In [230]:

```
# Importing required packages
from pandas import read_csv
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score
from sklearn.svm import SVC
from sklearn.compose import ColumnTransformer
from sklearn.preprocessing import OneHotEncoder
from sklearn.preprocessing import StandardScaler
import matplotlib.pyplot as plotPointer
import numpy as np
```

In [231]:

```
strokesData = read_csv('F:\ML\For Stroke Risk Dataset\healthcare-dataset-stroke-data.csv')
# strokesData.head()
strokesData
```

Out[231]:

| | id | gender | age | hypertension | heart_disease | ever_married | work_type | Residence_type | avg_glucose_level | bmi | smoking_status | stroke |
|------|-------|--------|------|--------------|---------------|--------------|---------------|----------------|-------------------|------|-----------------|--------|
| 0 | 9046 | Male | 67.0 | 0 | 1 | Yes | Private | Urban | 228.69 | 36.6 | formerly smoked | 1 |
| 1 | 51676 | Female | 61.0 | 0 | 0 | Yes | Self-employed | Rural | 202.21 | NaN | never smoked | 1 |
| 2 | 31112 | Male | 80.0 | 0 | 1 | Yes | Private | Rural | 105.92 | 32.5 | never smoked | 1 |
| 3 | 60182 | Female | 49.0 | 0 | 0 | Yes | Private | Urban | 171.23 | 34.4 | smokes | 1 |
| 4 | 1665 | Female | 79.0 | 1 | 0 | Yes | Self-employed | Rural | 174.12 | 24.0 | never smoked | 1 |
| | | | | | | | ••• | | | | | |
| 5105 | 18234 | Female | 80.0 | 1 | 0 | Yes | Private | Urban | 83.75 | NaN | never smoked | 0 |
| 5106 | 44873 | Female | 81.0 | 0 | 0 | Yes | Self-employed | Urban | 125.20 | 40.0 | never smoked | 0 |
| 5107 | 19723 | Female | 35.0 | 0 | 0 | Yes | Self-employed | Rural | 82.99 | 30.6 | never smoked | 0 |
| 5108 | 37544 | Male | 51.0 | 0 | 0 | Yes | Private | Rural | 166.29 | 25.6 | formerly smoked | 0 |
| 5109 | 44679 | Female | 44.0 | 0 | 0 | Yes | Govt_job | Urban | 85.28 | 26.2 | Unknown | 0 |

5110 rows x 12 columns

In [232]:

```
strokesData = strokesData[['hypertension', 'heart_disease', 'avg_glucose_level', 'bmi', 'smoking_status', 'stroke']]
# strokesData['smoking_status'] = strokesData['smoking_status'].replace('formerly smoked', 1)
# strokesData['smoking_status'] = strokesData['smoking_status'].replace('smokes', 2)
# strokesData['smoking_status'] = strokesData['smoking_status'].replace('never smoked', 0)
# strokesData['smoking_status'] = strokesData['smoking_status'].replace('Unknown', 0)
strokesData = strokesData.dropna()
```

In [233]:

```
if 'id' in strokesData :
    strokesData = strokesData.drop('id', axis=1)
strokesData
```

Out[233]:

| | hypertension | heart_disease | avg_glucose_level | bmi | smoking_status | stroke |
|------|--------------|---------------|-------------------|------|-----------------|--------|
| 0 | 0 | 1 | 228.69 | 36.6 | formerly smoked | 1 |
| 2 | 0 | 1 | 105.92 | 32.5 | never smoked | 1 |
| 3 | 0 | 0 | 171.23 | 34.4 | smokes | 1 |
| 4 | 1 | 0 | 174.12 | 24.0 | never smoked | 1 |
| 5 | 0 | 0 | 186.21 | 29.0 | formerly smoked | 1 |
| | | | | | | |
| 5104 | 0 | 0 | 103.08 | 18.6 | Unknown | 0 |
| 5106 | 0 | 0 | 125.20 | 40.0 | never smoked | 0 |
| 5107 | 0 | 0 | 82.99 | 30.6 | never smoked | 0 |
| 5108 | 0 | 0 | 166.29 | 25.6 | formerly smoked | 0 |
| 5109 | 0 | 0 | 85.28 | 26.2 | Unknown | 0 |
| | | | | | | |

4909 rows × 6 columns

```
Stroke-Risk-Generator-Six Features - Jupyter Notebook
In [234]:
# Spliting the data in training and testing subsets
D = strokesData.values
# iloc to select specefic rows
x = strokesData.iloc[:,:-1]
y = strokesData.iloc[:,-1]
 \texttt{\# x\_train, x\_test, y\_train, y\_test = train\_test\_split(x,y,test\_size=0.50, random\_state = 1) } 
In [235]:
transformingColoumn = ColumnTransformer(transformers = [('encoder', OneHotEncoder(), [0,4])], remainder='passthrough')
x = np.array(transformingColoumn.fit_transform(x))
x_train, x_test, y_train, y_test = train_test_split(x,y,test_size=0.50, random_state = 1)
In [236]:
sc = StandardScaler()
x_train_scaled = sc.fit_transform(x_train)
x_test_scaled = sc.fit_transform(x_test)
In [237]:
# SVM
modelSVM = SVC(random_state = 0, kernel = 'linear')
modelSVM.fit(x_train_scaled, y_train)
y_prediction_SVM = modelSVM.predict(x_test_scaled)
accOfSVM = accuracy_score(y_test, y_prediction_SVM)
accOfSVM
print("----")
print('The accuracy of the SVM for Testing is: {}'.format(accOfSVM.round(4)))
print("-----
# save the accuracy score
score = set()
score.add(('SVM', accOfSVM.round(4)))
The accuracy of the SVM for Testing is: 0.9605
In [238]:
from sklearn.tree import DecisionTreeClassifier #for using Decision Tree Algoithm
modelDT = DecisionTreeClassifier(criterion = 'entropy')
modelDT.fit(x_train_scaled, y_train) #train the model with the training dataset
y_prediction_DT = modelDT.predict(x_test_scaled)
accOfDT = accuracy_score(y_test, y_prediction_DT)
accOfDT
print("----")
print('The accuracy of the DT for Testing is: {}'.format(accOfDT.round(4)))
print("-----")
# save the accuracy score
score.add(('DT', accOfDT.round(4)))
-----
The accuracy of the DT for Testing is: 0.9214
In [239]:
# Naive Bayes
from sklearn.naive bayes import GaussianNB
modelNB = GaussianNB()
\verb|modelNB.fit(x_train_scaled, y_train)| \textit{\#train the model with the training dataset}
y_prediction_NB = modelNB.predict(x_test_scaled)
accOfNB = accuracy_score(y_prediction_NB, y_test)
accOfNB
print("----")
print('The accuracy of the NB for Testing is: {}'.format(accOfNB.round(4)))
print("----")
```

The accuracy of the NB for Testing is: 0.8721

save the accuracy score

score.add(('NB', accOfNB.round(4)))

```
In [240]:
# Linear Regression
from sklearn.linear_model import LogisticRegression # for Logistic Regression algorithm
model_LR = LogisticRegression()
model\_LR.fit(x\_train\_scaled, y\_train) #train the model with the training dataset
y_prediction_LR = model_LR.predict(x_test_scaled) #pass the testing data to the trained model
# checking the accuracy of the algorithm.
# by comparing predicted output by the model and the actual output
accOfLR = accuracy_score(y_prediction_LR, y_test)
acc0fLR
print("----")
print('The accuracy of the LR for Testing is: {}'.format(accOfLR.round(4)))
print("-----")
# save the accuracy score
score.add(('LR', accOfLR.round(4)))
The accuracy of the LR for Testing is: 0.9605
In [241]:
# KNN
from scipy import stats
{\bf from} \  \, {\bf sklearn.neighbors} \  \, {\bf import} \  \, {\bf KNeighborsClassifier} \  \, \# \  \, {\it for} \  \, {\it K} \  \, {\it nearest} \  \, {\it neighbours}
#from sklearn.linear_model import LogisticRegression # for Logistic Regression algorithm
model\_KNN = KNeighborsClassifier(n\_neighbors=3) \# 3 neighbours for putting the new data into a class
model\_KNN.fit(x\_train\_scaled, y\_train) #train the model with the training dataset
y_prediction_KNN = model_KNN.predict(x_test_scaled)
accOfKNN = accuracy_score(y_prediction_KNN, y_test)
accOfKNN
print('The accuracy of the KNN for Testing is: {}'.format(accOfKNN.round(4)))
print("----")
# save the accuracy score
score.add(('KNN', accOfKNN.round(4)))
The accuracy of the KNN for Testing is: 0.9523
\verb|C:\ProgramData\Anaconda3| lib\site-packages \\ sklearn \\ neighbors \\ \\ classification.py: 228: Future \\ Warning: Unlike other reduction \\ lib_{action} \\ l
functions (e.g. `skew`, `kurtosis`), the default behavior of `mode` typically preserves the axis it acts along. In SciPy 1. 11.0, this behavior will change: the default value of `keepdims` will become False, the `axis` over which the statistic is taken will be eliminated, and the value None will no longer be accepted. Set `keepdims` to True or False to avoid this warn
    mode, _ = stats.mode(_y[neigh_ind, k], axis=1)
In [242]:
```

```
print("The accuracy scores of different Models:")
print("----")
for s in score:
  print(s)
```

The accuracy scores of different Models: ('DT', 0.9214) ('LR', 0.9605) ('NB', 0.8721) ('KNN', 0.9523) ('SVM', 0.9605)

In []: