```
import numpy as np
import pandas as pd
df=pd.read_csv('/content/stroke.csv')
df
```

	id	gender	age	hypertension	heart_disease	ever_married	work_type	Resid		
0	9046	Male	67.0	0	1	Yes	Private			
1	51676	Female	61.0	0	0	Yes	Self- employed			
2	31112	Male	80.0	0	1	Yes	Private			
3	60182	Female	49.0	0	0	Yes	Private			
4	1665	Female	79.0	1	0	Yes	Self- employed			
5105	18234	Female	80.0	1	0	Yes	Private			
5106	44873	Female	81.0	0	0	Yes	Self- employed			
5107	19723	Female	35.0	0	0	Yes	Self- employed			
5108	37544	Male	51.0	0	0	Yes	Private			
5109	44679	Female	44.0	0	0	Yes	Govt_job			
5110 rows × 12 columns										

Next steps: Generate code with df View recommended plots

df.columns

df.drop(['id','Residence\_type'],axis=1,inplace=True)
df

	gender	age	hypertension	heart_disease	ever_married	work_type	avg_glucose_
0	Male	67.0	0	1	Yes	Private	
1	Female	61.0	0	0	Yes	Self- employed	
2	Male	80.0	0	1	Yes	Private	
3	Female	49.0	0	0	Yes	Private	
4	Female	79.0	1	0	Yes	Self- employed	
5105	Female	80.0	1	0	Yes	Private	
5106	Female	81.0	0	0	Yes	Self- employed	
5107	Female	35.0	0	0	Yes	Self- employed	
£100	Mala	51 N	0	n	Voc	Drivoto	<b>&gt;</b>

Next steps: Generate code with df View recommended plots

df.head()

```
gender
                 age hypertension heart_disease ever_married work_type avg_glucose_lev
                                  0
      n
           Male
                67.0
                                                              Yes
                                                                       Private
                                                                                          228
                                                                        Self-
                61.0
                                  0
                                                  0
                                                                                          202
      1 Female
                                                              Yes
                                                                    employed
      2
           Male
                 80.0
                                  0
                                                              Yes
                                                                       Private
                                                                                          105
                                  0
      3 Female
                49 0
                                                  0
                                                                       Private
                                                                                          171
                                                              Yes
              Generate code with df
                                       View recommended plots
 Next steps:
df.tail()
            gender
                         hypertension
                                        heart_disease ever_married work_type avg_glucose_
      5105 Female 80.0
                                                    0
                                                                 Yes
                                                                         Private
                                                                            Self-
      5106 Female 81.0
                                     0
                                                    0
                                                                 Yes
                                                                       employed
                                                                            Self-
      5107 Female 35.0
                                     0
                                                    0
                                                                 Yes
                                                                       employed
              Male 51.0
                                     0
                                                     0
                                                                         Private
      5108
                                                                 Yes
df.isna().sum()
df['bmi']=df['bmi'].fillna(df['bmi'].mean())
df.isna().sum()
     gender
                           0
                           0
     age
                           0
     hypertension
                           0
     heart_disease
     ever_married
                           0
     work_type
                           0
     avg_glucose_level
                           0
     bmi
                           0
     smoking_status
                           0
     stroke
     dtype: int64
df.dtypes
     gender
                             int64
     age
                           float64
     hypertension
                             int64
     heart_disease
                             int64
     ever_married
                             int64
                             int64
     work_type
     avg_glucose_level
                           float64
                           float64
     smoking_status
                             int64
     stroke
                             int64
     dtype: object
from sklearn.preprocessing import LabelEncoder
encode=LabelEncoder()
df['gender']=encode.fit_transform(df['gender'])
df['ever_married']=encode.fit_transform(df['ever_married'])
df['work_type']=encode.fit_transform(df['work_type'])
df['smoking_status']=encode.fit_transform(df['smoking_status'])
df.dtypes
    gender
\Box
                             int64
                           float64
     age
     hypertension
                             int64
                             int64
     heart_disease
                             int64
     ever_married
                             int64
     work_type
     avg_glucose_level
                           float64
     bmi
                           float64
     smoking_status
                             int64
     stroke
     dtype: object
x=df.iloc[:,:-1].values
Х
```

```
, 67.
, 1.
    array([[ 1.
                                       0.
                                                , ..., 228.69
            36.6
           [ 0.
                      , 61.
                                       0.
                                                , ..., 202.21
                                   ,
],
            28.89323691, 2.
           [ 1. , 80.
                                                , ..., 105.92
                    , 2.
            32.5
                                   ],
                                 , 0.
],
                    , 35.
          [ 0.
                                                , ..., 82.99
                  , 2.
, 51.
, 1.
            30.6
                                      0.
           [ 1.
                                                , ..., 166.29
            25.6
                                   ],
                                      0.
           [ 0.
                                                , ..., 85.28
                                   ,
]])
            26.2
                         0.
y=df.iloc[:,-1].values
У
    array([1, 1, 1, ..., 0, 0, 0])
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.30,random_state=1)
x_train
from sklearn.preprocessing import StandardScaler
scale=StandardScaler()
scale.fit(x train)
x_train=scale.transform(x_train)
x_test=scale.transform(x_test)
x train
from \ sklearn.neighbors \ import \ KNeighborsClassifier
from sklearn.naive_bayes import BernoulliNB
from sklearn.svm import SVC
from sklearn.metrics import confusion_matrix,accuracy_score,classification_report
knn=KNeighborsClassifier(n_neighbors=7)
base=BernoulliNB()
model=SVC()
1st=[knn,base,model]
for i in 1st:
 i.fit(x_train,y_train)
 y_pred=i.predict(x_test)
 print("\n")
 print("Model is", i)
 print('*'*100)
 print(confusion_matrix(y_test,y_pred))
 print("Score is",accuracy_score(y_test,y_pred))
  print(classification_report(y_test,y_pred))
    Model is KNeighborsClassifier(n neighbors=7)
    [[1449 1]
     [ 82
            1]]
    Score is 0.9458577951728636
                precision recall f1-score support
                           1.00
                                              1450
              0
                     0.95
                                       0.97
                              0.01
                     0.50
                                       0.02
                                                 83
                                       0.95
                                                1533
       accuracy
                           0.51
0.95
                     0.72
                                       0.50
                                                1533
       macro avg
                     0.92
                                       0.92
    weighted avg
                                                1533
    Model is BernoulliNB()
                        9]
    [[1441
     [ 80
             3]]
    Score is 0.9419439008480104
                 precision recall f1-score support
              0
                     0.95
                              0.99
                                       0.97
                                                1450
                                                83
              1
                     0.25
                              0.04
                                       0.06
                                       0.94
                                                1533
        accuracy
                     0.60
       macro avg
                           0.51
                                       0.52
                                                1533
    weighted avg
                     0.91
                              0.94
                                       0.92
                                                1533
```

```
Model is SVC()
[[1450 0]
 [ 83
        0]]
Score is 0.9458577951728636
            precision
                       recall f1-score support
          0
                 0.95
                          1.00
                                   0.97
                                             1450
                 0.00
                          0.00
                                   0.00
                                              83
   accuracy
                                   0.95
                                             1533
                 0.47
                          0.50
                                    0.49
                                             1533
  macro avg
weighted avg
                 0.89
                          0.95
                                    0.92
                                             1533
```

/usr/local/lib/python3.10/dist-packages/sklearn/metrics/\_classification.py:1344: UndefinedMetricWarning: Precision and F-score are i \_warn\_prf(average, modifier, msg\_start, len(result))

<sup>/</sup>usr/local/lib/python3.10/dist-packages/sklearn/metrics/\_classification.py:1344: UndefinedMetricWarning: Precision and F-score are i \_warn\_prf(average, modifier, msg\_start, len(result))

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