

```
In [3]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

```
In [4]: from sklearn import preprocessing, svm
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
```

```
In [5]: df=pd.read_csv(r"C:\Users\DELL E5490\Downloads\fiat500_VehicleSelection_Dataset.csv")
df
```

Out[5]:

	ID	model	engine_power	age_in_days	km	previous_owners	lat	lon	price
0	1	lounge	51	882	25000	1	44.907242	8.611560	8900
1	2	pop	51	1186	32500	1	45.666359	12.241890	8800
2	3	sport	74	4658	142228	1	45.503300	11.417840	4200
3	4	lounge	51	2739	160000	1	40.633171	17.634609	6000
4	5	pop	73	3074	106880	1	41.903221	12.495650	5700
...	...	...	...	...	...	...	...	...	...
1533	1534	sport	51	3712	115280	1	45.069679	7.704920	5200
1534	1535	lounge	74	3835	112000	1	45.845692	8.666870	4600
1535	1536	pop	51	2223	60457	1	45.481541	9.413480	7500
1536	1537	lounge	51	2557	80750	1	45.000702	7.682270	5990
1537	1538	pop	51	1766	54276	1	40.323410	17.568270	7900

1538 rows × 9 columns

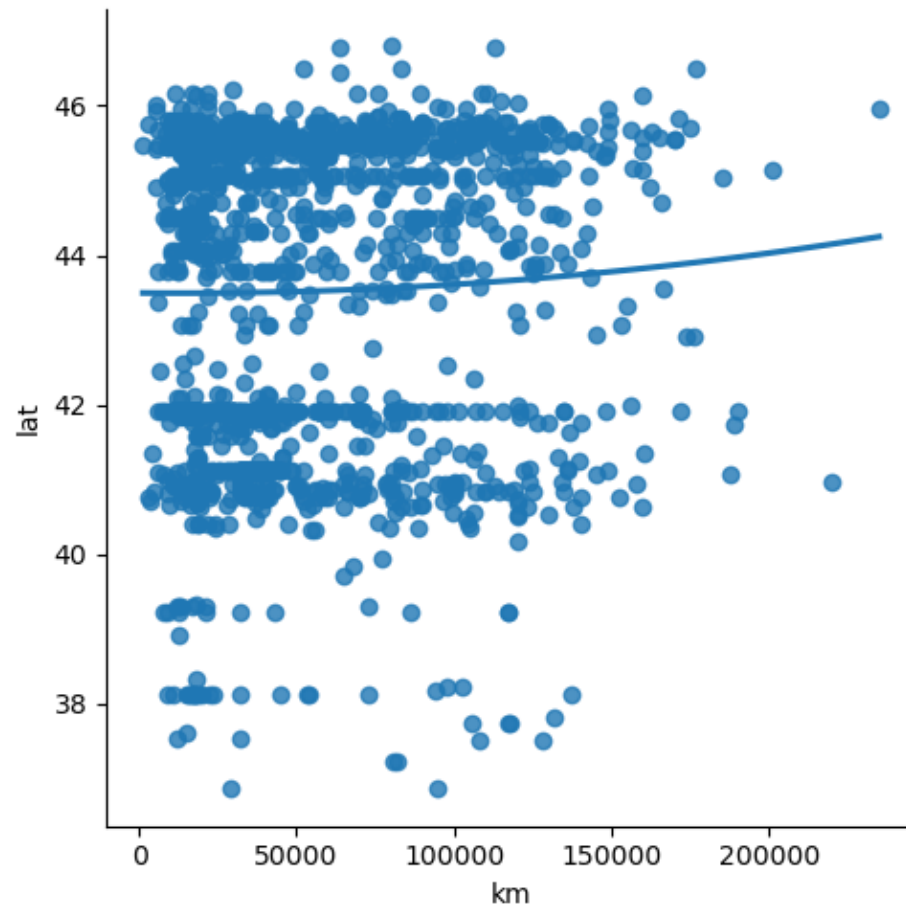
```
In [6]: df=df[['km', 'lat']]  
df.columns=['km', 'lat']  
df.head(10)
```

Out[6]:

	km	lat
0	25000	44.907242
1	32500	45.666359
2	142228	45.503300
3	160000	40.633171
4	106880	41.903221
5	70225	45.000702
6	11600	44.907242
7	49076	41.903221
8	76000	45.548000
9	89000	45.438301

```
In [7]: sns.lmplot(x="km",y="lat",data=df,order=2,ci=None)
```

```
Out[7]: <seaborn.axisgrid.FacetGrid at 0x231deff8940>
```



```
In [8]: df.describe()
```

```
Out[8]:
```

	km	lat
count	1538.000000	1538.000000
mean	53396.011704	43.541361
std	40046.830723	2.133518
min	1232.000000	36.855839
25%	20006.250000	41.802990
50%	39031.000000	44.394096
75%	79667.750000	45.467960
max	235000.000000	46.795612

```
In [9]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1538 entries, 0 to 1537
Data columns (total 2 columns):
#   Column  Non-Null Count  Dtype  
---  -
0    km      1538 non-null    int64  
1    lat      1538 non-null    float64
dtypes: float64(1), int64(1)
memory usage: 24.2 KB
```

```
In [10]: df.fillna(method = 'ffill',inplace = True)
```

C:\Users\DELL E5490\AppData\Local\Temp\ipykernel\_11448\3028625988.py:1: SettingWithCopyWarning:  
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) ([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 [11]: x=np.array(df['km']).reshape(-1,1)
y=np.array(df['lat']).reshape(-1,1)
```

```
In [12]: df.dropna(inplace = True)
```

C:\Users\DELL E5490\AppData\Local\Temp\ipykernel\_11448\1791587065.py:1: SettingWithCopyWarning:  
A value is trying to be set on a copy of a slice from a DataFrame

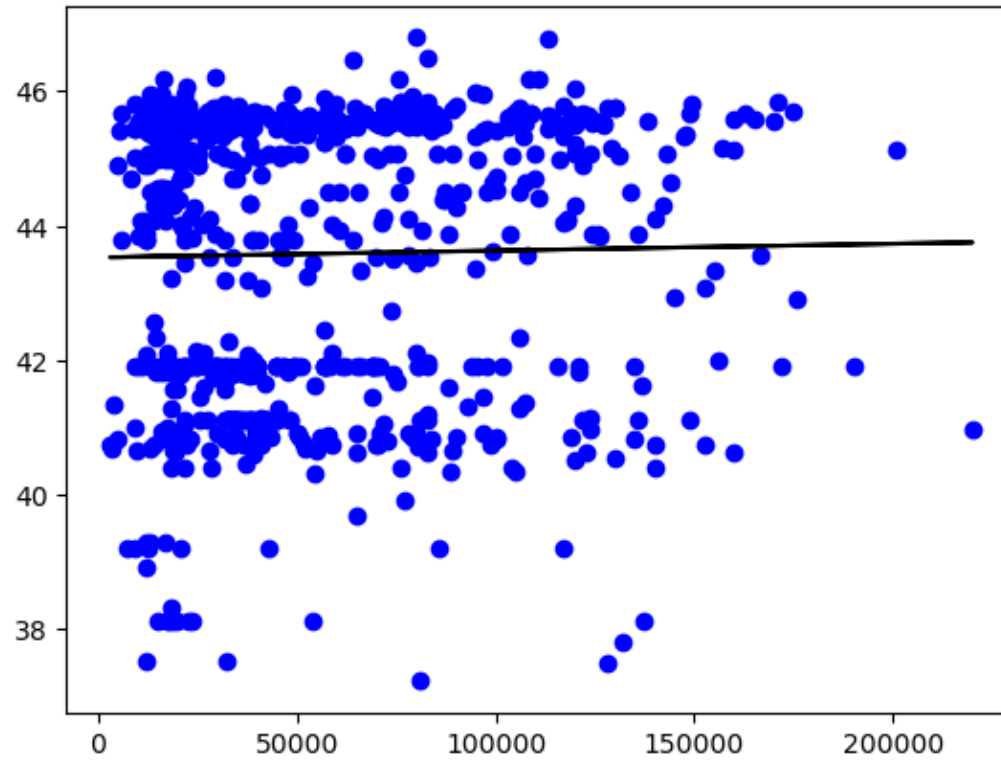
See the caveats in the documentation: [https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy) ([https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy))  
df.dropna(inplace = True)

```
In [13]: x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.5)
```

```
In [14]: regr=LinearRegression()
regr.fit(x_train,y_train)
print(regr.score(x_test,y_test))
```

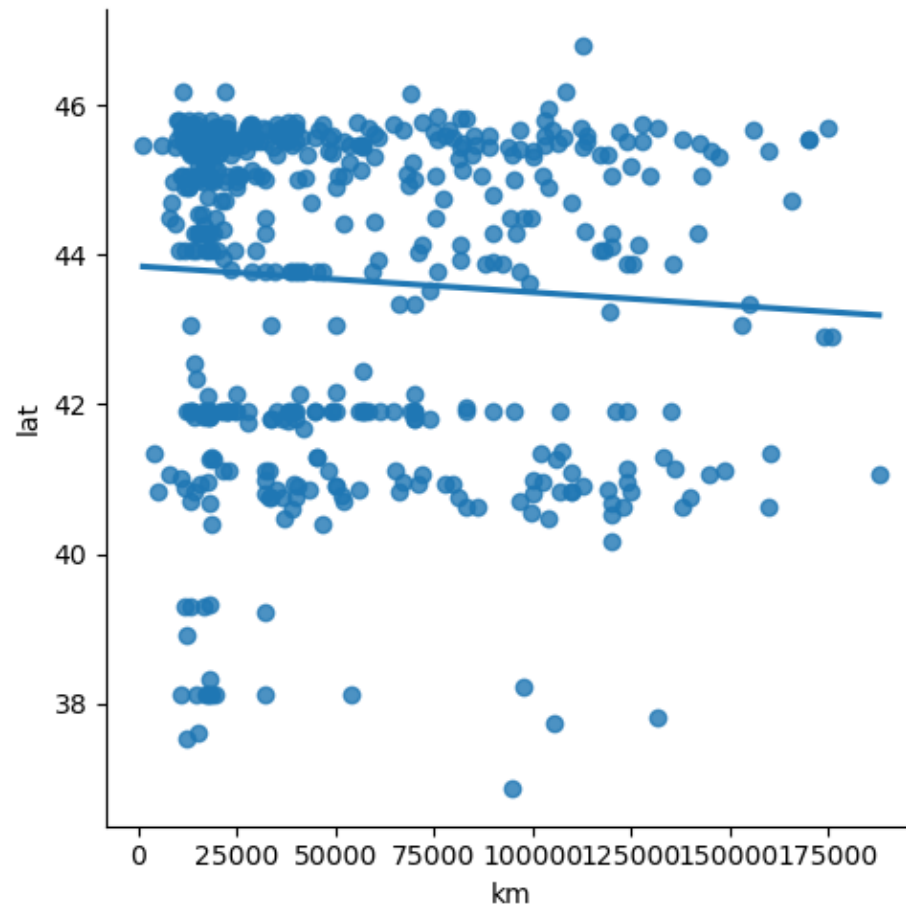
8.749783690176205e-05

```
In [15]: y_pred=regr.predict(x_test)
plt.scatter(x_test,y_test,color='b')
plt.plot(x_test,y_pred,color='k')
plt.show()
```



```
In [16]: df500=df[:500]  
sns.lmplot(x="km",y="lat",data=df500,order=1,ci=None)
```

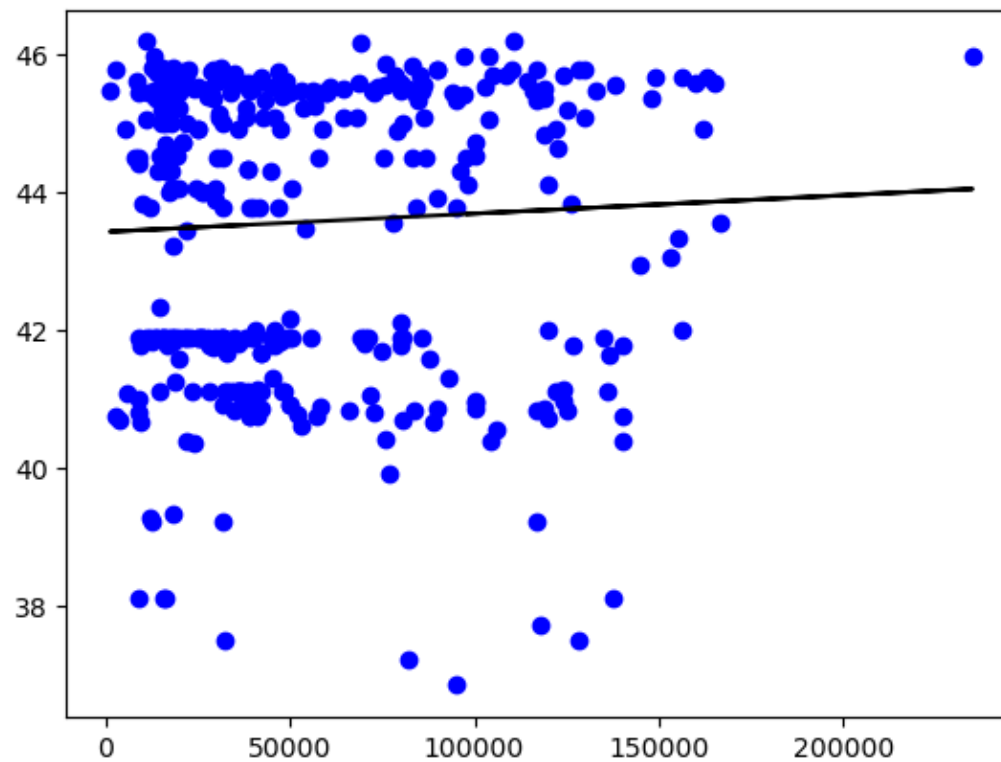
```
Out[16]: <seaborn.axisgrid.FacetGrid at 0x231e0844b50>
```



In [17]:

```
df500.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("Regression:",regr.score(x_test,y_test))
y_pred=regr.predict(x_test)
plt.scatter(x_test,y_test,color='b')
plt.plot(x_test,y_pred,color='k')
plt.show()
```

Regression: -0.00473157491442322





```
In [18]: 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)
```

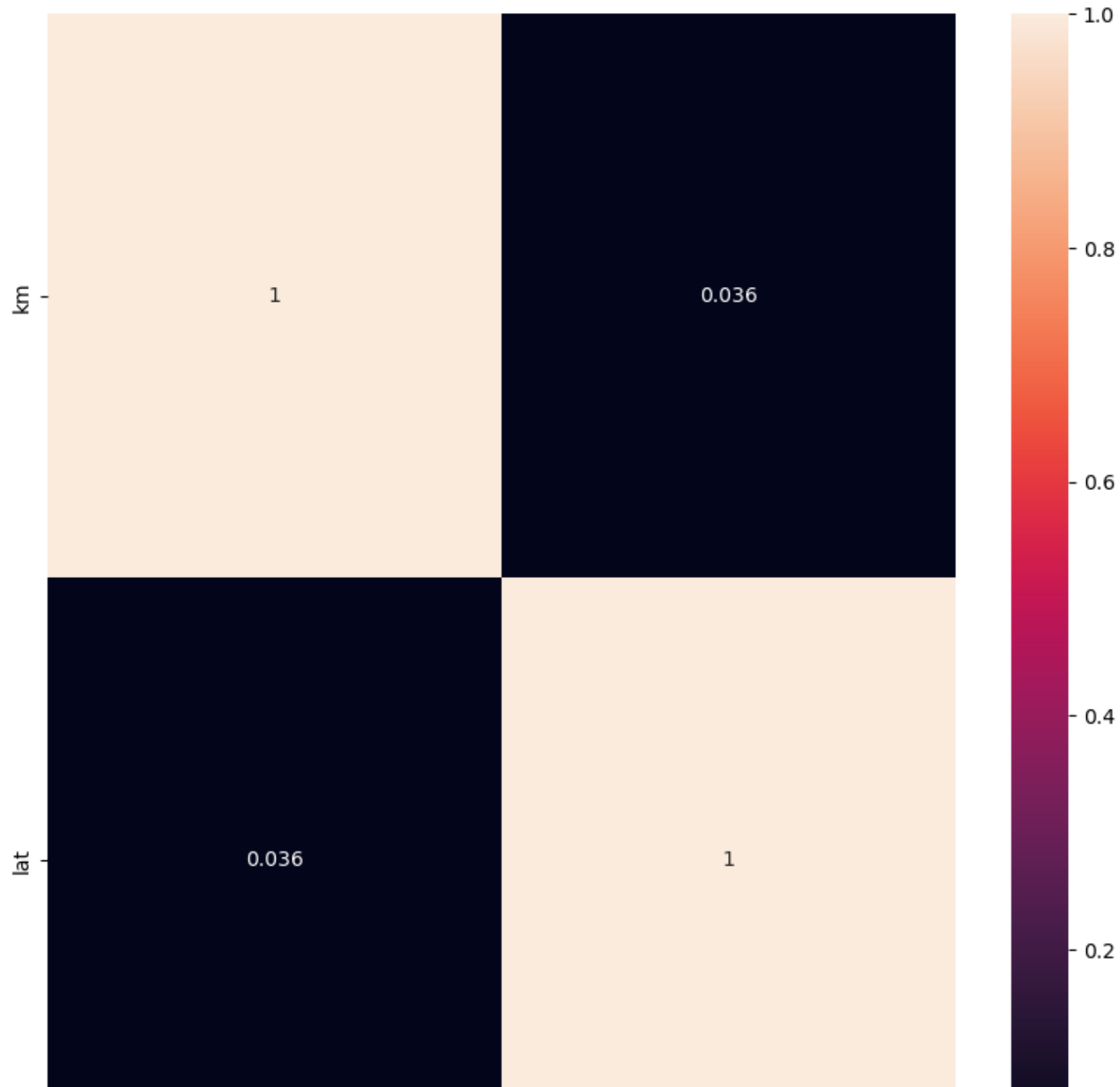
R2.score: -0.00473157491442322

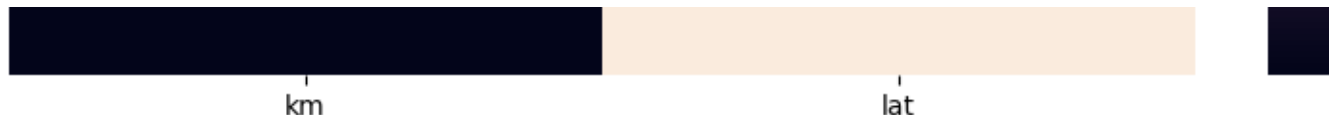
```
In [19]: from sklearn.linear_model import Ridge,RidgeCV,Lasso
from sklearn.preprocessing import StandardScaler
```

```
In [21]: plt.figure(figsize=(10,10))  
sns.heatmap(df.corr(),annot=True)
```

```
Out[21]: <Axes: >
```







```
In [22]: features=df.columns[0:2]
target=df.columns[-1]
#x and y values
x=df[features].values
y=df[target].values
#split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3,random_state=42)
print("The dimension of x_train is {}".format(x_train.shape))
print("The dimension of x_test is {}".format(x_test.shape))
#scale features
scaler=StandardScaler()
x_train=scaler.fit_transform(x_train)
x_test=scaler.transform(x_test)
```

The dimension of x\_train is (1076, 2)  
The dimension of x\_test is (462, 2)

```
In [23]: #model
lr=LinearRegression()
#fit model
lr.fit(x_train,y_train)
#predict
#prediction=lr.predict(x_test)
#actual
actual=y_test
train_score_lr=lr.score(x_train,y_train)
test_score_lr=lr.score(x_test,y_test)
print("\nLinear Regression Model:\n")
print("The train score for lr model is {}".format(train_score_lr))
print("The test score for lr model is {}".format(test_score_lr))
```

Linear Regression Model:

The train score for lr model is 1.0  
The test score for lr model is 1.0

```
In [25]: #ridge regression model
ridgeReg=Ridge(alpha=10)
ridgeReg.fit(x_train,y_train)
#train and test scorefor ridge regression
train_score_ridge=ridgeReg.score(x_train,y_train)
test_score_ridge=ridgeReg.score(x_test,y_test)
print("\nRidge Model:\n")
print("The train score for ridge model is {}".format(train_score_ridge))
print("The test score for ridge model is {}".format(test_score_ridge))
```

Ridge Model:

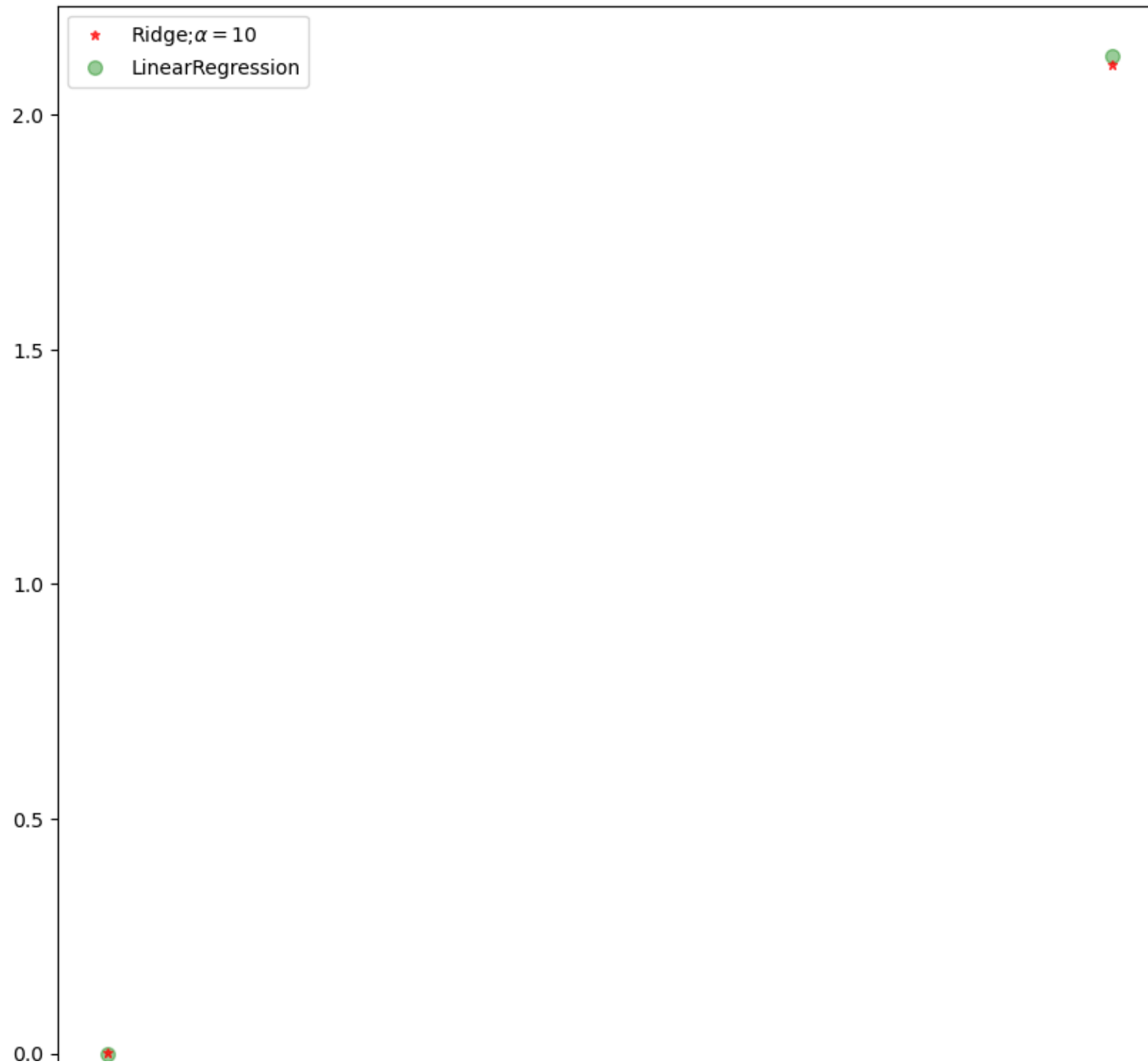
The train score for ridge model is 0.9999149781117884

The test score for ridge model is 0.9999142154121183

```
In [27]: .figure(figsize=(10,10))
         .plot(features,ridgeReg.coef_,alpha=0.7,linestyle='none',marker='*',markersize=5,color='red',label=r'Ridge;$\alpha=10$',zorder=1)
         .plot(features,lr.coef_,alpha=0.4,linestyle='none',marker='o',markersize=7,color='green',label='LinearRegression')
         .xticks(rotation=90)
         .legend()
         .show()
```









```
In [28]: #Lasso regression model
print("\nLasso Model:\n")
lasso=Lasso(alpha=10)
lasso.fit(x_train,y_train)
train_score_ls=lasso.score(x_train,y_train)
test_score_ls=lasso.score(x_test,y_test)
print("The train score for ls model is {}".format(train_score_ls))
print("The test score for ls model is {}".format(test_score_ls))
```

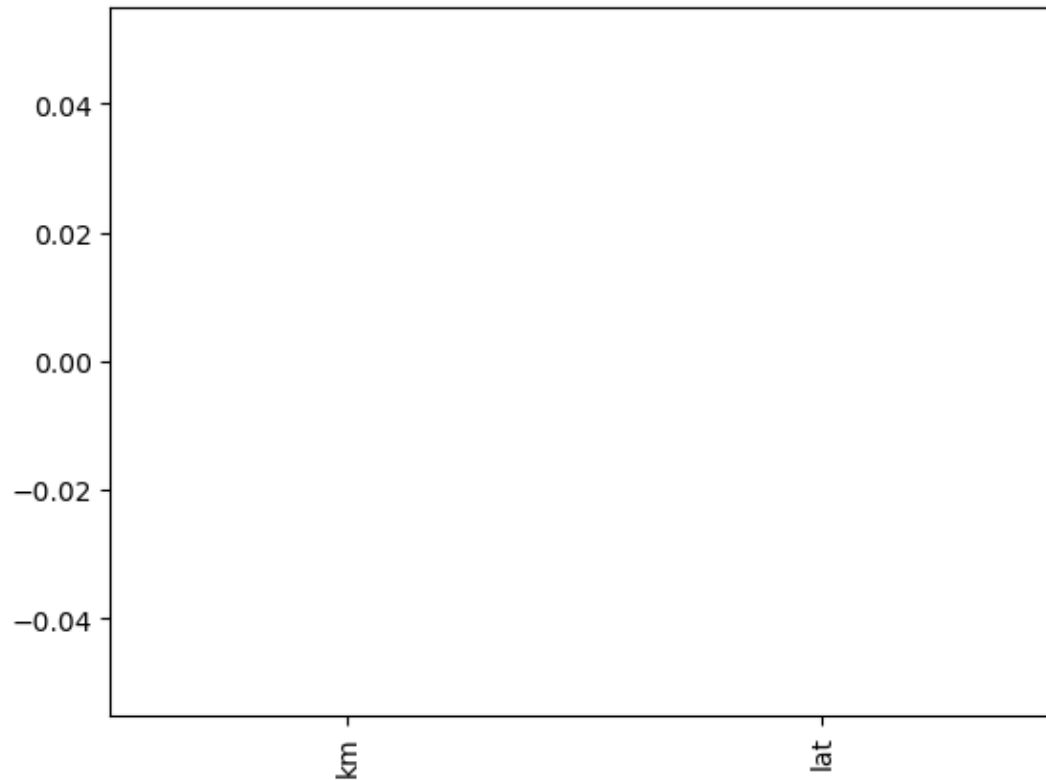
Lasso Model:

The train score for ls model is 0.0

The test score for ls model is -0.0027944198857072777

```
In [29]: pd.Series(lasso.coef_,features).sort_values(ascending=True).plot(kind="bar")
```

```
Out[29]: <Axes: >
```



```
In [30]: #using the linear CV model
from sklearn.linear_model import LassoCV
#Lasso Cross Validation
lasso_cv=LassoCV(alphas=[0.0001,0.001,0.01,0.1,1,10],random_state=0).fit(x_train,y_train)
#score
print(lasso_cv.score(x_train,y_train))
print(lasso_cv.score(x_test,y_test))
```

```
0.999999997786743
0.9999999977805583
```

```
In [33]: ot size
         .figure(figsize=(10,10))
         d plot for ridge regression
         .plot(features,ridgeReg.coef_,alpha=0.7,linestyle='none',marker='*',markersize=5,color='red',label=r'Ridge;$\alpha=10$',zorder=
         .plot(features,ridgeReg.coef_,alpha=0.4,linestyle='none',marker='o',markersize=7,color='green',label='LinearRegression')
         tate axis
         .xticks(rotation=90)
         .legend()
         .title("Comparision plot of Ridge,Lasso and Linear Regression Model")
         .show()
```







```
In [39]: #using the Linear CV model
from sklearn.linear_model import RidgeCV
#ridge Cross Validation
ridge_cv=RidgeCV(alphas=[0.0001,0.001,0.01,0.1,1,10]).fit(x_train,y_train)
#score
print("The train score for ridge model is {}".format(ridge_cv.score(x_train,y_train)))
print("The test score for ridge model is {}".format(ridge_cv.score(x_test,y_test)))
```

The train score for ridge model is 0.9999999999999999  
 The test score for ridge model is 0.9999999999999999

```
In [40]: from sklearn.linear_model import ElasticNet
regr=ElasticNet()
regr.fit(x,y)
print(regr.coef_)
print(regr.intercept_)
```

[3.74911416e-07 8.01713369e-01]  
 8.613651062148548

```
In [41]: y_pred_elastic=regr.predict(x_train)
```

```
In [42]: 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 1219.316769143576

```
In [ ]:
```

