

1. The faces in the wild data - face recognition
2. The KNN model
3. Random Forest
4. The Support Vector Machine (SVM)

Machine Learning Exercises

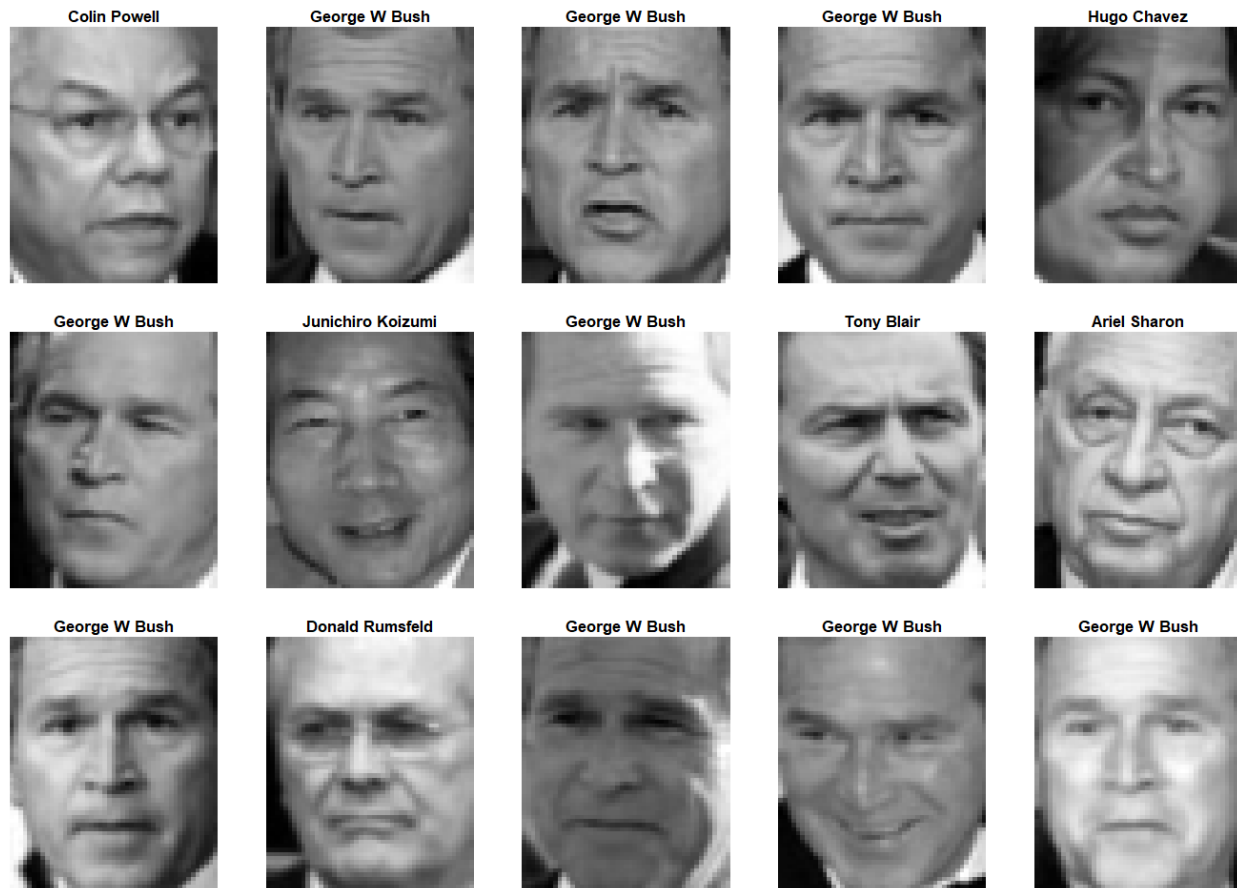
[Code ▾](#)

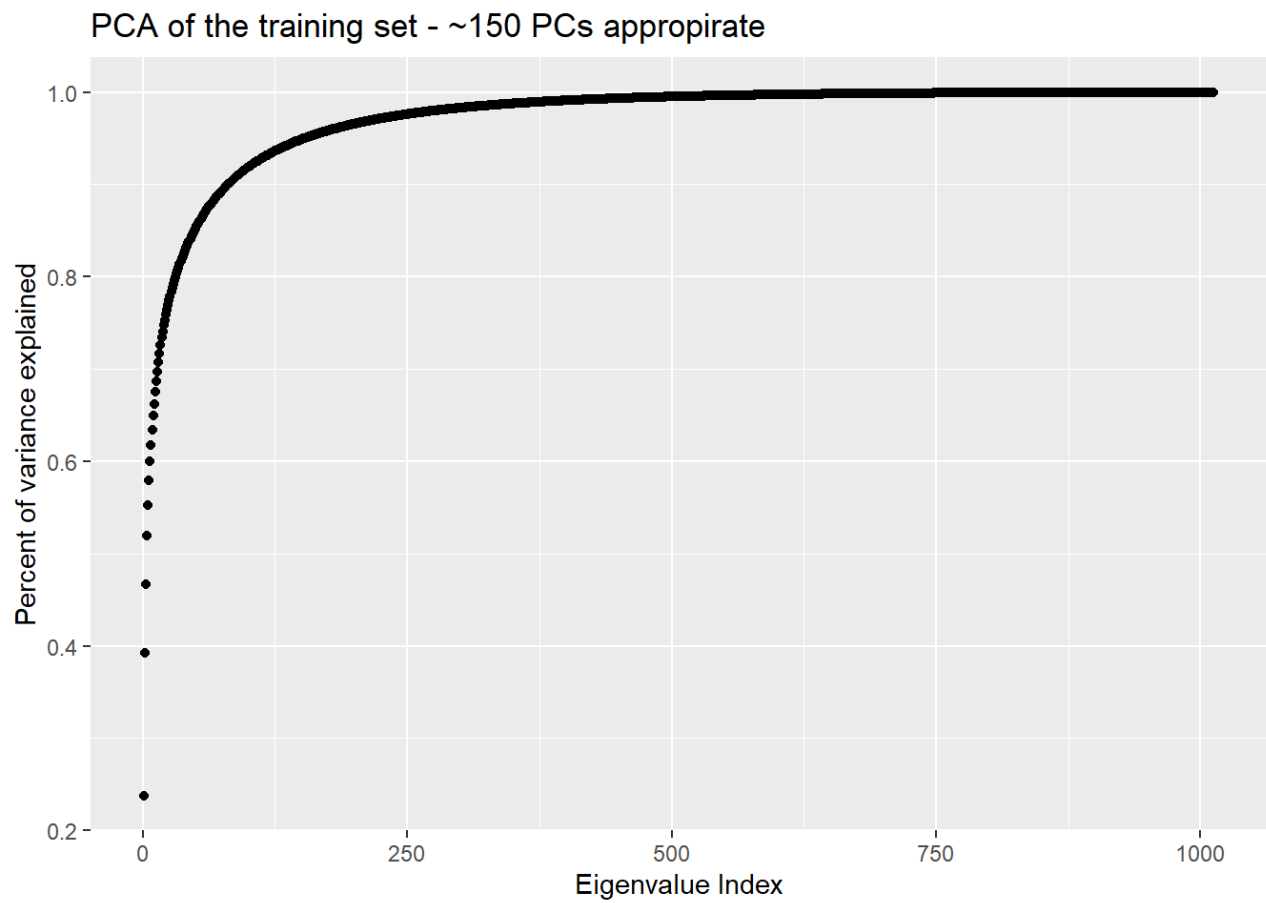
10/07/2021

[Code](#)

1. The faces in the wild data - face recognition

Data from scikit-learn package. Persons with more than 60 face figures were extracted. The following is the first few records in the data.

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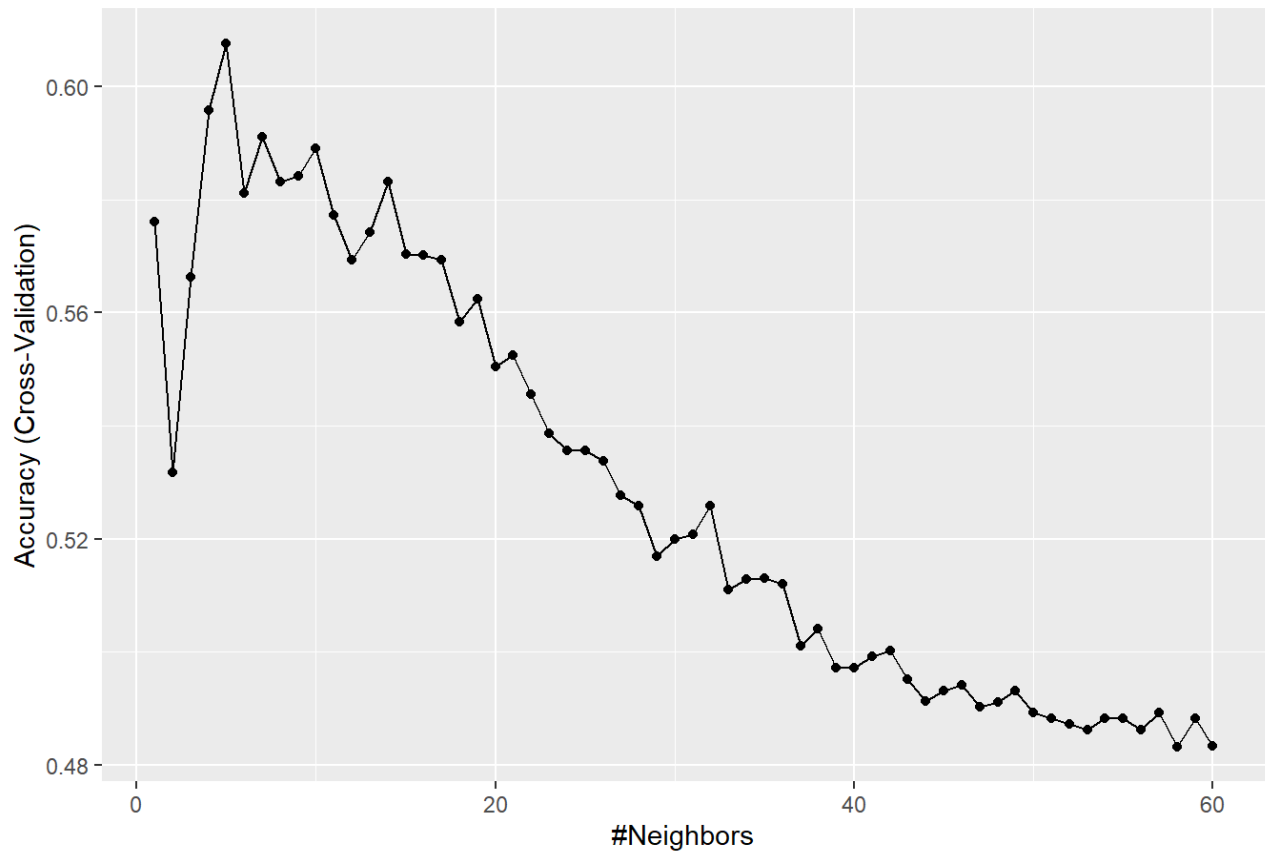
Each image contains 2914 pixels, and so use PCA to reduce dimension. A PCA analysis is as shown above. And, ~150 PCs should suffice.

2. The KNN model

2.1 The regular KNN

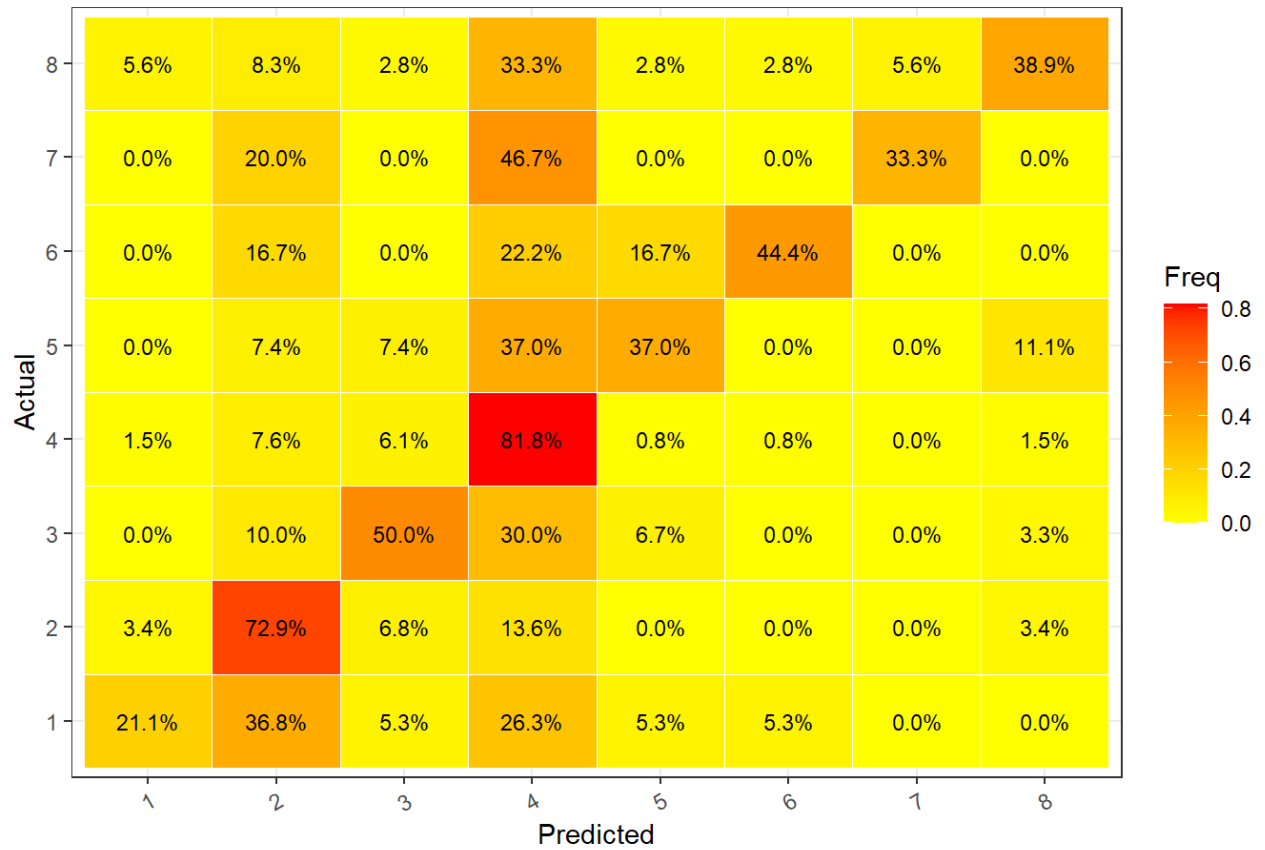
[Code](#)

Validatoin curve of KNN



Code

KNN confusion matrix on test set

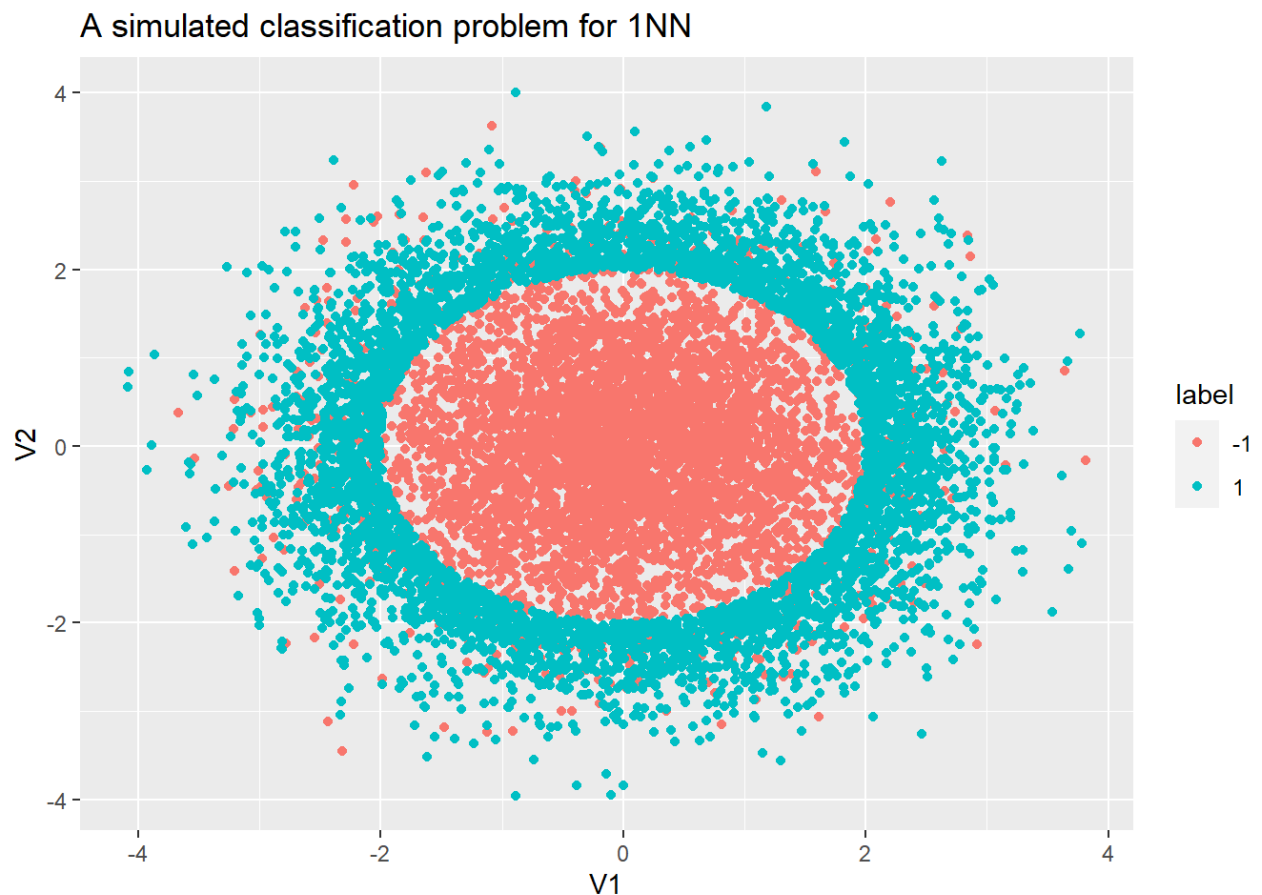


The accuracy of KNN is: 61.61%.

2.2 1NN to gain information about Bayes error

2.2.1 1NN to gain information about Bayes error - Simulated data

Asymptotically, 1NN error is never more than twice the Bayes error.

[Code](#)[Code](#)[Code](#)

The 1NN error on the test data is: 14.32%. So, the Bayes error must be more than 7.16%. The theoretical Bayes error (calculated from this trial) is 7.30%. Nonetheless, in practice, the asymptotic part is always in doubt.

2.2.2 1NN to gain information about Bayes error - face data

[Code](#)

The 1NN error on the test data is: 36.61%. So, the Bayes error must be more than 18.30%.

2.3 The Nested Cross Validation (NCV)

CV error is too optimistic. NCV is used to assess model error and stability. The following are NCV for KNN and SVM models using simulated data as earlier but with different sizes.

[Code](#)

Nested CV for the KNN model

trial	accuracy_test_ncv
1	0.91
2	0.90
3	0.91
4	0.92
5	0.91
6	0.91
7	0.92
8	0.93
9	0.91
10	0.93

[Code](#)

Nested CV for the SVM model

trial	accuracy_test_ncv
1	0.90
2	0.91
3	0.91
4	0.91
5	0.90

2.4 Naive Bayes

[Code](#)

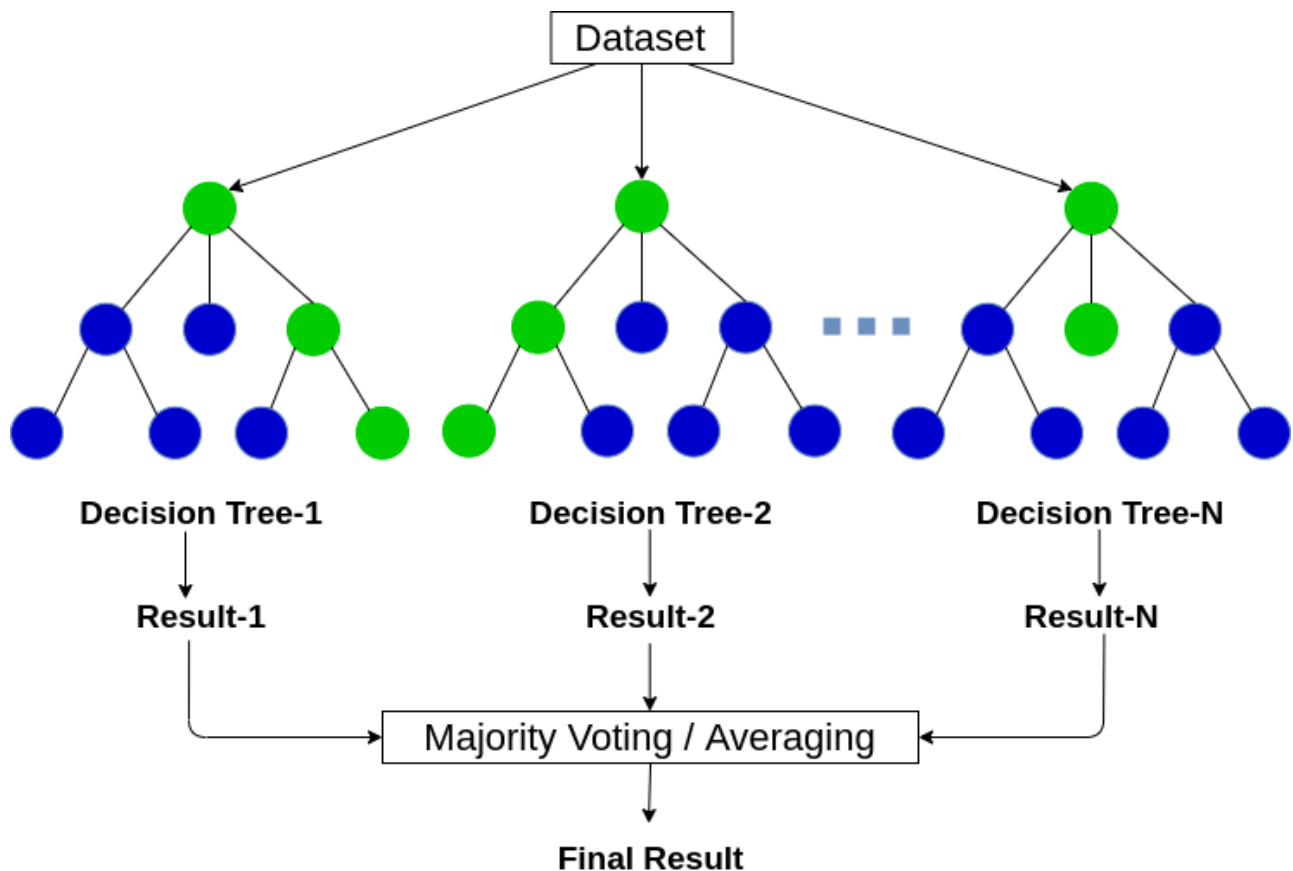
The accuracy of KNN is: 65.18%. It performs better than KNN as well as random forest.

3. Random Forest

The Random Forest:

- Unlike a single decision tree, which is notorious for its variability, random forest builds a multitude of trees.

- The forest makes a decision based on a collective majority vote of all the individual trees in a classification problem. The joint effort helps to reduce variance brought about by a single decision tree.
- Just a portion of the predictors are randomly selected to split a parent node, and this will help dampen the effect of the predominant predictors, decreasing variability in prediction.
- Lower tendency towards over-fitting.
- Small number of hyper-parameters to tune compared with Boosting/XGBoost.
- Random forests use out of bag sample and error (OOB error) to measure out of sample performance, and the favorable aspect is that the OOB errors can be produced during the training process.



Random Forest: Click to go to chart source

(<https://www.analyticsvidhya.com/blog/2020/05/decision-tree-vs-random-forest-algorithm/>)

3.1 Hyper-parameter tuning and model fitting

At the lowest OOB error, the best parameters are:

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Best Parameters for the Random Forest Model

mtry	OOBError	No_trees	Node_size
63	0.34	400	3

[Code](#)

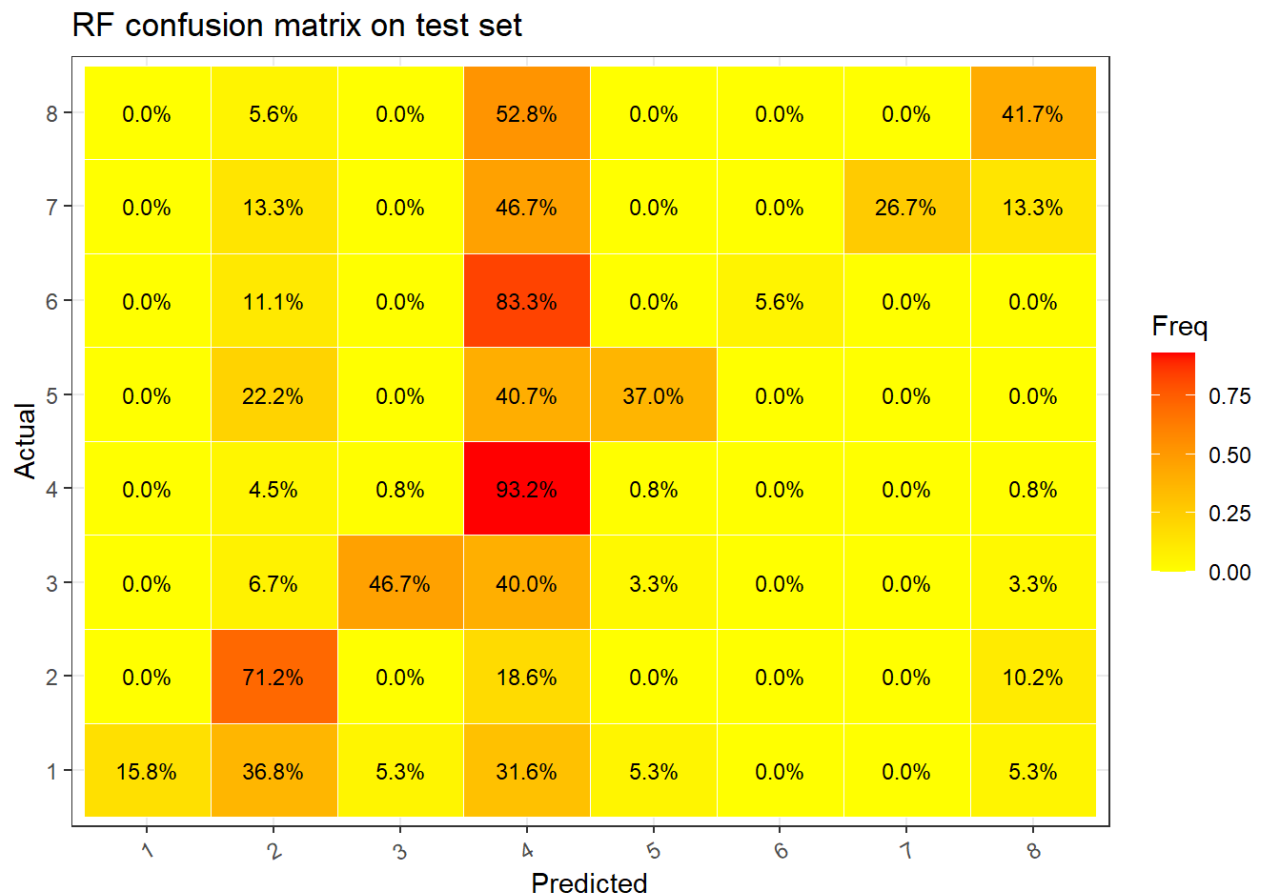
3.2 The Random Forest Model Performance

3.2.1 The Confusion Matrix

[Code](#)

The confusion matrix can show the model prediction vs the actual:

3.2.2 The notch difference plot

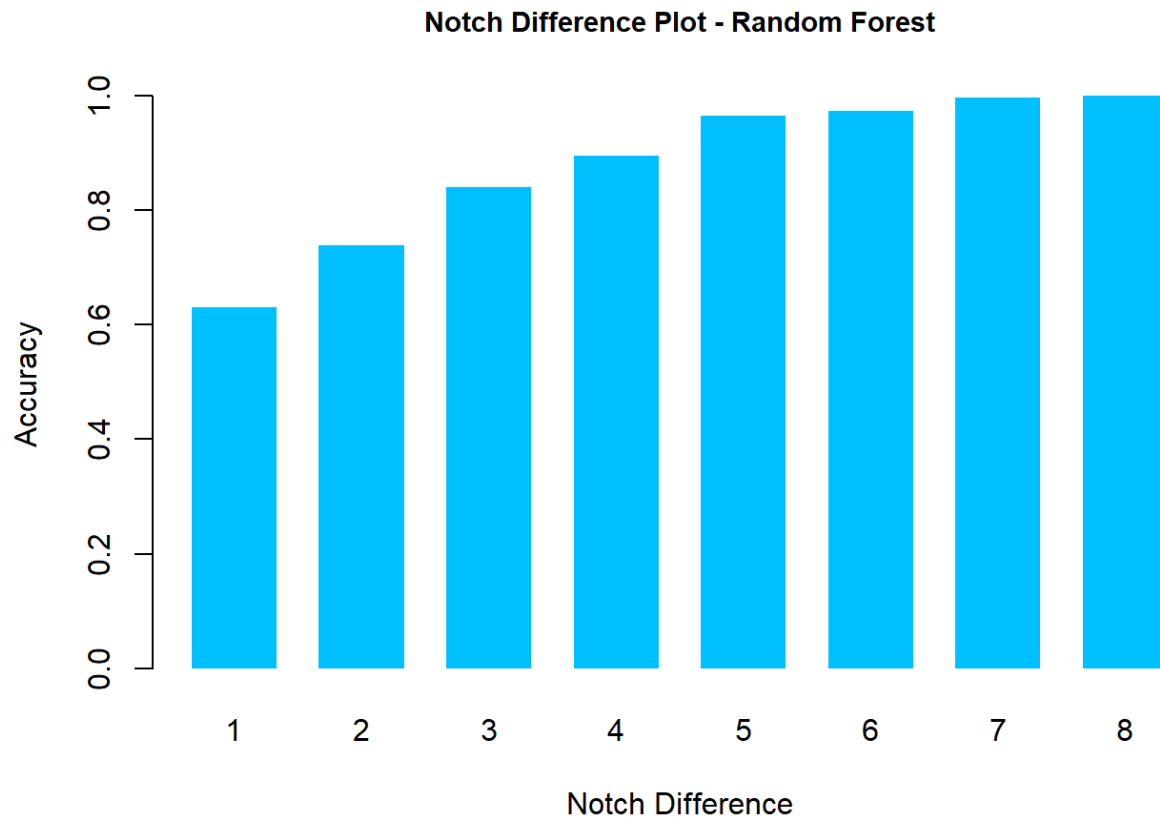
[Code](#)

There are totally 336 obligors in the test data set. Model prediction accuracy in test data is: 63.10%

[Code](#)

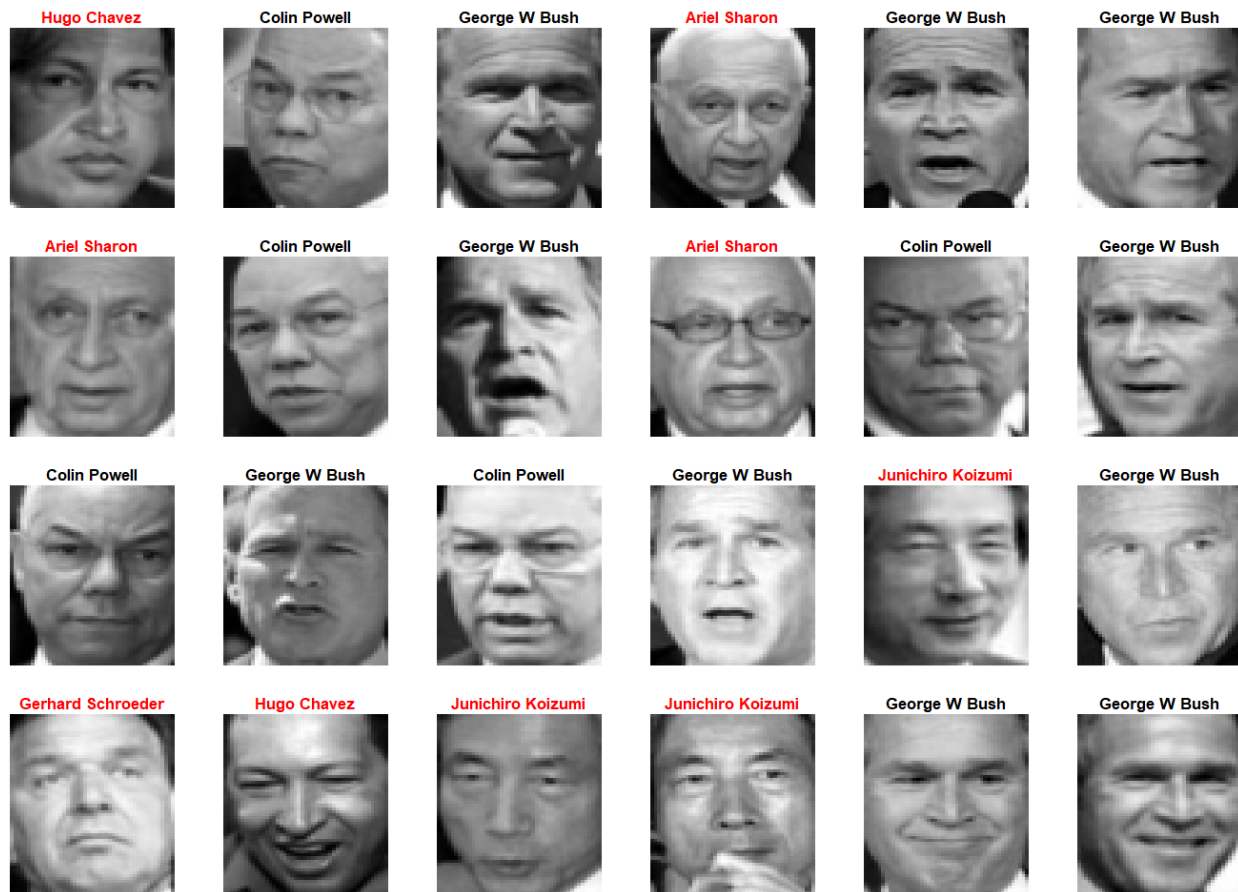
Notch difference table - Random Forest

1	2	3	4	5	6	7	8
0.631	0.7381	0.8393	0.8958	0.9643	0.9732	0.997	1

[Code](#)

3.2.3 The predicted vs actual images

[Code](#)



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4. The Support Vector Machine (SVM)

SVM improves substantially from the previous two learners.

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The accuracy of SVM is: 82.74%. The following shows the prediction among the first 24 images. Images labeled in red are miss-classified.

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SVM confusion matrix on test set

