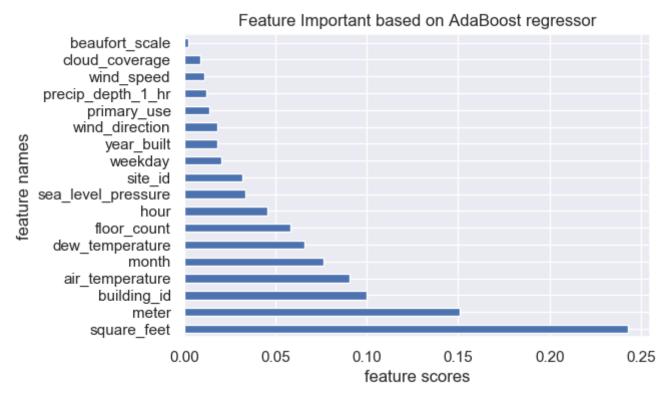
```
import pandas as pd
         import matplotlib
         import matplotlib.pyplot as plt
         import seaborn as sns
         import warnings
         import gc
         import datetime
         from sklearn.ensemble import AdaBoostRegressor
         from sklearn.model_selection import GridSearchCV
         from sklearn.model_selection import train_test_split
         from sklearn.model_selection import learning_curve
         from sklearn.metrics import SCORERS
         warnings.simplefilter('ignore')
         matplotlib.rcParams['figure.dpi'] = 100
         sns.set()
In [15]: # Extract a minibatch from the original training dataset
         def TrTeSplit(X_tr, y_tr, size):
             X_trspl, X_tespl, y_trspl, y_tespl = train_test_split(X_tr, y_tr, test_size=size, random_state=0)
             X_trspl = X_trspl.reset_index(drop = True)
             X_tespl = X_tespl.reset_index(drop = True)
             y_trspl = y_trspl.reset_index(drop = True)
             y_tespl = y_tespl.reset_index(drop = True)
             return X_trspl, X_tespl, y_trspl, y_tespl
         # Parameter Tunning using cross validation
         def model_param_select(X, y, nfolds):
             n_{estimators} = [45, 50, 55]
             learning_rate = [0.5, 1, 1.5]
             #the range is being reduced since after several runs
             #I have removed some redundant range value that will yield longer run time
             param_grid = {'n_estimators': n_estimators, 'learning_rate': learning_rate}
             grid_search = GridSearchCV(AdaBoostRegressor(base_estimator = DecisionTreeRegressor(max_depth=20)), param_grid, cv=nfolds)
             grid_search.fit(X, y)
             grid_search.best_params_
             return grid_search.best_params_
         def RMSE(y_test, y_pred):
             rss=((y_test-y_pred)**2).sum()
             mse=np.mean((y_test-y_pred)**2)
             rmse = np.sqrt(np.mean((y_test-y_pred)**2))
             return rmse
         def DataPre(new_train):
             y_tr = new_train['meter_reading']
             X_tr = new_train.drop(columns = ['meter_reading', 'timestamp', 'Unnamed: 0'])
             X_trspl, X_tespl, y_trspl, y_tespl = TrTeSplit(X_tr, y_tr, 0.33)
             del X_tr, y_tr
             return X_trspl, X_tespl, y_trspl, y_tespl
         def LearningVisual(model, X_trspl, y_trspl, X_tespl, y_tespl, percentSize):
             # Visualize learning curves
             plt.figure()
             train_sizes, train_scores, test_scores = \
             learning_curve(model, X_trspl, y_trspl, train_sizes=np.linspace(0.1, 1, percentSize),
                            scoring="neg_root_mean_squared_error", cv=10)
             train_sizes, val_train_scores, val_test_scores = \
             learning_curve(model, X_tespl, y_tespl, train_sizes=np.linspace(0.1, 1, percentSize),
                            scoring="neg_root_mean_squared_error", cv=10)
             plt.plot(train_sizes, -test_scores.mean(1), 'o-', color="r", label="train set")
             plt.plot(train_sizes, -val_test_scores.mean(1), 'o-', color ='b', label="validation set")
             plt.xlabel("Dataset size")
             plt.ylabel("Root Mean Squared Error")
             plt.title('Learning curves')
             plt.legend(loc="best")
             plt.show()
In [16]: # Load Data
         new_train = pd.read_csv('sample train tiny.csv')
         X trspl, X tespl, y trspl, y tespl = DataPre(new_train)
         del new_train
In [17]: #Implement Adaboost model with the choosen parameter
         model = AdaBoostRegressor(base_estimator = DecisionTreeRegressor(max_depth=20), n_estimators = 50, learning_rate = 1.0)
         model.fit(X_trspl, y_trspl)
         y_pred = model.predict(X_tespl)
         rmse = RMSE(y_tespl, y_pred)
         print(rmse)
         feat_importances = pd.Series(model.feature_importances_, index=X_trspl.columns)
         feat_importances.nlargest(20).plot(kind='barh')
         plt.title('Feature Important based on AdaBoost regressor')
         plt.xlabel('feature scores')
         plt.ylabel('feature names')
         plt.show()
         212.13184827079283
```



In [11]: import numpy as np

```
Learning curves

train set
validation set

275
250
2000 40000 60000 80000 100000 120000 140000 160000
Dataset size
```

```
In [ ]: X_train = pd.read_csv(r'Dataset/new_train.csv')
In [ ]: y_train = X_train['meter_reading']
        X_train = X_train.drop(columns = ['meter_reading', 'timestamp'])
        #X_test = X_test.drop(columns = ['meter_reading', 'timestamp'])
In [8]: X_trspl, X_tespl, y_trspl, y_tespl = TrTeSplit(X_train, y_train, 0.33)
In [ ]: #Implement Tree model with the choosen parameter
        \#numbers_sizes = (i*10**exp for exp in range(2, 9) for i in range(1, 10))
        #for i in numbers_sizes:
        model = AdaBoostRegressor(base_estimator = DecisionTreeRegressor(max_depth=20), n_estimators = 50, learning_rate = 1.0)
        model.fit(X_trspl, y_trspl)
        y_pred = model.predict(X_tespl)
        rmse = RMSE(y_tespl, y_pred)
        print(rmse)
        feat_importances = pd.Series(model.feature_importances_, index=X_trspl.columns)
        feat_importances.nlargest(20).plot(kind='barh')
        plt.title('Feature Important based on Decision Tree regressor')
        plt.xlabel('feature scores')
        plt.ylabel('feature names')
        plt.show()
```

In []: