## central test linear regression

## March 20, 2024

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[39]: # importing the necesary libraries
      import numpy as np
      import pandas as pd
      from sklearn.linear_model import LinearRegression,Ridge
      from sklearn.model_selection import train_test_split
      from sklearn.preprocessing import StandardScaler
      from sklearn.metrics import mean_squared_error, mean_absolute_error,r2_score
      import matplotlib.pyplot as plt
[40]: # Loading the dataset
      data=pd.read_csv('C:\\Users\\lynda\\Desktop\\company.csv')
      data.head()
[40]:
        R&D Spend Administration Marketing Spend
                                                          State
                                                                    Profit
      0 165349.20
                         136897.80
                                          471784.10
                                                      New York 192261.83
      1 162597.70
                         151377.59
                                          443898.53 California 191792.06
      2 153441.51
                                          407934.54
                                                        Florida 191050.39
                         101145.55
      3 144372.41
                         118671.85
                                          383199.62
                                                      New York 182901.99
      4 142107.34
                         91391.77
                                          366168.42
                                                       Florida 166187.94
[41]: #checking for null values
      data.isna().sum()
[41]: R&D Spend
                        0
      Administration
      Marketing Spend
      State
      Profit
                         0
      dtype: int64
[42]: # defining the x(independent valu)
      x=data[['R&D Spend','Administration','Marketing Spend']]
      x.head()
[42]:
        R&D Spend Administration Marketing Spend
      0 165349.20
                                          471784.10
                         136897.80
      1 162597.70
                         151377.59
                                          443898.53
      2 153441.51
                         101145.55
                                          407934.54
```

```
3 144372.41
                         118671.85
                                          383199.62
      4 142107.34
                          91391.77
                                          366168.42
[43]: # y(dependent values)
      y=data['Profit']
      y.head()
[43]: 0
          192261.83
          191792.06
      1
          191050.39
      3
          182901.99
          166187.94
     Name: Profit, dtype: float64
[44]: # spliting my data into training and testing set
      x_train,x_test,y_train,y_test = train_test_split(x,y, test_size=0.2,__
       →random_state=42)
[45]: # Building my Ordinary Linear regression
      model = LinearRegression()
      model.fit(x train, y train)
[45]: LinearRegression()
[47]: #calculating accuracy score
      y_pred = model_ord.predict(x_test)
      print('y_pred_ord',y_pred_ord)
     y_pred_ord [126703.02716461 84894.75081556 98893.41815974 46501.70815036
      129128.39734381 50992.69486261 109016.5536578 100878.4641454
       97700.59638629 113106.15292226]
[49]: accuracy = r2_score(y_test,y_pred)
      print('accuracy:',accuracy)
     accuracy: 0.900065308303732
[54]: # Optimization of Linear Regression
      scaler = StandardScaler()
      x train opt = scaler.fit transform(x train)
      x_test_opt = scaler.transform(x_test)
      # using ridge
      ridge_model=Ridge(alpha=0.5)
      ridge_model.fit(x_train_opt, y_train)
      y_pred_ridge= ridge_model.predict(x_test_opt)
      accuracy2= r2_score(y_test,y_pred_ridge)
      print('accuracy2',accuracy2)
```

## accuracy2 0.8983198084871222

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[55]: if accuracy>accuracy2:
        print('The ordinary model performs better than the optimized model')
    elif accuracy < accuracy2:
        print('The ordinary model performs better than the optimized')
    else:
        print('Both modes have the same accuracy')

The ordinary model performs better than the optimized model

[]:
[]:</pre>
```