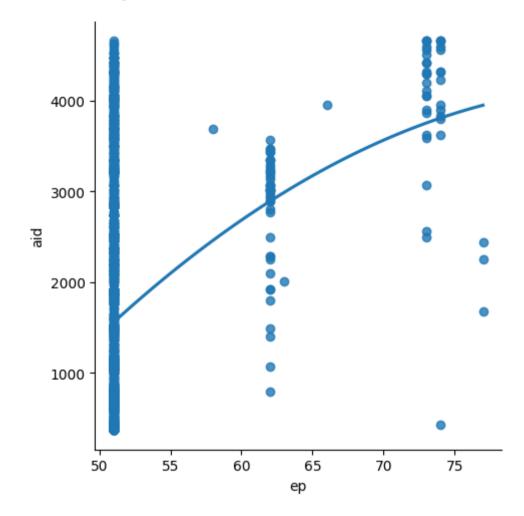
```
In [1]:
           1 import numpy as np
             import pandas as pd
             import seaborn as sns
             import matplotlib.pyplot as plt
           5 from sklearn import preprocessing, svm
             from sklearn.model selection import train test split
           7 from sklearn.linear model import LinearRegression
In [2]:
           1 df=pd.read csv(r"C:\Users\HP\OneDrive\Documents\fiat500 VehicleSelection Dataset.csv")
           2 df
Out[2]:
                                                        km previous_owners
                 ID model engine_power age_in_days
                                                                                  lat
                                                                                           Ion price
             0
                  1 lounge
                                     51
                                                882
                                                      25000
                                                                         1 44.907242
                                                                                      8.611560
                                                                                               8900
                  2
                                                      32500
                                                                         1 45.666359 12.241890
                                                                                               8800
                                     51
                                               1186
                       pop
             2
                                     74
                                               4658
                                                    142228
                                                                         1 45.503300 11.417840
                                                                                               4200
                  3
                      sport
                  4 lounge
                                     51
                                               2739
                                                    160000
                                                                           40.633171 17.634609
                                                                                               6000
                                     73
                                               3074
                                                    106880
                                                                         1 41.903221 12.495650
                                                                                               5700
                       pop
                                                                                      7.704920
          1533 1534
                      sport
                                     51
                                               3712 115280
                                                                         1 45.069679
                                                                                               5200
                                     74
                                                    112000
                                                                                      8.666870
          1534 1535 lounge
                                               3835
                                                                         1 45.845692
                                                                                               4600
          1535 1536
                                     51
                                               2223
                                                      60457
                                                                         1 45.481541
                                                                                      9.413480
                                                                                              7500
                       pop
                                     51
                                                      80750
                                                                                      7.682270
                                                                                               5990
          1536
               1537
                     lounge
                                               2557
                                                                         1 45.000702
          1537 1538
                                     51
                                               1766
                                                     54276
                                                                         1 40.323410 17.568270 7900
                       pop
         1538 rows × 9 columns
In [3]:
           1 df=df[['engine power', 'age in days']]
           2 df.columns=['ep','aid']
```

1 df.describe() In [4]: Out[4]: ер aid count 1538.000000 1538.000000 51.904421 1650.980494 mean std 3.988023 1289.522278 51.000000 366.000000 min 25% 51.000000 670.000000 50% 51.000000 1035.000000 51.000000 2616.000000 75% 77.000000 4658.000000 max In [5]: 1 df.head(10) Out[5]: aid **0** 51 882 **1** 51 1186 **2** 74 4658 **3** 51 2739 **4** 73 3074 **5** 74 3623 **6** 51 731 **7** 51 1521 **8** 73 4049 **9** 51 3653

```
In [6]:
           1 df.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 1538 entries, 0 to 1537
         Data columns (total 2 columns):
              Column Non-Null Count Dtype
                      1538 non-null int64
              ер
              aid
                      1538 non-null int64
         dtypes: int64(2)
         memory usage: 24.2 KB
           1 df.fillna(method='ffill')
In [24]:
Out[24]:
                    aid
               ер
                   882
             0 51
             1 51 1186
             2 74 4658
             3 51 2739
             4 73 3074
          1533 51 3712
          1534 74 3835
          1535 51 2223
          1536 51 2557
          1537 51 1766
         1538 rows × 2 columns
```

```
In [8]:
           1 x=np.array(df['ep']).reshape(-1,1)
           2 y=np.array(df['aid']).reshape(-1,1)
In [25]:
           1 df.dropna()
Out[25]:
                   aid
            0 51
                   882
            1 51 1186
            2 74 4658
            3 51 2739
            4 73 3074
          1533 51 3712
          1534 74 3835
          1535 51 2223
          1536 51 2557
          1537 51 1766
         1538 rows × 2 columns
```

Out[26]: <seaborn.axisgrid.FacetGrid at 0x247bfdeb9d0>



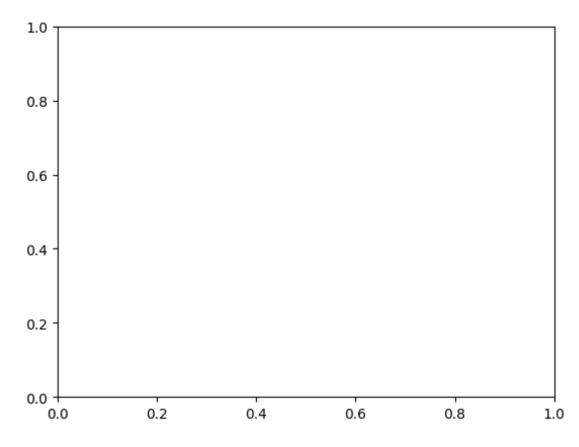
```
In [27]: 1 x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.25)
2 regr=LinearRegression()
3 regr.fit(x_train,y_train)
4 print(regr.score(x_test,y_test))
```

1.0

```
In [28]: 1 y_pred=regr.predict(x_test)
2 plt.scatter(x_test,y_test,color='b')
3 plt.plot(x_test,y_pred,color='k')
4 plt.show()
```

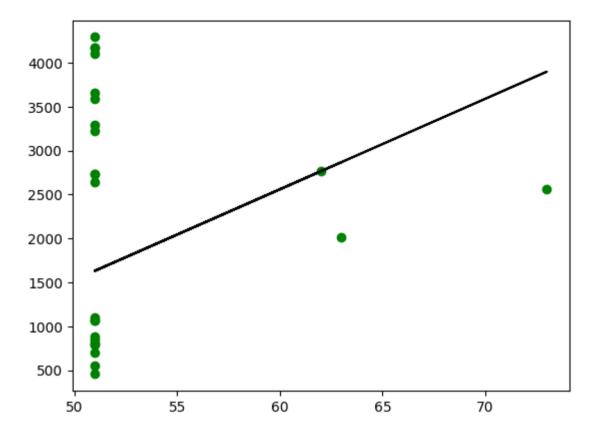
```
ValueError
                                          Traceback (most recent call last)
Cell In[28], line 2
      1 y pred=regr.predict(x test)
----> 2 plt.scatter(x test,y test,color='b')
      3 plt.plot(x test, v pred, color='k')
      4 plt.show()
File ~\AppData\Local\Programs\Python\Python310\lib\site-packages\matplotlib\pyplot.py:2862, in scatter(x, y, s, c, m
arker, cmap, norm, vmin, vmax, alpha, linewidths, edgecolors, plotnonfinite, data, **kwargs)
   2857 @ copy docstring and deprecators(Axes.scatter)
   2858 def scatter(
                x, y, s=None, c=None, marker=None, cmap=None, norm=None,
   2859
                vmin=None, vmax=None, alpha=None, linewidths=None, *,
   2860
   2861
                edgecolors=None, plotnonfinite=False, data=None, **kwargs):
            ret = gca().scatter(
-> 2862
                x, y, s=s, c=c, marker=marker, cmap=cmap, norm=norm,
   2863
                vmin=vmin, vmax=vmax, alpha=alpha, linewidths=linewidths,
   2864
                edgecolors=edgecolors, plotnonfinite=plotnonfinite,
   2865
                **({"data": data} if data is not None else {}), **kwargs)
   2866
   2867
            sci( ret)
            return ret
   2868
File ~\AppData\Local\Programs\Python\Python310\lib\site-packages\matplotlib\ init .py:1459, in preprocess data.<1
ocals>.inner(ax, data, *args, **kwargs)
   1456 @functools.wraps(func)
   1457 def inner(ax, *args, data=None, **kwargs):
   1458
            if data is None:
                return func(ax, *map(sanitize_sequence, args), **kwargs)
-> 1459
            bound = new sig.bind(ax, *args, **kwargs)
   1461
   1462
            auto label = (bound.arguments.get(label namer)
                          or bound.kwargs.get(label namer))
   1463
File ~\AppData\Local\Programs\Python\Python310\lib\site-packages\matplotlib\axes\ axes.py:4584, in Axes.scatter(sel
f, x, y, s, c, marker, cmap, norm, vmin, vmax, alpha, linewidths, edgecolors, plotnonfinite, **kwargs)
   4582 y = np.ma.ravel(y)
   4583 if x.size != y.size:
-> 4584
            raise ValueError("x and y must be the same size")
   4586 if s is None:
   4587
            s = (20 if mpl.rcParams[' internal.classic mode'] else
                 mpl.rcParams['lines.markersize'] ** 2.0)
   4588
```

ValueError: x and y must be the same size



-0.18690025900835017

Regression: -0.18690025900835017



R2_score: -0.18690025900835017

Conclusion:

```
In [31]:
           1 from sklearn.linear model import LinearRegression
           2 from sklearn.model selection import train test split
           3 feature=df.columns[0:3]
           4 target=df.columns[-1]
           5 x=df[feature].values
           6 y=df[target].values
          7 x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.25)
          8 lr = LinearRegression()
          9 lr.fit(x train,y train)
          10 print(lr.score(x test, y test))
          print(lr.score(x train,y train))
         1.0
         1.0
In [32]:
           1 from sklearn.linear model import Ridge,RidgeCV,Lasso,LassoCV
           2 ridge = Ridge(alpha=10)
           3 ridge.fit(x train,y train)
           4 train score ridge=ridge.score(x train,y train)
           5 test score ridge=ridge.score(x test,y test)
           6 print('\n Ridge method\n')
           7 print('The score of ridge method is {}'.format(train score ridge))
           8 print('The score of ridge method is {}'.format(test score ridge))
```

Ridge method

The score of ridge method is 1.0 The score of ridge method is 1.0

Lasso method

The score of lasso method is 0.999999999644759 The score of lasso method is 0.999999999644752

```
In [34]: 1 lasso_cv= LassoCV(alphas=[0.2,0.03,0.004,0.0001,1,20]).fit(x_train,y_train)
2 train_score_lasso_cv=lasso_cv.score(x_train,y_train)
3 test_score_lasso_cv=lasso_cv.score(x_test,y_test)
4 print('\n LassoCV method\n')
5 print('The score of Lasso method is {}'.format(train_score_lasso_cv))
6 print('The score of Lasso method is {}'.format(test_score_lasso_cv))
```

LassoCV method

The score of Lasso method is 1.0 The score of Lasso method is 1.0

```
In [35]: 1 ridge_cv=RidgeCV(alphas=[1,2.3,0.2,0.3,0.4,0.5,0.6]).fit(x_train,y_train)
2 print("\n RidgeCV Method\n")
3 print("The score of Ridge method is {}".format(ridge_cv.score(x_train,y_train)))
4 print("The score of Ridge method is {}".format(ridge_cv.score(x_test,y_test)))
```

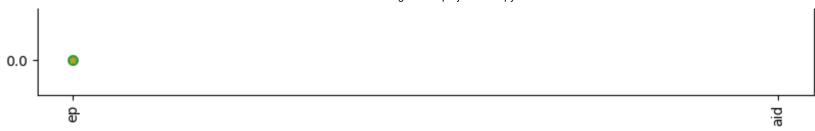
RidgeCV Method

The score of Ridge method is 0.999999994626445 The score of Ridge method is 0.999999994626342

```
In [36]:
1  plt.figure(figsize=(10,10))
2  plt.plot(feature,ridge.coef_,alpha=0.5,marker='*',markersize=5,linestyle='None',color='orange',label="Ridge = $\\
3  plt.plot(feature,lr.coef_,alpha=0.7,marker='o',markersize=7,color='green',linestyle='None',label="Linear Regressi plt.xticks(rotation=90)
5  plt.title("Comparision between Ridge and linear regression")
6  plt.legend()
7  plt.show()
```

Comparision between Ridge and linear regression

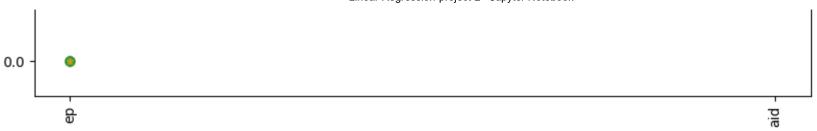




```
In [37]: 1 plt.figure(figsize=(10,10))
plt.plot(feature,ridge.coef_,alpha=0.5,marker='*',markersize=5,linestyle='None',color='orange',label="Ridge; $\ a plt.plot(feature,ridge.coef_,alpha=0.6,marker='d',markersize=6,linestyle='None',color='pink',label="Lasso; $\ alpha=0.7,marker='o',markersize=7,color='green',linestyle='None',label="Linear Regressi plt.xticks(rotation=90)
plt.title("Comparision between Ridge and linear regression")
plt.legend()
plt.show()
```

Comparision between Ridge and linear regression





ElasticNet

```
1 | from sklearn.linear_model import ElasticNet
In [39]:
           2 lr=ElasticNet()
           3 lr.fit(x,y)
           4 print("The score of ElasticNet is {}".format(lr.score(x train,y train)))
           5 print("The score of ElasticNet is {}".format(lr.score(x test,y test)))
           6 print(lr.coef )
           7 print(lr.intercept_)
         The score of ElasticNet is 0.9999999999996378
         The score of ElasticNet is 0.999999999999378
         ſ0.
                    0.99999941
         0.0009934970473750582
In [42]:
           1 y pred elastic = lr.predict(x test)
           2 mean_squared_error = np.mean(((y_pred_elastic)-y_test)**2)
           3 print("The mean squeared error is {}".format(mean squared error))
         The mean squeared error is 5.844007028719034e-07
 In [ ]:
```