

```
import pandas as pd
```

```
import numpy as np
```

```
import matplotlib.pyplot as plt
```

```
import seaborn as sns
```

```
boston = pd.read_csv('https://github.com/ybifoundation/Dataset/raw/main/Boston.csv')
```

```
iris = pd.read_csv('https://github.com/ybifoundation/Dataset/raw/main/IRIS.csv')
```

```
boston.head()
```

	CRIM	ZN	INDUS	CHAS	NX	RM	AGE	DIS	RAD	TAX	PTRATIO	B
0	0.00632	18.0	2.31	0	0.538	6.575	65.2	4.0900	1	296.0	15.3	396.90
1	0.02731	0.0	7.07	0	0.469	6.421	78.9	4.9671	2	242.0	17.8	396.90
2	0.02729	0.0	7.07	0	0.469	7.185	61.1	4.9671	2	242.0	17.8	392.83
3	0.03237	0.0	2.18	0	0.458	6.998	45.8	6.0622	3	222.0	18.7	394.63
4	0.06905	0.0	2.18	0	0.458	7.147	54.2	6.0622	3	222.0	18.7	396.90

```
iris.head()
```

	sepal_length	sepal_width	petal_length	petal_width	species	🔗
0	5.1	3.5	1.4	0.2	Iris-setosa	🔗
1	4.9	3.0	1.4	0.2	Iris-setosa	🔗
2	4.7	3.2	1.3	0.2	Iris-setosa	🔗
3	4.6	3.1	1.5	0.2	Iris-setosa	🔗
4	5.0	3.6	1.4	0.2	Iris-setosa	🔗

```
boston.columns
```

```
Index(['CRIM', 'ZN', 'INDUS', 'CHAS', 'NX', 'RM', 'AGE', 'DIS', 'RAD', 'TAX',  
       'PTRATIO', 'B', 'LSTAT', 'MEDV'],  
      dtype='object')
```

```
X = boston.drop(['MEDV'], axis=1)
```

```
X.head()
```

	<b>CRIM</b>	<b>ZN</b>	<b>INDUS</b>	<b>CHAS</b>	<b>NX</b>	<b>RM</b>	<b>AGE</b>	<b>DIS</b>	<b>RAD</b>	<b>TAX</b>	<b>PTRATIO</b>	<b>B</b>
<b>0</b>	0.00632	18.0	2.31	0	0.538	6.575	65.2	4.0900	1	296.0	15.3	396.90
<b>1</b>	0.02731	0.0	7.07	0	0.469	6.421	78.9	4.9671	2	242.0	17.8	396.90
<b>2</b>	0.02729	0.0	7.07	0	0.469	7.185	61.1	4.9671	2	242.0	17.8	392.83
<b>3</b>	0.03237	0.0	2.18	0	0.458	6.998	45.8	6.0622	3	222.0	18.7	394.63
<b>4</b>	0.06905	0.0	2.18	0	0.458	7.147	54.2	6.0622	3	222.0	18.7	396.90

```
Y = boston['MEDV']
```

```
Y
```

```
0      24.0
1      21.6
2      34.7
3      33.4
4      36.2
...
501     22.4
502     20.6
503     23.9
504     22.0
505     11.9
Name: MEDV, Length: 506, dtype: float64
```

```
iris.columns
```

```
Index(['sepal_length', 'sepal_width', 'petal_length', 'petal_width',
       'species'],
      dtype='object')
```

```
x = iris.drop(['species'], axis=1)
```

```
y = iris['species']
```

```
y
```

```
0      Iris-setosa
1      Iris-setosa
2      Iris-setosa
3      Iris-setosa
4      Iris-setosa
...
145     Iris-virginica
146     Iris-virginica
147     Iris-virginica
148     Iris-virginica
149     Iris-virginica
Name: species, Length: 150, dtype: object
```

```
from sklearn.model_selection import train_test_split
```

```
from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
a = le.fit_transform(y)
a
```

```
array([0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
       0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
       0, 0, 0, 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, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
       1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2,
       2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2,
       2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2])
```

```
x_train, x_test, y_train, y_test = train_test_split(X,Y, train_size=0.7, random_st
```

```
X_train, X_test, Y_train, Y_test = train_test_split(x,y, train_size=0.7, random_st
```

```
x_train.shape, x_test.shape, y_train.shape, y_test.shape
```

```
((354, 13), (152, 13), (354,), (152,))
```

```
X_train.shape, X_test.shape, Y_train.shape, Y_test.shape
```

```
((105, 4), (45, 4), (105,), (45,))
```

```
x_train
```

	SEPAL LENGTH	SEPAL WIDTH	PETAL LENGTH	PETAL WIDTH	EDIT
84	5.4	3.0	4.5	1.5	
13	4.3	3.0	1.1	0.1	
64	5.6	2.9	3.6	1.3	
39	5.1	3.4	1.5	0.2	
79	5.7	2.6	3.5	1.0	
...	...	...	...	...	
72	6.3	2.5	4.9	1.5	
143	6.8	3.2	5.9	2.3	
60	5.0	2.0	3.5	1.0	
50	7.0	3.2	4.7	1.4	
96	5.7	2.9	4.2	1.3	

105 rows × 4 columns

```
from sklearn.preprocessing import StandardScaler
```

```
from sklearn.preprocessing import MinMaxScaler
```

```
ss = StandardScaler()
```

```
x_train = ss.fit_transform(x_train)
```

```
X_train = ss.fit_transform(X_train)
```

```
x_train
```

```
array([[-0.14113619, -0.48175769, -0.19860022, ..., 0.00438903,
       -0.05084503, -0.01555641],
      [-0.42121529,  3.02166196, -1.33410259, ..., -1.68641979,
       0.42969249, -1.33650784],
      [-0.41266839, -0.48175769,  0.22414717, ..., 0.14148164,
       0.19739169, -0.10842497],
      ...,
      [-0.38944304, -0.48175769, -0.19860022, ..., 0.00438903,
       0.37963873,  0.77313338],
      [-0.41404001,  0.41002186, -0.81324318, ..., -0.72677154,
       0.43161763,  0.09671754],
      [-0.41578561,  2.06618387, -1.3831586 , ..., -0.04130851,
       0.39707198, -0.68781395]])
```

## X\_train

```
[ -0.97514097,  0.62399509, -1.31675446, -1.28941139],  
[ -1.7587364 , -0.27972194, -1.31675446, -1.28941139],  
[  0.72264911, -0.73158046,  0.77175337,  0.91650626],  
[ -1.36693868,  0.39806584, -1.19741116, -1.28941139],  
[  1.24504606,  0.17213658,  0.65241006,  0.5028967 ],  
[  0.3308514 , -1.86122675,  0.2347085 , -0.18645256],  
[  2.55103844, -0.05379268,  1.4878132 ,  1.60585552],  
[  0.98384759, -0.05379268,  1.30879824,  1.46798567],  
[  1.11444683, -0.05379268,  0.47339511,  0.36502685],  
[  0.85324835, -0.27972194,  0.41372346,  0.227157 ],  
  
[  0.06965293, -0.73158046,  0.89109667,  1.05437611],  
[ -0.06094631, -0.5056512 ,  0.29438015,  0.227157 ],  
[  1.89804225,  0.39806584,  1.42814154,  0.91650626],  
[  1.24504606, -0.05379268,  0.95076833,  1.60585552],  
[  0.3308514 , -1.86122675,  0.83142502,  0.5028967 ],  
[  0.85324835,  0.17213658,  1.12978328,  0.91650626],  
[  0.72264911, -1.63529749,  0.47339511,  0.227157 ],  
[ -1.23633945,  0.84992435, -1.19741116, -1.28941139],  
[ -0.97514097, -1.63529749, -0.18299307, -0.18645256],  
[ -0.45274402,  1.97957065, -1.37642611, -1.01367168],  
[  0.59204988, -0.5056512 ,  0.71208172,  0.91650626],  
[  1.24504606,  0.17213658,  0.47339511,  0.36502685],  
[ -0.19154555, -0.73158046,  0.3540518 ,  0.227157 ],  
[  0.3308514 , -0.27972194,  0.53306676,  0.5028967 ],  
[ -0.45274402,  1.97957065, -1.1377395 , -1.01367168],  
[  0.46145064, -0.05379268,  0.77175337,  0.91650626],  
[  1.3756453 , -0.05379268,  1.12978328,  1.33011582],  
[  0.72264911, -0.5056512 ,  0.89109667,  0.5028967 ],  
[  0.3308514 , -0.73158046,  0.89109667,  0.64076655],  
[ -1.23633945,  0.17213658, -1.19741116, -1.28941139],  
[  0.46145064, -0.5056512 ,  0.65241006,  0.08928714],  
[  0.06965293,  2.2054999 , -1.43609776, -1.28941139],  
[ -0.45274402,  0.84992435, -1.25708281, -1.01367168],  
[ -0.32214479, -1.40936823,  0.05569354, -0.18645256],  
[ -1.49753792,  1.30178287, -1.55544107, -1.28941139],  
[ -1.62813716, -1.63529749, -1.37642611, -1.15154153],  
[  0.72264911,  0.62399509,  0.65241006,  0.64076655],  
[ -0.32214479,  1.07585361, -1.37642611, -1.28941139],  
[  0.06965293, -0.73158046,  0.29438015, -0.18645256],  
[ -1.7587364 ,  0.39806584, -1.37642611, -1.28941139],  
[ -1.36693868,  0.39806584, -1.37642611, -1.28941139],  
[ -0.97514097,  0.84992435, -1.25708281, -1.28941139],  
[  0.06965293, -0.95750971,  0.2347085 ,  0.08928714],  
[  0.72264911,  0.84992435,  1.18945493,  1.74372538],  
[  0.46145064, -0.5056512 ,  0.2347085 ,  0.227157 ],  
[ -0.06094631, -0.5056512 ,  0.53306676,  0.227157 ],  
[ -1.10574021, -1.18343897,  0.53306676,  0.77863641],  
[ -1.10574021,  0.17213658, -1.25708281, -1.42728124],  
[ -0.97514097,  1.07585361, -1.37642611, -1.15154153],  
[  1.11444683, -0.27972194,  0.59273841,  0.227157 ],  
[ -0.97514097,  1.07585361, -1.19741116, -0.73793197],  
[  0.3308514 ,  0.84992435,  0.53306676,  0.64076655],  
[  1.24504606,  0.62399509,  1.24912659,  1.33011582],  
[  0.72264911, -1.18343897,  0.77175337,  0.5028967 ],  
[  1.3756453 ,  0.39806584,  1.36846989,  1.60585552],  
[ -0.97514097, -2.31308526, -0.06364976, -0.18645256],  
[  1.63684377,  0.39806584,  0.65241006,  0.36502685],  
[  0.06094631,  0.27972194,  0.2540510 ,  0.227157 ]]
```

```
from sklearn.neighbors import KNeighborsRegressor
```

```
from sklearn.neighbors import KNeighborsClassifier
```

```
knnreg = KNeighborsRegressor(n_neighbors=3)
```

```
knncla = KNeighborsClassifier()
```

```
knnreg.fit(x_train, y_train)
```

```
KNeighborsRegressor(n_neighbors=3)
```

```
knncla.fit(X_train, Y_train)
```

KNeighborsClassifier()

```
y_pred = knnreg.predict(x_test)
```

/usr/local/lib/python3.7/dist-packages/sklearn/base.py:444: UserWarning: X has feature names, but {self.\_\_class\_\_.\_\_name\_\_} was fitted without"

```
Y_pred = knncla.predict(X_test)
```

```
/usr/local/lib/python3.7/dist-packages/sklearn/base.py:444: UserWarning: X has feature names, but {self.class_name} was fitted without"
```

```
from sklearn.metrics import mean absolute error, mean absolute percentage error
```

```
from sklearn.metrics import classification_report, confusion_matrix
```

mean absolute error(y\_test,y\_pred)

7.4596491228070185

mean absolute percentage error(y test,y pred)

0.2993812726776956

```
confusion matrix(Y test,Y pred)
```

```
array([[ 0,  0, 14],
```

```
[ 0, 0, 9],  
[ 0, 0, 22]])
```

```
print(classification_report(Y_test,Y_pred))
```

	precision	recall	f1-score	support
Iris-setosa	0.00	0.00	0.00	14
Iris-versicolor	0.00	0.00	0.00	9
Iris-virginica	0.49	1.00	0.66	22
accuracy			0.49	45
macro avg	0.16	0.33	0.22	45
weighted avg	0.24	0.49	0.32	45

```
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:131
    _warn_prf(average, modifier, msg_start, len(result))
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:131
    _warn_prf(average, modifier, msg_start, len(result))
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:131
    _warn_prf(average, modifier, msg_start, len(result))
```

```
error_rate = []

for i in range(1,40):
    knn = KNeighborsRegressor(n_neighbors=i)
    knn.fit(x_train,y_train)
    pred_i = knn.predict(x_test)
    error_rate.append(mean_absolute_error(y_test,pred_i))
```

Error rate = [1]

```
for i in range(1,40):
    knn = KNeighborsClassifier(n_neighbors=i)
    knn.fit(X_train,Y_train)
    pred_i = knn.predict(X_test)
    Error_rate.append(classification_report(Y_test,pred_i))

    /usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1:
        _warn_prf(average, modifier, msg_start, len(result))
/usr/local/lib/python3.7/dist-packages/sklearn/base.py:444: UserWarning: X |
    f"X has feature names, but {self.__class__.__name__} was fitted without"
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1:
        _warn_prf(average, modifier, msg_start, len(result))
/usr/local/lib/python3.7/dist-packages/sklearn/base.py:444: UserWarning: X |
    f"X has feature names, but {self.__class__.__name__} was fitted without"
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1:
        _warn_prf(average, modifier, msg_start, len(result))
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        _warn_prf(average, modifier, msg_start, len(result))
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1:
        warn_prf(average, modifier, msg_start, len(result))
```

```
/usr/local/lib/python3/dist-packages/sklearn/base.py:444: UserWarning: X
  f"X has feature names, but {self.__class__.__name__} was fitted without"
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1:
  _warn_prf(average, modifier, msg_start, len(result))
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  _warn_prf(average, modifier, msg_start, len(result))
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1:
  _warn_prf(average, modifier, msg_start, len(result))
/usr/local/lib/python3.7/dist-packages/sklearn/base.py:444: UserWarning: X
  f"X has feature names, but {self.__class__.__name__} was fitted without"
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/usr/local/lib/python3.7/dist-packages/sklearn/base.py:444: UserWarning: X
  f"X has feature names, but {self.__class__.__name__} was fitted without"
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  _warn_prf(average, modifier, msg_start, len(result))
/usr/local/lib/python3.7/dist-packages/sklearn/base.py:444: UserWarning: X
  f"X has feature names, but {self.__class__.__name__} was fitted without"
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  _warn_prf(average, modifier, msg_start, len(result))
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1:
  _warn_prf(average, modifier, msg_start, len(result))
/usr/local/lib/python3.7/dist-packages/sklearn/base.py:444: UserWarning: X
  f"X has feature names, but {self.__class__.__name__} was fitted without"
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1:
  _warn_prf(average, modifier, msg_start, len(result))
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1:
  _warn_prf(average, modifier, msg_start, len(result))
/usr/local/lib/python3.7/dist-packages/sklearn/base.py:444: UserWarning: X
  f"X has feature names, but {self.__class__.__name__} was fitted without"
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1:
  _warn_prf(average, modifier, msg_start, len(result))
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1:
  _warn_prf(average, modifier, msg_start, len(result))
```

error rate

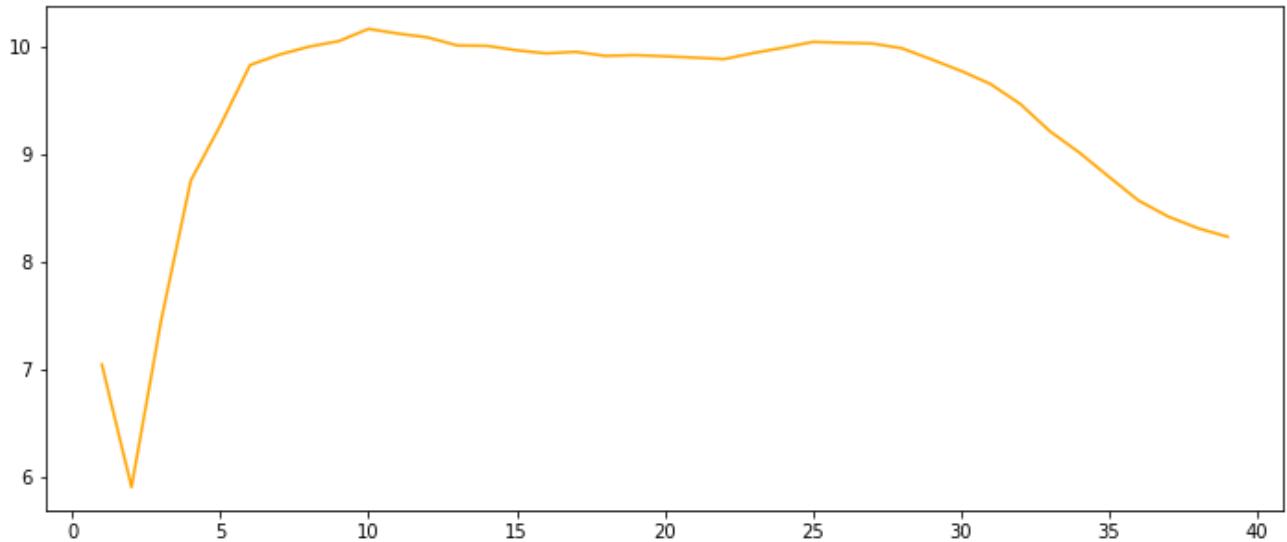
[7.050657894736842,  
5.910197368421054,  
7.4596491228070185,  
8.759868421052632,  
9.273684210526318,  
9.833552631578948,  
9.931203007518798,  
10.004605263157895,  
10.056140350877193,  
10.16953947368421,  
10.125657894736843,  
10.09018640350877,  
10.015435222672064,  
10.011983082706767,  
9.970921052631578,  
9.942228618421053

9. 956114551083592,  
9. 918421052631578,  
9. 925969529085872,  
9. 914901315789473,  
9. 902443609022558,  
9. 888397129186604,  
9. 945823798627002,  
9. 994024122807017,  
10. 048947368421052,  
10. 04046052631579,  
10. 033893762183236,  
9. 989943609022555,  
9. 886910163339383,  
9. 779495614035088,  
9. 65689728353141,  
9. 472615131578948,  
9. 218799840510366,  
9. 019736842105264,  
8. 793928571428571,  
8. 573410087719298,  
8. 425142247510667,  
8. 317122576177285,  
8. 237955465587044]

### Error\_rate

```
'          precision    recall   f1-score  support\n\n'
'Iris-setosa
```

```
fig, ax = plt.subplots(figsize = (12,5))
ax.plot(range(1,40), error_rate, color= 'orange')
plt.show()
```



```
fig, ax = plt.subplots(figsize = (12,5))
ax.plot(range(1,40), Error_rate, color= 'magenta')
plt.show()
```

	precision	recall	f1-score	support
Iris-setosa	0.00	0.00	0.00	14
Iris-versicolor	0.00	0.00	0.00	9
Iris-virginica	0.49	1.00	0.66	22
accuracy			0.49	45
macro avg	0.16	0.33	0.22	45
weighted avg	0.24	0.49	0.32	45



0s completed at 23:10

