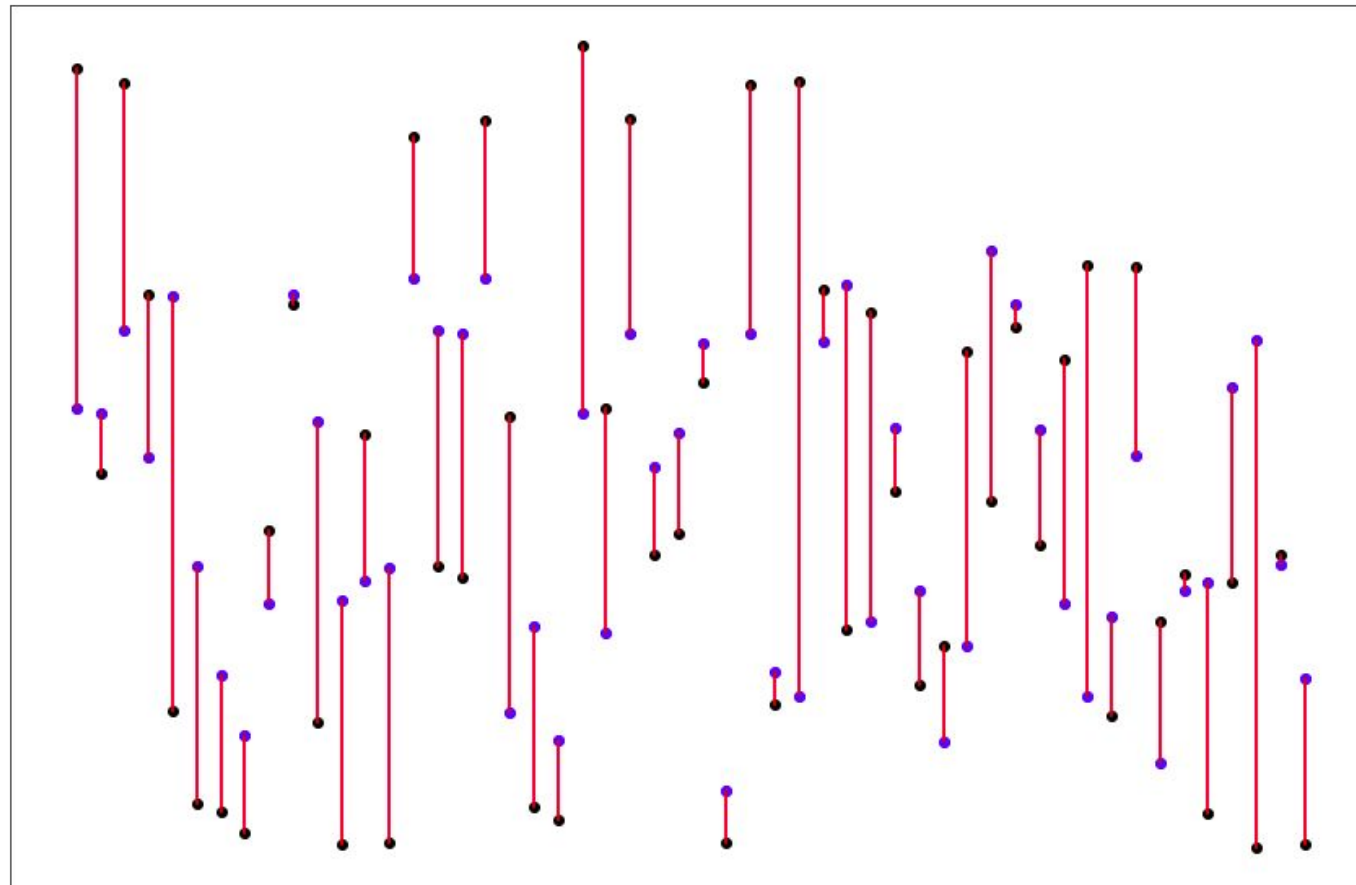


Bonus: ElasticNet

- Regression



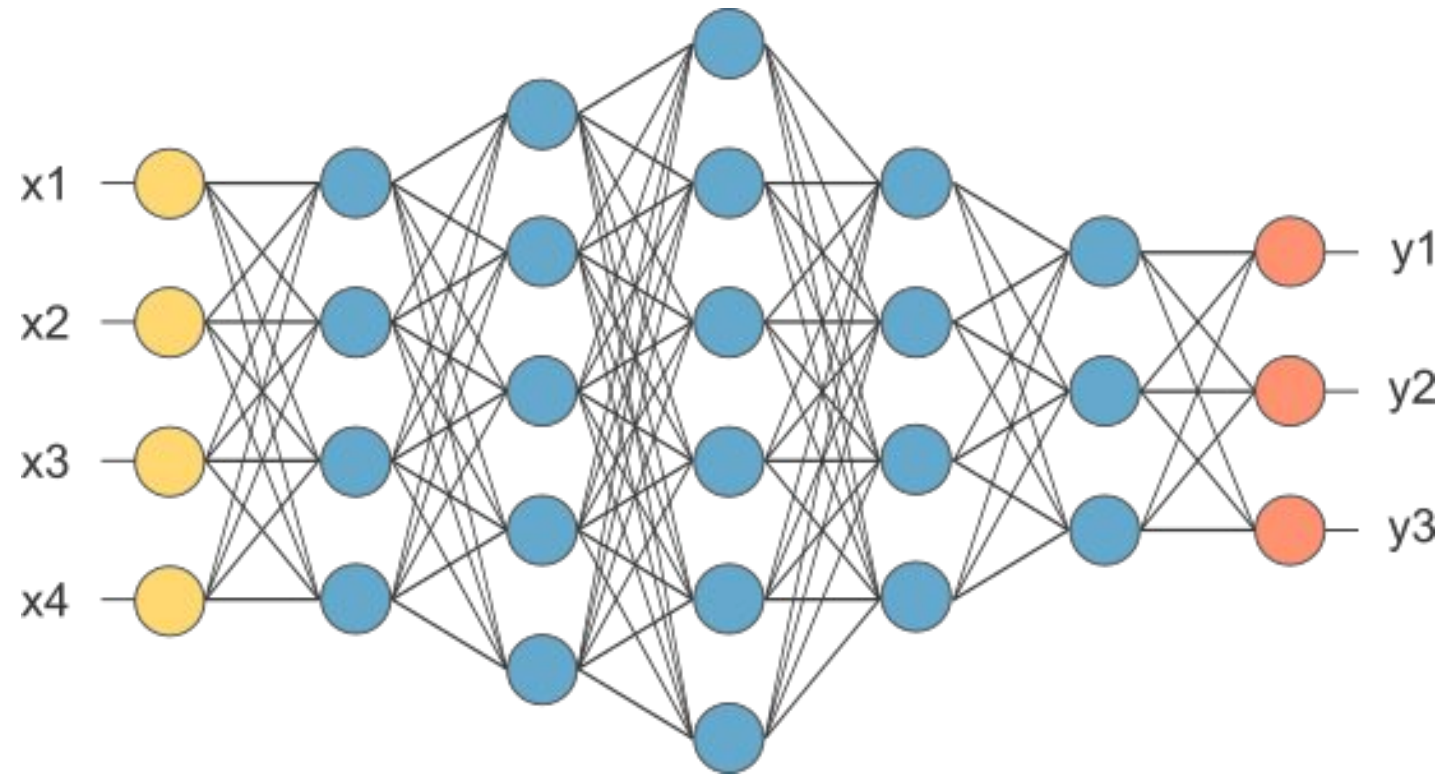
Bonus: ElasticNet



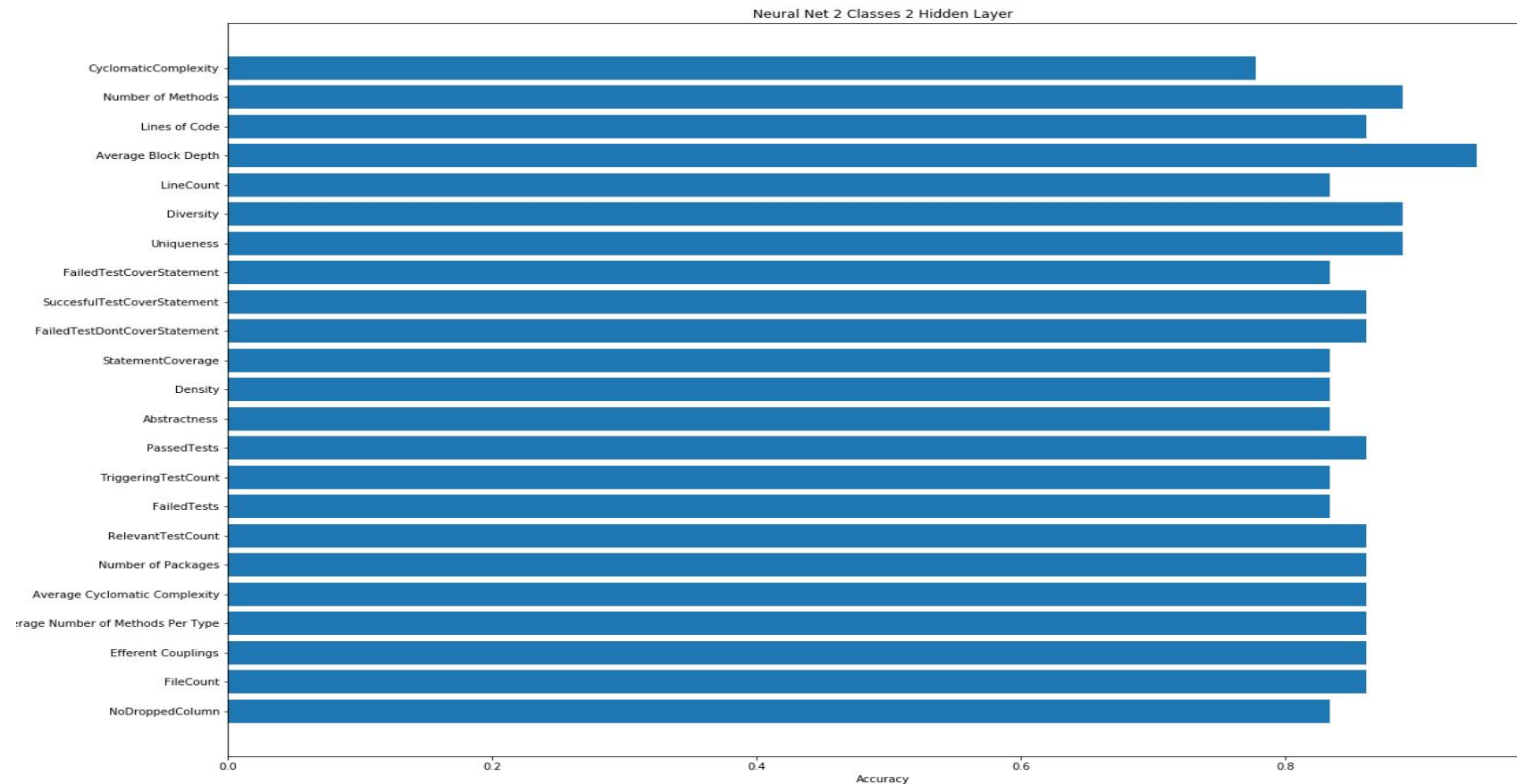
MSE = 0.128

Neural Network - MLP

- Layers
- Activation

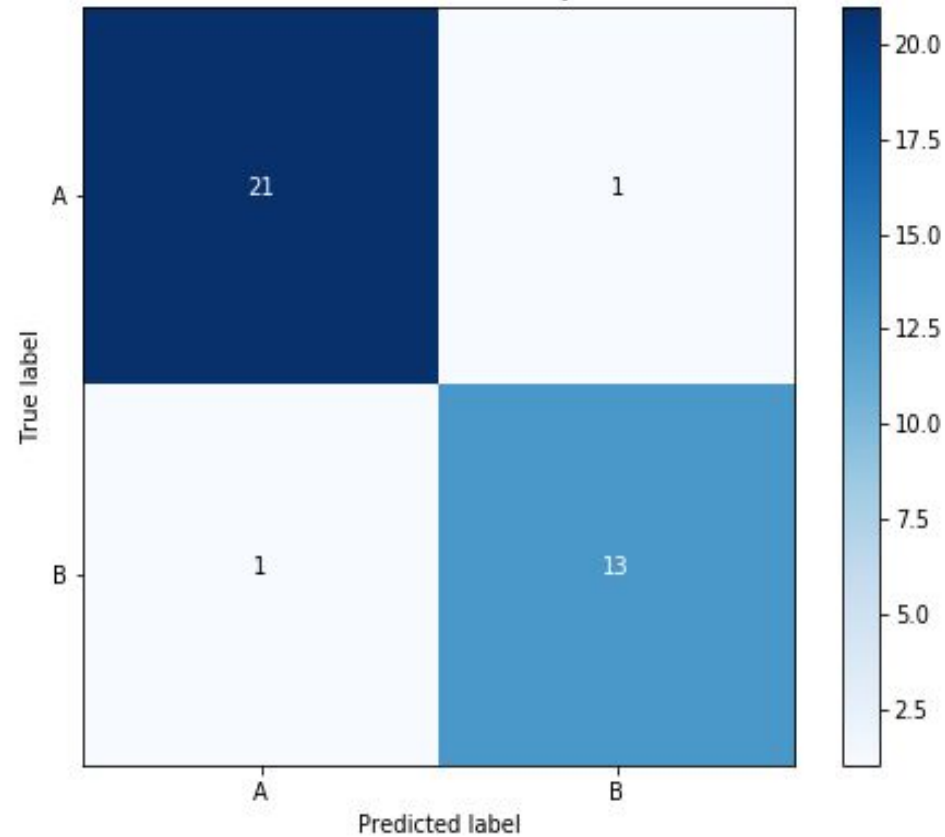


Neural Network - MLP (2 hidden layer)



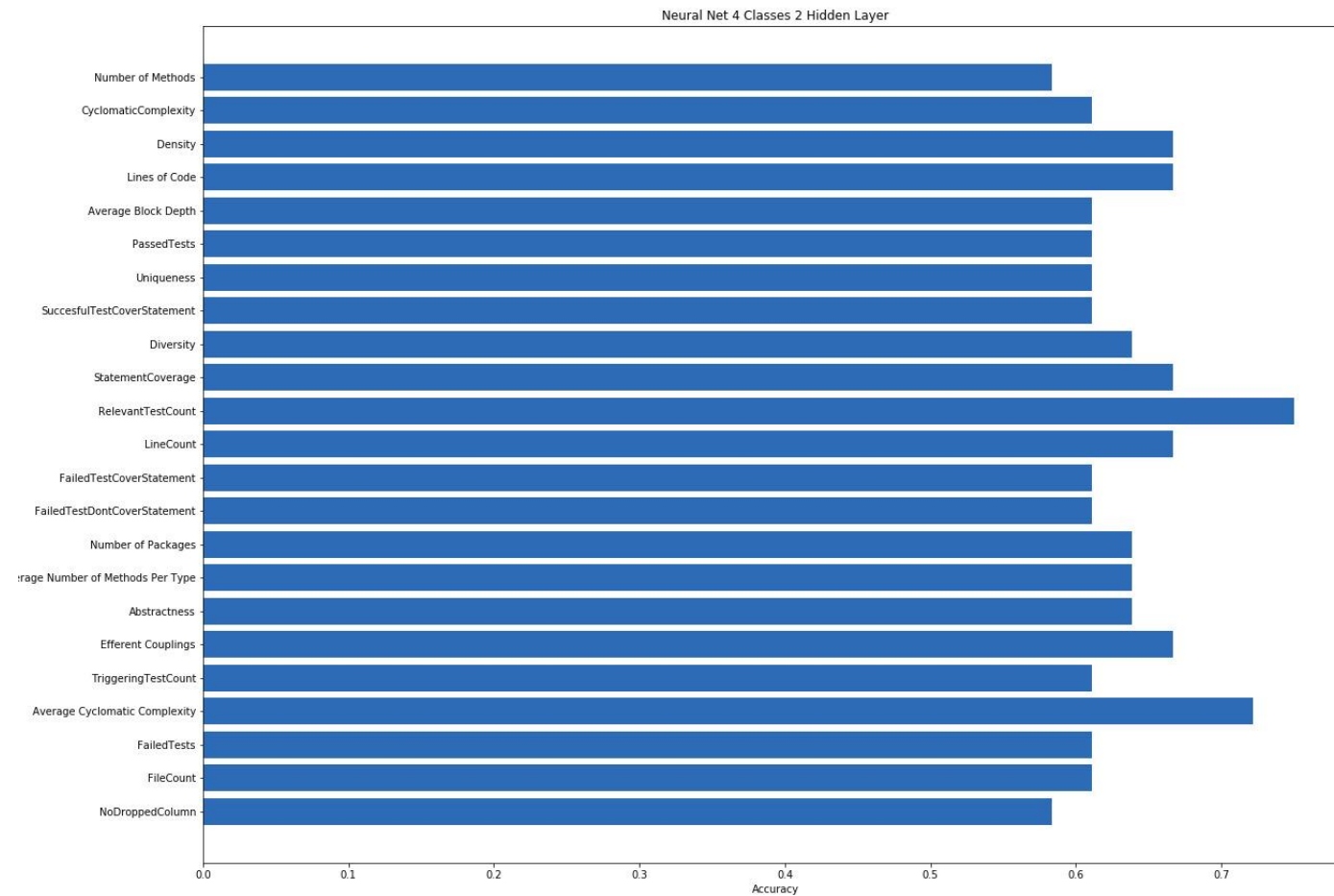
Neural Network – MLP (2 hidden layer)

Confusion matrix for class size 2 HiddenLayer = (525, 5) LR: 1e-10

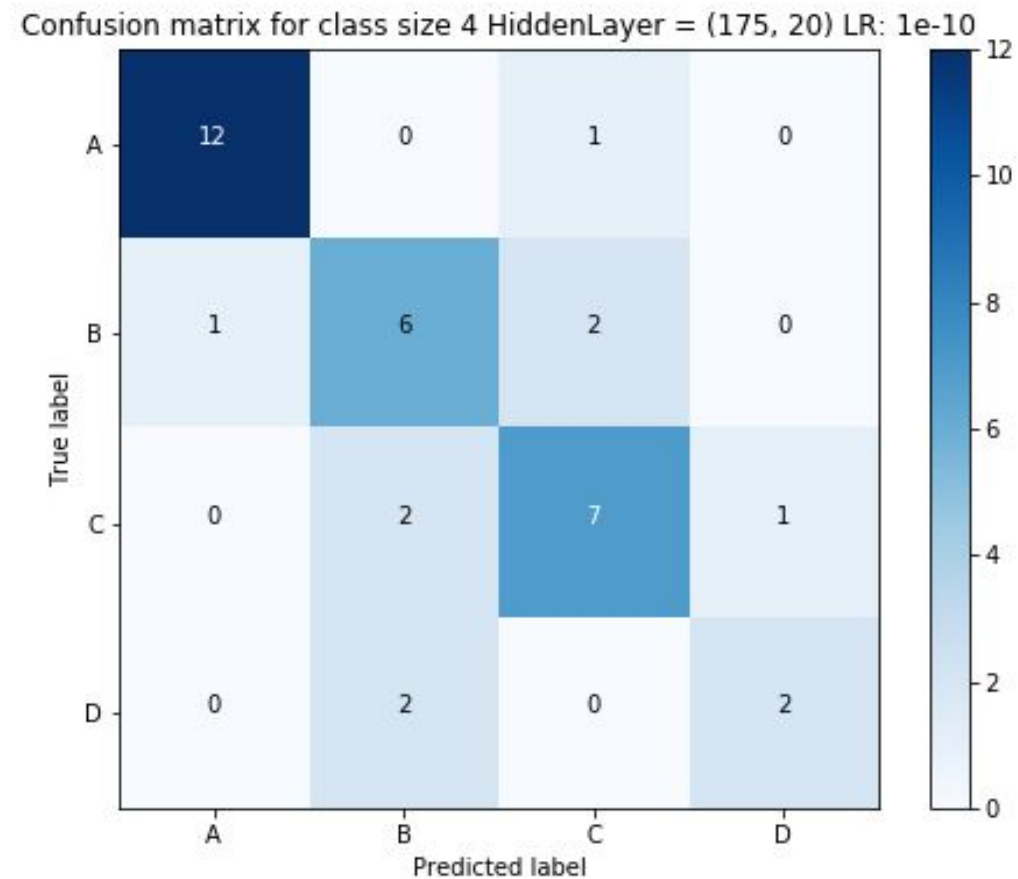


Acc. = 94.4%

Neural Network - (4 classes)

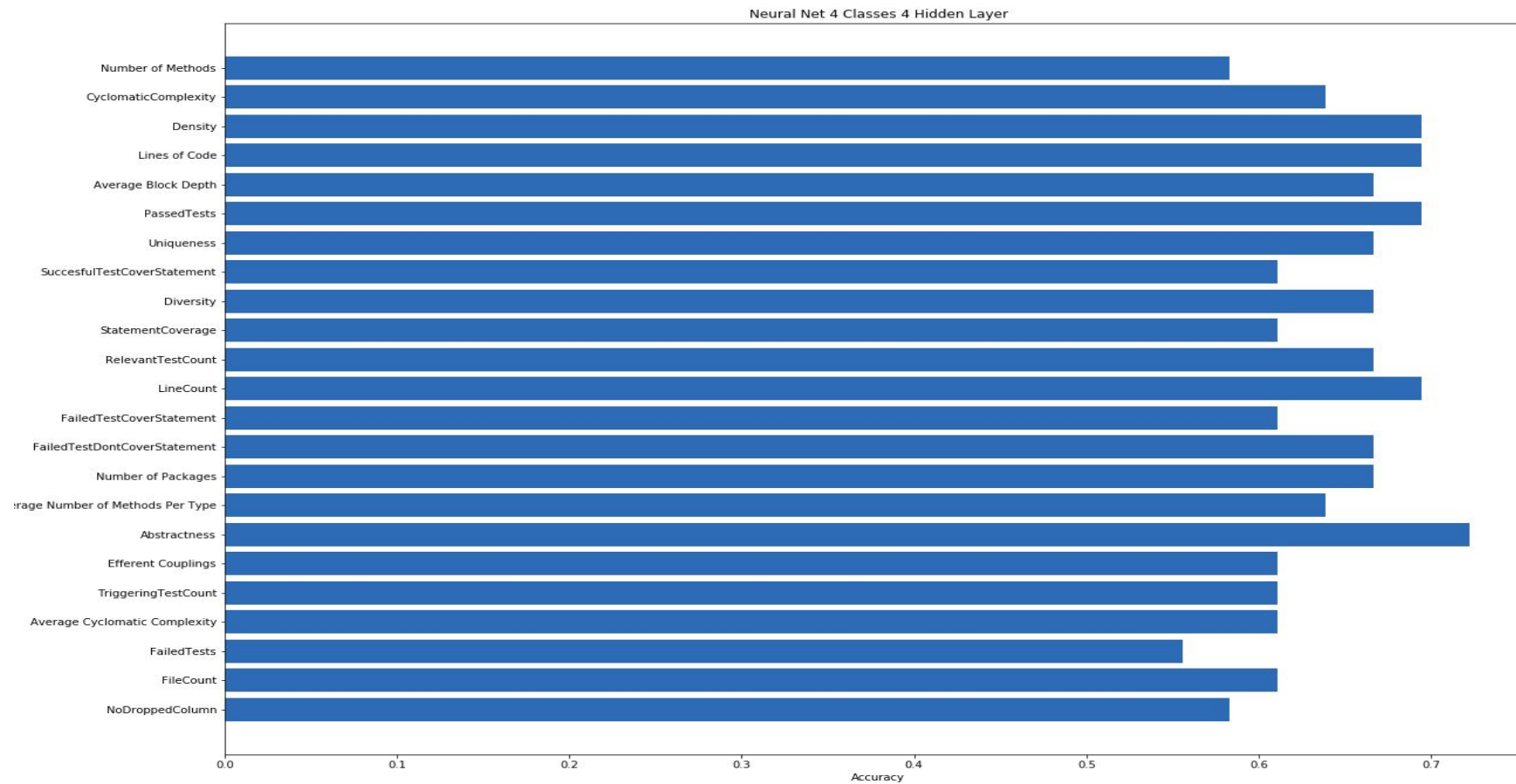


Neural Network - (4 classes)



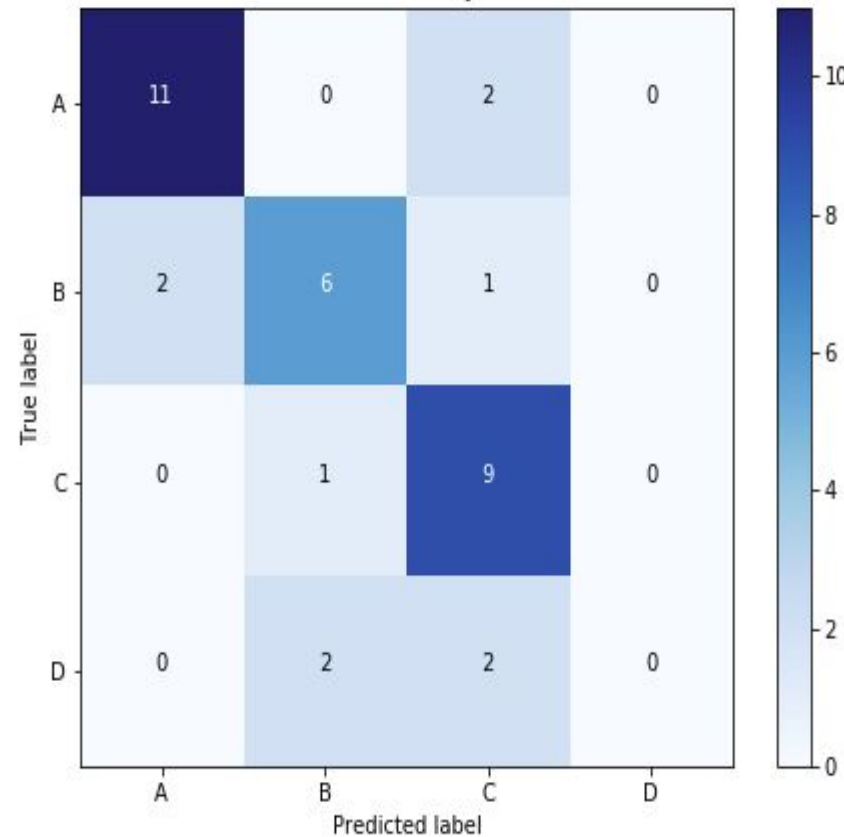
Acc. = 75%

Neural Network - MLP(4 C 4 L)



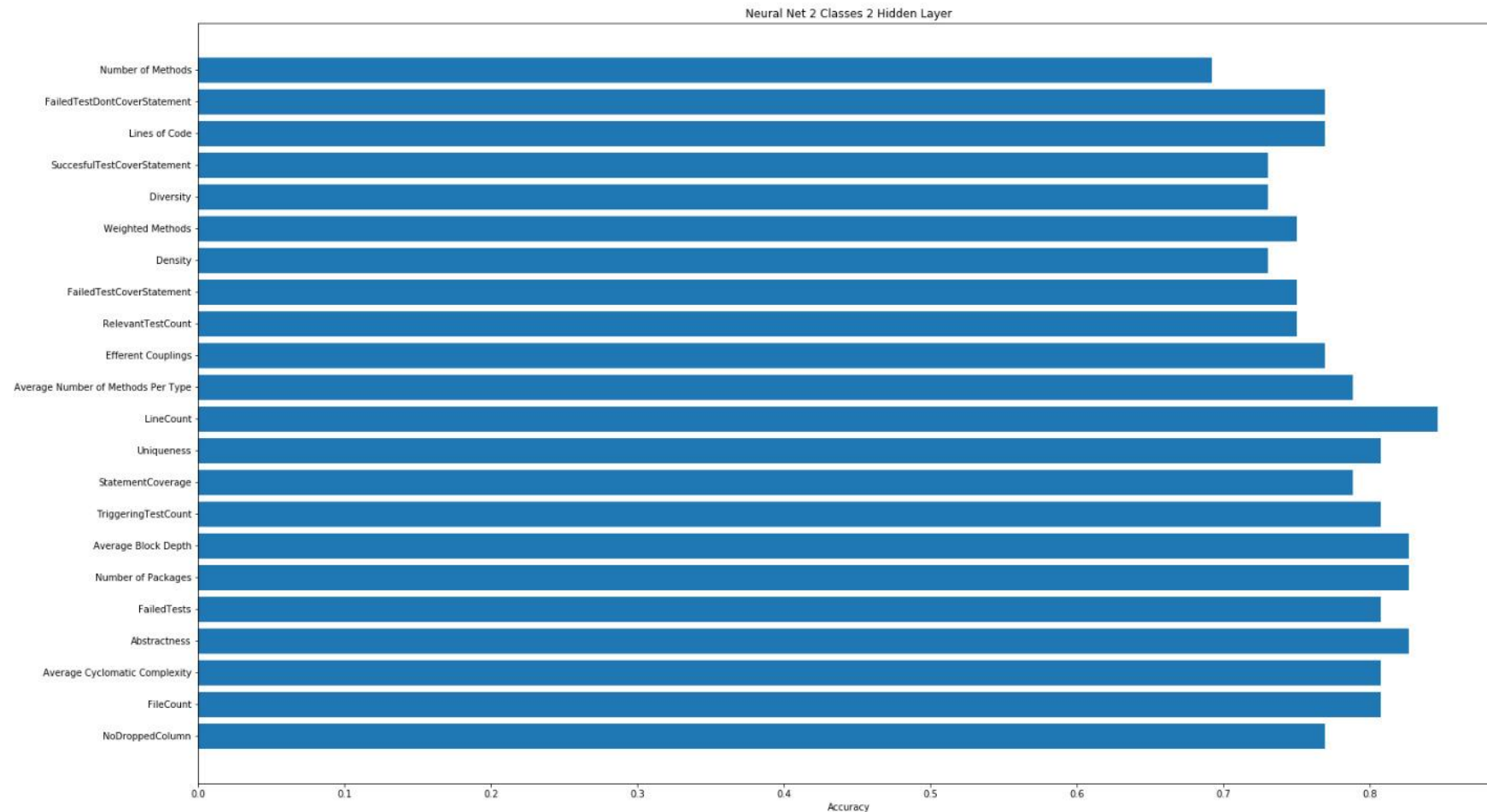
Neural Network - MLP (4 classes 4 Layer)

Confusion matrix for class size 4 HiddenLayer = (425, 55, 5, 350) LR: 1e-08



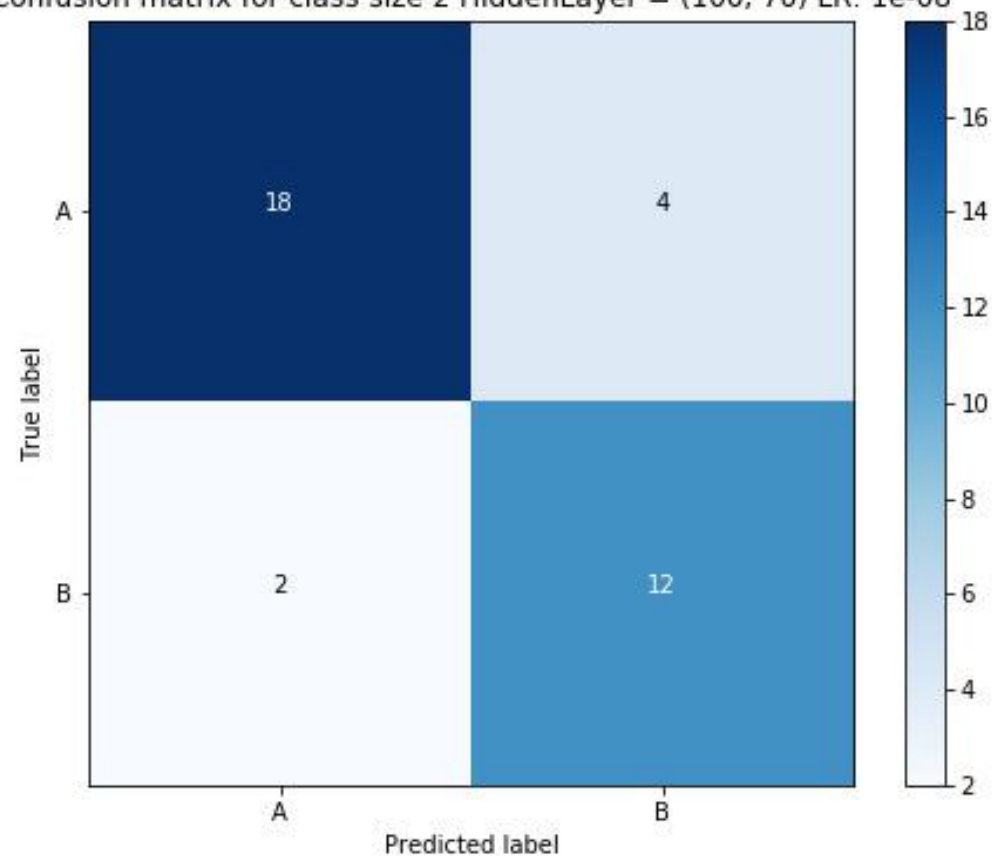
Acc. = 72%

Neural Network – MLP with Math



Neural Network – MLP with Math

Confusion matrix for class size 2 HiddenLayer = (100, 70) LR: 1e-08



Acc. = 86.6%

Final Results

	2 Class	3 Class	4 Class	5 Class	6 Class
Adaboost	0.83	0.61	0.55	0.5	0.58
SVM	0.8	0.58	0.63	0.58	0.47
ExtraTrees	0.8	0.55	0.61	0.61	0.55
MLP	0.94	0.72	0.75	0.66	0.61

Figure 1: Final accuracy results with less data more features

	2 Class	3 Class	4 Class	5 Class	6 Class
2 Layers	0.94	0.75	0.75	0.61	0.61
3 Layers	0.88	0.72	0.69	0.66	0.61
4 Layers	0.88	0.72	0.72	0.52	0.61
5 Layers	0.88	-	-	-	-

Figure 2: MLP's accuracy with respect to layers

	2 Class	3 Class
ExtraTrees	0.75	0.57
MLP	0.84	0.71

Figure 3: Final accuracy results with less features more data

Best Features so Far

With More Features

- Weighted Methods
- Cyclomatic Complexity
- Number of Methods
- Line Count
- Density

With More Data

- Number of Methods
- Weighted Methods
- Passed Tests
- Density
- FailedTestsDontCoverStatemen
t

What have we learned?

- Parsing data
- Technologies
- Target representation
- Correlation helps
- Dynamic metrics - JDCall Graph Problems
- Visualization!
- Neural Network
- More data is better than more features?
- Cross validation
- Suggestions



Thank you! Questions?

