

```
In [1]: import pandas as pd
import numpy as np
from sklearn.linear_model import Ridge
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
```

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In [2]: # Load the dataset
data = pd.read_csv('1M_ahead_dataset.csv')
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In [3]: # Assume that 'Yt.1M' is the target and the rest are predictors
X = data.drop(['Yt.1M'], axis=1)
y = data['Yt.1M']
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In [4]: # Define candidate values for alpha ( $\lambda$ ) to test.
alphas = [305, 306, 307, 308, 309]
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In [5]: # Set up 5-Fold cross-validation
kf = KFold(n_splits=5, shuffle=True, random_state=42)
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In [6]: # To store the average performance for each alpha candidate.
results = []
```

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In [7]: # To store coefficients for each fold
coefficient_results = []
```

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In [8]: # Loop over each candidate alpha
for alpha in alphas:
    fold_metrics = []
    print(f"\n--- Alpha {alpha} ---")
    fold_counter = 1
    for train_index, test_index in kf.split(X):
        # Split into training and testing 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 that scales the data and then applies Ridge regression
        model = Pipeline([
            ('scaler', StandardScaler()),
            ('ridge', Ridge(alpha=alpha))
        ])

        # Fit the model on training data
        model.fit(X_train, y_train)

        # Extract coefficients and intercept from the Ridge model
        # (Note: These coefficients correspond to scaled features)
        coef = model.named_steps['ridge'].coef_
        intercept = model.named_steps['ridge'].intercept_

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

        # Save the coefficients info
        coefficient_results.append({
            'alpha': alpha,
            'fold': fold_counter,
            'coefficients': coef,
            'intercept': intercept
        })

        # Make predictions for 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})

        # Increment fold counter
        fold_counter += 1

    # Compute average metrics over 5 folds for 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 305 ---

Fold 1 Coefficients:

Coefficients: [0.0112779 0.00093962 0.00136772 0.01437968 -0.00086332 0.00807467
-0.00037825 -0.00395971 -0.00531684 0.00616475 -0.00403575 0.00904481
0.00059337]

Intercept: 0.04288559670781893

Fold 2 Coefficients:

Coefficients: [0.0103322 0.00015042 0.00231984 0.01553257 -0.00331228 0.00598527
-0.00053489 -0.00323713 -0.00436739 0.00518027 -0.0013217 0.01092383
0.00067608]

Intercept: 0.044819341563786

Fold 3 Coefficients:

Coefficients: [0.00672094 0.00164141 0.00278518 0.01492215 -0.00065732 0.00590635
-0.000569 -0.00499564 -0.00884486 0.00665053 -0.00124285 0.00560227
0.00311413]

Intercept: 0.0430213698630137

Fold 4 Coefficients:

Coefficients: [0.00626842 0.00261658 0.00316873 0.01559258 -0.00117358 0.00930485
-0.00056288 -0.00352702 -0.00881027 0.00544224 -0.00405974 0.00719789
0.00245405]

Intercept: 0.04924054794520548

Fold 5 Coefficients:

Coefficients: [8.81535025e-03 4.02550616e-03 2.59285965e-03 1.70487439e-02
-7.05436424e-04 5.39575640e-03 3.45429370e-05 -2.93855546e-03
-9.07738401e-03 6.35666389e-03 -2.85887959e-03 9.64138675e-03
2.13041167e-03]

Intercept: 0.04469027397260275

Alpha 305 -- Avg MSE: 0.0154, Avg RMSE: 0.1235, Avg MAE: 0.0728

--- Alpha 306 ---

Fold 1 Coefficients:

Coefficients: [0.01126527 0.00094307 0.00136623 0.01436736 -0.00086125 0.00806736
-0.00037808 -0.00395621 -0.00530667 0.00616025 -0.00403123 0.00903558
0.00059421]

Intercept: 0.04288559670781893

Fold 2 Coefficients:

Coefficients: [0.01032076 0.00015674 0.0023158 0.01551802 -0.00330701 0.0059808
-0.00053465 -0.00323414 -0.00435935 0.00517551 -0.00132091 0.01091288
0.00067752]

Intercept: 0.044819341563786

Fold 3 Coefficients:

Coefficients: [0.00671415 0.00164292 0.00278068 0.01490788 -0.00065548 0.00590198
-0.00056881 -0.00499307 -0.00883115 0.00664469 -0.00124115 0.00559491
0.00311182]

Intercept: 0.0430213698630137

Fold 4 Coefficients:

Coefficients: [0.00626104 0.0026179 0.00316545 0.01557896 -0.0011713 0.0092966
-0.00056256 -0.00352715 -0.00879601 0.00543873 -0.00405474 0.00719061
0.00245306]

Intercept: 0.04924054794520548

Fold 5 Coefficients:

Coefficients: [8.80650298e-03 4.02322032e-03 2.59230475e-03 1.70334528e-02
-7.03498478e-04 5.39114549e-03 4.12700740e-05 -2.93928145e-03
-9.06167650e-03 6.35177744e-03 -2.85530018e-03 9.62921460e-03
2.12932500e-03]

Intercept: 0.04469027397260275

Alpha 306 -- Avg MSE: 0.0154, Avg RMSE: 0.1235, Avg MAE: 0.0728

--- Alpha 307 ---

Fold 1 Coefficients:

Coefficients: [0.01125268 0.0009465 0.00136474 0.01435507 -0.00085919 0.00806007
-0.00037791 -0.00395271 -0.00529654 0.00615575 -0.00402672 0.00902638
0.00059504]

Intercept: 0.04288559670781893

Fold 2 Coefficients:

Coefficients: [0.01030935 0.00016302 0.00231179 0.01550349 -0.00330176 0.00597633

-0.00053442 -0.00323116 -0.00435134 0.00517075 -0.00132011 0.01090194
0.00067895]
Intercept: 0.044819341563786

Fold 3 Coefficients:
Coefficients: [0.00670737 0.0016444 0.00277619 0.01489363 -0.00065365 0.00589761
-0.00056863 -0.00499049 -0.00881748 0.00663885 -0.00123945 0.00558757
0.00310952]
Intercept: 0.0430213698630137

Fold 4 Coefficients:
Coefficients: [0.00625368 0.0026192 0.00316218 0.01556535 -0.00116903 0.00928836
-0.00056224 -0.00352727 -0.0087818 0.00543522 -0.00404976 0.00718334
0.00245208]
Intercept: 0.04924054794520548

Fold 5 Coefficients:
Coefficients: [8.79767713e-03 4.02093714e-03 2.59174711e-03 1.70181880e-02
-7.01565878e-04 5.38654775e-03 4.79644501e-05 -2.93998579e-03
-9.04602747e-03 6.34689763e-03 -2.85173242e-03 9.61708172e-03
2.12824012e-03]
Intercept: 0.04469027397260275

Alpha 307 -- Avg MSE: 0.0154, Avg RMSE: 0.1235, Avg MAE: 0.0728

--- Alpha 308 ---

Fold 1 Coefficients:
Coefficients: [0.01124011 0.00094991 0.00136327 0.01434279 -0.00085713 0.0080528
-0.00037773 -0.00394921 -0.00528644 0.00615126 -0.00402222 0.00901719
0.00059587]
Intercept: 0.04288559670781893

Fold 2 Coefficients:
Coefficients: [0.01029797 0.00016925 0.0023078 0.01548899 -0.00329652 0.00597187
-0.00053418 -0.00322817 -0.00434336 0.005166 -0.00131932 0.01089103
0.00068038]
Intercept: 0.044819341563786

Fold 3 Coefficients:
Coefficients: [0.00670062 0.00164586 0.00277173 0.01487942 -0.00065182 0.00589326
-0.00056844 -0.0049879 -0.00880386 0.00663303 -0.00123777 0.00558025
0.00310722]
Intercept: 0.0430213698630137

Fold 4 Coefficients:
Coefficients: [0.00624634 0.00262048 0.00315893 0.01555178 -0.00116676 0.00928014
-0.00056192 -0.00352737 -0.00876764 0.00543172 -0.0040448 0.0071761
0.00245109]
Intercept: 0.04924054794520548

Fold 5 Coefficients:
Coefficients: [8.78887261e-03 4.01865661e-03 2.59118676e-03 1.70029494e-02
-6.99638608e-04 5.38196308e-03 5.46262614e-05 -2.94066869e-03
-9.03043657e-03 6.34202446e-03 -2.84817624e-03 9.60498791e-03
2.12715703e-03]
Intercept: 0.04469027397260275

Alpha 308 -- Avg MSE: 0.0154, Avg RMSE: 0.1235, Avg MAE: 0.0728

--- Alpha 309 ---

Fold 1 Coefficients:
Coefficients: [0.01122758 0.00095328 0.00136181 0.01433054 -0.00085507 0.00804554
-0.00037756 -0.00394571 -0.00527638 0.00614677 -0.00401773 0.00900803
0.00059669]
Intercept: 0.04288559670781893

Fold 2 Coefficients:
Coefficients: [0.01028662 0.00017544 0.00230385 0.01547452 -0.00329129 0.00596742
-0.00053394 -0.00322518 -0.0043354 0.00516126 -0.00131852 0.01088014
0.0006818]
Intercept: 0.044819341563786

Fold 3 Coefficients:
Coefficients: [0.00669388 0.0016473 0.00276729 0.01486523 -0.00065 0.00588891
-0.00056825 -0.0049853 -0.00879028 0.00662722 -0.00123608 0.00557295
0.00310492]
Intercept: 0.0430213698630137

Fold 4 Coefficients:
Coefficients: [0.00623903 0.00262174 0.00315569 0.01553822 -0.00116449 0.00927193
-0.0005616 -0.00352745 -0.00875353 0.00542822 -0.00403985 0.00716886

0.0024501]
Intercept: 0.04924054794520548

Fold 5 Coefficients:

Coefficients: [8.78008933e-03 4.01637874e-03 2.59062373e-03 1.69877371e-02
-6.97716650e-04 5.37739141e-03 6.12557025e-05 -2.94133035e-03
-9.01490344e-03 6.33715791e-03 -2.84463156e-03 9.59293294e-03
2.12607570e-03]

Intercept: 0.04469027397260275

Alpha 309 -- Avg MSE: 0.0154, Avg RMSE: 0.1235, Avg MAE: 0.0728

```
In [9]: # Summarize the overall results 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	305	0.015383	0.123498	0.072762
1	306	0.015382	0.123498	0.072767
2	307	0.015382	0.123498	0.072773
3	308	0.015382	0.123498	0.072778
4	309	0.015382	0.123498	0.072783

```
In [10]: # 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: 308.0

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