

I did some outlier removal along with missing value replacement. I also standardized and normalized the data using the choosing buttons. Since this dataset has many features, I applied principal component analysis to remove all of the unnecessary dimensions and get a better dataset. Since PCA in Weka only uses numerical values, I first had to clean the dataset accordingly.

Step 2:

1. KNN

```
-0.8238urvival Months-0.45ERegional Node Examined+0.245Tumor Size+0.215Age-0.084Reginol Node Positive
-0.655Reginol Node Positive-0.557Regional Node Examined+0.425Tumor Size+0.266Survival Months+0.063Age
0.738Tumor Size+0.268Survival Months+0.374Age-0.291Regional Node Examined-OReginol Node Positive
Status
Test mode: 10-fold cross-validation
=== Classifier model (full training set) ===

TBl instance-based classifier
using 1 nearest neighbour(s) for classification

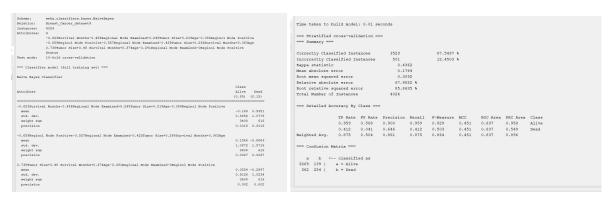
Time taken to build model: 0 seconds
=== Stratified cross-validation ===
=== Summary ===

Correctly Classified Instances 659 16.6252 %
Rappa statistic 0.3558
Man absolute error 0.1664
Root mean squared error 0.4076
Relative absolute error 64.1552 %
Root relative squared error 113.2092 %
Total Number of Instances 4024
=== Detailed Accuracy By Class ===

TP Rate FP Rate Precision Recall F-Measure MCC ROC Area PRC Area Class 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503 0.503
```

KNNs main hyperparameters are cross-validation folds along with test split. It has a classification accuracy of around 83 percent. Its pros are it does not make assumptions about the data making it versatile. But it is not good at large datasets as it can be computationally intensive.

# 2. Naive Bayes



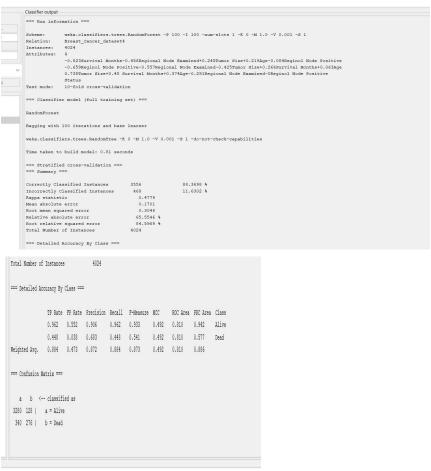
Naive Bayes is fast and efficient and is a very ideal algorithm for categorization-based data. However, it assumes that features are independent of each other which is not usually the case. Since this is categorization-based, the accuracy of classification is also higher and error is also lower than KNN. Batchsize is its hyperparameter.

#### 3. Decision Tree

```
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```

Hyperparameters are cross validation folds, batch size, and split percentage. Decision trees avoid overfitting by using pruning, also good for continuous and categorical data. This algorithm is not good for imbalance data. We get almost a 90 percent accuracy here.

### 4. Random Forest



A random forest handles both numeric and categorical data in a well-suited way and it is good at handling outliers. However, tuning the wrong number of trees can cause slow down. Its

hyperparameters are the number of trees, the maximum depth, and the number of cross-validation folds. As seen above, it correctly classifies all features.

# 5. Gradient Boosting

Gradient boosting is known for high accuracy in both classification and regression based problems. However, it can be sensitive to overfitting especially when tuning is not right. We can tune the maximum depth of the tree and learning rate. We get an 87 percent accuracy of classifiers here.

#### 6. Neural Networks

Neural networks are very good at handling high-dimensional data and for numerous machine learning tasks. However, with so many hyperparameters(learning rate, number of layers, neurons per layer, etc), tuning them the wrong way can lead to wrong results. Results here show almost a 90 percent accurate classification instance.

### **Step 3: Hyperparameter Tuning [15]**

```
Test output

Available resultsets
(1) trees.RandomForest '-F 100 -I 100 -num-slots 1 -K 0 -M 1.0 -V 0.001 -S 1' 1116839470751428740

Test output

Available resultsets
(1) trees.RandomForest '-F 100 -I 100 -num-slots 1 -K 0 -M 1.0 -V 0.001 -S 1' 1116839470751428740
(2) functions.MultilayerFerceptron '-L 0.3 -M 0.2 -N 500 -V 0 -S 0 -E 20 -H a' -5990607817048210400
```

I picked randomForest and the neural networks as both have more than 2 hyperparameters. The results are in the pictures, and the RandomForest seemed to work better for this.

Model	Accuracy (%)	Key Features for Classification	Comments
KNN	83	Distance metrics (cross- validation folds)	Versatile but computationally intensive on large datasets
Naive Bayes	Higher than KNN (no exact %)	Feature independence assumption, Batch size	Fast and efficient for categorization-based data
Decision Tree	~90	Split percentage, Pruning	Avoids overfitting, good for categorical/continuous data but struggles with imbalanced data
Random Forest	100	Number of trees, Maximum depth, Cross- validation folds	Handles outliers well, slow with incorrect number of trees
Gradient Boosting	87	Maximum depth, Learning rate	High accuracy, but prone to overfitting
Neural Networks	~90	Learning rate, Number of layers, Neurons per layer	Effective for high-dimensional data, sensitive to hyperparameter tuning