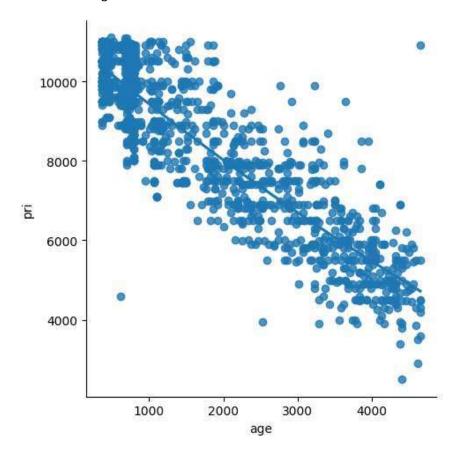
```
In [ ]: ▶
```

LineaRegression-vehicle selection

```
In [19]:
           ▶ # Step-1 :Importing all the required libraries
              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 [24]:
              df=pd.read_csv(r"C:\Users\jyothi reddy\Downloads\fiat500_VehicleSelection_Dataset (1).csv"
    Out[24]:
                          model engine_power age_in_days
                       ID
                                                              km
                                                                   previous_owners
                                                                                         lat
                                                                                                  Ion price
                        1
                          lounge
                                           51
                                                            25000
                                                                                  44.907242
                                                                                              8.611560
                                                                                                       8900
                  0
                                                       882
                        2
                                           51
                                                      1186
                                                            32500
                                                                                   45.666359
                                                                                             12.241890
                                                                                                       8800
                  1
                             pop
                        3
                                           74
                                                      4658
                                                           142228
                                                                                   45.503300
                                                                                             11.417840
                                                                                                       4200
                  2
                            sport
                  3
                                           51
                                                           160000
                                                                                   40.633171
                                                                                                       6000
                        4
                          lounge
                                                      2739
                                                                                            17.634609
                        5
                                           73
                                                      3074
                                                           106880
                                                                                   41.903221
                                                                                             12.495650
                                                                                                       5700
                             qoq
                                                                                              7.704920
               1533 1534
                            sport
                                           51
                                                      3712 115280
                                                                                   45.069679
                                                                                                       5200
                    1535
                                           74
                                                      3835 112000
                                                                                   45.845692
                                                                                              8.666870
                                                                                                       4600
               1534
                          lounge
               1535 1536
                                                      2223
                                                            60457
                                                                                  45.481541
                                                                                              9.413480
                                                                                                       7500
                                           51
                             pop
                                                      2557
                                                            80750
                                                                                   45.000702
               1536
                     1537
                          lounge
                                           51
                                                                                              7.682270
                                                                                                       5990
               1537 1538
                                                      1766
                                                            54276
                                                                                   40.323410 17.568270
                                           51
                             gog
               1538 rows × 9 columns
In [25]:
              df=df[['age_in_days','price']]
              df.columns=['age','pri']
```

In [26]: ▶ sns.lmplot(x="age",y="pri", data = df, order = 2, ci = None)

Out[26]: <seaborn.axisgrid.FacetGrid at 0x133c1803790>



Out[27]:

	age	at
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

```
Untitled5 - Jupyter Notebook

▶ df.describe()
In [28]:
   Out[28]:
                                       at
                          age
              count 1538.000000
                               1538.000000
              mean 1650.980494
                               8576.003901
                std 1289.522278
                               1939.958641
                    366.000000
                               2500.000000
               min
                    670.000000
                               7122.500000
               25%
               50% 1035.000000
                               9000.000000
               75% 2616.000000
                              10000.000000
               max 4658.000000 11100.000000
In [29]:

    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
                  age
                                          int64
              1 pri 1538 non-null int64
             dtypes: int64(2)
             memory usage: 24.2 KB
In [30]:
          C:\Users\jyothi reddy\AppData\Local\Temp\ipykernel_13408\48824337.py:1: SettingWithCopyWa
             rning:
             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_g
             uide/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)
          # Step-5: Training Our Model
In [31]:
             X = np.array(df['age']).reshape(-1, 1)
             y = np.array(df['pri']).reshape(-1, 1)
             #Seperating the data into independent and dependent variables and convert
             #Now each dataset contains only one column
In [32]:

    df.dropna(inplace = True)

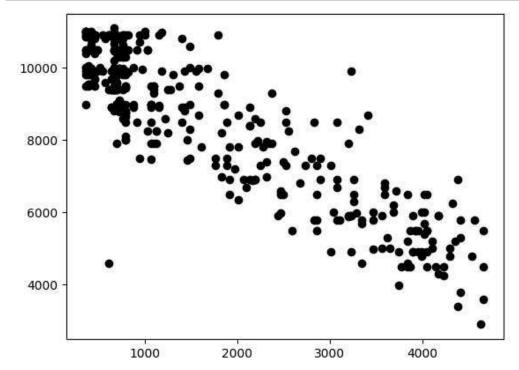
             C:\Users\jyothi reddy\AppData\Local\Temp\ipykernel 13408\1791587065.py:1: SettingWithCopy
             Warning:
             A value is trying to be set on a copy of a slice from a DataFrame
```

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_g uide/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)

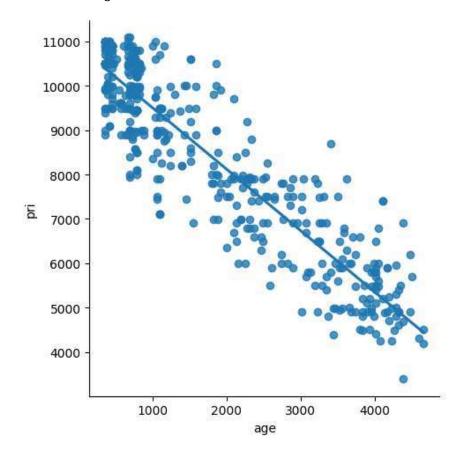
```
In [33]: N X_train,X_test,y_train,y_test = train_test_split(X, y, test_size = 0.25)
# Splitting the data into training data and test data
regr = LinearRegression()
regr.fit(X_train, y_train)
print(regr.score(X_test, y_test))
```

0.7988385840550731



```
In [35]:  # Step-7: Working with a smaller Dataset
df500 = df[:][:500]
# Selecting the 1st 500 rows of the data
sns.lmplot(x ="age", y ="pri", data = df500, order = 1, ci = None)
```

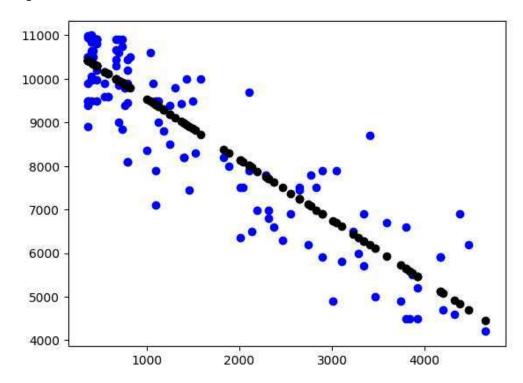
Out[35]: <seaborn.axisgrid.FacetGrid at 0x133c3e7fd50>



```
In [36]: M

df500.fillna(method = 'ffill', inplace = True)
X = np.array(df500['age']).reshape(-1, 1)
y = np.array(df500['pri']).reshape(-1, 1)
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.scatter(X_test, y_pred, color = 'k')
plt.show()
```

Regression: 0.8055415735693708



R2 score: 0.8055415735693708

Step 9-conclusion: Data set we have taken is poor for linear model but with the smaller data works well with Linear model

```
In [ ]: • N
```