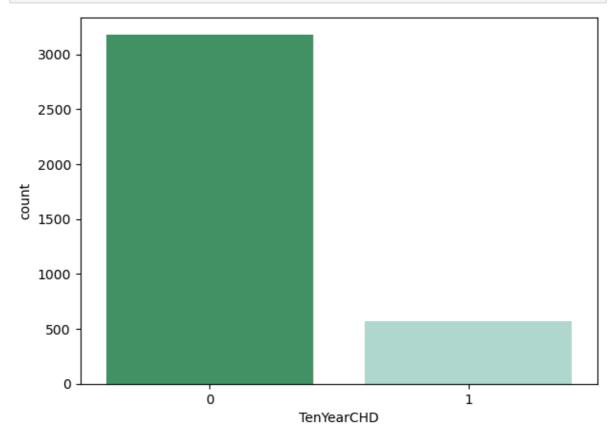
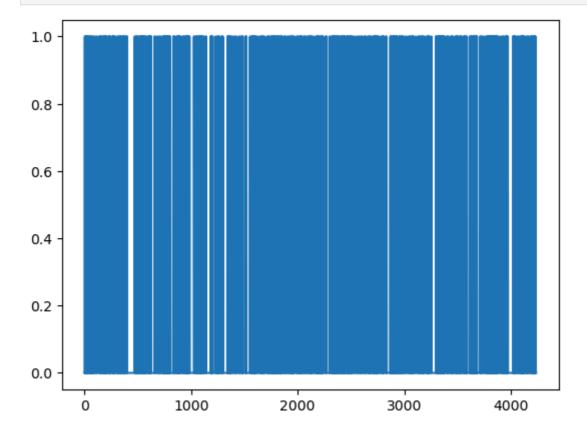
```
In [1]:
        import pandas as pd
        import pylab as pl
        import numpy as np
        import scipy.optimize as opt
        import statsmodels.api as sm
        from sklearn import preprocessing
        'exec(% matplotlib inline)'
        import matplotlib.pyplot as plt
        import matplotlib.mlab as mlab
        import seaborn as sns
In [2]: disease_df = pd.read_csv("heartD.csv")
        disease_df.drop(['education'], inplace = True, axis = 1)
        disease_df.rename(columns ={'male':'Sex_male'}, inplace = True)
        disease_df.dropna(axis = 0, inplace = True)
In [3]:
        print(disease_df.head(), disease_df.shape)
        print(disease_df.TenYearCHD.value_counts())
           Sex_male age currentSmoker cigsPerDay
                                                    BPMeds prevalentStroke \
        a
                 1
                                     0
                                               0.0
                                                       0.0
        1
                    46
                                      0
                                               0.0
                                                       0.0
        2
                  1
                     48
                                     1
                                              20.0
                                                       0.0
                                                                          0
                                              30.0
        3
                  0
                      61
                                     1
                                                       0.0
                                                                          0
        4
                  0
                     46
                                      1
                                              23.0
                                                       0.0
                                                                          0
           prevalentHyp diabetes totChol sysBP diaBP
                                                           BMI heartRate glucose \
        0
                            0 195.0 106.0 70.0 26.97
                                                                80.0
                                                                              77.0
                     0
        1
                      0
                               0
                                    250.0 121.0 81.0 28.73
                                                                     95.0
                                                                              76.0
                                     245.0 127.5
                                                  80.0 25.34
                                                                     75.0
        2
                      0
                               0
                                                                              70.0
        3
                      1
                               0
                                     225.0 150.0
                                                   95.0 28.58
                                                                     65.0
                                                                             103.0
        4
                                     285.0 130.0
                                                  84.0 23.10
                                                                     85.0
                                                                              85.0
           TenYearCHD
        0
        1
                    0
        2
                    0
        3
                    1
        4
                        (3751, 15)
        0
             3179
              572
        1
        Name: TenYearCHD, dtype: int64
In [4]: X = np.asarray(disease_df[['age', 'Sex_male', 'cigsPerDay',
                                   'totChol', 'sysBP', 'glucose']])
        y = np.asarray(disease_df['TenYearCHD'])
        X = preprocessing.StandardScaler().fit(X).transform(X)
        from sklearn.model_selection import train_test_split
        X_train, X_test, y_train, y_test = train_test_split(
                X, y, test_size = 0.3, random_state = 4)
        print ('Train set:', X_train.shape, y_train.shape)
        print ('Test set:', X_test.shape, y_test.shape)
        Train set: (2625, 6) (2625,)
        Test set: (1126, 6) (1126,)
        plt.figure(figsize=(7, 5))
        sns.countplot(x='TenYearCHD', data=disease_df,
```

```
palette="BuGn_r")
plt.show()
```



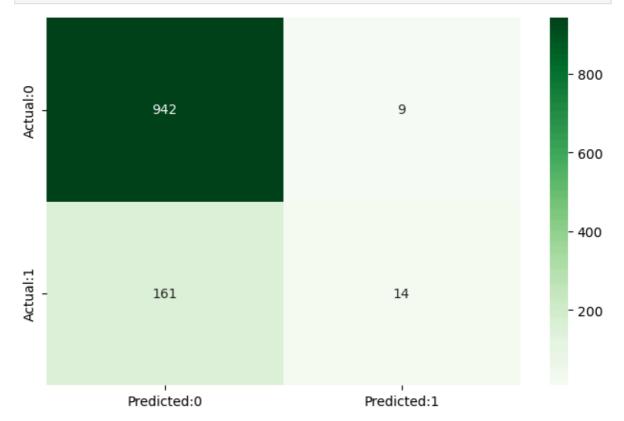
In [7]: lasten = disease_df['TenYearCHD'].plot()
plt.show(lasten)



```
In [8]: #Logistic Regression
    from sklearn.linear_model import LogisticRegression
    logreg = LogisticRegression()
```

```
logreg.fit(X_train, y_train)
y_pred = logreg.predict(X_test)
```

Accuracy of the model is = 0.8490230905861457



The details for confusion matrix is = recall f1-score precision support 0 0.85 0.99 0.92 951 1 0.61 0.08 0.14 175 accuracy 0.85 1126 0.73 0.54 0.53 1126 macro avg

0.85

0.82

```
In [32]: #Random Forest
from sklearn.ensemble import RandomForestClassifier
    rf_classifier = RandomForestClassifier(n_estimators=100, random_state=4)
```

0.80

1126

weighted avg

```
rf_classifier.fit(X_train, y_train)
         y_pred_rf = rf_classifier.predict(X_test)
In [33]:
         accuracy_rf = accuracy_score(y_test, y_pred_rf)
         print("Random Forest Classifier Accuracy:", accuracy_rf)
         Random Forest Classifier Accuracy: 0.8268974700399467
         cm_rf = confusion_matrix(y_test, y_pred_rf)
In [47]:
          conf_matrix_rf = pd.DataFrame(data = cm_rf,
                                     columns = ['Predicted:0', 'Predicted:1'],
                                     index =['Actual:0', 'Actual:1'])
          plt.figure(figsize = (8, 5))
          sns.heatmap(conf_matrix_rf, annot = True, fmt = 'd', cmap = "Reds")
          plt.show()
          print('The details for confusion matrix is =')
          print (classification_report(y_test, y_pred_rf))
                                                                                          600
                                                                                         - 500
          Actual:0
                             612
                                                               10
                                                                                         - 400
                                                                                         - 300
                                                                                         - 200
                                                                9
                             120
                                                                                        - 100
                         Predicted:0
                                                           Predicted:1
         The details for confusion matrix is =
                        precision
                                     recall f1-score
                                                         support
                     0
                             0.84
                                       0.98
                                                 0.90
                                                             622
                     1
                             0.47
                                       0.07
                                                 0.12
                                                             129
                                                 0.83
                                                             751
             accuracy
                             0.65
                                       0.53
                                                 0.51
                                                             751
            macro avg
         weighted avg
                             0.77
                                       0.83
                                                  0.77
                                                             751
         #SVM
In [40]:
         from sklearn.svm import SVC
          svm_classifier = SVC(kernel='rbf', random_state=4)
          svm classifier.fit(X train, y train)
         y_pred_svm = svm_classifier.predict(X_test)
```

In [41]:

accuracy_svm = accuracy_score(y_test, y_pred_svm)

print("Accuracy:", accuracy_svm)

```
Accuracy: 0.833555259653795
```

```
In [48]:
         cm_svm = confusion_matrix(y_test, y_pred_svm)
          conf_matrix_svm = pd.DataFrame(data = cm_svm,
                                     columns = ['Predicted:0', 'Predicted:1'],
                                     index =['Actual:0', 'Actual:1'])
          plt.figure(figsize = (8, 5))
         sns.heatmap(conf_matrix_svm, annot = True, fmt = 'd', cmap = "Blues")
          plt.show()
          print('The details for confusion matrix is =')
         print (classification_report(y_test, y_pred_svm))
                                                                                          600
                                                                                         - 500
                            622
                                                                0
                                                                                         - 400
                                                                                         - 300
                                                                                         - 200
                             125
                                                                4
                                                                                        - 100
                                                                                        - 0
                         Predicted:0
                                                           Predicted:1
         The details for confusion matrix is =
                        precision
                                     recall f1-score
                                                        support
                     0
                             0.83
                                       1.00
                                                 0.91
                                                             622
                             1.00
                                       0.03
                                                 0.06
                                                             129
             accuracy
                                                 0.83
                                                             751
                             0.92
                                       0.52
                                                 0.48
                                                             751
            macro avg
                                                 0.76
         weighted avg
                             0.86
                                       0.83
                                                             751
In [25]:
         from sklearn.neighbors import KNeighborsClassifier
          knn classifier = KNeighborsClassifier(n neighbors=5)
          knn_classifier.fit(X_train, y_train)
         y_pred_knn = knn_classifier.predict(X_test)
In [26]:
         accuracy_knn = accuracy_score(y_test, y_pred_knn)
          print("k-NN Accuracy:", accuracy_knn)
         k-NN Accuracy: 0.8135818908122503
         cm_knn = confusion_matrix(y_test, y_pred_knn)
In [51]:
          conf_matrix_knn = pd.DataFrame(data = cm_knn,
                                     columns = ['Predicted:0', 'Predicted:1'],
```

```
index =['Actual:0', 'Actual:1'])

plt.figure(figsize = (8, 5))
sns.heatmap(conf_matrix_knn, annot = True, fmt = 'd', cmap = "Purples")

plt.show()
print('The details for confusion matrix is =')
print (classification_report(y_test, y_pred_knn))
```



The details for confusion matrix is =

	precision	recall	f1-score	support
0	0.84	0.96	0.89	622
1	0.37	0.12	0.18	129
accuracy			0.81	751
macro avg	0.60	0.54	0.54	751
weighted avg	0.76	0.81	0.77	751

In []: