# In [1]:

```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn import preprocessing,svm
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
```

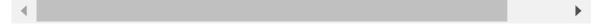
# In [2]:

```
df=pd.read_csv(r"C:\Users\HP\OneDrive\Documents\fiat500_VehicleSelection_Dataset.cs
df
```

# Out[2]:

|      | ID   | model  | engine_power | age_in_days | km     | previous_owners | lat       |        |
|------|------|--------|--------------|-------------|--------|-----------------|-----------|--------|
| 0    | 1    | lounge | 51           | 882         | 25000  | 1               | 44.907242 | 8.61   |
| 1    | 2    | рор    | 51           | 1186        | 32500  | 1               | 45.666359 | 12.24  |
| 2    | 3    | sport  | 74           | 4658        | 142228 | 1               | 45.503300 | 11.417 |
| 3    | 4    | lounge | 51           | 2739        | 160000 | 1               | 40.633171 | 17.634 |
| 4    | 5    | рор    | 73           | 3074        | 106880 | 1               | 41.903221 | 12.49  |
|      |      |        |              |             |        |                 |           |        |
| 1533 | 1534 | sport  | 51           | 3712        | 115280 | 1               | 45.069679 | 7.70₄  |
| 1534 | 1535 | lounge | 74           | 3835        | 112000 | 1               | 45.845692 | 8.66€  |
| 1535 | 1536 | рор    | 51           | 2223        | 60457  | 1               | 45.481541 | 9.413  |
| 1536 | 1537 | lounge | 51           | 2557        | 80750  | 1               | 45.000702 | 7.682  |
| 1537 | 1538 | pop    | 51           | 1766        | 54276  | 1               | 40.323410 | 17.568 |

#### 1538 rows × 9 columns



# In [3]:

```
1 df=df[['engine_power','age_in_days']]
2 df.columns=['ep','aid']
```

```
In [4]:
```

```
1 df.describe()
```

## Out[4]:

|       | ер          | aid         |
|-------|-------------|-------------|
| count | 1538.000000 | 1538.000000 |
| mean  | 51.904421   | 1650.980494 |
| std   | 3.988023    | 1289.522278 |
| min   | 51.000000   | 366.000000  |
| 25%   | 51.000000   | 670.000000  |
| 50%   | 51.000000   | 1035.000000 |
| 75%   | 51.000000   | 2616.000000 |
| max   | 77.000000   | 4658.000000 |

# In [5]:

```
1 df.head(10)
```

# Out[5]:

|   | ер | 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
--- 0 ep 1538 non-null int64
1 aid 1538 non-null int64
dtypes: int64(2)
memory usage: 24.2 KB
```

```
In [7]:
```

```
1 df.fillna(method='ffill',inplace=True)
```

C:\Users\HP\AppData\Local\Temp\ipykernel\_22436\4116506308.py:1: SettingWi
thCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy)

df.fillna(method='ffill',inplace=True)

## In [8]:

```
1 x=np.array(df['ep']).reshape(-1,1)
2 y=np.array(df['aid']).reshape(-1,1)
```

# In [9]:

```
1 df.dropna(inplace=True)
```

C:\Users\HP\AppData\Local\Temp\ipykernel\_22436\1379821321.py:1: SettingWi
thCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy)

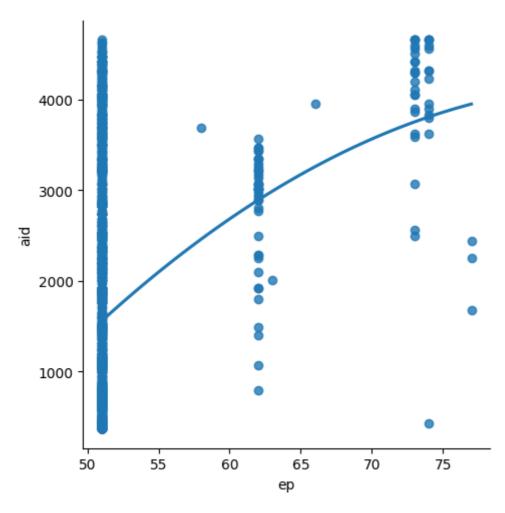
df.dropna(inplace=True)

# In [10]:

```
#Exploring the data scatter_plotting the data scatter
sns.lmplot(x = "ep", y = "aid", data = df, order = 2, ci = None)
```

# Out[10]:

<seaborn.axisgrid.FacetGrid at 0x20e4319bb50>



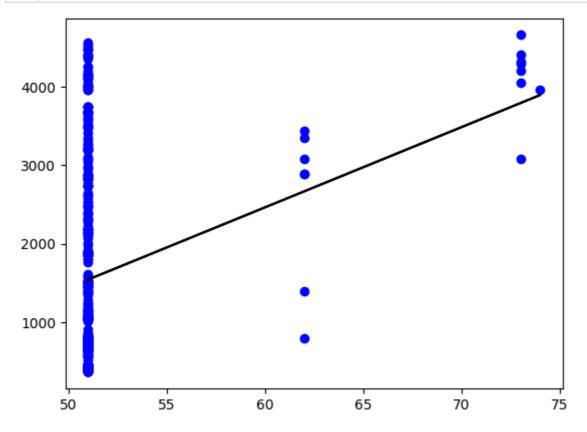
# In [11]:

```
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))
```

#### 0.08266478450766102

# In [12]:

```
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()
```

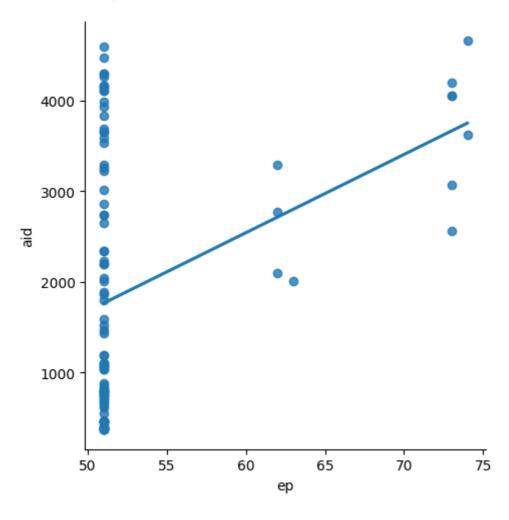


# In [13]:

```
1 df100=df[:][:100]
2 sns.lmplot(x='ep',y='aid',data=df100,order=1,ci=None)
3
```

# Out[13]:

<seaborn.axisgrid.FacetGrid at 0x20e28fa1ab0>



#### In [14]:

```
df100.fillna(method='ffill',inplace=True)

X=np.array(df100['ep']).reshape(-1,1)

y=np.array(df100['aid']).reshape(-1,1)

df100.dropna(inplace=True)

X_train,x_test,y_train,y_test=train_test_split(X,y,test_size=0.25)

regr=LinearRegression()

regr.fit(X_train,y_train)

print(regr.score(x_test,y_test))

print("Regression: ",regr.score(x_test,y_test))

y_pred=regr.predict(x_test)

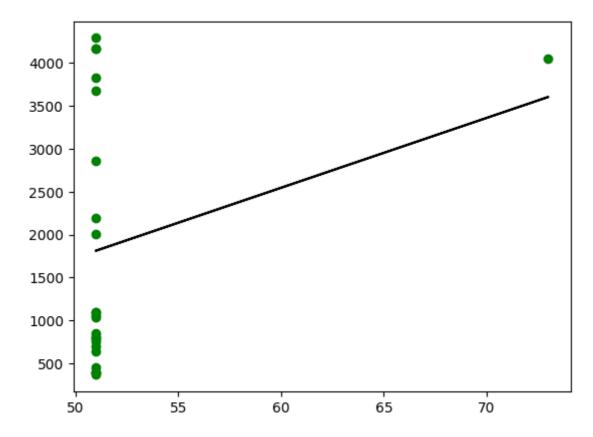
plt.scatter(x_test,y_test,color='g')

plt.plot(x_test,y_pred,color='k')

plt.show()
```

#### 0.08993389468303326

Regression: 0.08993389468303326



# In [15]:

```
from sklearn.linear_model import LinearRegression
from sklearn.metrics import r2_score
model=LinearRegression()
model.fit(X_train,y_train)
y_pred=model.predict(x_test)
r2=r2_score(y_test,y_pred)
print("R2_score: ",r2)
```

R2\_score: 0.08993389468303326

# **Conclusion:**

Dataset we have taken is poor for linear model but with the smaller data w orks well with linear model

## In [16]:

```
from sklearn.linear_model import LinearRegression
from sklearn.model_selection import train_test_split
feature=df.columns[0:3]
target=df.columns[-1]

x=df[feature].values
y=df[target].values
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.25)
lr = LinearRegression()
lr.fit(x_train,y_train)
print(lr.score(x_test,y_test))
print(lr.score(x_test,y_train))
```

1.0

1.0

## In [17]:

```
from sklearn.linear_model import Ridge,RidgeCV,Lasso,LassoCV
ridge = Ridge(alpha=10)
ridge.fit(x_train,y_train)
train_score_ridge=ridge.score(x_train,y_train)
test_score_ridge=ridge.score(x_test,y_test)
print('\n Ridge method\n')
print('The score of ridge method is {}'.format(train_score_ridge))
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
```

## In [18]:

```
1  lasso = Lasso(alpha=10)
2  lasso.fit(x_train,y_train)
3  train_score_lasso=lasso.score(x_train,y_train)
4  test_score_lasso=lasso.score(x_test,y_test)
5  print('\n Lasso method\n')
6  print('The score of lasso method is {}'.format(train_score_lasso))
7  print('The score of lasso method is {}'.format(test_score_lasso))
```

## Lasso method

```
The score of lasso method is 0.9999999999632855
The score of lasso method is 0.9999999999632833
```

#### In [19]:

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

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

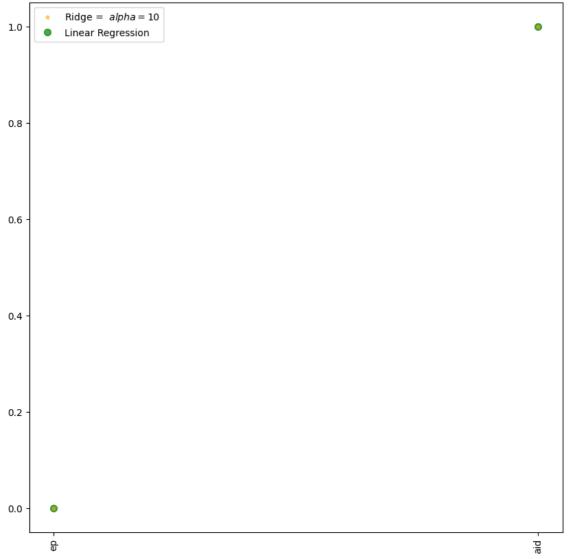
## RidgeCV Method

```
The score of Ridge method is 0.999999996466192
The score of Ridge method is 0.999999996465979
```

# In [21]:

```
plt.figure(figsize=(10,10))
plt.plot(feature,ridge.coef_,alpha=0.5,marker='*',markersize=5,linestyle='None',cole
plt.plot(feature,lr.coef_,alpha=0.7,marker='o',markersize=7,color='green',linestyle
plt.xticks(rotation=90)
plt.title("Comparision between Ridge and linear regression")
plt.legend()
plt.show()
```

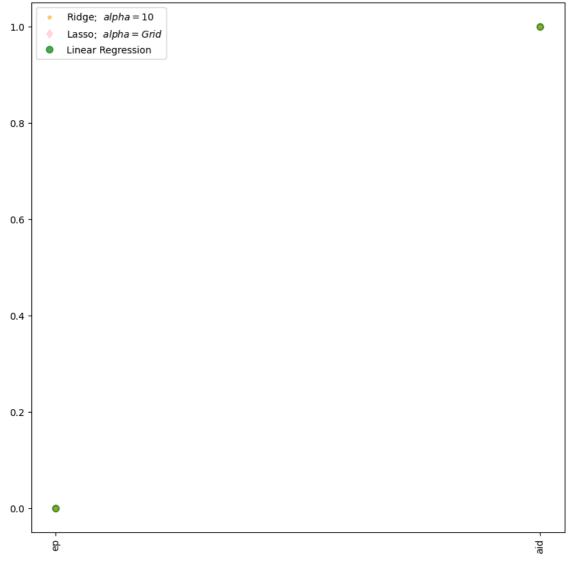
# Comparision between Ridge and linear regression



#### In [22]:

```
plt.figure(figsize=(10,10))
plt.plot(feature,ridge.coef_,alpha=0.5,marker='*',markersize=5,linestyle='None',cole
plt.plot(feature,ridge.coef_,alpha=0.6,marker='d',markersize=6,linestyle='None',cole
plt.plot(feature,lr.coef_,alpha=0.7,marker='o',markersize=7,color='green',linestyle='none',cole
plt.xticks(rotation=90)
plt.title("Comparision between Ridge and linear regression")
plt.legend()
plt.show()
```

#### Comparision between Ridge and linear regression



# In [ ]:

1