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Assignment Day 2 : KKN Pipelines Date 12/12/2022

<https://github.com/ijazkhan101/Machine-Learning-with-Python>

```
import matplotlib.pyplot as plt
from sklearn.metrics import mean_squared_error
from sklearn.datasets import load_boston
from sklearn.preprocessing import StandardScaler
from sklearn.pipeline import Pipeline
from sklearn.neighbors import KNeighborsRegressor

X,y = load_boston(return_X_y=True)

KNN
pipe =Pipeline([("scaler:",StandardScaler()),
 ("Algo",KNeighborsRegressor())])

pipe

Pipeline(steps=[('scaler:', StandardScaler()), ('Algo',
KNeighborsRegressor())])

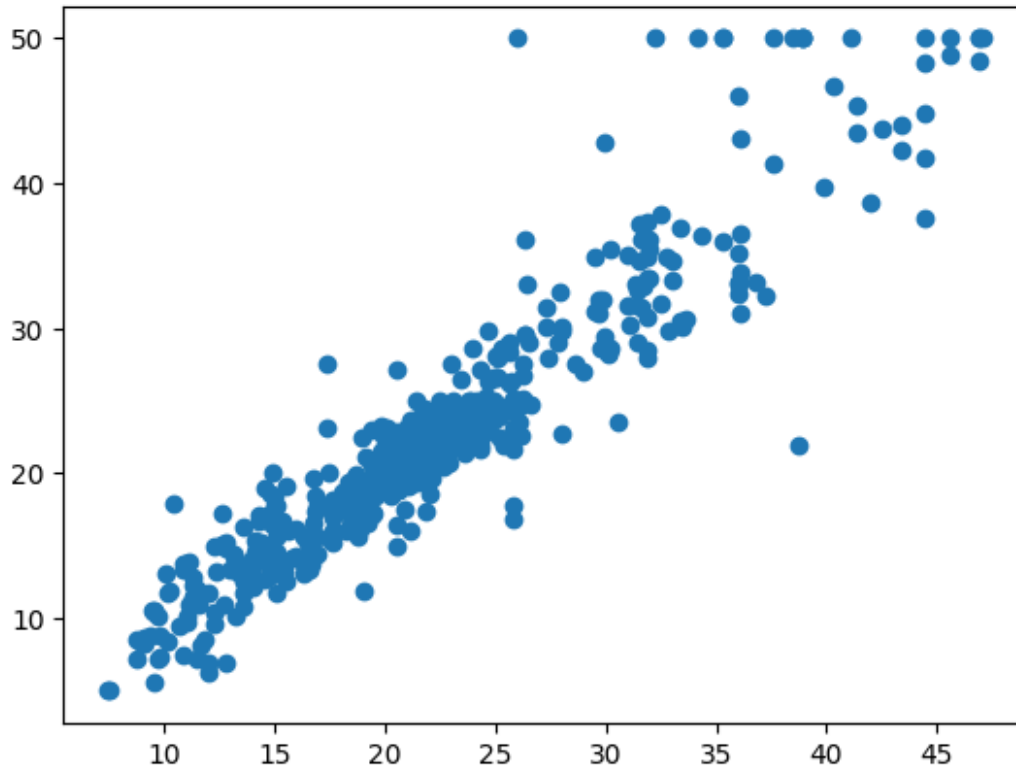
pipe.fit(X,y)

Pipeline(steps=[('scaler:', StandardScaler()), ('Algo',
KNeighborsRegressor())])

predicted_y=pipe.predict(X)

plt.scatter(predicted_y,y)

<matplotlib.collections.PathCollection at 0x212bfde8160>
```



2nd Assigment RandomizedSearchCV

```
import matplotlib.pyplot as plt
from sklearn.linear_model import LinearRegression
from sklearn.datasets import load_boston
from sklearn.preprocessing import StandardScaler
from sklearn.pipeline import Pipeline
from sklearn.model_selection import RandomizedSearchCV
from sklearn.neighbors import KNeighborsRegressor
from scipy.stats import uniform
import warnings
warnings.filterwarnings('ignore')

X,y = load_boston(return_X_y=True)

pipe =Pipeline([
    ('scaler',StandardScaler()),
    ("algo",KNeighborsRegressor())
])

pipe.get_params()

{'memory': None,
 'steps': [('scaler', StandardScaler()), ('algo',
KNeighborsRegressor())],
 'verbose': False,
 'scaler': StandardScaler(),
 'algo': KNeighborsRegressor(),
```

```

'scaler__copy': True,
'scaler__with_mean': True,
'scaler__with_std': True,
'algo__algorithm': 'auto',
'algo__leaf_size': 30,
'algo__metric': 'minkowski',
'algo__metric_params': None,
'algo__n_jobs': None,
'algo__n_neighbors': 5,
'algo__p': 2,
'algo__weights': 'uniform'}

# using RandomizedSearchCV
model =RandomizedSearchCV(
    estimator=pipe,
    param_distributions ={'algo__n_neighbors':
[1,2,3,4,5,6,7,8,9,10]},
    cv=5
)

model

RandomizedSearchCV(cv=5,
                    estimator=Pipeline(steps=[('scaler',
StandardScaler()),
                                              ('algo',
KNeighborsRegressor())]),
                    param_distributions={'algo__n_neighbors': [1, 2, 3,
4, 5, 6,
7, 8, 9,
10]})

model.fit(X, y)

RandomizedSearchCV(cv=5,
                    estimator=Pipeline(steps=[('scaler',
StandardScaler()),
                                              ('algo',
KNeighborsRegressor())]),
                    param_distributions={'algo__n_neighbors': [1, 2, 3,
4, 5, 6,
7, 8, 9,
10]})

model.predict(X)

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```

```

import pandas as pd
pd.DataFrame(model.cv_results_)

```

	mean_fit_time	std_fit_time	mean_score_time	std_score_time	\
0	0.004998	0.002531	0.004200	0.001166	
1	0.007800	0.004400	0.011601	0.010052	
2	0.007598	0.005315	0.011001	0.010899	
3	0.004597	0.001622	0.010401	0.008593	
4	0.023999	0.022564	0.007799	0.002786	
5	0.009199	0.003430	0.008600	0.007735	
6	0.007000	0.004427	0.012601	0.009457	
7	0.007597	0.002155	0.016400	0.005678	
8	0.006799	0.002317	0.007404	0.002503	
9	0.004399	0.001198	0.004599	0.001498	

param_algo__n_neighbors	params
split0_test_score \	
0	1 {'algo__n_neighbors': 1}
0.339313	
1	2 {'algo__n_neighbors': 2}
0.441649	
2	3 {'algo__n_neighbors': 3}
0.520304	
3	4 {'algo__n_neighbors': 4}
0.547088	
4	5 {'algo__n_neighbors': 5}
0.560895	
5	6 {'algo__n_neighbors': 6}
0.582450	
6	7 {'algo__n_neighbors': 7}
0.602434	
7	8 {'algo__n_neighbors': 8}
0.615090	
8	9 {'algo__n_neighbors': 9}
0.625314	
9	10 {'algo__n_neighbors': 10}
0.614446	

split1_test_score	split2_test_score	split3_test_score	
split4_test_score \			
0	0.423779	0.534566	0.486373
1.623928			-
1	0.547962	0.474980	0.496794
0.548699			-
2	0.593339	0.547746	0.513891
0.002980			
3	0.606925	0.509770	0.490452
0.211278			
4	0.619174	0.486619	0.469869
0.231330			
5	0.621194	0.509111	0.446859
0.250417			
6	0.636185	0.516102	0.442088

0.245749			
7	0.631185	0.551340	0.440117
0.239072			
8	0.630621	0.564464	0.429107
0.279376			
9	0.652489	0.555555	0.420648
0.261128			

	mean_test_score	std_test_score	rank_test_score
0	0.032020	0.830549	10
1	0.282537	0.417052	9
2	0.435652	0.218139	8
3	0.473103	0.136807	7
4	0.473577	0.132431	6
5	0.482006	0.130434	5
6	0.488512	0.139022	4
7	0.495361	0.144674	3
8	0.505776	0.134503	1
9	0.500853	0.143381	2