

Amey

December 12, 2021

0.1 Approach

- load Pandas DataFrame containing electricity data
- split the data in train, test and validation sets (+ normalise independent variables if required)
- fit model parameters using GridSearchCV [scikit-learn](#)
- evaluate estimator performance by means of 5 fold ‘shuffled’ nested cross-validation
- predict cross validated estimates of y for each data point and plot on scatter diagram vs true y
- find the best model and fit to validation set to find electricity demand

0.2 Packages required

- [Python 3.8](#)
- [Matplotlib](#)
- [Pandas](#)
- [Numpy](#)
- [scikit-learn](#)

0.3 Implement

Install packages

```
[1]: !pip install scikit-learn
      # !pip install xgboost
```

```
Requirement already satisfied: scikit-learn in
/Users/maido/PycharmProjects/pythonProject1/venv/lib/python3.8/site-packages
(1.0.1)
```

```
Requirement already satisfied: joblib>=0.11 in
/Users/maido/PycharmProjects/pythonProject1/venv/lib/python3.8/site-packages
(from scikit-learn) (1.1.0)
```

```
Requirement already satisfied: scipy>=1.1.0 in
/Users/maido/PycharmProjects/pythonProject1/venv/lib/python3.8/site-packages
(from scikit-learn) (1.7.3)
```

```
Requirement already satisfied: threadpoolctl>=2.0.0 in
/Users/maido/PycharmProjects/pythonProject1/venv/lib/python3.8/site-packages
(from scikit-learn) (3.0.0)
```

Requirement already satisfied: numpy>=1.14.6 in
/Users/maido/PycharmProjects/pythonProject1/venv/lib/python3.8/site-packages
(from scikit-learn) (1.21.4)

```
[2]: import warnings
warnings.filterwarnings('ignore')
# warnings.filterwarnings(action='once')
```

```
[15]: import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
from sklearn.ensemble import RandomForestRegressor
from sklearn.linear_model import LinearRegression, Lasso, Ridge
from sklearn.neighbors import KNeighborsRegressor
from sklearn.neural_network import MLPRegressor
from sklearn.tree import DecisionTreeRegressor
from sklearn.model_selection import train_test_split, GridSearchCV, \
    cross_val_score, cross_val_predict, KFold
from sklearn.preprocessing import StandardScaler, PolynomialFeatures
from sklearn.pipeline import Pipeline
# import xgboost as xgb
```

Preprocessing

- Read the dataset

```
[16]: file = pd.read_csv('Data.csv')
file = pd.DataFrame(file)
file
```

- Split to train, test and validation datasets

```
[17]: df = file[file.demand.notnull()]
df
```

```
[17]:
```

	period	temperature	hours before sunrise	hours before sunset	demand
0	1	8.4	6.016667	17.633333	496.0
1	2	8.1	5.516667	17.133333	535.0
2	3	7.8	5.016667	16.633333	511.0
3	4	7.5	4.516667	16.133333	496.0
4	5	7.3	4.016667	15.633333	490.0
...
48235	48236	13.2	-17.666667	-1.183333	998.0
48236	48237	12.1	-18.166667	-1.683333	867.0
48237	48238	12.1	-18.666667	-2.183333	730.0
48238	48239	12.1	-19.166667	-2.683333	608.0
48239	48240	12.0	-19.666667	-3.183333	517.0

[48240 rows x 5 columns]

```
[18]: y = df.demand
      y
```

```
[18]: 0      496.0
      1      535.0
      2      511.0
      3      496.0
      4      490.0
      ...
      48235    998.0
      48236    867.0
      48237    730.0
      48238    608.0
      48239    517.0
      Name: demand, Length: 48240, dtype: float64
```

- Drop period and demand

```
[19]: X = df.drop('period', axis=1).drop('demand', axis = 1)
      X
```

```
[19]:      temperature  hours before sunrise  hours before sunset
0           8.4           6.016667           17.633333
1           8.1           5.516667           17.133333
2           7.8           5.016667           16.633333
3           7.5           4.516667           16.133333
4           7.3           4.016667           15.633333
...
48235        13.2          -17.666667           -1.183333
48236        12.1          -18.166667           -1.683333
48237        12.1          -18.666667           -2.183333
48238        12.1          -19.166667           -2.683333
48239        12.0          -19.666667           -3.183333
```

[48240 rows x 3 columns]

- Train/test split

```
[20]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.33,
      ↪random_state=0)
```

- Validation set

```
[21]: df_vali = file[file.demand.isnull()]
```

```
[22]: vali = df_vali.drop('period', axis=1).drop('demand', axis = 1)
      vali
```

```
[22]:      temperature  hours before sunrise  hours before sunset
48240           11.9           3.833333           20.316667
48241           12.0           3.333333           19.816667
48242           12.1           2.833333           19.316667
48243           12.0           2.333333           18.816667
48244           11.9           1.833333           18.316667
...
52555           12.4          -15.516667           -3.800000
52556           12.3          -16.016667           -4.300000
52557           12.2          -16.516667           -4.800000
52558           11.9          -17.016667           -5.300000
52559           11.9          -17.516667           -5.800000
```

[4320 rows x 3 columns]

Define the pipeline models

- define pipeline
- cross validation
- show model coefficients or feature importances
- plot predicted demand vs actual demand
- fit the validation set

```
[35]: def model(pipeline, parameters, X_train, y_train, X, y):

      grid_obj = GridSearchCV(estimator=pipeline,
                              param_grid=parameters,
                              cv=3,
                              scoring='r2',
                              verbose=2,
                              n_jobs=1,
                              refit=True)

      grid_obj.fit(X_train, y_train)

      grid_obj.predict(vali)

      '''Results'''

      results = pd.DataFrame(pd.DataFrame(grid_obj.cv_results_))
      # results_sorted = results.sort_values(by=['mean_test_score'],
      ↪ascending=False)
      results_vali = grid_obj.predict(vali)
```

```

print("#### Results")
# print(results_sorted)
print(results)
print(results_vali)

print("best_index", grid_obj.best_index_)
print("best_score", grid_obj.best_score_)
print("best_params", grid_obj.best_params_)

'''Cross Validation'''
# Cross-validation is a resampling procedure used to evaluate machine_
→ learning models on a limited data sample.

estimator = grid_obj.best_estimator_
'''
if estimator.named_steps['sc1'] == True:
    X = (X - X.mean()) / (X.std())
    y = (y - y.mean()) / (y.std())
'''
shuffle = KFold(n_splits=5,
                shuffle=True,
                random_state=0)
cv_scores = cross_val_score(estimator,
                             X,
                             y.values.ravel(),
                             cv=shuffle,
                             scoring='r2')

print("#### CV Results")
print("mean_score", cv_scores.mean())

'''Show model coefficients or feature importances'''
# Feature importance refers to how useful a feature is at predicting a_
→ target variable.
# A coefficient refers to a number or quantity placed with a variable.

try:
    print("Model coefficients: ", list(zip(list(X), estimator.
→ named_steps['clf'].coef_)))
except:
    print("Model does not support model coefficients")

try:
    print("Feature importances: ", list(zip(list(X), estimator.
→ named_steps['clf'].feature_importances_)))

```

```

except:
    print("Model does not support feature importances")

    '''Predict y vs y_predicted in scatter'''

y_pred = cross_val_predict(estimator, X, y, cv=shuffle)

plt.scatter(y, y_pred)
xmin, xmax = plt.xlim()
ymin, ymax = plt.ylim()
plt.plot([xmin, xmax], [ymin, ymax], "g--", lw=1, alpha=0.4)
plt.xlabel("True demand")
plt.ylabel("Predicted demand")
plt.annotate(' R-squared CV = {}'.format(round(float(cv_scores.mean()),
↪3)), size=9,
            xy=(xmin,ymax), xytext=(10, -15), textcoords='offset points')
plt.annotate(grid_obj.best_params_, size=9,
            xy=(xmin, ymax), xytext=(10, -35), textcoords='offset points',
↪wrap=True)
plt.title('predicted demand vs actual demand')
plt.show()

    '''Fit the validation set'''

# convert array to serial
vali_series = pd.Series(results_vali)

df_vali.iloc[:,4] = vali_series.values
print(df_vali)

```

Pipeline and Parameters

- Linear Regression

```

[36]: pipe_ols = Pipeline([('scl', StandardScaler()),
                          ('clf', LinearRegression())])

param_ols = {}

```

- XGBoost

```

[37]: # # - XGBoost

# pipe_xgb = Pipeline([('clf', xgb.XGBRegressor())])

```

```
# param_xgb = {'clf__max_depth': [5],
#              'clf__min_child_weight': [6],
#              'clf__gamma': [0.01],
#              'clf__subsample': [0.7],
#              'clf__colsample_bytree': [1]}
```

- KNN

```
[38]: pipe_knn = Pipeline([('clf', KNeighborsRegressor())])

param_knn = {'clf__n_neighbors': [5, 10, 15, 25, 30]}
```

- Lasso

```
[39]: pipe_lasso = Pipeline([('scl', StandardScaler()),
                             ('clf', Lasso(max_iter=1500))])

param_lasso = {'clf__alpha': [0.01, 0.1, 1, 10]}
```

- Ridge

```
[40]: pipe_ridge = Pipeline([('scl', StandardScaler()),
                              ('clf', Ridge())])

param_ridge = {'clf__alpha': [0.01, 0.1, 1, 10]}
```

- Polynomial Regression

```
[41]: pipe_poly = Pipeline([('scl', StandardScaler()),
                             ('polynomial', PolynomialFeatures()),
                             ('clf', LinearRegression())])

param_poly = {'polynomial__degree': [2, 4, 6]}
```

- Decision Tree Regression

```
[42]: pipe_tree = Pipeline([('clf', DecisionTreeRegressor())])

param_tree = {'clf__max_depth': [2, 5, 10],
              'clf__min_samples_leaf': [5, 10, 50, 100]}
```

- Random Forest

```
[43]: pipe_forest = Pipeline([('clf', RandomForestRegressor())])

param_forest = {'clf__n_estimators': [10, 20, 50],
                'clf__max_features': [None, 1, 2],
                'clf__max_depth': [1, 2, 5]}
```

- MLP Regression

```
[44]: pipe_neural = Pipeline([('scl', StandardScaler()),
                              ('clf', MLPRegressor())])

param_neural = {'clf__alpha': [0.001, 0.01, 0.1, 1, 10, 100],
                 'clf__hidden_layer_sizes': [(5),(10,10),(7,7,7)],
                 'clf__solver': ['lbfgs'],
                 'clf__activation': ['relu', 'tanh'],
                 'clf__learning_rate' : ['constant', 'invscaling']}
```

Execute model hyperparameter tuning and crossvalidation

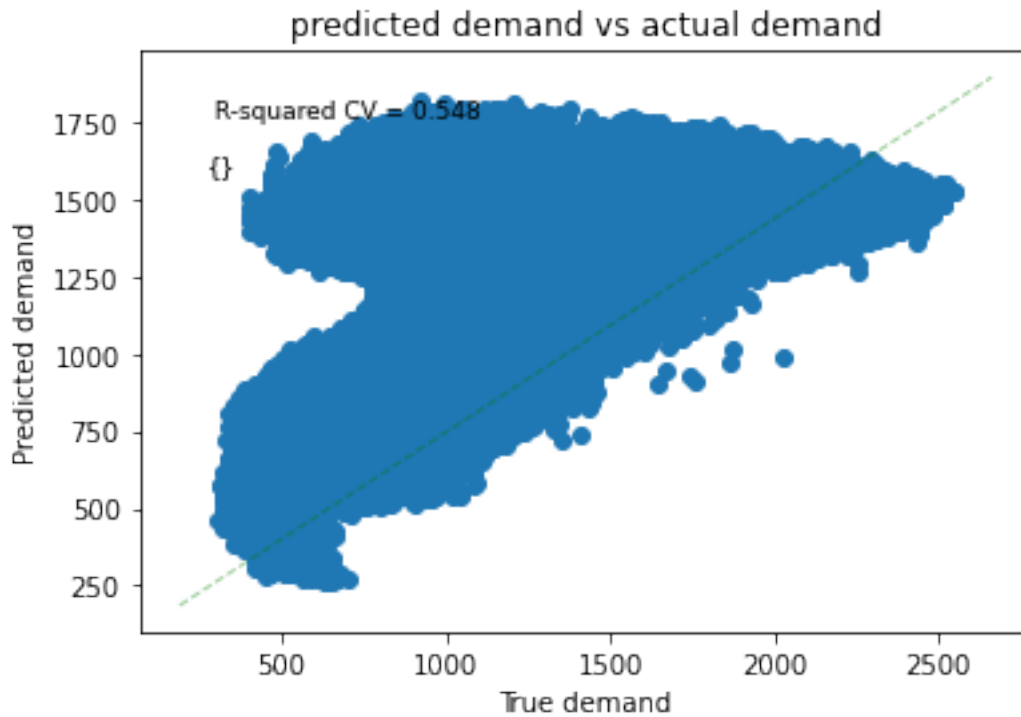
- Linear Regression

```
[45]: model(pipe_ols, param_ols, X_train, y_train, X, y)
```

```
Fitting 3 folds for each of 1 candidates, totalling 3 fits
[CV] END ... total time= 0.0s
[CV] END ... total time= 0.0s
[CV] END ... total time= 0.0s
##### Results
    mean_fit_time  std_fit_time  mean_score_time  std_score_time  params  \
0      0.00709      0.000543      0.002471      0.000276      {}

    split0_test_score  split1_test_score  split2_test_score  mean_test_score  \
0      0.547489      0.552356      0.545854      0.548566

    std_test_score  rank_test_score
0      0.002762      1
[ 321.4481177  342.15489827  362.86167883 ... 1489.06422134 1513.34921922
 1534.95055411]
best_index 0
best_score 0.548566322512026
best_params {}
##### CV Results
mean_score 0.5478994436082716
Model coefficients: [('temperature', -55.73826968410802), ('hours before
sunrise', 126.97687725007376), ('hours before sunset', -433.2156285659945)]
Model does not support feature importances
```

	period	temperature	hours before sunrise	hours before sunset	\
48240	48241	11.9	3.833333	20.316667	
48241	48242	12.0	3.333333	19.816667	
48242	48243	12.1	2.833333	19.316667	
48243	48244	12.0	2.333333	18.816667	
48244	48245	11.9	1.833333	18.316667	
...	
52555	52556	12.4	-15.516667	-3.800000	
52556	52557	12.3	-16.016667	-4.300000	
52557	52558	12.2	-16.516667	-4.800000	
52558	52559	11.9	-17.016667	-5.300000	
52559	52560	11.9	-17.516667	-5.800000	
	demand				
48240	321.448118				
48241	342.154898				
48242	362.861679				
48243	385.357568				
48244	407.853457				
...	...				
52555	1444.072443				
52556	1466.568332				
52557	1489.064221				
52558	1513.349219				

52559 1534.950554

[4320 rows x 5 columns]

- XGBoost

```
[46]: # model(pipe_xgb, param_xgb, X_train, y_train, X, y)
```

- KNN

```
[47]: model(pipe_knn, param_knn, X_train, y_train, X, y)
```

Fitting 3 folds for each of 5 candidates, totalling 15 fits

```
[CV] END ...clf__n_neighbors=5; total time= 0.1s
[CV] END ...clf__n_neighbors=5; total time= 0.0s
[CV] END ...clf__n_neighbors=5; total time= 0.0s
[CV] END ...clf__n_neighbors=10; total time= 0.1s
[CV] END ...clf__n_neighbors=10; total time= 0.1s
[CV] END ...clf__n_neighbors=10; total time= 0.1s
[CV] END ...clf__n_neighbors=15; total time= 0.1s
[CV] END ...clf__n_neighbors=15; total time= 0.1s
[CV] END ...clf__n_neighbors=15; total time= 0.1s
[CV] END ...clf__n_neighbors=25; total time= 0.1s
[CV] END ...clf__n_neighbors=25; total time= 0.1s
[CV] END ...clf__n_neighbors=25; total time= 0.1s
[CV] END ...clf__n_neighbors=30; total time= 0.1s
[CV] END ...clf__n_neighbors=30; total time= 0.1s
[CV] END ...clf__n_neighbors=30; total time= 0.1s
```

Results

	mean_fit_time	std_fit_time	mean_score_time	std_score_time	\
0	0.009826	0.000417	0.038892	0.001207	
1	0.010220	0.000744	0.052401	0.000680	
2	0.009880	0.000536	0.062440	0.002019	
3	0.009335	0.000051	0.082684	0.001527	
4	0.009375	0.000065	0.091887	0.002703	

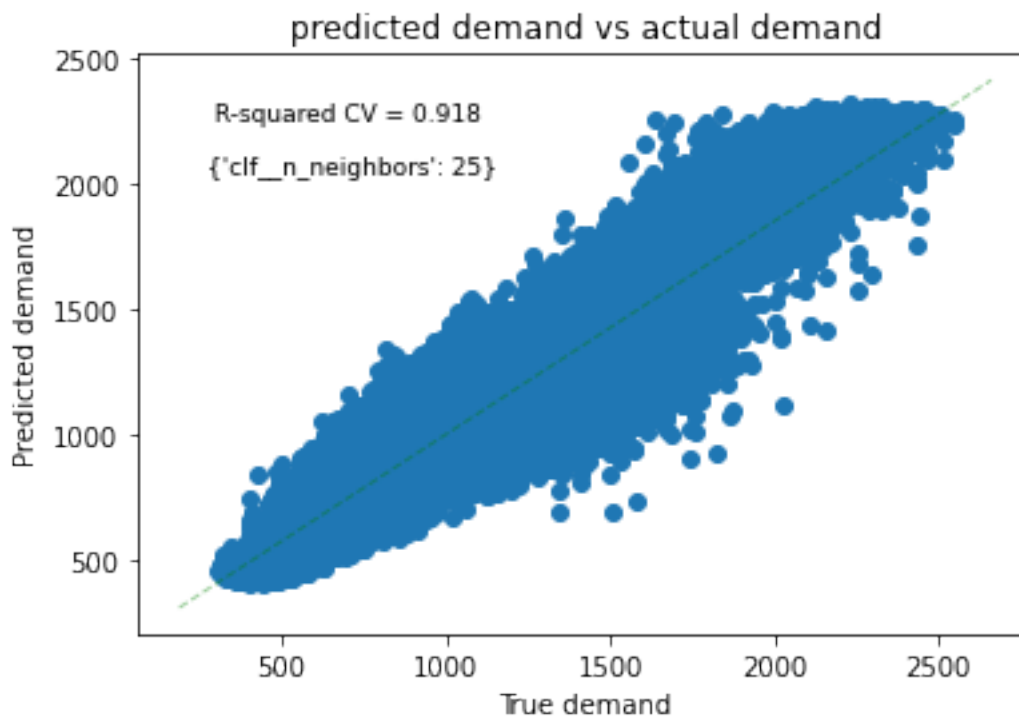
	param_clf__n_neighbors	params	split0_test_score	\
0	5	{'clf__n_neighbors': 5}	0.909001	
1	10	{'clf__n_neighbors': 10}	0.915740	
2	15	{'clf__n_neighbors': 15}	0.917366	
3	25	{'clf__n_neighbors': 25}	0.918470	
4	30	{'clf__n_neighbors': 30}	0.917932	

	split1_test_score	split2_test_score	mean_test_score	std_test_score	\
0	0.905546	0.908140	0.907563	0.001468	
1	0.913639	0.914163	0.914514	0.000893	
2	0.915896	0.916517	0.916593	0.000602	
3	0.916293	0.917259	0.917341	0.000891	
4	0.915944	0.917189	0.917022	0.000820	

```

rank_test_score
0          5
1          4
2          3
3          1
4          2
[530.16 517.68 487.36 ... 846.08 662.6  591.56]
best_index 3
best_score 0.9173405968230345
best_params {'clf__n_neighbors': 25}
##### CV Results
mean_score 0.9175986294004446
Model does not support model coefficients
Model does not support feature importances

```



	period	temperature	hours before sunrise	hours before sunset	demand
48240	48241	11.9	3.833333	20.316667	530.16
48241	48242	12.0	3.333333	19.816667	517.68
48242	48243	12.1	2.833333	19.316667	487.36
48243	48244	12.0	2.333333	18.816667	477.96
48244	48245	11.9	1.833333	18.316667	468.40
...
52555	52556	12.4	-15.516667	-3.800000	1156.68

52556	52557	12.3	-16.016667	-4.300000	1051.48
52557	52558	12.2	-16.516667	-4.800000	846.08
52558	52559	11.9	-17.016667	-5.300000	662.60
52559	52560	11.9	-17.516667	-5.800000	591.56

[4320 rows x 5 columns]

- Lasso

```
[48]: model(pipe_lasso, param_lasso, X_train, y_train, X, y)
```

Fitting 3 folds for each of 4 candidates, totalling 12 fits

```
[CV] END ...clf__alpha=0.01; total time= 0.0s
[CV] END ...clf__alpha=0.01; total time= 0.0s
[CV] END ...clf__alpha=0.01; total time= 0.0s
[CV] END ...clf__alpha=0.1; total time= 0.0s
[CV] END ...clf__alpha=0.1; total time= 0.0s
[CV] END ...clf__alpha=0.1; total time= 0.0s
[CV] END ...clf__alpha=1; total time= 0.0s
[CV] END ...clf__alpha=1; total time= 0.0s
[CV] END ...clf__alpha=1; total time= 0.0s
[CV] END ...clf__alpha=10; total time= 0.0s
[CV] END ...clf__alpha=10; total time= 0.0s
[CV] END ...clf__alpha=10; total time= 0.0s
```

Results

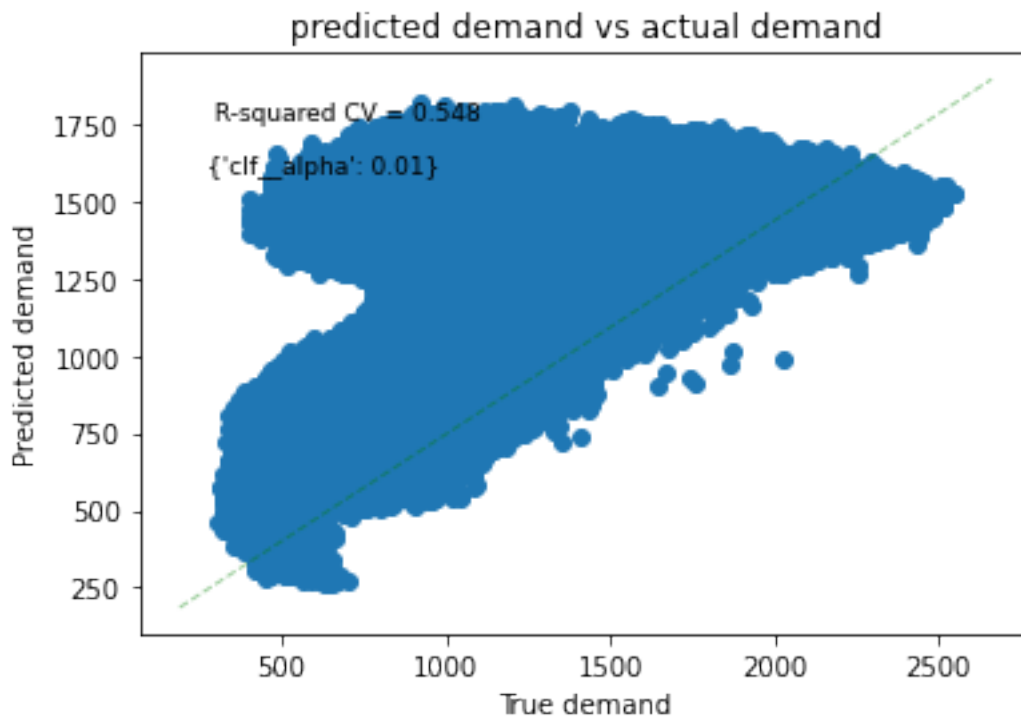
	mean_fit_time	std_fit_time	mean_score_time	std_score_time	\
0	0.011720	0.000095	0.002458	0.000129	
1	0.011157	0.000765	0.002509	0.000053	
2	0.009200	0.000532	0.002295	0.000022	
3	0.006961	0.000784	0.002316	0.000040	

	param_clf__alpha	params	split0_test_score	\
0	0.01	{'clf__alpha': 0.01}	0.547490	
1	0.1	{'clf__alpha': 0.1}	0.547501	
2	1	{'clf__alpha': 1}	0.547481	
3	10	{'clf__alpha': 10}	0.538635	

	split1_test_score	split2_test_score	mean_test_score	std_test_score	\
0	0.552357	0.545851	0.548566	0.002763	
1	0.552367	0.545827	0.548565	0.002774	
2	0.552332	0.545458	0.548424	0.002885	
3	0.543320	0.535048	0.539001	0.003387	

	rank_test_score
0	1
1	2
2	3
3	4

```
[ 321.52964041  342.23628052  362.94292063 ... 1489.05529693 1513.34207834
 1534.94375378]
best_index 0
best_score 0.5485663040582126
best_params {'clf__alpha': 0.01}
##### CV Results
mean_score 0.5478994297681836
Model coefficients: [('temperature', -55.76823988289015), ('hours before
sunrise', 126.81872246453133), ('hours before sunset', -433.0625589484212)]
Model does not support feature importances
```



	period	temperature	hours before sunrise	hours before sunset	\
48240	48241	11.9	3.833333	20.316667	
48241	48242	12.0	3.333333	19.816667	
48242	48243	12.1	2.833333	19.316667	
48243	48244	12.0	2.333333	18.816667	
48244	48245	11.9	1.833333	18.316667	
...	
52555	52556	12.4	-15.516667	-3.800000	
52556	52557	12.3	-16.016667	-4.300000	
52557	52558	12.2	-16.516667	-4.800000	
52558	52559	11.9	-17.016667	-5.300000	
52559	52560	11.9	-17.516667	-5.800000	

	demand
48240	321.529640
48241	342.236281
48242	362.942921
48243	385.439631
48244	407.936342
...	...
52555	1444.061875
52556	1466.558586
52557	1489.055297
52558	1513.342078
52559	1534.943754

[4320 rows x 5 columns]

- Ridge

```
[49]: model(pipe_ridge, param_ridge, X_train, y_train, X, y)
```

Fitting 3 folds for each of 4 candidates, totalling 12 fits

```
[CV] END ...clf__alpha=0.01; total time= 0.0s
[CV] END ...clf__alpha=0.01; total time= 0.0s
[CV] END ...clf__alpha=0.01; total time= 0.0s
[CV] END ...clf__alpha=0.1; total time= 0.0s
[CV] END ...clf__alpha=0.1; total time= 0.0s
[CV] END ...clf__alpha=0.1; total time= 0.0s
[CV] END ...clf__alpha=1; total time= 0.0s
[CV] END ...clf__alpha=1; total time= 0.0s
[CV] END ...clf__alpha=1; total time= 0.0s
[CV] END ...clf__alpha=10; total time= 0.0s
[CV] END ...clf__alpha=10; total time= 0.0s
[CV] END ...clf__alpha=10; total time= 0.0s
```

Results

	mean_fit_time	std_fit_time	mean_score_time	std_score_time	\
0	0.006983	0.001807	0.002522	0.000431	
1	0.006365	0.001351	0.002424	0.000353	
2	0.007045	0.001271	0.002121	0.000306	
3	0.005093	0.000227	0.001704	0.000032	

	param_clf__alpha	params	split0_test_score	\
0	0.01	{'clf__alpha': 0.01}	0.547489	
1	0.1	{'clf__alpha': 0.1}	0.547489	
2	1	{'clf__alpha': 1}	0.547491	
3	10	{'clf__alpha': 10}	0.547506	

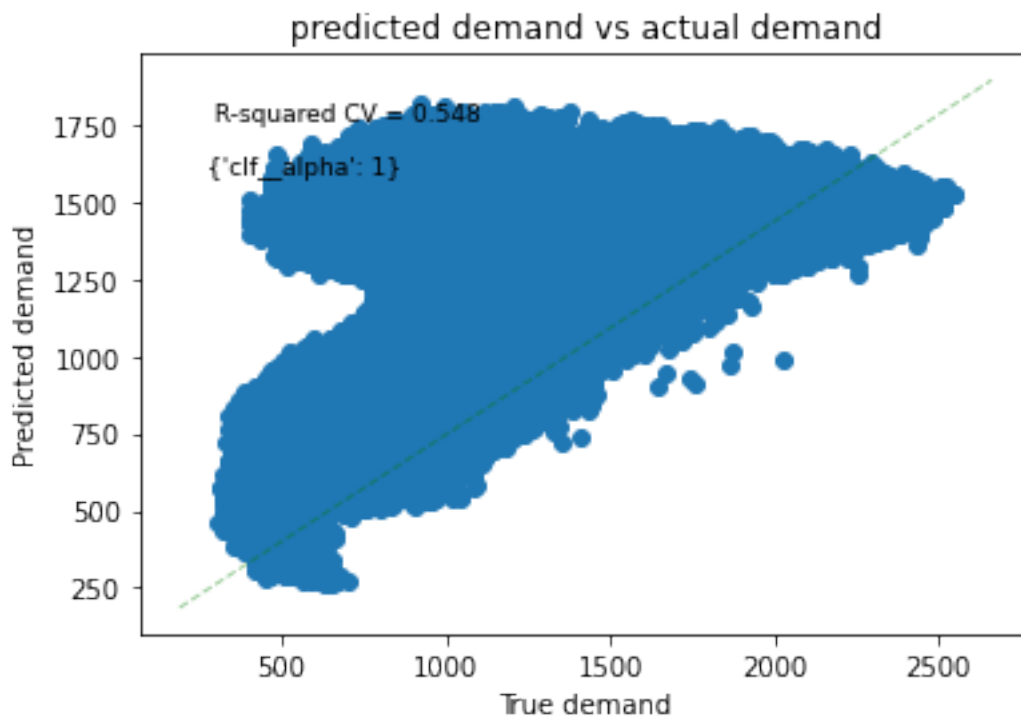
	split1_test_score	split2_test_score	mean_test_score	std_test_score	\
0	0.552356	0.545854	0.548566	0.002762	
1	0.552356	0.545853	0.548566	0.002762	

```

2          0.552358          0.545850          0.548566          0.002763
3          0.552368          0.545818          0.548564          0.002777

rank_test_score
0          3
1          2
2          1
3          4
[ 321.53000135  342.2362621  362.94252284 ... 1489.04783699 1513.33458552
 1534.93596821]
best_index 2
best_score 0.5485663301961063
best_params {'clf__alpha': 1}
##### CV Results
mean_score 0.547899440883825
Model coefficients: [('temperature', -55.77363688870504), ('hours before
sunrise', 126.82937131627627), ('hours before sunset', -433.0690376983775)]
Model does not support feature importances

```



	period	temperature	hours before sunrise	hours before sunset \
48240	48241	11.9	3.833333	20.316667
48241	48242	12.0	3.333333	19.816667
48242	48243	12.1	2.833333	19.316667
48243	48244	12.0	2.333333	18.816667

48244	48245	11.9	1.833333	18.316667
...
52555	52556	12.4	-15.516667	-3.800000
52556	52557	12.3	-16.016667	-4.300000
52557	52558	12.2	-16.516667	-4.800000
52558	52559	11.9	-17.016667	-5.300000
52559	52560	11.9	-17.516667	-5.800000

	demand
48240	321.530001
48241	342.236262
48242	362.942523
48243	385.439027
48244	407.935532
...	...
52555	1444.054828
52556	1466.551332
52557	1489.047837
52558	1513.334586
52559	1534.935968

[4320 rows x 5 columns]

- Polynomial Regression

```
[50]: model(pipe_poly, param_poly, X_train, y_train, X, y)
```

Fitting 3 folds for each of 3 candidates, totalling 9 fits

```
[CV] END ...polynomial__degree=2; total time= 0.0s
[CV] END ...polynomial__degree=2; total time= 0.0s
[CV] END ...polynomial__degree=2; total time= 0.0s
[CV] END ...polynomial__degree=4; total time= 0.0s
[CV] END ...polynomial__degree=4; total time= 0.0s
[CV] END ...polynomial__degree=4; total time= 0.0s
[CV] END ...polynomial__degree=6; total time= 0.1s
[CV] END ...polynomial__degree=6; total time= 0.1s
[CV] END ...polynomial__degree=6; total time= 0.1s
```

Results

	mean_fit_time	std_fit_time	mean_score_time	std_score_time	\
0	0.014971	0.002945	0.004138	0.000539	
1	0.036599	0.001038	0.006828	0.000498	
2	0.110401	0.005728	0.010378	0.001034	

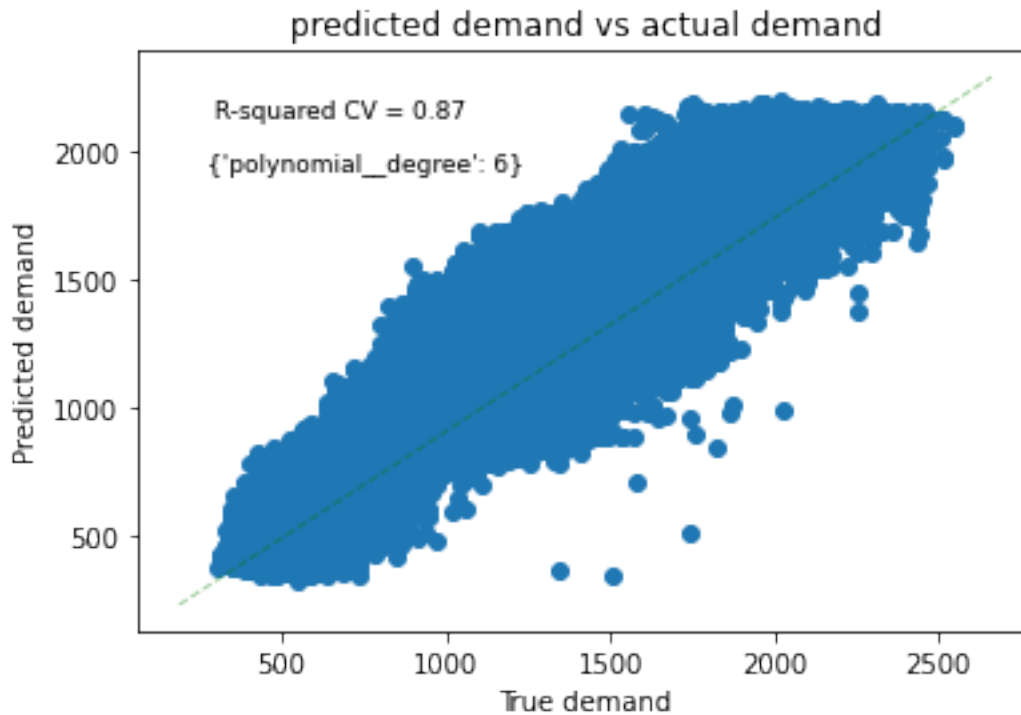
	param_polynomial__degree	params	split0_test_score	\
0	2	{'polynomial__degree': 2}	0.642505	
1	4	{'polynomial__degree': 4}	0.809408	
2	6	{'polynomial__degree': 6}	0.871406	


```

split1_test_score  split2_test_score  mean_test_score  std_test_score  \
0          0.649147          0.644655          0.645436          0.002767
1          0.806470          0.810984          0.808954          0.001870
2          0.868596          0.869231          0.869744          0.001203

rank_test_score
0          3
1          2
2          1
[652.1328581  516.0687481  436.67898674 ... 922.87818015  667.18504386
 356.47292377]
best_index 2
best_score 0.8697440760087667
best_params {'polynomial__degree': 6}
##### CV Results
mean_score 0.8695894465627841
Model coefficients: [('temperature', 8.495889079381798e-08), ('hours before
sunrise', -197.60258201034532), ('hours before sunset', -3.7160343507902347)]
Model does not support feature importances

```



```

period  temperature  hours before sunrise  hours before sunset  \
48240   48241         11.9          3.833333          20.316667
48241   48242         12.0          3.333333          19.816667
48242   48243         12.1          2.833333          19.316667

```

48243	48244	12.0	2.333333	18.816667
48244	48245	11.9	1.833333	18.316667
...
52555	52556	12.4	-15.516667	-3.800000
52556	52557	12.3	-16.016667	-4.300000
52557	52558	12.2	-16.516667	-4.800000
52558	52559	11.9	-17.016667	-5.300000
52559	52560	11.9	-17.516667	-5.800000

	demand
48240	652.132858
48241	516.068748
48242	436.678987
48243	399.641621
48244	395.362234
...	...
52555	1293.980206
52556	1131.264377
52557	922.878180
52558	667.185044
52559	356.472924

[4320 rows x 5 columns]

- Decision Tree Regression

```
[51]: model(pipe_tree, param_tree, X_train, y_train, X, y)
```

```
Fitting 3 folds for each of 12 candidates, totalling 36 fits
[CV] END ...clf__max_depth=2, clf__min_samples_leaf=5; total time= 0.0s
[CV] END ...clf__max_depth=2, clf__min_samples_leaf=5; total time= 0.0s
[CV] END ...clf__max_depth=2, clf__min_samples_leaf=5; total time= 0.0s
[CV] END ...clf__max_depth=2, clf__min_samples_leaf=10; total time= 0.0s
[CV] END ...clf__max_depth=2, clf__min_samples_leaf=10; total time= 0.0s
[CV] END ...clf__max_depth=2, clf__min_samples_leaf=10; total time= 0.0s
[CV] END ...clf__max_depth=2, clf__min_samples_leaf=50; total time= 0.0s
[CV] END ...clf__max_depth=2, clf__min_samples_leaf=50; total time= 0.0s
[CV] END ...clf__max_depth=2, clf__min_samples_leaf=50; total time= 0.0s
[CV] END ...clf__max_depth=2, clf__min_samples_leaf=100; total time= 0.0s
[CV] END ...clf__max_depth=2, clf__min_samples_leaf=100; total time= 0.0s
[CV] END ...clf__max_depth=2, clf__min_samples_leaf=100; total time= 0.0s
[CV] END ...clf__max_depth=5, clf__min_samples_leaf=5; total time= 0.0s
[CV] END ...clf__max_depth=5, clf__min_samples_leaf=5; total time= 0.0s
[CV] END ...clf__max_depth=5, clf__min_samples_leaf=5; total time= 0.0s
[CV] END ...clf__max_depth=5, clf__min_samples_leaf=10; total time= 0.0s
[CV] END ...clf__max_depth=5, clf__min_samples_leaf=10; total time= 0.0s
[CV] END ...clf__max_depth=5, clf__min_samples_leaf=10; total time= 0.0s
[CV] END ...clf__max_depth=5, clf__min_samples_leaf=50; total time= 0.0s
```

```

[CV] END ...clf__max_depth=5, clf__min_samples_leaf=50; total time= 0.0s
[CV] END ...clf__max_depth=5, clf__min_samples_leaf=50; total time= 0.0s
[CV] END ...clf__max_depth=5, clf__min_samples_leaf=100; total time= 0.0s
[CV] END ...clf__max_depth=5, clf__min_samples_leaf=100; total time= 0.0s
[CV] END ...clf__max_depth=5, clf__min_samples_leaf=100; total time= 0.0s
[CV] END ...clf__max_depth=10, clf__min_samples_leaf=5; total time= 0.0s
[CV] END ...clf__max_depth=10, clf__min_samples_leaf=5; total time= 0.0s
[CV] END ...clf__max_depth=10, clf__min_samples_leaf=5; total time= 0.0s
[CV] END ...clf__max_depth=10, clf__min_samples_leaf=10; total time= 0.0s
[CV] END ...clf__max_depth=10, clf__min_samples_leaf=10; total time= 0.0s
[CV] END ...clf__max_depth=10, clf__min_samples_leaf=10; total time= 0.0s
[CV] END ...clf__max_depth=10, clf__min_samples_leaf=50; total time= 0.0s
[CV] END ...clf__max_depth=10, clf__min_samples_leaf=50; total time= 0.0s
[CV] END ...clf__max_depth=10, clf__min_samples_leaf=50; total time= 0.0s
[CV] END ...clf__max_depth=10, clf__min_samples_leaf=100; total time= 0.0s
[CV] END ...clf__max_depth=10, clf__min_samples_leaf=100; total time= 0.0s
[CV] END ...clf__max_depth=10, clf__min_samples_leaf=100; total time= 0.0s

```

Results

	mean_fit_time	std_fit_time	mean_score_time	std_score_time	\
0	0.011134	0.000591	0.001933	0.000194	
1	0.010282	0.000426	0.001859	0.000120	
2	0.010123	0.000309	0.001702	0.000056	
3	0.010011	0.000217	0.001866	0.000076	
4	0.020652	0.000517	0.002883	0.000683	
5	0.020783	0.000362	0.002559	0.000708	
6	0.020420	0.000293	0.002072	0.000078	
7	0.020775	0.000612	0.002293	0.000237	
8	0.037138	0.000265	0.002567	0.000155	
9	0.038484	0.000864	0.002583	0.000017	
10	0.034270	0.001573	0.002573	0.000113	
11	0.031923	0.001029	0.002432	0.000038	

	param_clf__max_depth	param_clf__min_samples_leaf	\
0	2	5	
1	2	10	
2	2	50	
3	2	100	
4	5	5	
5	5	10	
6	5	50	
7	5	100	
8	10	5	
9	10	10	
10	10	50	
11	10	100	

	params	split0_test_score	\
0	{'clf__max_depth': 2, 'clf__min_samples_leaf': 5}	0.614903	

```

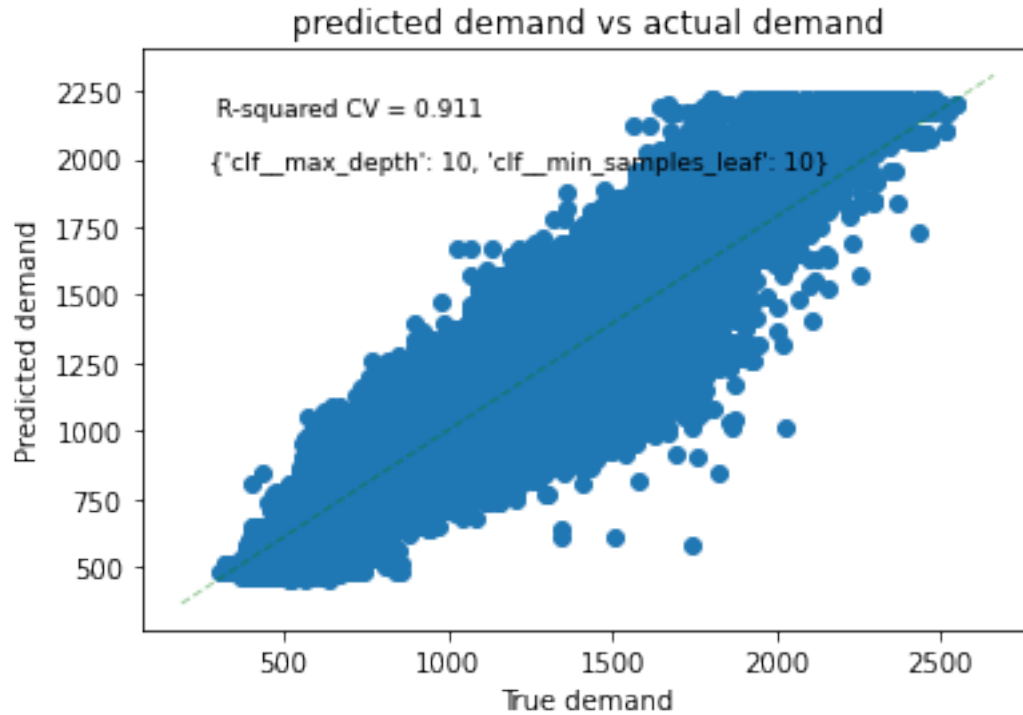
1  {'clf__max_depth': 2, 'clf__min_samples_leaf':...      0.614903
2  {'clf__max_depth': 2, 'clf__min_samples_leaf':...      0.614903
3  {'clf__max_depth': 2, 'clf__min_samples_leaf':...      0.614903
4  {'clf__max_depth': 5, 'clf__min_samples_leaf': 5}      0.865191
5  {'clf__max_depth': 5, 'clf__min_samples_leaf':...      0.865191
6  {'clf__max_depth': 5, 'clf__min_samples_leaf':...      0.865008
7  {'clf__max_depth': 5, 'clf__min_samples_leaf':...      0.864936
8  {'clf__max_depth': 10, 'clf__min_samples_leaf'...      0.907996
9  {'clf__max_depth': 10, 'clf__min_samples_leaf'...      0.909336
10 {'clf__max_depth': 10, 'clf__min_samples_leaf'...      0.906331
11 {'clf__max_depth': 10, 'clf__min_samples_leaf'...      0.897609

      split1_test_score  split2_test_score  mean_test_score  std_test_score  \
0          0.597650          0.599382          0.603978          0.007757
1          0.597650          0.599382          0.603978          0.007757
2          0.597650          0.599382          0.603978          0.007757
3          0.597650          0.599382          0.603978          0.007757
4          0.862205          0.862235          0.863210          0.001401
5          0.862205          0.862235          0.863210          0.001401
6          0.861967          0.862067          0.863014          0.001411
7          0.861967          0.861958          0.862954          0.001402
8          0.906542          0.907004          0.907181          0.000606
9          0.907549          0.907204          0.908030          0.000935
10         0.905102          0.904737          0.905390          0.000682
11         0.896899          0.897394          0.897301          0.000297

      rank_test_score
0          9
1          9
2          9
3          9
4          5
5          5
6          7
7          8
8          2
9          1
10         3
11         4
[ 519.18090452  519.18090452  482.85922684 ... 1016.53846154  706.15789474
  588.6          ]
best_index 9
best_score 0.9080296736083269
best_params {'clf__max_depth': 10, 'clf__min_samples_leaf': 10}
##### CV Results
mean_score 0.9110851130156965
Model does not support model coefficients
Feature importances: [('temperature', 0.049639106318395375), ('hours before

```

```
sunrise', 0.2254406166609969), ('hours before sunset', 0.7249202770206077)]
```



	period	temperature	hours before sunrise	hours before sunset	\
48240	48241	11.9	3.833333	20.316667	
48241	48242	12.0	3.333333	19.816667	
48242	48243	12.1	2.833333	19.316667	
48243	48244	12.0	2.333333	18.816667	
48244	48245	11.9	1.833333	18.316667	
...	
52555	52556	12.4	-15.516667	-3.800000	
52556	52557	12.3	-16.016667	-4.300000	
52557	52558	12.2	-16.516667	-4.800000	
52558	52559	11.9	-17.016667	-5.300000	
52559	52560	11.9	-17.516667	-5.800000	

	demand
48240	519.180905
48241	519.180905
48242	482.859227
48243	482.859227
48244	482.859227
...	...
52555	1211.777778
52556	946.500000

```
52557 1016.538462
52558 706.157895
52559 588.600000
```

```
[4320 rows x 5 columns]
```

- Random Forest

```
[52]: model(pipe_forest, param_forest, X_train, y_train, X, y)
```

```
Fitting 3 folds for each of 27 candidates, totalling 81 fits
```

```
[CV] END clf__max_depth=1, clf__max_features=None, clf__n_estimators=10; total
time= 0.0s
[CV] END clf__max_depth=1, clf__max_features=None, clf__n_estimators=10; total
time= 0.0s
[CV] END clf__max_depth=1, clf__max_features=None, clf__n_estimators=10; total
time= 0.0s
[CV] END clf__max_depth=1, clf__max_features=None, clf__n_estimators=20; total
time= 0.1s
[CV] END clf__max_depth=1, clf__max_features=None, clf__n_estimators=20; total
time= 0.1s
[CV] END clf__max_depth=1, clf__max_features=None, clf__n_estimators=20; total
time= 0.1s
[CV] END clf__max_depth=1, clf__max_features=None, clf__n_estimators=50; total
time= 0.2s
[CV] END clf__max_depth=1, clf__max_features=None, clf__n_estimators=50; total
time= 0.2s
[CV] END clf__max_depth=1, clf__max_features=None, clf__n_estimators=50; total
time= 0.2s
[CV] END clf__max_depth=1, clf__max_features=1, clf__n_estimators=10; total
time= 0.0s
[CV] END clf__max_depth=1, clf__max_features=1, clf__n_estimators=10; total
time= 0.0s
[CV] END clf__max_depth=1, clf__max_features=1, clf__n_estimators=10; total
time= 0.0s
[CV] END clf__max_depth=1, clf__max_features=1, clf__n_estimators=20; total
time= 0.1s
[CV] END clf__max_depth=1, clf__max_features=1, clf__n_estimators=20; total
time= 0.1s
[CV] END clf__max_depth=1, clf__max_features=1, clf__n_estimators=20; total
time= 0.1s
[CV] END clf__max_depth=1, clf__max_features=1, clf__n_estimators=50; total
time= 0.1s
[CV] END clf__max_depth=1, clf__max_features=1, clf__n_estimators=50; total
time= 0.1s
[CV] END clf__max_depth=1, clf__max_features=1, clf__n_estimators=50; total
time= 0.1s
[CV] END clf__max_depth=1, clf__max_features=2, clf__n_estimators=10; total
```

[illegible]


```

time= 0.1s
[CV] END clf__max_depth=5, clf__max_features=1, clf__n_estimators=20; total
time= 0.1s
[CV] END clf__max_depth=5, clf__max_features=1, clf__n_estimators=20; total
time= 0.1s
[CV] END clf__max_depth=5, clf__max_features=1, clf__n_estimators=50; total
time= 0.4s
[CV] END clf__max_depth=5, clf__max_features=1, clf__n_estimators=50; total
time= 0.4s
[CV] END clf__max_depth=5, clf__max_features=1, clf__n_estimators=50; total
time= 0.4s
[CV] END clf__max_depth=5, clf__max_features=2, clf__n_estimators=10; total
time= 0.1s
[CV] END clf__max_depth=5, clf__max_features=2, clf__n_estimators=10; total
time= 0.1s
[CV] END clf__max_depth=5, clf__max_features=2, clf__n_estimators=10; total
time= 0.1s
[CV] END clf__max_depth=5, clf__max_features=2, clf__n_estimators=20; total
time= 0.2s
[CV] END clf__max_depth=5, clf__max_features=2, clf__n_estimators=20; total
time= 0.2s
[CV] END clf__max_depth=5, clf__max_features=2, clf__n_estimators=20; total
time= 0.2s
[CV] END clf__max_depth=5, clf__max_features=2, clf__n_estimators=50; total
time= 0.6s
[CV] END clf__max_depth=5, clf__max_features=2, clf__n_estimators=50; total
time= 0.6s
[CV] END clf__max_depth=5, clf__max_features=2, clf__n_estimators=50; total
time= 0.6s

```

Results

	mean_fit_time	std_fit_time	mean_score_time	std_score_time	\
0	0.043469	0.001117	0.004038	0.000080	
1	0.081017	0.001199	0.006647	0.000657	
2	0.207130	0.014803	0.014670	0.002282	
3	0.032488	0.000681	0.004917	0.000344	
4	0.055928	0.001697	0.007117	0.001071	
5	0.130357	0.000924	0.013099	0.000121	
6	0.037700	0.000903	0.004355	0.000388	
7	0.071457	0.002586	0.006725	0.000156	
8	0.178148	0.004057	0.014855	0.002646	
9	0.068519	0.000558	0.005609	0.000515	
10	0.139546	0.000575	0.008218	0.000451	
11	0.336381	0.001428	0.016573	0.000542	
12	0.039052	0.001773	0.004993	0.000096	
13	0.076120	0.000994	0.007939	0.000334	
14	0.188770	0.004812	0.017353	0.000791	
15	0.058623	0.002477	0.005386	0.000376	
16	0.109966	0.003355	0.009021	0.000288	

17	0.256334	0.005952	0.016754	0.000893
18	0.138388	0.002347	0.006782	0.000059
19	0.274116	0.003119	0.011248	0.000087
20	0.667539	0.002830	0.025380	0.000221
21	0.070038	0.001236	0.007524	0.000711
22	0.136199	0.002251	0.011544	0.000192
23	0.334877	0.003614	0.025403	0.000575
24	0.104141	0.001957	0.007346	0.000586
25	0.215473	0.004304	0.013363	0.001382
26	0.539003	0.002780	0.026809	0.000914

	param_clf__max_depth	param_clf__max_features	param_clf__n_estimators	\
0	1	None	10	
1	1	None	20	
2	1	None	50	
3	1	1	10	
4	1	1	20	
5	1	1	50	
6	1	2	10	
7	1	2	20	
8	1	2	50	
9	2	None	10	
10	2	None	20	
11	2	None	50	
12	2	1	10	
13	2	1	20	
14	2	1	50	
15	2	2	10	
16	2	2	20	
17	2	2	50	
18	5	None	10	
19	5	None	20	
20	5	None	50	
21	5	1	10	
22	5	1	20	
23	5	1	50	
24	5	2	10	
25	5	2	20	
26	5	2	50	

	params	split0_test_score	\
0	{'clf__max_depth': 1, 'clf__max_features': Non...	0.409458	
1	{'clf__max_depth': 1, 'clf__max_features': Non...	0.427587	
2	{'clf__max_depth': 1, 'clf__max_features': Non...	0.413499	
3	{'clf__max_depth': 1, 'clf__max_features': 1, ...	0.424206	
4	{'clf__max_depth': 1, 'clf__max_features': 1, ...	0.321034	
5	{'clf__max_depth': 1, 'clf__max_features': 1, ...	0.396814	
6	{'clf__max_depth': 1, 'clf__max_features': 2, ...	0.461713	

7	{'clf__max_depth': 1, 'clf__max_features': 2, ...	0.457805
8	{'clf__max_depth': 1, 'clf__max_features': 2, ...	0.459235
9	{'clf__max_depth': 2, 'clf__max_features': Non...	0.622426
10	{'clf__max_depth': 2, 'clf__max_features': Non...	0.630156
11	{'clf__max_depth': 2, 'clf__max_features': Non...	0.629646
12	{'clf__max_depth': 2, 'clf__max_features': 1, ...	0.472733
13	{'clf__max_depth': 2, 'clf__max_features': 1, ...	0.580846
14	{'clf__max_depth': 2, 'clf__max_features': 1, ...	0.586879
15	{'clf__max_depth': 2, 'clf__max_features': 2, ...	0.658512
16	{'clf__max_depth': 2, 'clf__max_features': 2, ...	0.661222
17	{'clf__max_depth': 2, 'clf__max_features': 2, ...	0.661863
18	{'clf__max_depth': 5, 'clf__max_features': Non...	0.879036
19	{'clf__max_depth': 5, 'clf__max_features': Non...	0.879085
20	{'clf__max_depth': 5, 'clf__max_features': Non...	0.878044
21	{'clf__max_depth': 5, 'clf__max_features': 1, ...	0.798301
22	{'clf__max_depth': 5, 'clf__max_features': 1, ...	0.847905
23	{'clf__max_depth': 5, 'clf__max_features': 1, ...	0.830141
24	{'clf__max_depth': 5, 'clf__max_features': 2, ...	0.879659
25	{'clf__max_depth': 5, 'clf__max_features': 2, ...	0.881775
26	{'clf__max_depth': 5, 'clf__max_features': 2, ...	0.883930

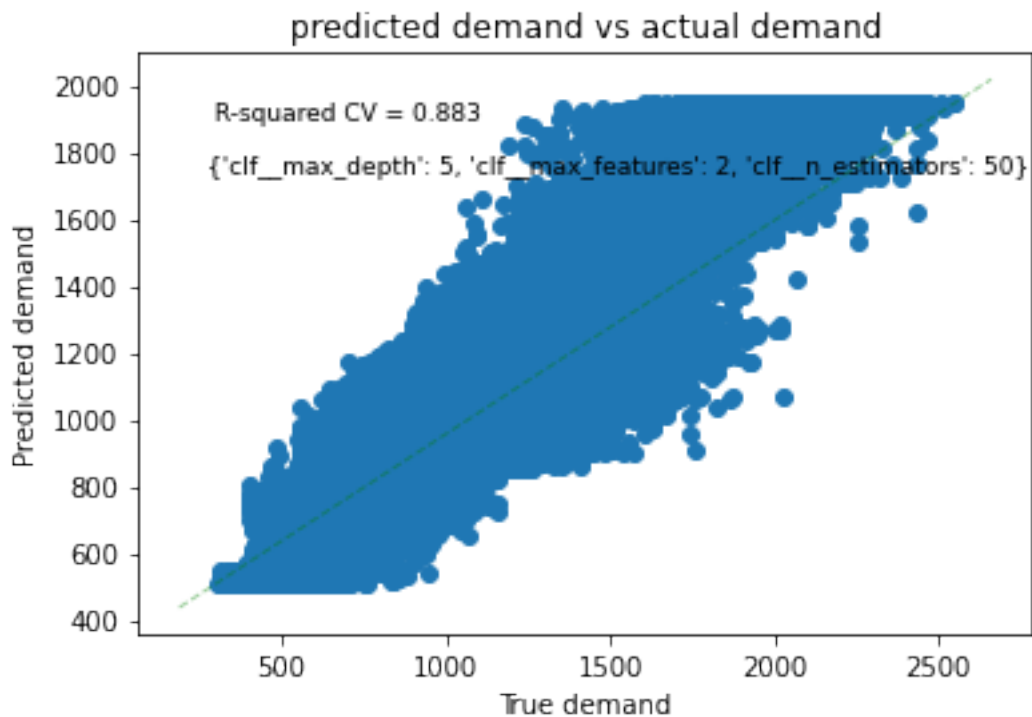
	split1_test_score	split2_test_score	mean_test_score	std_test_score \
0	0.405206	0.422511	0.412392	0.007363
1	0.409134	0.417465	0.418062	0.007545
2	0.410072	0.418733	0.414101	0.003561
3	0.394326	0.335597	0.384710	0.036808
4	0.402534	0.370365	0.364645	0.033517
5	0.372575	0.414443	0.394611	0.017163
6	0.465179	0.471537	0.466143	0.004068
7	0.456934	0.470286	0.461675	0.006099
8	0.457877	0.467334	0.461482	0.004175
9	0.613380	0.624200	0.620002	0.004738
10	0.610882	0.627085	0.622708	0.008456
11	0.615067	0.632286	0.625666	0.007572
12	0.625005	0.511639	0.536459	0.064595
13	0.607795	0.592538	0.593726	0.011034
14	0.597784	0.597701	0.594121	0.005121
15	0.644056	0.654388	0.652319	0.006080
16	0.648701	0.661102	0.657008	0.005875
17	0.652278	0.669181	0.661107	0.006921
18	0.872339	0.876674	0.876016	0.002773
19	0.873873	0.878721	0.877226	0.002376
20	0.875601	0.878667	0.877438	0.001323
21	0.803076	0.807701	0.803026	0.003837
22	0.821741	0.811561	0.827069	0.015308
23	0.827082	0.828521	0.828582	0.001249
24	0.882614	0.874783	0.879019	0.003229
25	0.875167	0.885630	0.880857	0.004321

26	0.881778	0.883998	0.883235	0.001031
----	----------	----------	----------	----------

```

rank_test_score
0      24
1      22
2      23
3      26
4      27
5      25
6      19
7      20
8      21
9      15
10     14
11     13
12     18
13     17
14     16
15     12
16     11
17     10
18      6
19      5
20      4
21      9
22      8
23      7
24      3
25      2
26      1
[ 508.04086737  508.04086737  508.04086737 ... 1051.37809806  867.34975045
 752.3030571 ]
best_index 26
best_score 0.8832349910787602
best_params {'clf__max_depth': 5, 'clf__max_features': 2, 'clf__n_estimators':
50}
##### CV Results
mean_score 0.8826708441241793
Model does not support model coefficients
Feature importances: [('temperature', 0.07674538201360276), ('hours before
sunrise', 0.34922074801091474), ('hours before sunset', 0.5740338699754826)]

```



	period	temperature	hours before sunrise	hours before sunset	\
48240	48241	11.9	3.833333	20.316667	
48241	48242	12.0	3.333333	19.816667	
48242	48243	12.1	2.833333	19.316667	
48243	48244	12.0	2.333333	18.816667	
48244	48245	11.9	1.833333	18.316667	
...	
52555	52556	12.4	-15.516667	-3.800000	
52556	52557	12.3	-16.016667	-4.300000	
52557	52558	12.2	-16.516667	-4.800000	
52558	52559	11.9	-17.016667	-5.300000	
52559	52560	11.9	-17.516667	-5.800000	
demand					
48240	508.040867				
48241	508.040867				
48242	508.040867				
48243	508.040867				
48244	508.040867				
...	...				
52555	1188.241934				
52556	1172.252128				
52557	1051.378098				
52558	867.349750				

52559 752.303057

[4320 rows x 5 columns]

- Multi-layer Perceptron (MLP) Regression

```
[53]: model(pipe_neural, param_neural, X_train, y_train, X, y)
```

Fitting 3 folds for each of 72 candidates, totalling 216 fits

```
[CV] END clf__activation=relu, clf__alpha=0.001, clf__hidden_layer_sizes=5,
clf__learning_rate=constant, clf__solver=lbfgs; total time= 0.5s
[CV] END clf__activation=relu, clf__alpha=0.001, clf__hidden_layer_sizes=5,
clf__learning_rate=constant, clf__solver=lbfgs; total time= 0.5s
[CV] END clf__activation=relu, clf__alpha=0.001, clf__hidden_layer_sizes=5,
clf__learning_rate=constant, clf__solver=lbfgs; total time= 0.6s
[CV] END clf__activation=relu, clf__alpha=0.001, clf__hidden_layer_sizes=5,
clf__learning_rate=invscaling, clf__solver=lbfgs; total time= 0.2s
[CV] END clf__activation=relu, clf__alpha=0.001, clf__hidden_layer_sizes=5,
clf__learning_rate=invscaling, clf__solver=lbfgs; total time= 0.6s
[CV] END clf__activation=relu, clf__alpha=0.001, clf__hidden_layer_sizes=5,
clf__learning_rate=invscaling, clf__solver=lbfgs; total time= 0.6s
[CV] END clf__activation=relu, clf__alpha=0.001, clf__hidden_layer_sizes=(10,
10), clf__learning_rate=constant, clf__solver=lbfgs; total time= 1.6s
[CV] END clf__activation=relu, clf__alpha=0.001, clf__hidden_layer_sizes=(10,
10), clf__learning_rate=constant, clf__solver=lbfgs; total time= 1.7s
[CV] END clf__activation=relu, clf__alpha=0.001, clf__hidden_layer_sizes=(10,
10), clf__learning_rate=constant, clf__solver=lbfgs; total time= 1.8s
[CV] END clf__activation=relu, clf__alpha=0.001, clf__hidden_layer_sizes=(10,
10), clf__learning_rate=invscaling, clf__solver=lbfgs; total time= 1.7s
[CV] END clf__activation=relu, clf__alpha=0.001, clf__hidden_layer_sizes=(10,
10), clf__learning_rate=invscaling, clf__solver=lbfgs; total time= 1.7s
[CV] END clf__activation=relu, clf__alpha=0.001, clf__hidden_layer_sizes=(10,
10), clf__learning_rate=invscaling, clf__solver=lbfgs; total time= 1.8s
[CV] END clf__activation=relu, clf__alpha=0.001, clf__hidden_layer_sizes=(7, 7,
7), clf__learning_rate=constant, clf__solver=lbfgs; total time= 1.9s
[CV] END clf__activation=relu, clf__alpha=0.001, clf__hidden_layer_sizes=(7, 7,
7), clf__learning_rate=constant, clf__solver=lbfgs; total time= 2.0s
[CV] END clf__activation=relu, clf__alpha=0.001, clf__hidden_layer_sizes=(7, 7,
7), clf__learning_rate=constant, clf__solver=lbfgs; total time= 2.2s
[CV] END clf__activation=relu, clf__alpha=0.001, clf__hidden_layer_sizes=(7, 7,
7), clf__learning_rate=invscaling, clf__solver=lbfgs; total time= 2.2s
[CV] END clf__activation=relu, clf__alpha=0.001, clf__hidden_layer_sizes=(7, 7,
7), clf__learning_rate=invscaling, clf__solver=lbfgs; total time= 2.2s
[CV] END clf__activation=relu, clf__alpha=0.001, clf__hidden_layer_sizes=(7, 7,
7), clf__learning_rate=invscaling, clf__solver=lbfgs; total time= 1.9s
[CV] END clf__activation=relu, clf__alpha=0.01, clf__hidden_layer_sizes=5,
clf__learning_rate=constant, clf__solver=lbfgs; total time= 0.5s
[CV] END clf__activation=relu, clf__alpha=0.01, clf__hidden_layer_sizes=5,
```

[illegible]

[illegible]

[illegible]

[illegible]

```

7), clf__learning_rate=constant, clf__solver=lbfgs; total time= 3.8s
[CV] END clf__activation=tanh, clf__alpha=100, clf__hidden_layer_sizes=(7, 7,
7), clf__learning_rate=constant, clf__solver=lbfgs; total time= 4.0s
[CV] END clf__activation=tanh, clf__alpha=100, clf__hidden_layer_sizes=(7, 7,
7), clf__learning_rate=invscaling, clf__solver=lbfgs; total time= 4.3s
[CV] END clf__activation=tanh, clf__alpha=100, clf__hidden_layer_sizes=(7, 7,
7), clf__learning_rate=invscaling, clf__solver=lbfgs; total time= 5.3s
[CV] END clf__activation=tanh, clf__alpha=100, clf__hidden_layer_sizes=(7, 7,
7), clf__learning_rate=invscaling, clf__solver=lbfgs; total time= 5.0s

```

Results

	mean_fit_time	std_fit_time	mean_score_time	std_score_time	\
0	0.554897	0.020260	0.003289	0.000616	
1	0.454644	0.173347	0.002676	0.000400	
2	1.694442	0.104064	0.003267	0.000093	
3	1.759638	0.060039	0.003641	0.000214	
4	2.033563	0.090458	0.003836	0.000625	
..	
67	0.742136	0.210942	0.003533	0.000382	
68	3.210097	0.282265	0.005056	0.000679	
69	3.032816	0.291888	0.004489	0.000313	
70	2.822582	1.543837	0.005725	0.000546	
71	4.852538	0.392508	0.006293	0.000469	

	param_clf__activation	param_clf__alpha	param_clf__hidden_layer_sizes	\
0	relu	0.001	5	
1	relu	0.001	5	
2	relu	0.001	(10, 10)	
3	relu	0.001	(10, 10)	
4	relu	0.001	(7, 7, 7)	
..	
67	tanh	100	5	
68	tanh	100	(10, 10)	
69	tanh	100	(10, 10)	
70	tanh	100	(7, 7, 7)	
71	tanh	100	(7, 7, 7)	

	param_clf__learning_rate	param_clf__solver	\
0	constant	lbfgs	
1	invscaling	lbfgs	
2	constant	lbfgs	
3	invscaling	lbfgs	
4	constant	lbfgs	
..	
67	invscaling	lbfgs	
68	constant	lbfgs	
69	invscaling	lbfgs	
70	constant	lbfgs	
71	invscaling	lbfgs	

	params	split0_test_score \
0	{'clf__activation': 'relu', 'clf__alpha': 0.00...	0.831908
1	{'clf__activation': 'relu', 'clf__alpha': 0.00...	0.574359
2	{'clf__activation': 'relu', 'clf__alpha': 0.00...	0.894616
3	{'clf__activation': 'relu', 'clf__alpha': 0.00...	0.894166
4	{'clf__activation': 'relu', 'clf__alpha': 0.00...	0.895319
..
67	{'clf__activation': 'tanh', 'clf__alpha': 100,...	0.641083
68	{'clf__activation': 'tanh', 'clf__alpha': 100,...	0.632171
69	{'clf__activation': 'tanh', 'clf__alpha': 100,...	0.580067
70	{'clf__activation': 'tanh', 'clf__alpha': 100,...	-0.000020
71	{'clf__activation': 'tanh', 'clf__alpha': 100,...	0.006785

	split1_test_score	split2_test_score	mean_test_score	std_test_score \
0	0.818977	0.850786	0.833890	0.013061
1	0.774649	0.831894	0.726968	0.110412
2	0.897433	0.892455	0.894834	0.002038
3	0.895533	0.902007	0.897235	0.003420
4	0.899472	0.898038	0.897610	0.001722
..
67	0.645847	0.564585	0.617172	0.037235
68	0.724033	0.752965	0.703056	0.051496
69	0.589566	0.507371	0.559002	0.036714
70	0.552069	0.473564	0.341871	0.243869
71	0.452576	0.391018	0.283460	0.197246

	rank_test_score
0	20
1	31
2	5
3	3
4	1
..	...
67	50
68	32
69	55
70	62
71	67

[72 rows x 16 columns]

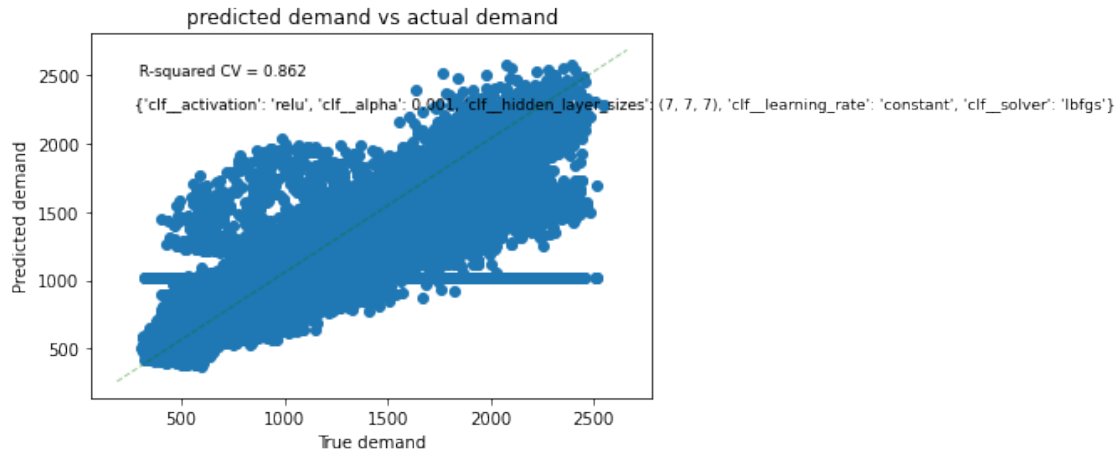
[537.0052627 521.63028325 506.25530381 ... 938.2405219 810.57541918
676.93547165]

best_index 4

best_score 0.8976099133817449

best_params {'clf__activation': 'relu', 'clf__alpha': 0.001,
'clf__hidden_layer_sizes': (7, 7, 7), 'clf__learning_rate': 'constant',
'clf__solver': 'lbfgs'}


```
##### CV Results
mean_score 0.8620766955405965
Model does not support model coefficients
Model does not support feature importances
```



	period	temperature	hours before sunrise	hours before sunset \
48240	48241	11.9	3.833333	20.316667
48241	48242	12.0	3.333333	19.816667
48242	48243	12.1	2.833333	19.316667
48243	48244	12.0	2.333333	18.816667
48244	48245	11.9	1.833333	18.316667
...
52555	52556	12.4	-15.516667	-3.800000
52556	52557	12.3	-16.016667	-4.300000
52557	52558	12.2	-16.516667	-4.800000
52558	52559	11.9	-17.016667	-5.300000
52559	52560	11.9	-17.516667	-5.800000

	demand
48240	537.005263
48241	521.630283
48242	506.255304
48243	491.621076
48244	476.986848
...	...
52555	1201.537187
52556	1069.888854
52557	938.240522
52558	810.575419
52559	676.935472

[4320 rows x 5 columns]

0.4 Conclusion

KNN fits the best

1. KNN

* Parameters: clf__n_neighbors: 25 * Score: 0.918

2. Decision Tree Regression

* Parameters: clf__max_depth: 10, clf__min_samples_leaf: 10 * Score: 0.911

3. Polynomial Regression

* Parameters: polynomial__degree: 6 * Score: 0.87

4. Random Forest

* Parameters: clf__max_depth: 5, clf__max_features: 2, clf__n_estimators: 50 * Score: 0.883

5. Linear Regression

* Parameters: non * Score: 0.548

6. Lasso

* Parameters: clf__alpha: 0.01 * Score: 0.548

7. Ridge

* Parameters: clf__alpha: 1 * Score: 0.548

8. XGBoost

* Parameters: clf_colsample_bytree: 1, clf_gamma: 0.01, clf_max_depth: 5, clf_min_child_weight: 6, clf_subsample: 0.7 * Score: 0.918

9. Multi-layer Perceptron (MLP) Regression

* Parameters: 'clf__activation': 'relu', 'clf__alpha': 0.001, 'clf__hidden_layer_sizes': (7, 7, 7), 'clf__learning_rate': 'constant', 'clf__solver': 'lbfgs' * Score: 0.862