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

	mean_radius	mean_texture	mean_perimeter	mean_area	mean_smoothness	diagnosi
0	17.99	10.38	122.80	1001.0	0.11840	(
1	20.57	17.77	132.90	1326.0	0.08474	(
2	19.69	21.25	130.00	1203.0	0.10960	(
3	11.42	20.38	77.58	386.1	0.14250	(
4	20.29	14.34	135.10	1297.0	0.10030	(
						••
564	21.56	22.39	142.00	1479.0	0.11100	(
565	20.13	28.25	131.20	1261.0	0.09780	(
566	16.60	28.08	108.30	858.1	0.08455	(
567	20.60	29.33	140.10	1265.0	0.11780	(
568	7.76	24.54	47.92	181.0	0.05263	
569 rows × 6 columns						

Next steps:

Generate code with df

View recommended plots

df.groupby('diagnosis')['diagnosis'].count()

diagnosis 0 212 1 357

Name: diagnosis, dtype: int64

df.head()

	mean_radius	mean_texture	mean_perimeter	mean_area	mean_smoothness	diagnosis
0	17.99	10.38	122.80	1001.0	0.11840	0
1	20.57	17.77	132.90	1326.0	0.08474	0
2	19.69	21.25	130.00	1203.0	0.10960	0
3	11.42	20.38	77.58	386.1	0.14250	0
4	20.29	14.34	135.10	1297.0	0.10030	0
4						<b>•</b>

Next steps:

Generate code with df

View recommended plots

df.tail()

	mean_radius	mean_texture	mean_perimeter	mean_area	mean_smoothness	diagnosi
564	21.56	22.39	142.00	1479.0	0.11100	
565	20.13	28.25	131.20	1261.0	0.09780	(
566	16.60	28.08	108.30	858.1	0.08455	(
567	20.60	29.33	140.10	1265.0	0.11780	(
568	7.76	24.54	47.92	181.0	0.05263	
4						<b>)</b>

```
df.isna().sum()
    mean_radius
    mean_texture
                      0
                      0
    mean_perimeter
    mean_area
                      a
    mean smoothness
                      0
    diagnosis
                      0
    dtype: int64
df.dtypes
    mean_radius
                      float64
                      float64
    mean texture
                      float64
    mean perimeter
                      float64
    mean area
    mean_smoothness
                      float64
                        int64
    diagnosis
    dtype: object
x=df.iloc[:,:-1].values
    array([[1.799e+01, 1.038e+01, 1.228e+02, 1.001e+03, 1.184e-01],
           [2.057e+01, 1.777e+01, 1.329e+02, 1.326e+03, 8.474e-02],
           [1.969e+01, 2.125e+01, 1.300e+02, 1.203e+03, 1.096e-01],
           [1.660e+01, 2.808e+01, 1.083e+02, 8.581e+02, 8.455e-02],
           [2.060e+01, 2.933e+01, 1.401e+02, 1.265e+03, 1.178e-01],
           [7.760e+00, 2.454e+01, 4.792e+01, 1.810e+02, 5.263e-02]])
y=df.iloc[:,-1].values
У
    0,\ 0,\ 1,\ 0,\ 1,\ 1,\ 1,\ 1,\ 1,\ 0,\ 0,\ 1,\ 0,\ 0,\ 1,\ 1,\ 1,\ 1,\ 0,\ 1,\ 0,\ 0,
           1, 1, 1, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0,
           1, 1, 1, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 1, 1, 1, 0, 1, 1, 0, 1,
           1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 1, 0,
           0, 1, 0, 0, 1, 1, 0, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1,
             1, 0, 1, 1, 1, 1, 0, 0, 1, 0, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1, 1,
           1, 0, 1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1, 1, 0, 0, 1, 0, 0,
           0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0,
           1, 1, 1, 0, 1, 1, 1, 1, 0, 0, 1, 1, 0, 1, 1, 0, 0, 1, 0, 1, 1,
           1, 1, 0, 1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
           0, 0, 1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 0, 1, 0, 0, 1, 1,
           1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 1, 1,
           1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1, 0, 0,
           0, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0,
           0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0,
           1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 1, 0, 0, 1, 1,
           1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 0,
           1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 0, 1, 0, 1, 1, 1, 1,
           1, 0, 1, 1, 0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0,
```

```
1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1,
            1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 1, 0, 1, 0, 1, 1,
            1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1,
            1, 1, 1, 1, 1, 0, 1, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
            1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 1])
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=42)
y train
     array([1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 1, 0, 1, 0, 1, 1,
            1, 1, 0, 1, 1, 0, 1, 1, 1, 0, 1, 0, 0, 1, 1, 1, 0, 1, 1, 1, 0,
            1, 1, 1, 1, 1, 0, 1, 0, 0, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0,
            1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 1, 0, 1, 1, 0, 1, 1, 1, 0, 1, 0, 1,
            0, 1, 0, 1, 1, 0, 1, 1, 1, 0, 1, 1, 1, 0, 1, 1, 0, 1, 1, 0, 1,
            1, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 0, 1, 0, 0, 1, 1,
            0,\ 1,\ 0,\ 0,\ 0,\ 1,\ 1,\ 1,\ 0,\ 1,\ 1,\ 0,\ 1,\ 1,\ 1,\ 0,\ 1,\ 1,
            0, 0, 1, 1, 0, 1, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 0, 1,
            0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 1, 1, 0, 1, 1, 1, 1,
            0, 1, 1, 0, 0, 1, 0, 1, 0, 1, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 1, 1,
            1, 1, 1, 0, 1, 1, 0, 1, 0, 0, 0, 1, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1,
            1, 1, 1, 0, 1, 1, 1, 0, 1, 1, 0, 0, 1, 0, 1, 0, 0, 1, 1, 0, 1, 0,
            0, 1, 0, 0, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 0, 0, 0, 1, 1, 1, 0, 0,
            1, 0, 0, 1, 1, 1, 0, 1, 0, 0, 0, 1, 1, 0, 1, 1, 0, 0, 0, 0,
            0, 1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 1, 1, 1, 1, 0, 1, 0, 1, 1, 1,
            1, 1, 0, 0, 0, 1, 1, 0, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0, 1, 1, 1, 0,
            0, 0, 1, 1, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
            1, 1, 1, 1, 0, 1, 1, 0, 1, 1, 0, 0, 0, 1, 0, 0, 1, 0, 1, 1, 1, 1,
            0, 1])
from sklearn.preprocessing import MinMaxScaler
norm=MinMaxScaler()
norm.fit(x_train)
x_train=norm.transform(x_train)
x_test=norm.transform(x_test)
x train
     array([[0.29624369, 0.27730808, 0.28381849, 0.1778941, 0.16780652],
            [0.27812332, 0.22590463, 0.26940639, 0.16437827, 0.08563782],
            [0.34276899, 0.14440311, 0.355879 , 0.20840127, 0.40231936],
            [0.32317939, 0.2404464 , 0.29937215, 0.19831803, 0.01764298],
            [0.30799745, 0.33513696, 0.3052226, 0.18411568, 0.43106353],
            [0.21984426, 0.36557322, 0.20605023, 0.12370205, 0.17464565]])
from sklearn.neighbors import KNeighborsClassifier
from sklearn.naive_bayes import GaussianNB
from sklearn.svm import SVC
from sklearn.metrics import confusion_matrix,accuracy_score
from sklearn.metrics import classification report
from sklearn.metrics import ConfusionMatrixDisplay
model1=KNeighborsClassifier(n neighbors=7)
model2=GaussianNB()
model3=SVC()
lst=[model1,model2,model3]
for i in 1st:
  i.fit(x_train,y_train)
 y_pred=i.predict(x_test)
  print("Model is",i)
 print(y pred)
  print(confusion_matrix(y_test,y_pred))
  print("Score is",accuracy_score(y_test,y_pred))
  print(classification_report(y_test,y_pred))
  print('\n')
```

```
Model is KNeighborsClassifier(n_neighbors=7)
0 1 0 1 1 1 1 0 0 1 1 1 1 1 1 1 0 1 1 1 1 1 1
[[ 53 10]
[ 4 104]]
Score is 0.9181286549707602
      precision
           recall f1-score
                    support
    0
        0.93
            0.84
                 0.88
                      63
    1
        0.91
            0.96
                 0.94
                      108
                 0.92
                      171
 accuracy
        0.92
            0.90
 macro avg
                 0.91
                      171
        0.92
            0.92
                 0.92
weighted avg
                      171
Model is GaussianNB()
0 1 0 1 1 1 1 0 0 1 1 1 1 1 1 1 0 1 1 1 1 1 1
[[ 53 10]
[ 0 108]]
Score is 0.9415204678362573
           recall f1-score
      precision
                    support
    0
        1.00
            0.84
                 0.91
                      63
    1
        0.92
            1.00
                 0.96
                      108
                 0.94
                      171
 accuracy
        0.96
            0.92
                 0.93
                      171
 macro avg
        0.95
            0.94
weighted avg
                 0.94
                      171
Model is SVC()
[[ 44 19]
[ 0 108]]
precision
           recall f1-score
                    support
    0
        1.00
            0.70
                 0.82
                      63
    1
        0.85
            1.00
                 0.92
                      108
                 0.89
                      171
 accuracy
        0.93
            0.85
                 0.87
                      171
 macro avg
weighted avg
        0.91
            0.89
                 0.88
                      171
```