1. Import Packages and Observe Dataset

```
In [ ]: import pandas as pd
         import numpy as np
         import seaborn as sns
         import matplotlib.pyplot as plt
         %matplotlib inline
 In [9]: from sklearn import preprocessing
         from sklearn.model_selection import train_test_split
         from sklearn.linear_model import LinearRegression, Ridge, Lasso
         from sklearn.metrics import r2_score
In [10]: data = pd.read_csv(r"C:\Users\admin\Downloads\car-mpg.csv")
         data.head()
Out[10]:
                       disp
                              hp
                                        acc yr origin car_type
            mpg cyl
                                    wt
                                                                              car name
             18.0
                    8 307.0
                            130 3504 12.0
                                             70
                                                              0 chevrolet chevelle malibu
             15.0
                    8 350.0
                            165
                                  3693 11.5
                                                                        buick skylark 320
             18.0
                    8 318.0 150
                                  3436 11.0 70
                                                              0
                                                     1
                                                                        plymouth satellite
             16.0
                    8 304.0
                            150
                                  3433 12.0 70
                                                              0
                                                                           amc rebel sst
             17.0
                    8 302.0 140 3449 10.5 70
                                                     1
                                                              0
                                                                             ford torino
In [24]: # 1. Drop 'car_name' if it exists
         if 'car_name' in data.columns:
             data = data.drop(['car_name'], axis=1)
         # 2. Replace numbers in 'origin' with names and make dummy columns
         if 'origin' in data.columns:
             data['origin'] = data['origin'].replace({1: 'america', 2: 'europe', 3: 'asia'})
             data = pd.get_dummies(data, columns=['origin'])
         # 3. Replace '?' with NaN
         data = data.replace('?', np.nan)
         # 4. Make sure all numbers are numeric (fix warning)
         data = data.apply(pd.to_numeric, errors='coerce')
         # 5. Fill missing numbers with median
         data = data.fillna(data.median())
In [25]: data.head()
```

```
Out[25]:
             mpg cyl
                         disp
                                 hp
                                        wt
                                             acc
                                                  yr car_type origin_america origin_asia origin_eur
              18.0
                        307.0
                              130.0 3504
                                            12.0
                                                  70
                                                             0
                                                                          True
                                                                                     False
                                                                                                    F
          0
                     8
              15.0
                     8 350.0
                              165.0
                                     3693
                                            11.5
                                                 70
                                                             0
                                                                                     False
                                                                                                    F
                                                                          True
          2
              18.0
                     8 318.0
                              150.0 3436
                                            11.0
                                                 70
                                                             0
                                                                          True
                                                                                     False
                                                                                                    F
              16.0
                     8 304.0
                              150.0 3433
                                            12.0
                                                                                                    F
                                                 70
                                                             0
                                                                          True
                                                                                     False
              17.0
                     8 302.0 140.0 3449
                                            10.5 70
                                                             0
                                                                          True
                                                                                     False
                                                                                                    F
In [26]: X=data.drop(['mpg'],axis=1)
          Y=data[['mpg']]
In [30]: from sklearn import preprocessing
          # Scale X
          X_s = preprocessing.scale(X)
          X_s = pd.DataFrame(X_s, columns=X.columns)
          # Scale Y
          Y_s = preprocessing.scale(Y)
          Y s = pd.DataFrame(Y s, columns=Y.columns)
In [31]: X_s
Out[31]:
                      cyl
                               disp
                                            hp
                                                      wt
                                                                 acc
                                                                                 car_type origin_ame
                                                                            yr
                1.498191
             0
                           1.090604
                                      0.673118
                                                 0.630870
                                                          -1.295498 -1.627426 -1.062235
                                                                                                 0.773
                 1.498191
                           1.503514
                                      1.589958
                                                 0.854333
                                                          -1.477038 -1.627426 -1.062235
                                                                                                 0.773
                                                          -1.658577 -1.627426 -1.062235
             2
                 1.498191
                           1.196232
                                      1.197027
                                                 0.550470
                                                                                                 0.773
             3
                 1.498191
                           1.061796
                                      1.197027
                                                 0.546923
                                                          -1.295498 -1.627426
                                                                                -1.062235
                                                                                                 0.773
                 1.498191
                           1.042591
                                      0.935072
                                                 0.565841
                                                          -1.840117 -1.627426
                                                                                -1.062235
                                                                                                 0.773
             4
                -0.856321
                           -0.513026
                                     -0.479482
                                                -0.213324
          393
                                                            0.011586
                                                                      1.621983
                                                                                 0.941412
                                                                                                 0.773
               -0.856321 -0.925936 -1.370127 -0.993671
                                                            3.279296
                                                                      1.621983
                                                                                 0.941412
                                                                                                -1.292
          394
                -0.856321
                           -0.561039 -0.531873
                                                          -1.440730
                                                                                                 0.773
          395
                                                -0.798585
                                                                      1.621983
                                                                                 0.941412
               -0.856321 -0.705077 -0.662850 -0.408411
                                                           1.100822
                                                                      1.621983
                                                                                                 0.773
          396
                                                                                 0.941412
               -0.856321 -0.714680 -0.584264 -0.296088
                                                            1.391285
                                                                      1.621983
                                                                                 0.941412
                                                                                                 0.773
         398 rows × 10 columns
```

```
In [36]: from sklearn.model_selection import train_test_split

X_train, X_test, Y_train, Y_test = train_test_split(X_s, Y_s, test_size=0.2, random X_train.shape, X_test.shape, Y_train.shape, Y_test.shape

Out[36]: ((318, 10), (80, 10), (318, 1), (80, 1))
```

2.a Simple Linear Model

```
In [39]: regression_model=LinearRegression()
         regression model.fit(X train,Y train)
         for idX,col_name in enumerate(X_train.columns):
             print('The coefficient for {} is {}'.format(col_name,regression_model.coef_[0][
         intercept=regression_model.intercept_[0]
         print('The intercept is {}'.format(intercept))
        The coefficient for cyl is 0.3079552263085646
        The coefficient for disp is 0.2749461027215073
        The coefficient for hp is -0.2003169751506641
        The coefficient for wt is -0.6755882897308452
        The coefficient for acc is 0.03311218992661376
        The coefficient for yr is 0.3720678277926433
        The coefficient for car_type is 0.37184733079847426
        The coefficient for origin_america is -0.07869861984419448
        The coefficient for origin_asia is 0.05552985672521807
        The coefficient for origin_europe is 0.04186331320884182
        The intercept is 0.004215857290129611
```

2.b Regularized Ridge Regression

2.c Regularized Lasso Regression

```
In [43]: from sklearn.linear_model import Lasso

# Create and train Lasso model
lasso_model = Lasso(alpha=0.1)
lasso_model.fit(X_train, Y_train) # use .fit(), not .fir()
```