

```
In [1]: import numpy as np  
import pandas as pd  
import matplotlib.pyplot as plt
```

```
In [2]: df=pd.read_csv(r"C:\Users\user\Downloads\5_Instagram data.csv")
df.fillna(0,inplace=True)
df
```

Out[2]:

	Impressions	From Home	From Hashtags	From Explore	From Other	Saves	Comments	Shares	Likes	Profile Visits	Follower
0	3920	2586	1028	619	56	98	9	5	162	35	
1	5394	2727	1838	1174	78	194	7	14	224	48	
2	4021	2085	1188	0	533	41	11	1	131	62	
3	4528	2700	621	932	73	172	10	7	213	23	
4	2518	1704	255	279	37	96	5	4	123	8	
...
114	13700	5185	3041	5352	77	573	2	38	373	73	
115	5731	1923	1368	2266	65	135	4	1	148	20	
116	4139	1133	1538	1367	33	36	0	1	92	34	
117	32695	11815	3147	17414	170	1095	2	75	549	148	

	Impressions	From Home	From Hashtags	From Explore	From Other	Saves	Comments	Shares	Likes	Profile Visits	F
118	36919	13473	4176	16444	2547	653	5	26	443	611	

119 rows × 13 columns

In [3]: df.head()

Out[3]:

	Impressions	From Home	From Hashtags	From Explore	From Other	Saves	Comments	Shares	Likes	Profile Visits	Foll
0	3920	2586	1028	619	56	98	9	5	162	35	
1	5394	2727	1838	1174	78	194	7	14	224	48	
2	4021	2085	1188	0	533	41	11	1	131	62	
3	4528	2700	621	932	73	172	10	7	213	23	
4	2518	1704	255	279	37	96	5	4	123	8	

In [4]: `df.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 119 entries, 0 to 118
Data columns (total 13 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Impressions           119 non-null    int64
1   From Home              119 non-null    int64
2   From Hashtags          119 non-null    int64
3   From Explore           119 non-null    int64
4   From Other             119 non-null    int64
5   Saves                  119 non-null    int64
6   Comments               119 non-null    int64
7   Shares                 119 non-null    int64
8   Likes                  119 non-null    int64
9   Profile Visits         119 non-null    int64
10  Follows                119 non-null    int64
11  Caption                 119 non-null    object
12  Hashtags                119 non-null    object
dtypes: int64(11), object(2)
memory usage: 12.2+ KB
```

In [5]: `import seaborn as sns`

In [6]: `df.describe()`

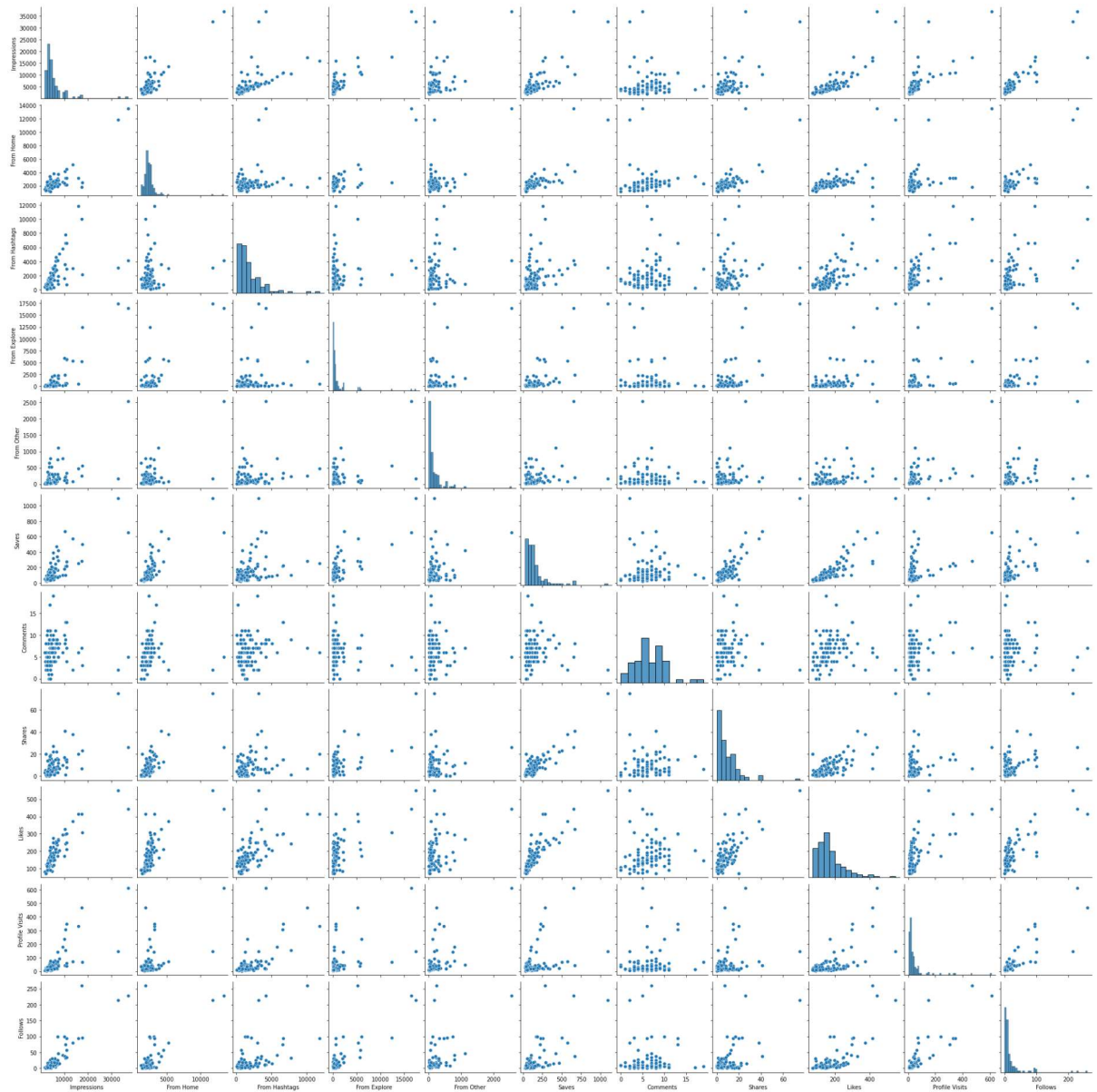
Out[6]:

	Impressions	From Home	From Hashtags	From Explore	From Other	Saves	Comm
count	119.000000	119.000000	119.000000	119.000000	119.000000	119.000000	119.00
mean	5703.991597	2475.789916	1887.512605	1078.100840	171.092437	153.310924	6.66
std	4843.780105	1489.386348	1884.361443	2613.026132	289.431031	156.317731	3.54
min	1941.000000	1133.000000	116.000000	0.000000	9.000000	22.000000	0.00
25%	3467.000000	1945.000000	726.000000	157.500000	38.000000	65.000000	4.00
50%	4289.000000	2207.000000	1278.000000	326.000000	74.000000	109.000000	6.00
75%	6138.000000	2602.500000	2363.500000	689.500000	196.000000	169.000000	8.00
max	36919.000000	13473.000000	11817.000000	17414.000000	2547.000000	1095.000000	19.00

In [7]: `df=pd.read_csv("5_Instagram data.csv")`

```
In [8]: sns.pairplot(df)
```

```
Out[8]: <seaborn.axisgrid.PairGrid at 0x26f4590ea90>
```

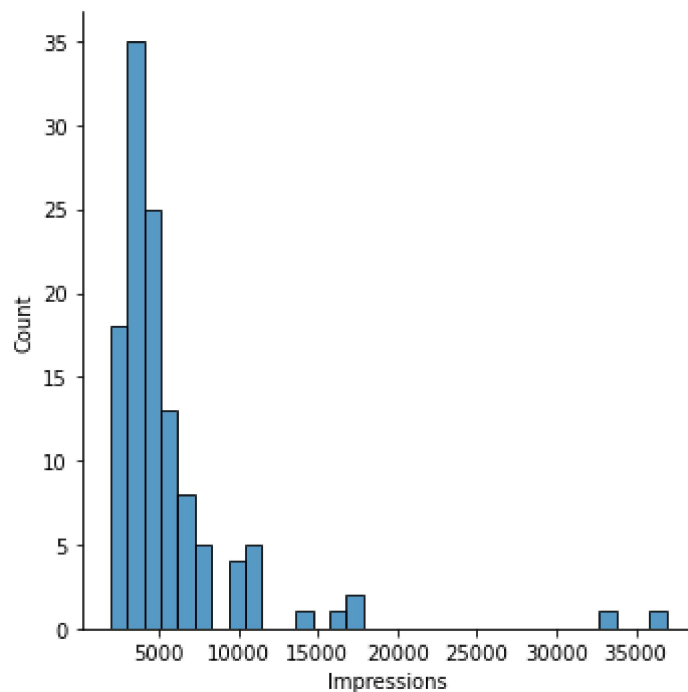


```
In [9]: df1=df.drop(['Comments'],axis=1)
df1
df1=df1.drop(df1.index[1537:])
df1.isna().sum()
```

```
Out[9]: Impressions      0
From Home      0
From Hashtags  0
From Explore   0
From Other     0
Saves          0
Shares         0
Likes          0
Profile Visits 0
Follows        0
Caption        0
Hashtags       0
dtype: int64
```

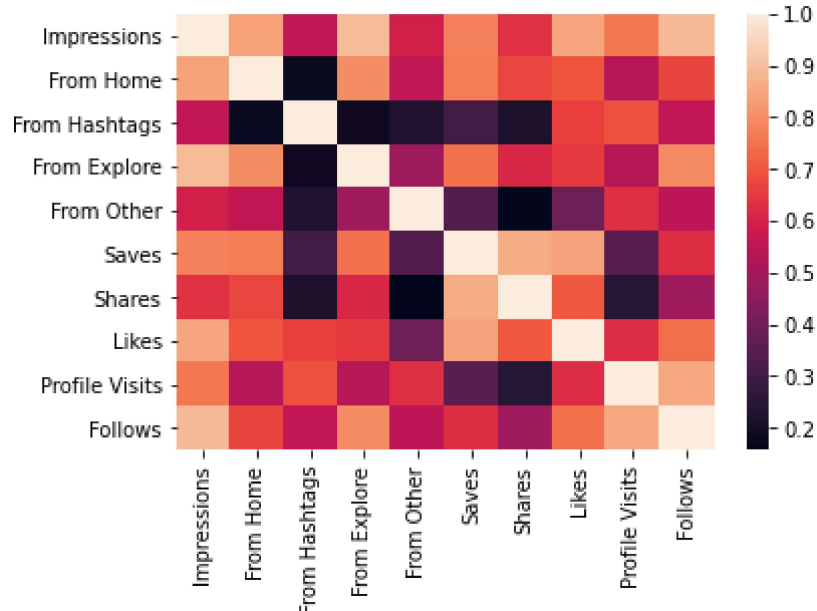
```
In [10]: sns.displot(df['Impressions'])
```

```
Out[10]: <seaborn.axisgrid.FacetGrid at 0x26f4a488910>
```



```
In [11]: sns.heatmap(df1.corr())
```

```
Out[11]: <AxesSubplot:>
```



```
In [12]: from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
```

```
In [13]: df1.isna().sum()
```

```
Out[13]: Impressions      0
From Home      0
From Hashtags  0
From Explore   0
From Other     0
Saves          0
Shares         0
Likes         0
Profile Visits 0
Follows        0
Caption        0
Hashtags       0
dtype: int64
```



```
In [14]: y=df1['Likes']
x=df1.drop(['Caption','Hashtags'],axis=1)
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
print(x_train)
```

	Impressions	From Home	From Hashtags	From Explore	From Other	Saves	\
35	2523	1659	796	29	21	34	
72	3606	2509	183	446	300	174	
83	4002	3401	278	128	73	111	
61	6339	2190	4036	48	27	171	
24	4628	2406	1260	861	26	144	
..	
0	3920	2586	1028	619	56	98	
52	2941	1716	1058	84	48	48	
7	3541	2071	628	500	60	135	
64	7571	3717	841	1716	1115	421	
74	6559	2225	4041	158	72	179	

	Shares	Likes	Profile Visits	Follows
35	0	86	4	2
72	15	138	17	12
83	18	205	16	2
61	5	248	21	10
24	3	160	10	4
..
0	5	162	35	2
52	1	99	12	4
7	9	124	12	6
64	12	269	50	46
74	6	257	22	12

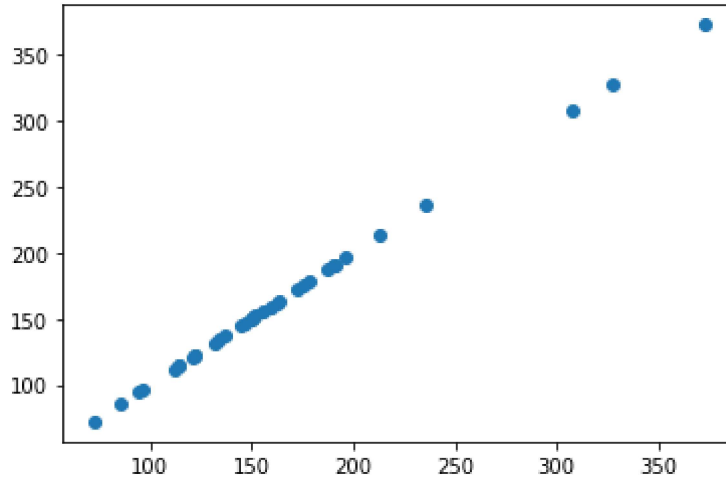
[83 rows x 10 columns]

```
In [15]: model=LinearRegression()
model.fit(x_train,y_train)
model.intercept_
```

Out[15]: -7.105427357601002e-13

```
In [16]: prediction=model.predict(x_test)
plt.scatter(y_test,prediction)
```

Out[16]: <matplotlib.collections.PathCollection at 0x26f4e0e5670>



```
In [17]: model.score(x_test,y_test)
```

Out[17]: 1.0

```
In [18]: from sklearn.linear_model import Ridge,Lasso
```

```
In [19]: rr=Ridge(alpha=10)
rr.fit(x_train,y_train)
```

Out[19]: Ridge(alpha=10)

```
In [20]: rr.score(x_test,y_test)
```

Out[20]: 0.9999999935804778

```
In [21]: la =Lasso(alpha=10)
la.fit(x_train,y_train)
```

Out[21]: Lasso(alpha=10)

```
In [22]: la.score(x_test,y_test)
```

Out[22]: 0.9999845944911459

```
In [23]: from sklearn.linear_model import ElasticNet
en=ElasticNet()
en.fit(x_train,y_train)
print(en.coef_)
print(en.intercept_)
print(en.predict(x_test))
print(en.score(x_test,y_test))
from sklearn import metrics
print("Mean Absolute Error:",metrics.mean_absolute_error(y_test,prediction))
print("Mean Squared Error:",metrics.mean_squared_error(y_test,prediction))
print("Root Mean Squared Error:",np.sqrt(metrics.mean_squared_error(y_test,prediction)))
```

```
[-2.73324381e-04  3.08785864e-04  3.28143840e-04  2.65893595e-04
 2.82219150e-04  7.65557477e-04  0.00000000e+00  9.97521227e-01
-0.00000000e+00  0.00000000e+00]
```

```
0.1545091620847927
```

```
[328.14083722 121.0123474 236.3265904 121.99532989 195.97189891
114.03007253 162.96346015 189.951583 186.94081757 151.99972609
113.98532621 72.0612509 149.99184938 132.01725967 122.08085335
171.97136112 94.05261907 133.95819648 178.02259982 112.04785196
144.97808057 85.05833179 174.94968083 307.8467465 147.07154204
190.87793596 372.96778075 149.99398787 137.06604274 155.99257904
161.93446632 96.07518261 158.96005402 150.96164369 151.02266212
212.82650758]
```

```
0.9999982366421484
```

```
Mean Absolute Error: 6.663311877572495e-13
```

```
Mean Squared Error: 7.599766755251328e-25
```

```
Root Mean Squared Error: 8.717664111016969e-13
```