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In [1]: import pandas as pd
import numpy as np
from sklearn.linear_model import Lasso
from sklearn.metrics import mean_squared_error, mean_absolute_error
from sklearn.model_selection import KFold
from sklearn.preprocessing import StandardScaler
from sklearn.pipeline import Pipeline

In [2]: # Load the dataset
data = pd.read_csv('1M_ahead_dataset.csv')

In [3]: # Separate predictors (X) and the target (Yt.1M)
X = data.drop(['Yt.1M'], axis=1)
y = data['Yt.1M']

In [4]: # Define candidate values for alpha ( $\lambda$ ) for Lasso Regression
alphas = [0.00392, 0.00393, 0.00394, 0.00395, 0.00396]

In [5]: # Set up 5-Fold cross-validation
kf = KFold(n_splits=5, shuffle=True, random_state=42)

In [6]: # To store average performance for each alpha candidate
results = []
# To store coefficients and intercepts for each fold
coefficient_results = []

In [7]: # Loop over each candidate alpha value
for alpha in alphas:
    fold_metrics = []
    print(f"\n--- Alpha {alpha} ---")
    fold_counter = 1
    for train_index, test_index in kf.split(X):
        # Split data into train and test sets for this fold
        X_train, X_test = X.iloc[train_index], X.iloc[test_index]
        y_train, y_test = y.iloc[train_index], y.iloc[test_index]

        # Create a pipeline: first scaling the features, then applying Lasso regression
        model = Pipeline([
            ('scaler', StandardScaler()),
            ('lasso', Lasso(alpha=alpha, max_iter=10000))
        ])

        # Train the Lasso regression model
        model.fit(X_train, y_train)

        # Extract coefficients and the intercept from the model
        coef = model.named_steps['lasso'].coef_
        intercept = model.named_steps['lasso'].intercept_

        print(f"\nFold {fold_counter} Coefficients:")
        print("Coefficients:", coef)
        print("Intercept:", intercept)

        # Store coefficient information for the current fold
        coefficient_results.append({
            'alpha': alpha,
            'fold': fold_counter,
            'coefficients': coef,
            'intercept': intercept
        })

        # Make predictions on the test set
        y_pred = model.predict(X_test)

        # Compute evaluation metrics
        mse = mean_squared_error(y_test, y_pred)
        mae = mean_absolute_error(y_test, y_pred)
        rmse = np.sqrt(mse)
        fold_metrics.append({'MSE': mse, 'RMSE': rmse, 'MAE': mae})

        fold_counter += 1

    # Average metrics over the 5 folds for the current alpha
    avg_mse = np.mean([m['MSE'] for m in fold_metrics])
    avg_rmse = np.mean([m['RMSE'] for m in fold_metrics])
    avg_mae = np.mean([m['MAE'] for m in fold_metrics])

    results.append({'alpha': alpha, 'avg_MSE': avg_mse, 'avg_RMSE': avg_rmse, 'avg_MAE': avg_mae})
    print(f"\nAlpha {alpha} -- Avg MSE: {avg_mse:.4f}, Avg RMSE: {avg_rmse:.4f}, Avg MAE: {avg_mae:.4f}")

```

--- Alpha 0.00392 ---

Fold 1 Coefficients:

Coefficients: [0.01193566 0. 0. 0.01660468 -0. 0.00626665
-0. -0.00230171 -0.00424899 0.00454103 -0.00203568 0.00882164
-0.]

Intercept: 0.04288559670781893

Fold 2 Coefficients:

Coefficients: [0.01035196 0. 0. 0.01840813 -0.001732 0.00308755
-0. -0.0021666 -0.00214098 0.00305171 -0. 0.01135243
0.]

Intercept: 0.044819341563786

Fold 3 Coefficients:

Coefficients: [0.00609366 0. 0.00165883 0.01792982 -0. 0.00409314
-0. -0.00167882 -0.00940134 0.00493291 -0. 0.00408763
0.]

Intercept: 0.0430213698630137

Fold 4 Coefficients:

Coefficients: [0.00527816 0. 0.0030764 0.01882981 -0. 0.00884039
-0. -0. -0.01006121 0.00315207 -0.00194534 0.00685442
0.]

Intercept: 0.04924054794520548

Fold 5 Coefficients:

Coefficients: [0.00853421 0.00389471 0. 0.02127411 -0. 0.00223461
-0. -0. -0.01003145 0.00507371 -0. 0.00994989
0.]

Intercept: 0.044690273972602744

Alpha 0.00392 -- Avg MSE: 0.0153, Avg RMSE: 0.1233, Avg MAE: 0.0729

--- Alpha 0.00393 ---

Fold 1 Coefficients:

Coefficients: [0.01192048 0. 0. 0.01659853 -0. 0.00625298
-0. -0.00229901 -0.0042274 0.00453259 -0.00202408 0.0088098
-0.]

Intercept: 0.04288559670781893

Fold 2 Coefficients:

Coefficients: [0.01033735 0. 0. 0.01839986 -0.00172037 0.00307595
-0. -0.00216344 -0.00212277 0.00303939 -0. 0.01134077
0.]

Intercept: 0.044819341563786

Fold 3 Coefficients:

Coefficients: [0.00608388 0. 0.00164808 0.01792139 -0. 0.00408158
-0. -0.00167439 -0.00938129 0.00492026 -0. 0.00407371
0.]

Intercept: 0.0430213698630137

Fold 4 Coefficients:

Coefficients: [0.0052649 0. 0.00306816 0.01882409 -0. 0.00882699
-0. -0. -0.01003925 0.00314289 -0.00193334 0.00684442
0.]

Intercept: 0.04924054794520548

Fold 5 Coefficients:

Coefficients: [0.0085226 0.00388743 0. 0.02126984 -0. 0.00222107
-0. -0. -0.01001002 0.00506107 -0. 0.00993914
0.]

Intercept: 0.044690273972602744

Alpha 0.00393 -- Avg MSE: 0.0153, Avg RMSE: 0.1233, Avg MAE: 0.0729

--- Alpha 0.00394 ---

Fold 1 Coefficients:

Coefficients: [0.01190529 0. 0. 0.01659238 -0. 0.00623932
-0. -0.00229631 -0.00420581 0.00452415 -0.00201247 0.00879796
-0.]

Intercept: 0.04288559670781893

Fold 2 Coefficients:

Coefficients: [0.01032273 0. 0. 0.01839159 -0.00170874 0.00306436
-0. -0.00216028 -0.00210456 0.00302707 -0. 0.01132912
0.]

Intercept: 0.044819341563786

Fold 3 Coefficients:

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Coefficients: [ 0.0060741  0.          0.00163733  0.01791295 -0.          0.00407002
-0.          -0.00166996 -0.00936124  0.0049076  -0.          0.00405979
0.          ]
Intercept: 0.0430213698630137

Fold 4 Coefficients:
Coefficients: [ 0.00525164  0.          0.00305992  0.01881837 -0.          0.00881359
-0.          -0.          -0.01001729  0.00313372 -0.00192134  0.00683442
0.          ]
Intercept: 0.04924054794520548

Fold 5 Coefficients:
Coefficients: [ 0.00851099  0.00388015  0.          0.02126556 -0.          0.00220752
0.          -0.          -0.0099886  0.00504844 -0.          0.0099284
0.          ]
Intercept: 0.044690273972602744

Alpha 0.00394 -- Avg MSE: 0.0153, Avg RMSE: 0.1233, Avg MAE: 0.0729

--- Alpha 0.00395 ---

Fold 1 Coefficients:
Coefficients: [ 0.0118901  0.          0.          0.01658623 -0.          0.00622566
-0.          -0.00229362 -0.00418422  0.00451572 -0.00200086  0.00878612
-0.          ]
Intercept: 0.04288559670781893

Fold 2 Coefficients:
Coefficients: [ 0.01030812  0.          0.          0.01838333 -0.0016971  0.00305276
-0.          -0.00215713 -0.00208636  0.00301476 -0.          0.01131747
0.          ]
Intercept: 0.044819341563786

Fold 3 Coefficients:
Coefficients: [ 0.00606431  0.          0.00162658  0.01790452 -0.          0.00405845
-0.          -0.00166553 -0.00934119  0.00489494 -0.          0.00404587
0.          ]
Intercept: 0.0430213698630137

Fold 4 Coefficients:
Coefficients: [ 0.00523838  0.          0.00305169  0.01881265 -0.          0.00880019
-0.          -0.          -0.00999534  0.00312455 -0.00190935  0.00682443
0.          ]
Intercept: 0.04924054794520548

Fold 5 Coefficients:
Coefficients: [ 0.00849939  0.00387286  0.          0.02126129 -0.          0.00219398
0.          -0.          -0.00996718  0.0050358  -0.          0.00991765
0.          ]
Intercept: 0.044690273972602744

Alpha 0.00395 -- Avg MSE: 0.0153, Avg RMSE: 0.1233, Avg MAE: 0.0729

--- Alpha 0.00396 ---

Fold 1 Coefficients:
Coefficients: [ 0.01187492  0.          0.          0.01658008 -0.          0.00621199
-0.          -0.00229092 -0.00416262  0.00450728 -0.00198925  0.00877429
-0.          ]
Intercept: 0.04288559670781893

Fold 2 Coefficients:
Coefficients: [ 0.0102935  0.          0.          0.01837506 -0.00168547  0.00304117
-0.          -0.00215397 -0.00206815  0.00300244 -0.          0.01130581
0.          ]
Intercept: 0.044819341563786

Fold 3 Coefficients:
Coefficients: [ 0.00605453  0.          0.00161583  0.01789609 -0.          0.00404689
-0.          -0.0016611  -0.00932114  0.00488228 -0.          0.00403196
0.          ]
Intercept: 0.0430213698630137

Fold 4 Coefficients:
Coefficients: [ 0.00522512  0.          0.00304345  0.01880693 -0.          0.00878679
-0.          -0.          -0.00997338  0.00311537 -0.00189735  0.00681443
0.          ]
Intercept: 0.04924054794520548

Fold 5 Coefficients:
Coefficients: [ 0.00848708  0.00386586  0.          0.02125733 -0.          0.00218018
0.          -0.          -0.00994549  0.00502306 -0.          0.00990686
0.          ]
```

Intercept: 0.044690273972602744

Alpha 0.00396 -- Avg MSE: 0.0153, Avg RMSE: 0.1233, Avg MAE: 0.0729

```
In [8]: # Summarize the cross-validation performance for each candidate alpha in a DataFrame
results_df = pd.DataFrame(results)
print("\nOverall Cross-Validation Results:")
print(results_df)
```

Overall Cross-Validation Results:

	alpha	avg_MSE	avg_RMSE	avg_MAE
0	0.00392	0.015325	0.123307	0.072916
1	0.00393	0.015325	0.123307	0.072918
2	0.00394	0.015325	0.123308	0.072921
3	0.00395	0.015325	0.123308	0.072924
4	0.00396	0.015325	0.123308	0.072926

```
In [9]: # Identify the best alpha based on average MSE
best_alpha = results_df.loc[results_df['avg_MSE'].idxmin()]['alpha']
print("\nBest alpha based on average MSE:", best_alpha)
```

Best alpha based on average MSE: 0.00395

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