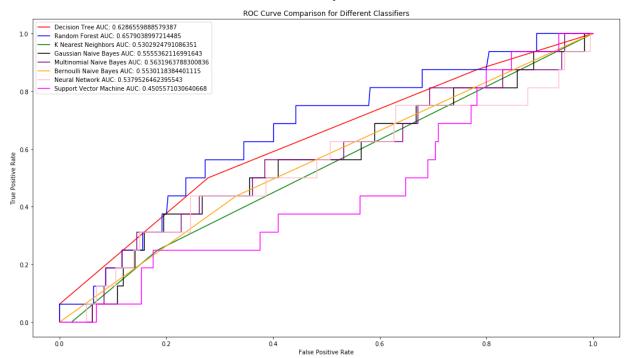
```
In [1]: import numpy as np
         from sklearn.svm import SVC
         from sklearn.model selection import train test split
         from sklearn.tree import DecisionTreeClassifier
         import matplotlib.pyplot as plt
         from sklearn.metrics import roc curve
         from sklearn.metrics import auc
         from sklearn.ensemble import RandomForestClassifier
         from sklearn.neighbors import KNeighborsClassifier
         from sklearn.naive bayes import GaussianNB, MultinomialNB, BernoulliNB
         from sklearn.neural network import MLPClassifier
         from sklearn.metrics import confusion matrix
In [2]: features = np.genfromtxt("./Aggregated_Data.csv", delimiter=",", usecols=(1
         targets = np.genfromtxt("./Aggregated Data.csv", delimiter=",", usecols=8)
         trainingFeatures, testingFeatures, trainingTargets, testingTargets = train_
In [10]: #follows the Confusion Matrix information page on Scikit Learn to visualize
         #https://scikit-learn.org/stable/auto examples/model selection/plot confusi
         def plotMatrix(confusionMatrix, title, color):
             classLabels = ["No Outbreak", "Outbreak"]
             plt.rcParams['figure.figsize'] = (5.0, 5.0)
             plt.imshow(confusionMatrix, cmap=plt.cm.Blues)
             plt.xticks(ticks=[0,1], labels=classLabels)
             plt.yticks(ticks=[0,1], labels=classLabels)
             plt.ylabel("Predicted Labels")
             plt.xlabel("True Labels")
             plt.title(title)
             maximum = confusionMatrix.max()/2
             plt.text(0, 0, str(confusionMatrix[0][0]), horizontalalignment="center"
             plt.text(0, 1, str(confusionMatrix[0][1]), horizontalalignment="center"
             plt.text(1, 0, str(confusionMatrix[1][0]), horizontalalignment="center"
             plt.text(1, 1, str(confusionMatrix[1][1]), horizontalalignment="center"
             plt.show()
In [9]: def thresholdBasedClassification(threshold, classifier, label, testingFeatu
             predictions = []
             probabilities = classifier.predict_proba(testingFeatures)
             for j in range(0, probabilities.shape[0]):
                 if (probabilities[j][1] <= threshold):</pre>
                     predictions.append(0)
                 else:
                     predictions.append(1)
             ##https://scikit-learn.org/stable/modules/generated/sklearn.metrics.com
             #uses the confusion matrix documentation on Scikit Learn to calculate t
             confusionMatrix = confusion matrix(testingTargets, predictions)
             #uses the Confusion Matrix information page on Scikit Learn to find out
             #https://scikit-learn.org/stable/auto examples/model selection/plot cor
             normalizedConfusionMatrix = confusionMatrix.astype('float') / confusion
             plotMatrix(confusionMatrix, label + (" Confusion Matrix (Threshold: %.4
```

plotMatrix(normalizedConfusionMatrix, label + (" Normalized Confusion M

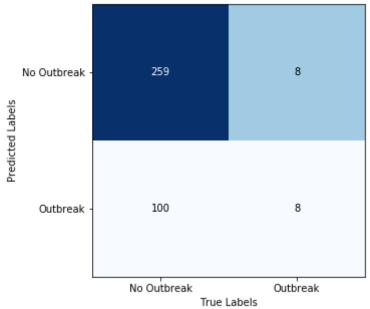
In [5]:

```
#The citations for the classifiers can be found in their respective files
decisionTree = DecisionTreeClassifier(criterion = "entropy", max_depth = 4)
decisionTree.fit(trainingFeatures, trainingTargets)
randomForest = RandomForestClassifier(n estimators=100, max depth=4, randomForestClassifier(n estimato
randomForest.fit(trainingFeatures, trainingTargets)
knn = KNeighborsClassifier(n neighbors=6)
knn.fit(trainingFeatures, trainingTargets)
gaussianNaiveBayes = GaussianNB()
gaussianNaiveBayes.fit(trainingFeatures, trainingTargets)
multinomialNaiveBayes = MultinomialNB()
multinomialNaiveBayes.fit(trainingFeatures, trainingTargets)
bernoulliNaiveBayes = BernoulliNB(binarize=0.1)
bernoulliNaiveBayes.fit(trainingFeatures, trainingTargets)
network = MLPClassifier(solver='lbfgs', alpha=1e-5, hidden layer sizes=(3,)
network.fit(trainingFeatures, trainingTargets)
svm = SVC(kernel='poly', degree=3, random_state=0, gamma='auto', probabilit
svm.fit(trainingFeatures, trainingTargets)
classifiers = [decisionTree, randomForest, knn, gaussianNaiveBayes, multing
colors = ['red', 'blue', 'green', 'black', 'purple', 'orange', 'pink', 'mag
labels = ["Decision Tree", "Random Forest", 'K Nearest Neighbors', "Gaussia
thresholds = []
plt.rcParams['figure.figsize'] = (18.0, 10.0)
for i in range(0,8):
       outbreakProbabilities = classifiers[i].predict proba(testingFeatures)[;
       #uses the documentation on Scikit Learn for roc curve to calculate the
       #https://scikit-learn.org/stable/modules/generated/sklearn.metrics.roc
       fpr, tpr, tresh = roc curve(testingTargets, outbreakProbabilities)
       thresholds.append(np.median(outbreakProbabilities))
       #uses the documentation on Scikit Learn for auc on how to calculate auc
       #https://scikit-learn.org/stable/modules/generated/sklearn.metrics.auc.
       area = auc(fpr, tpr)
       plt.plot(fpr, tpr, color=colors[i], label=labels[i]+" AUC: "+str(area))
       plt.legend()
       plt.xlabel("False Positive Rate")
       plt.ylabel("True Positive Rate")
       plt.title("ROC Curve Comparison for Different Classifiers")
plt.show()
```

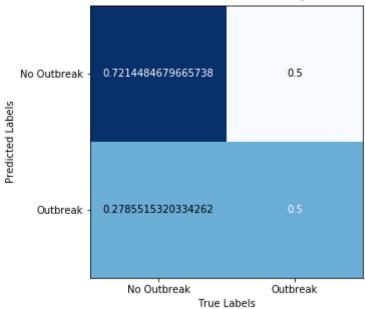


In [11]: thresholdBasedClassification(thresholds[0], classifiers[0],labels[0], testi



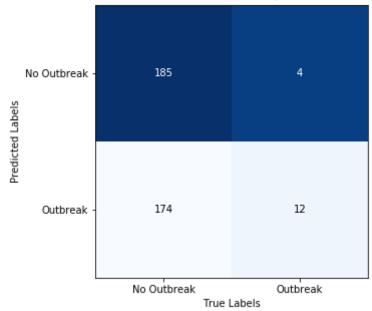


Decision Tree Normalized Confusion Matrix (Threshold: 0.0463)

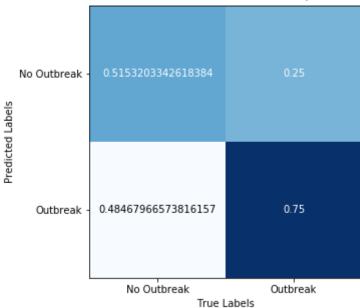


In [12]: thresholdBasedClassification(thresholds[1], classifiers[1],labels[1], testi

Random Forest Confusion Matrix (Threshold: 0.0450)

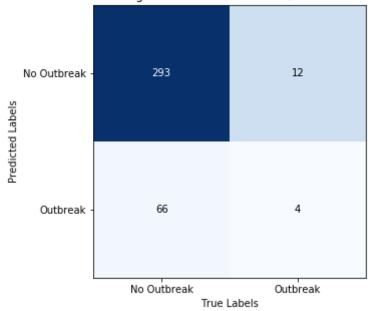


Random Forest Normalized Confusion Matrix (Threshold: 0.0450)

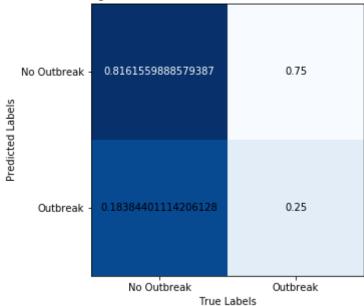


In [13]: thresholdBasedClassification(thresholds[2], classifiers[2],labels[2], testi

K Nearest Neighbors Confusion Matrix (Threshold: 0.0000)

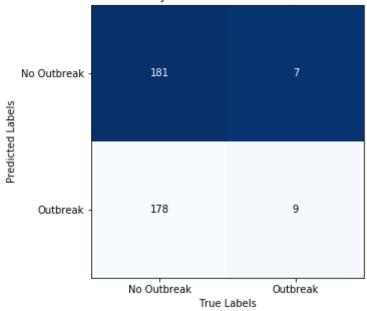


K Nearest Neighbors Normalized Confusion Matrix (Threshold: 0.0000)

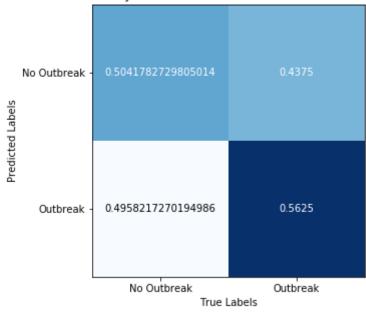


In [14]: thresholdBasedClassification(thresholds[3], classifiers[3],labels[3], testi

Gaussian Naive Bayes Confusion Matrix (Threshold: 0.0429)

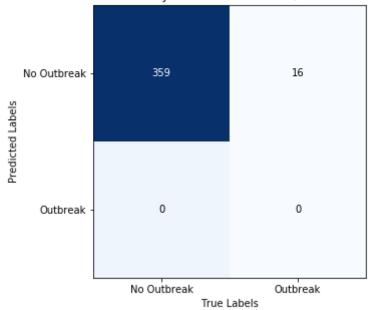


Gaussian Naive Bayes Normalized Confusion Matrix (Threshold: 0.0429)

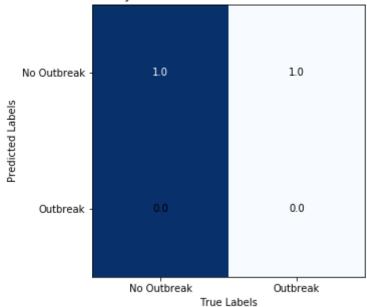


In [22]: thresholdBasedClassification(0.5, classifiers[3],labels[3], testingFeatures

Gaussian Naive Bayes Confusion Matrix (Threshold: 0.5000)

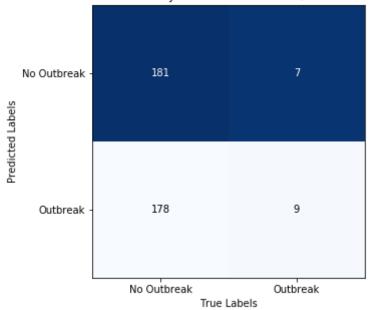


Gaussian Naive Bayes Normalized Confusion Matrix (Threshold: 0.5000)

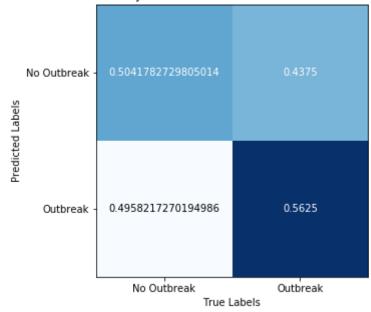


In [15]: thresholdBasedClassification(thresholds[4], classifiers[4],labels[4], testi

Multinomial Naive Bayes Confusion Matrix (Threshold: 0.0434)

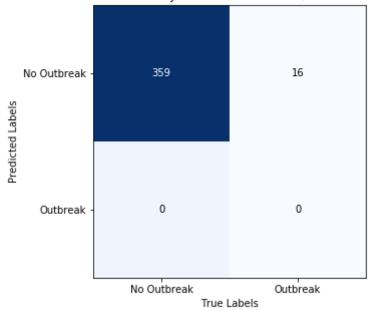


Multinomial Naive Bayes Normalized Confusion Matrix (Threshold: 0.0434)

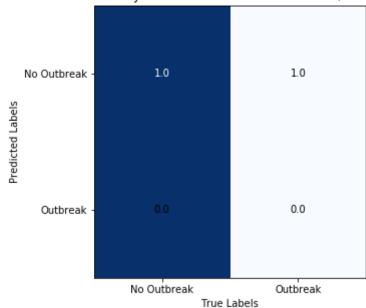


In [23]: thresholdBasedClassification(0.5, classifiers[4],labels[4], testingFeatures

Multinomial Naive Bayes Confusion Matrix (Threshold: 0.5000)

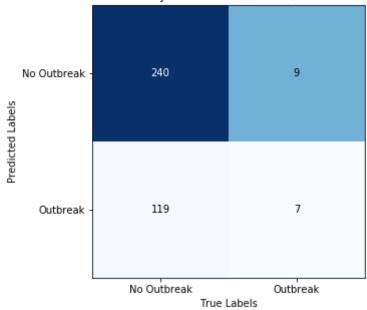


Multinomial Naive Bayes Normalized Confusion Matrix (Threshold: 0.5000)

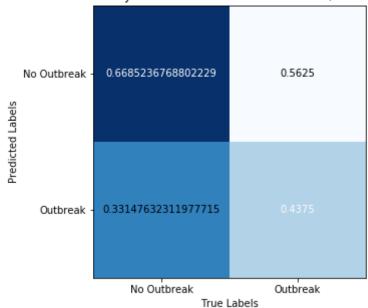


In [16]: thresholdBasedClassification(thresholds[5], classifiers[5],labels[5], testi

Bernoulli Naive Bayes Confusion Matrix (Threshold: 0.0392)

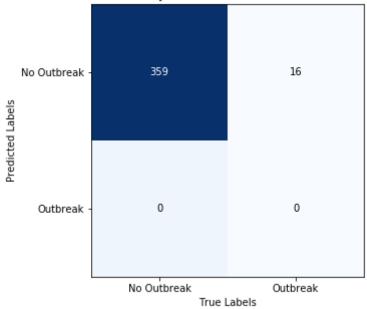


Bernoulli Naive Bayes Normalized Confusion Matrix (Threshold: 0.0392)

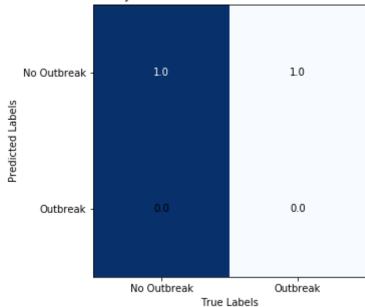


In [19]: thresholdBasedClassification(0.5, classifiers[5],labels[5], testingFeatures

Bernoulli Naive Bayes Confusion Matrix (Threshold: 0.5000)

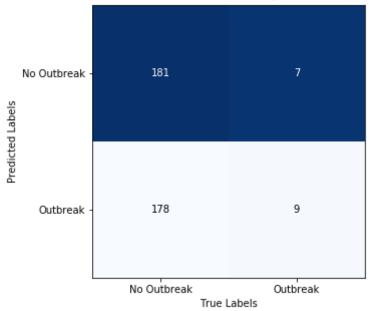


Bernoulli Naive Bayes Normalized Confusion Matrix (Threshold: 0.5000)

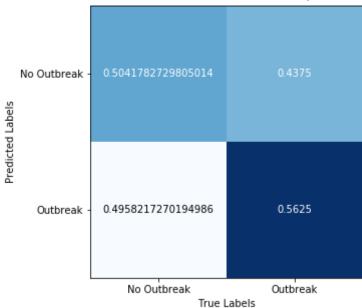


In [17]: thresholdBasedClassification(thresholds[6], classifiers[6],labels[6], testi

Neural Network Confusion Matrix (Threshold: 0.0429)

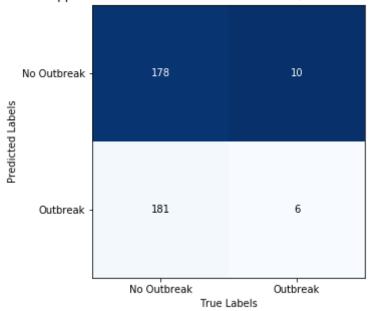


Neural Network Normalized Confusion Matrix (Threshold: 0.0429)

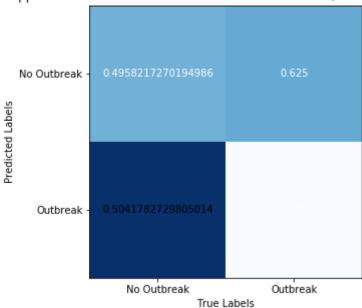


In [18]: thresholdBasedClassification(thresholds[7], classifiers[7],labels[7], testi

Support Vector Machine Confusion Matrix (Threshold: 0.0507)



Support Vector Machine Normalized Confusion Matrix (Threshold: 0.0507)



In []: