

COE_CatA_Model

October 5, 2025

1 1. Import libraries and data

```
[1]: from sklearn.linear_model import LinearRegression, Lasso, Ridge
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score, precision_score, recall_score, r2_score
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import plotly.express as px
```

1.1 1.1: Load initial CSV

```
[2]: coe_df = pd.read_csv(r"F:\Data Science Grand Track\Machine_Learning\datasets\coe_pricing\COEBiddingResultsPrices.csv")
```

```
[3]: display(coe_df)
```

	month	bidding_no	vehicle_class	quota	bids_success	bids_received	\
0	2010-01	1	Category A	1152	1145	1342	
1	2010-01	1	Category B	687	679	883	
2	2010-01	1	Category C	173	173	265	
3	2010-01	1	Category D	373	365	509	
4	2010-01	1	Category E	586	567	1011	
...	
1855	2025-09	2	Category A	1272	1266	2521	
1856	2025-09	2	Category B	792	772	1273	
1857	2025-09	2	Category C	276	274	413	
1858	2025-09	2	Category D	539	539	617	
1859	2025-09	2	Category E	256	252	540	

premium

0	18502
1	19190
2	19001
3	889
4	19889

...

```
1855    119003
1856    136890
1857     72501
1858      9209
1859    140502
```

[1860 rows x 7 columns]

1.2 1.2. Get Cat A data

```
[4]: coe_catA_df = coe_df[coe_df['vehicle_class'] == 'Category A']
```

```
[5]: display(coe_catA_df)
```

	month	bidding_no	vehicle_class	quota	bids_success	bids_received	\
0	2010-01	1	Category A	1152	1145	1342	
5	2010-01	2	Category A	1151	1149	1673	
10	2010-02	1	Category A	1154	1153	1326	
15	2010-02	2	Category A	1148	1148	1493	
20	2010-03	1	Category A	1148	1141	1758	
...	
1835	2025-07	2	Category A	1253	1,253	1,874	
1840	2025-08	1	Category A	1268	1,257	1,755	
1845	2025-08	2	Category A	1264	1,257	1,922	
1850	2025-09	1	Category A	1275	1,271	1,844	
1855	2025-09	2	Category A	1272	1266	2521	
		premium					
0		18502					
5		20501					
10		19989					
15		20340					
20		20802					
...		...					
1835		101102					
1840		102009					
1845		104524					
1850		107889					
1855		119003					

[372 rows x 7 columns]

1.3 1.3: Drop month column, keeping only numeric data

```
[6]: coe_catA_df = coe_catA_df.drop(['vehicle_class'], axis=1)
```

```
[7]: coe_catA_df = coe_catA_df.drop(['bidding_no', 'month'], axis=1)
```

```
[8]: display(coe_catA_df)
```

	quota	bids_success	bids_received	premium
0	1152	1145	1342	18502
5	1151	1149	1673	20501
10	1154	1153	1326	19989
15	1148	1148	1493	20340
20	1148	1141	1758	20802
...
1835	1253	1,253	1,874	101102
1840	1268	1,257	1,755	102009
1845	1264	1,257	1,922	104524
1850	1275	1,271	1,844	107889
1855	1272	1266	2521	119003

[372 rows x 4 columns]

1.4 1.4: Cleaning data:

1.4.1 1.4.1: Function to remove the , in numbers and convert it

```
[9]: def remove_extra_comma(num: str) -> int:  
    num = num.replace(",", "")  
    return int(num)
```

1.4.2 1.4.2: Apply function

```
[10]: coe_catA_df['bids_received'] = coe_catA_df['bids_received'].  
      ↪apply(remove_extra_comma)  
coe_catA_df['bids_success'] = coe_catA_df['bids_success'].  
      ↪apply(remove_extra_comma)
```

```
[11]: display(coe_catA_df.dtypes)
```

quota	int64
bids_success	int64
bids_received	int64
premium	int64
dtype: object	

2 2. Prepare data for training

2.1 2.1: Set X and y

```
[12]: X = coe_catA_df[['quota', 'bids_success', 'bids_received']].values  
y = coe_catA_df['premium'].values
```

```
[13]: print(X.shape)
      print(y.shape)
```

```
(372, 3)
(372,)
```

2.2 2.2 Split into training and testing data

```
[14]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.25,
                                                       random_state=42)
```

3 3. Train model using Lasso

3.1 3.1: Train model using training data

```
[15]: # Import LassoCV for automatic cross validation -> auto alpha
      from sklearn.linear_model import LassoCV
```

```
[ ]: # Creating a object:
predict_classA_lasso = LassoCV(
    cv=10, # Set number of cross validation folds to 10
    random_state = 42, # Split data randomly
    max_iter=5000
).fit(X_train, y_train)

# Print out the best
print("Best alpha: {best_alpha}".format(best_alpha = predict_classA_lasso.
                                         alpha_))
```

```
Best alpha: 5626.363897393403
```

3.2 3.2: Predicting price:

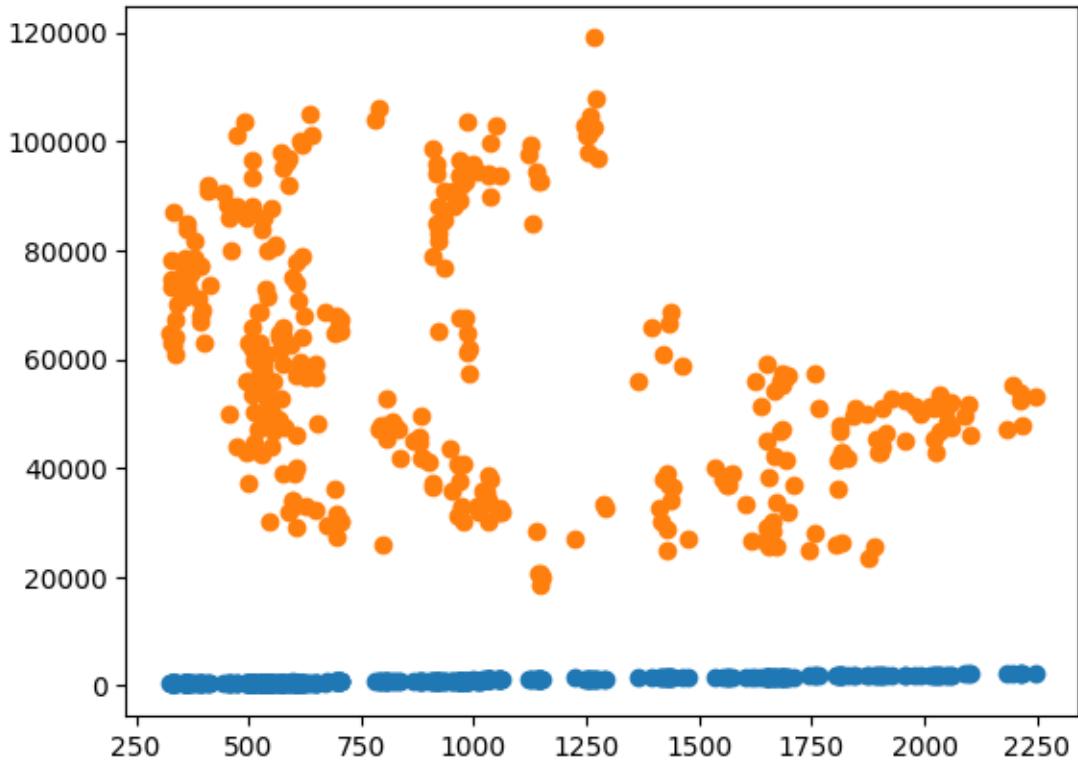
```
[17]: # Predicting y (label; COE price from testing data)
y_pred = predict_classA_lasso.predict(X_test)
```

```
[ ]: display(y_pred)
      display(y_pred.shape)
```

4 4. Calculating metrics

4.1 4.1: Plot between

```
[ ]: plt.scatter(coe_catA_df['bids_success'], coe_catA_df['quota'])
      plt.show()
```



4.2 4.2: Calculate metrics

4.2.1 4.2.1: Calculate MSE

```
[23]: from sklearn.metrics import mean_squared_error
```

```
[24]: mse = mean_squared_error(y_pred, y_test)
print(f"Mean square error is: {mse}")
```

Mean square error is: 462220339.8743456

4.2.2 4.2.2: Calculate R_squared

```
[33]: r2 = predict_classA_lasso.score(X_test, y_test)
print(f"R^2 square score: {r2}")
```

R^2 square score: 0.12553682062655902

4.2.3 4.2.3: Calculate feature importance

```
[ ]: feature_coef = predict_classA_lasso = LassoCV(
    cv=10, # Set number of cross validation folds to 10
    random_state = 42, # Split data randomly
```

```

    max_iter=5000
).fit(X_train, y_train).coef_

[-107.38520317    83.96685537    6.06366876]

[44]: # Get names of
column_names = ['quota', 'bids_success', 'bids_received']

# Combine into a DataFrame for easier plotting
coef_df = pd.DataFrame({
    'Feature': column_names,
    'Coefficient': feature_coef
})

# Create the bar plot
fig = px.bar(
    coef_df,
    x='Feature',
    y='Coefficient',
    color='Coefficient', # color based on coefficient value
    color_continuous_scale=['red', 'white', 'green'], # negative to positive
    title='Lasso Coefficients of Factors to COE Price (Cat A)',
    labels={'Coefficient': 'Lasso Weight'}
)

fig.show()

```