In []:	M	
In []:	M	

Linear Regression- USA_Housing

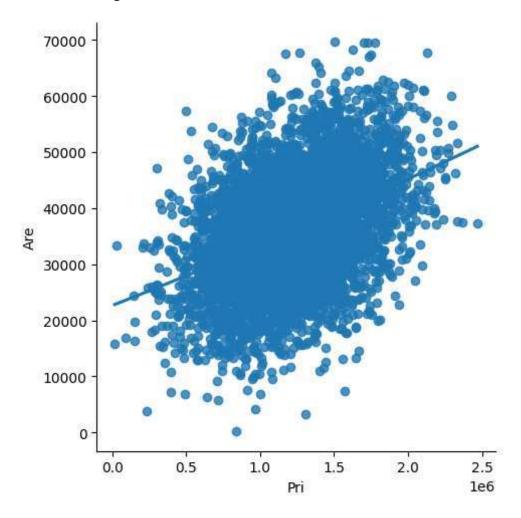
Out[25]:

	Price	Area Population	Avg. Area Number of Bedrooms	Avg. Area Number of Rooms	Avg. Area House Age	Avg. Area Income	
208 Micha 674\nLa	1.059034e+06	23086.800503	4.09	7.009188	5.682861	79545.458574	0
188 Jo Suit Ka	1.505891e+06	40173.072174	3.09	6.730821	6.002900	79248.642455	1
91 Stravenue\	1.058988e+06	36882.159400	5.13	8.512727	5.865890	61287.067179	2
USS Barn	1.260617e+06	34310.242831	3.26	5.586729	7.188236	63345.240046	3
USNS Ray	6.309435e+05	26354.109472	4.23	7.839388	5.040555	59982.197226	4
USNS W AP	1.060194e+06	22837.361035	3.46	6.137356	7.830362	60567.944140	4995
PS 8489\nAP	1.482618e+06	25616.115489	4.02	6.576763	6.999135	78491.275435	4996
4215 T Suite 076\r	1.030730e+06	33266.145490	2.13	4.805081	7.250591	63390.686886	4997
USS Walla	1.198657e+06	42625.620156	5.44	7.130144	5.534388	68001.331235	4998
37778 G€ Apt. 509	1.298950e+06	46501.283803	4.07	6.792336	5.992305	65510.581804	4999
5000 rows × 7 columns							

5000 rows × 7 columns

In [27]: N sns.lmplot(x="Pri",y="Are", data = df, order = 2, ci = None)

Out[27]: <seaborn.axisgrid.FacetGrid at 0x22c0a0bd8d0>



	Pri	Are
0	1.059034e+06	23086.800503
1	1.505891e+06	40173.072174
2	1.058988e+06	36882.159400
3	1.260617e+06	34310.242831
4	6.309435e+05	26354.109472
5	1.068138e+06	26748.428425
6	1.502056e+06	60828.249085
7	1.573937e+06	36516.358972
8	7.988695e+05	29387.396003
	1 2 3 4 5 6 7	 1.059034e+06 1.505891e+06 1.058988e+06 1.260617e+06 6.309435e+05 1.068138e+06 1.502056e+06 1.573937e+06

9 1.545155e+06 40149.965749

```
In [29]:
             df.describe()
    Out[29]:
                            Pri
                                        Are
              count 5.000000e+03
                                 5000.000000
              mean 1,232073e+06 36163,516039
                std 3.531176e+05
                                 9925.650114
                min 1.593866e+04
                                  172.610686
               25% 9.975771e+05 29403.928702
               50% 1.232669e+06 36199.406689
               75% 1.471210e+06 42861.290769
               max 2.469066e+06 69621.713378
In [30]:
             df.info()
             <class 'pandas.core.frame.DataFrame'>
             RangeIndex: 5000 entries, 0 to 4999
             Data columns (total 2 columns):
                  Column Non-Null Count Dtype
              0
                  Pri
                           5000 non-null
                                           float64
              1
                  Are
                           5000 non-null
                                           float64
             dtypes: float64(2)
             memory usage: 78.3 KB
             df.fillna(method ='ffill', inplace = True)
In [31]:
             C:\Users\jyothi reddy\AppData\Local\Temp\ipykernel_11344\48824337.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-d
             ocs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (http
             s://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#return
             ing-a-view-versus-a-copy)
               df.fillna(method ='ffill', inplace = True)
In [32]:
          # Step-5: Training Our Model
             X = np.array(df['Pri']).reshape(-1, 1)
             y = np.array(df['Are']).reshape(-1, 1)
             #Seperating the data into independent and dependent variables and convert
             #Now each dataset contains only one column
```

C:\Users\jyothi reddy\AppData\Local\Temp\ipykernel_11344\1791587065.py:
1: SettingWithCopyWarning:

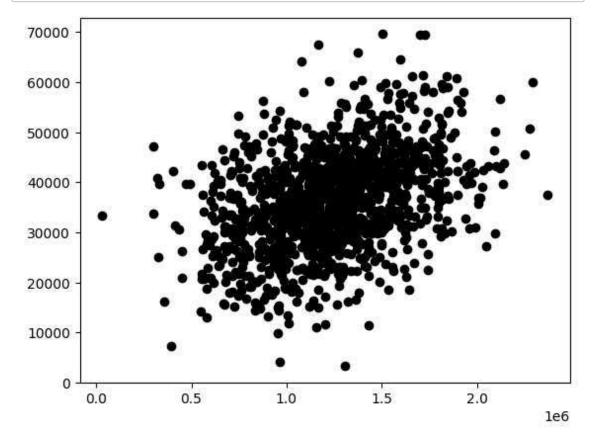
A value is trying to be set on a copy of a slice from a DataFrame

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)

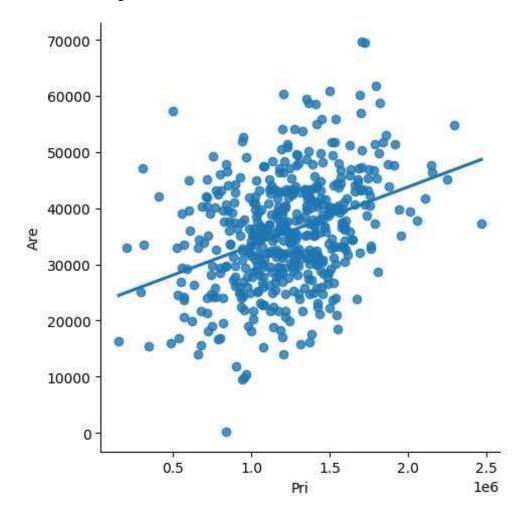
df.dropna(inplace = True)

0.15220040967892345

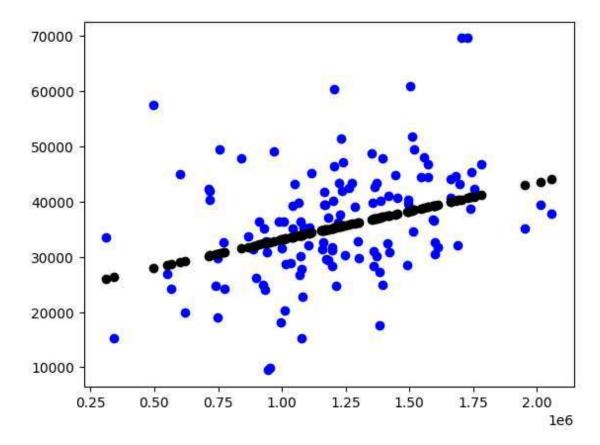
```
In [35]: #step-6: Exploring Our Results
    y_pred = regr.predict(X_test)
    plt.scatter(X_test, y_test, color = 'k')
    plt.show()
# Data scatter of predicted values
```



Out[36]: <seaborn.axisgrid.FacetGrid at 0x22c09ee9190>



Regression: 0.12069840400358822



R2 score: 0.12069840400358822

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

In []: ▶	