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

Out[2]:

	Impressions	From Home	From Hashtags	From Explore	From Other	Saves	Comments	Shares	Likes	Profile Visits	F
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.describe()

Out[3]:

	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 [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 [6]: df1=df.head(100)
df1

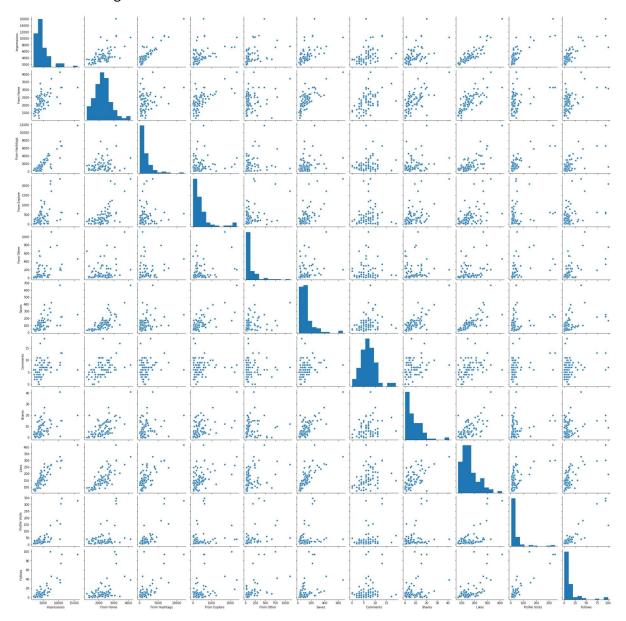
Out[6]:

	Impressions	From Home	From Hashtags	From Explore	From Other	Saves	Comments	Shares	Likes	Profile Visits	Fo
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	
95	5394	2275	2975	45	65	61	19	6	147	69	
96	2766	2541	116	51	9	40	10	4	114	11	
97	3924	2244	1278	326	34	139	11	3	151	19	
98	3015	2034	771	115	41	52	11	4	92	9	
99	5409	2643	2006	1068	230	393	10	27	275	38	

100 rows × 13 columns

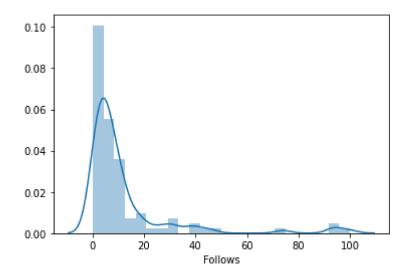
In [7]: sns.pairplot(df1)

Out[7]: <seaborn.axisgrid.PairGrid at 0x2b784c7eee0>



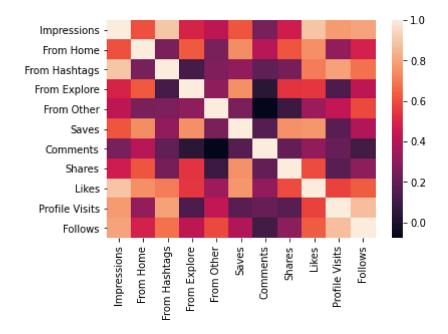
```
In [8]: | sns.distplot(df1['Follows'])
```

Out[8]: <matplotlib.axes._subplots.AxesSubplot at 0x2b7897bf8b0>



In [9]: sns.heatmap(df1.corr())

Out[9]: <matplotlib.axes._subplots.AxesSubplot at 0x2b788b8ab80>



```
In [11]: from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.3)
```

```
In [12]: | from sklearn.linear_model import LinearRegression
         lr= LinearRegression()
         lr.fit(x_train,y_train)
Out[12]: LinearRegression()
         prediction= lr.predict(x_test)
In [13]:
         plt.scatter(y_test,prediction)
Out[13]: <matplotlib.collections.PathCollection at 0x2b78bbf3940>
          50
          40
          30
          20
          10
                      20
                               40
                                       60
                                               80
                                                       100
In [14]:
         print(lr.intercept_)
          -9.620335915442526
In [15]:
         print(lr.score(x_train,y_train))
         0.9510973612275092
In [16]: from sklearn.linear model import Ridge,Lasso
In [17]: rr=Ridge(alpha=10)
         rr.fit(x_train,y_train)
Out[17]: Ridge(alpha=10)
In [18]: |rr.score(x_test,y_test)
Out[18]: 0.7050821670628811
```

```
In [19]: la=Lasso(alpha=10)
         la.fit(x_train,y_train)
         C:\Users\Admin\anaconda3\lib\site-packages\sklearn\linear_model\_coordinate_d
         escent.py:529: ConvergenceWarning: Objective did not converge. You might want
         to increase the number of iterations. Duality gap: 3.4795880848457728, tolera
         nce: 2.430314285714286
           model = cd fast.enet coordinate descent(
Out[19]: Lasso(alpha=10)
In [20]: la.score(x_test,y_test)
Out[20]: 0.7124336007520684
In [21]: from sklearn.linear_model import ElasticNet
         en=ElasticNet()
         en.fit(x train,y train)
         C:\Users\Admin\anaconda3\lib\site-packages\sklearn\linear model\ coordinate d
         escent.py:529: ConvergenceWarning: Objective did not converge. You might want
         to increase the number of iterations. Duality gap: 337.5874168339527, toleran
         ce: 2.430314285714286
           model = cd fast.enet coordinate descent(
Out[21]: ElasticNet()
In [22]: |print(en.intercept )
         -9.164250297735885
In [23]: |print(en.predict(x test))
         [51.07863436 -1.56964157
                                   5.24247963 9.28851698 21.2550892
                                                                        5.48816099
           9.38940229 -3.17578915 53.19370088 7.64379639 1.25053779 2.37106854
           2.30884024 6.89068846 8.18485441 8.88515564 8.43948527 21.35807918
          -3.17578915   0.79890081   3.66524313   15.23829695   0.38069227   11.85622776
           6.49074679 6.5888629 15.61559998 11.97125521 1.40367014 7.39040191
In [24]: |print(en.score(x_test,y_test))
         0.7078054633636812
```

Evalution

```
In [25]: from sklearn import metrics
In [26]: print("Mean Absolute Error", metrics.mean_absolute_error(y_test, prediction))
```

Mean Absolute Error 5.500448590180688