

```
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\pappu\Downloads\fiat500_VehicleSelection_Dataset.csv")
df
```

Out[2]:

	ID	model	engine_power	age_in_days	km	previous_owners	lat	lon
0	1	lounge	51	882	25000	1	44.907242	8.611560
1	2	pop	51	1186	32500	1	45.666359	12.241890
2	3	sport	74	4658	142228	1	45.503300	11.417840
3	4	lounge	51	2739	160000	1	40.633171	17.634609
4	5	pop	73	3074	106880	1	41.903221	12.495650
...
1533	1534	sport	51	3712	115280	1	45.069679	7.704920
1534	1535	lounge	74	3835	112000	1	45.845692	8.666870
1535	1536	pop	51	2223	60457	1	45.481541	9.413480
1536	1537	lounge	51	2557	80750	1	45.000702	7.682270
1537	1538	pop	51	1766	54276	1	40.323410	17.568270

1538 rows × 9 columns

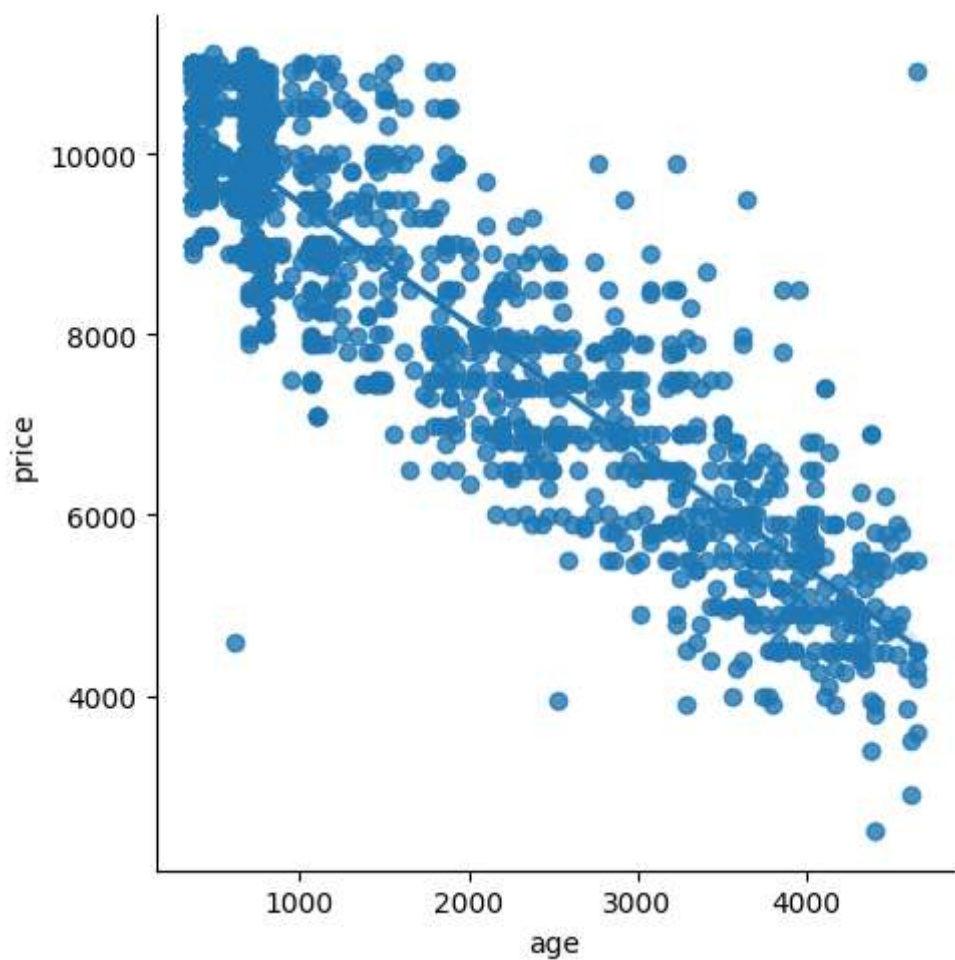
```
In [3]: df=df[['age_in_days','price']]  
df.columns=['age','price']  
df.head(10)
```

Out[3]:

	age	price
0	882	8900
1	1186	8800
2	4658	4200
3	2739	6000
4	3074	5700
5	3623	7900
6	731	10750
7	1521	9190
8	4049	5600
9	3653	6000

```
In [4]: sns.lmplot(x='age',y='price',data=df)
```

Out[4]: <seaborn.axisgrid.FacetGrid at 0x219f8ff90f0>



In [5]: `df.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1538 entries, 0 to 1537
Data columns (total 2 columns):
 #   Column  Non-Null Count  Dtype  
---  -
 0   age      1538 non-null   int64  
 1   price    1538 non-null   int64  
dtypes: int64(2)
memory usage: 24.2 KB
```

In [6]: `x=df[['age']]`
`y=df['price']`

In [7]: `x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2)`

In [8]: `lr=LinearRegression()`

In [9]: `lr.fit(x_train,y_train)`

Out[9]:

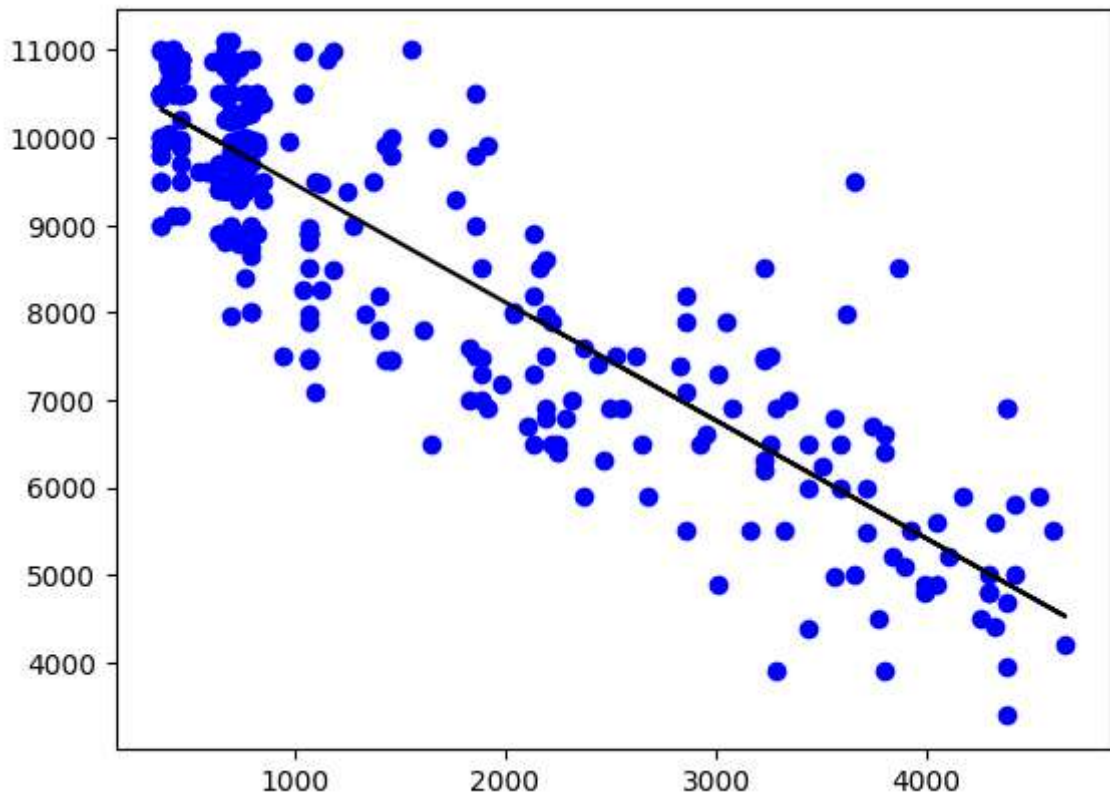
▼ LinearRegression

LinearRegression()

In [10]: `lr.score(x_test,y_test)`

Out[10]: 0.7682404327577492

```
In [11]: y_pred=lr.predict(x_test)
plt.scatter(x_test,y_test,color='b')
plt.plot(x_test,y_pred,color='k')
plt.show()
```



Ridge Regression

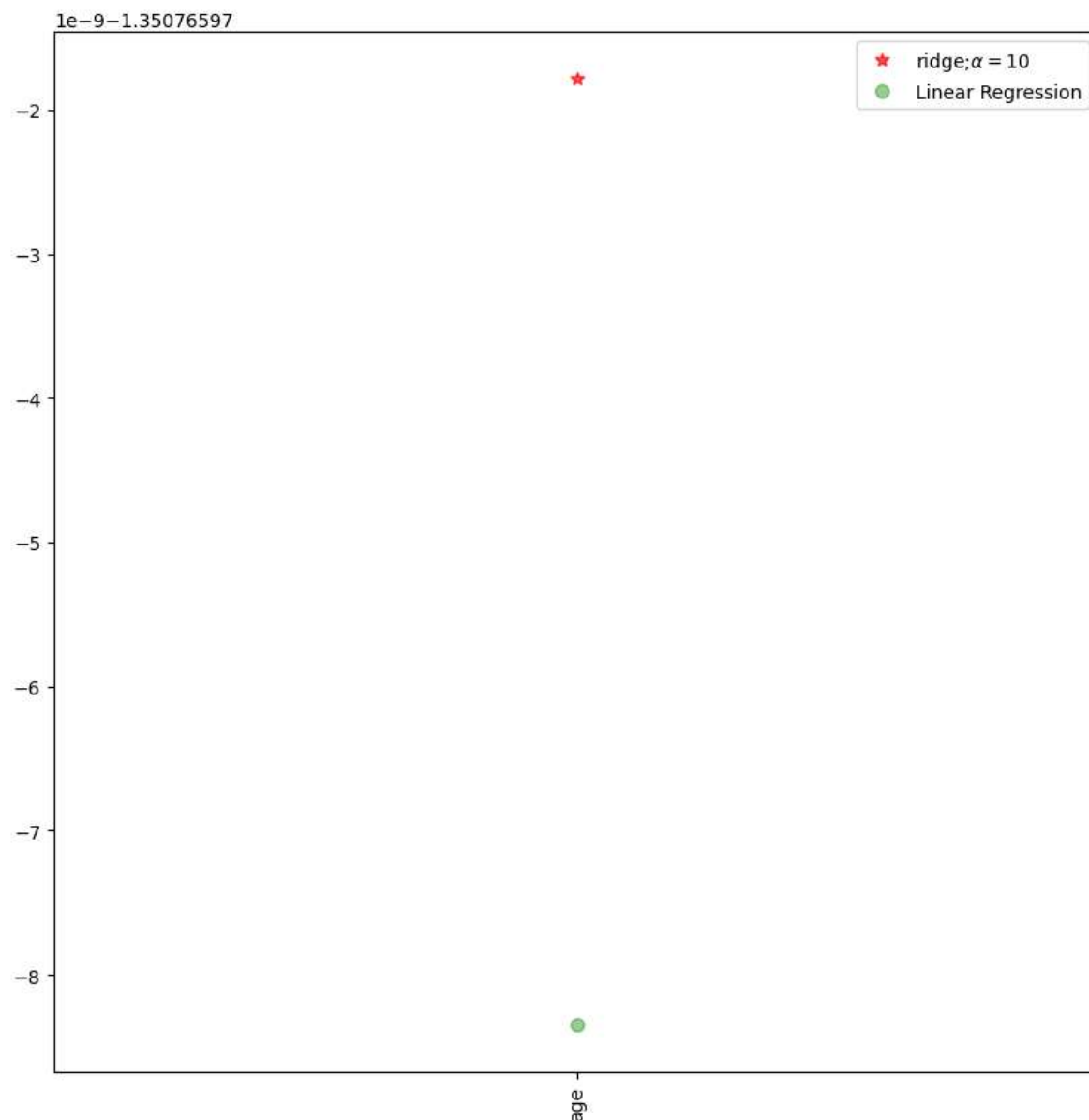
```
In [12]: from sklearn.linear_model import Ridge,RidgeCV
from sklearn.linear_model import Lasso
from sklearn.preprocessing import StandardScaler
```

```
In [13]: ridgereg=Ridge(alpha=10)
ridgereg.fit(x_train,y_train)
train_score_ridge=ridgereg.score(x_train,y_train)
test_score_ridge=ridgereg.score(x_test,y_test)
print('\nRidgeModel:')
print("Train score of Ridge model is {}".format(train_score_ridge))
print("Test score of Ridge model is {}".format(test_score_ridge))
```

```
RidgeModel:
Train score of Ridge model is 0.8050178332327796
Test score of Ridge model is 0.7682404329678096
```

```
In [14]: features=['age']  
target=['price']
```

```
In [15]: plt.figure(figsize=(10,10))  
plt.plot(features,ridgereg.coef_,alpha=0.7,linestyle='none',marker='*',markersize=7,  
plt.plot(features,lr.coef_,alpha=0.4,linestyle='none',marker='o',markersize=7,  
plt.xticks(rotation=90)  
plt.legend()  
plt.show()
```



Lasso Regression

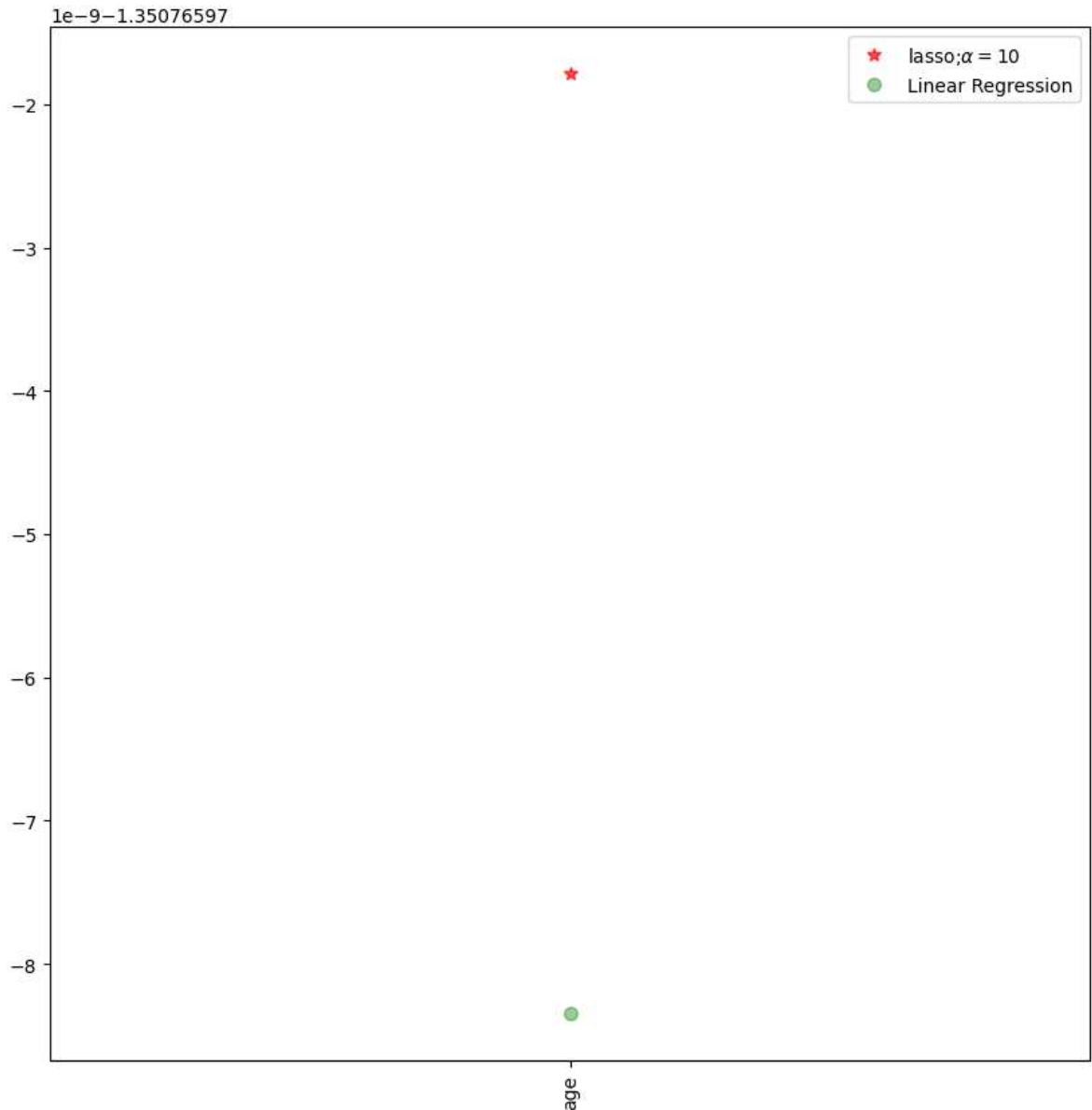
```
In [16]: lasso = Ridge(alpha=10)
lasso.fit(x_train, y_train)
train_score_lasso = lasso.score(x_train, y_train)
test_score_lasso = lasso.score(x_test, y_test)
print('\nLassoModel:')
print("Train score of lasso model is {}".format(train_score_lasso))
print("Test score of lasso model is {}".format(test_score_lasso))
```

LassoModel:

Train score of lasso model is 0.8050178332327796

Test score of lasso model is 0.7682404329678096

```
In [17]: plt.figure(figsize=(10,10))
plt.plot(features,ridgereg.coef_,alpha=0.7,linestyle='none',marker='*',markersize=7,
plt.plot(features,lr.coef_,alpha=0.4,linestyle='none',marker='o',markersize=7,
plt.xticks(rotation=90)
plt.legend()
plt.show()
```



```
In [18]: from sklearn.linear_model import LassoCV
lasso_cv=LassoCV(alphas=[0.0001,0.001,0.01,0.1,1,10],random_state=0).fit(x_train,y_train)
print(lasso_cv.score(x_train,y_train))
print(lasso_cv.score(x_test,y_test))
```

```
0.8050178332170441
0.7682406240217069
```

Elastic Net

```
In [19]: from sklearn.linear_model import ElasticNet
regr=ElasticNet()
regr.fit(x,y)
print(regr.coef_)
print(regr.intercept_)
```

```
[-1.34392217]
10794.7931859617
```

```
In [20]: y_pred_elastic=regr.predict(x_train)
```

```
In [21]: mean_squared_error=np.mean((y_pred_elastic-y_train)**2)
print("Mean Squared Error on test set",mean_squared_error)
```

```
Mean Squared Error on test set 740216.9493362805
```

```
In [ ]:
```