```
In [1]: #fixing OPEN BLAS max pthreads reached issue
        import os
        os.environ['OPENBLAS NUM THREADS'] = '1'
        os.environ['GOTO NUM THREADS'] = '1'
        os.environ['OMP NUM THREADS'] = '1'
        #imports
        import pandas as pd
        import numpy as np
        import autosklearn.regression as autocl
        from sklearn.model selection import train test split
        import sklearn.metrics
        import sklearn.datasets
        import matplotlib.pyplot as plt
        from scipy import stats
        import xgboost as xgb
        from autosklearn.metrics import balanced_accuracy, precision, recall, f1
        from sklearn import (manifold, datasets, preprocessing, model_selection, decomposit
        from sklearn.preprocessing import StandardScaler
        import sklearn.preprocessing
        CONSTRAINT = 0.3
In [2]: # select user input
        # user_input = int(input("Which user's result do you want to look at?: "))
        user input = 0
        # read the dataframe
        data = pd.read_csv('data.csv')
        df = data[data['user number'] == user input]
        # drop all Null data (filtering null values)
        df.dropna()
        # gather the data
        X = df.iloc[:,:-2].values
        y = df["symptom_value"].values
        n_samples = X.shape[0]
        n_train = int(n_samples * 0.8)
        indices = np.arange(n samples)
        np.random.shuffle(indices)
In [3]: def preprocessing(X):
            from sklearn.preprocessing import StandardScaler
            import sklearn.preprocessing
            std scaler = StandardScaler()
            X = std_scaler.fit_transform(X)
            X = sklearn.preprocessing.normalize(X, norm='12')
            return X
        X = preprocessing(X)
```

```
In [4]: # split the dataset into training and test data
        train indices = indices[:n train]
        test_indices = indices[n_train:]
        X train = X[train indices]
        y_train = y[train_indices]
        X_test = X[test_indices]
        y_test = y[test_indices]
In [5]: # train the module
        automl = autocl.AutoSklearnRegressor(
            time left for this task = 120,
            per_run_time_limit = 30,
            disable evaluator output = False,
            resampling_strategy = "cv",
            resampling_strategy_arguments = {"folds": 10},
In [6]: automl.fit(X_train, y_train)
        print("Before re-fit")
        predictions = automl.predict(X_test)
        print("Test R2 score:", sklearn.metrics.r2_score(y_test, predictions))
        print("#" * 80)
        print("After re-fit")
        automl.refit(X_train.copy(), y_train.copy())
        # pickle.dump(automl, open('model.sav', 'wb'))
        predictions = automl.predict(X_test)
        print("Test R2 score:", sklearn.metrics.r2_score(y_test, predictions))
        print("#" * 80)
        # automl.fit(X_train, y_train)
```

```
[WARNING] [2023-04-16 18:39:42,839:Client-EnsembleBuilder] No runs were available to
      build an ensemble from
      [WARNING] [2023-04-16 18:39:44.650:Client-EnsembleBuilder] No models better than ran
      dom - using Dummy losses!
             Models besides current dummy model: 0
             Dummy models: 1
      [WARNING] [2023-04-16 18:39:47,509:Client-EnsembleBuilder] No models better than ran
      dom - using Dummy losses!
             Models besides current dummy model: 0
             Dummv models: 1
      [WARNING] [2023-04-16 18:39:47,769:Client-EnsembleBuilder] No models better than ran
      dom - using Dummy losses!
             Models besides current dummy model: 0
             Dummv models: 1
      [WARNING] [2023-04-16 18:39:50,060:Client-EnsembleBuilder] No models better than ran
      dom - using Dummy losses!
             Models besides current dummy model: 0
             Dummy models: 1
      [WARNING] [2023-04-16 18:39:51,699:Client-EnsembleBuilder] No models better than ran
      dom - using Dummy losses!
             Models besides current dummy model: 0
             Dummy models: 1
      [WARNING] [2023-04-16 18:39:56,475:Client-EnsembleBuilder] No models better than ran
      dom - using Dummy losses!
             Models besides current dummy model: 0
             Dummv models: 1
      [WARNING] [2023-04-16 18:39:59,311:Client-EnsembleBuilder] No models better than ran
      dom - using Dummy losses!
             Models besides current dummy model: 0
             Dummy models: 1
      [WARNING] [2023-04-16 18:40:01,953:Client-EnsembleBuilder] No models better than ran
      dom - using Dummy losses!
             Models besides current dummy model: 0
             Dummy models: 1
      [WARNING] [2023-04-16 18:40:02,218:Client-EnsembleBuilder] No models better than ran
      dom - using Dummy losses!
             Models besides current dummy model: 0
             Dummy models: 1
      [WARNING] [2023-04-16 18:40:02,532:Client-EnsembleBuilder] No models better than ran
      dom - using Dummy losses!
             Models besides current dummy model: 0
             Dummy models: 1
      [WARNING] [2023-04-16 18:40:02,828:Client-EnsembleBuilder] No models better than ran
      dom - using Dummy losses!
             Models besides current dummy model: 0
             Dummy models: 1
      Before re-fit
      Test R2 score: 0.733966672783318
      After re-fit
      Test R2 score: 0.7398337410800013
      In [7]: # print the results
        print(automl.leaderboard())
```

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# print the accuracy
predictions = automl.predict(X_test)

print("Test R2 score:", sklearn.metrics.r2_score(y_test, predictions))
```

	rank	ensemble_weight	type	cost	duration
$model_id$					
14	1	0.46	liblinear_svr	6.738509	1.562520
49	2	0.04	k_nearest_neighbors	9.433448	1.999790
22	3	0.50	gaussian_process	17.008425	1.969288
Test R2 s	core:	0.739833741080001	3		