DATA COLLECTION ¶

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

```
In [2]: # To Import Dataset
sd=pd.read_csv(r"c:\Users\user\Downloads\Instagram.csv")
sd
```

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	

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]: # to display top 10 rows
 sd.head(10)

Out	[3]:		

[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	
	5	3884	2046	1214	329	43	74	7	10	144	9	
	6	2621	1543	599	333	25	22	5	1	76	26	
	7	3541	2071	628	500	60	135	4	9	124	12	
	8	3749	2384	857	248	49	155	6	8	159	36	
	9	4115	2609	1104	178	46	122	6	3	191	31	

DATA CLEANING AND PRE PROCESSING

```
In [4]: | sd.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 119 entries, 0 to 118
         Data columns (total 13 columns):
              Column
                               Non-Null Count
                                                 Dtvpe
                               -----
              ____
                                                 int64
          0
              Impressions
                               119 non-null
                               119 non-null
          1
              From Home
                                                 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
              Shares
          7
                              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]: # to display summary of statistics
         sd.describe()
Out[5]:
                                               From
                                                            From
                 Impressions
                              From Home
                                                                  From Other
                                                                                  Saves Comm
                                                          Explore
                                            Hashtags
          count
                  119.000000
                              119.000000
                                           119.000000
                                                       119.000000
                                                                   119.000000
                                                                               119.000000 119.00
                 5703.991597
                             2475.789916
                                          1887.512605
                                                      1078.100840
                                                                              153.310924
          mean
                                                                   171.092437
                                                                                           6.66
                 4843.780105
                                                                              156.317731
            std
                             1489.386348
                                          1884.361443
                                                      2613.026132
                                                                   289.431031
                                                                                           3.54
                                                                                           0.00
           min
                 1941.000000
                             1133.000000
                                           116.000000
                                                         0.000000
                                                                    9.000000
                                                                               22.000000
           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
                36919.000000 13473.000000 11817.000000 17414.000000 2547.000000
                                                                             1095.000000
                                                                                          19.00
```

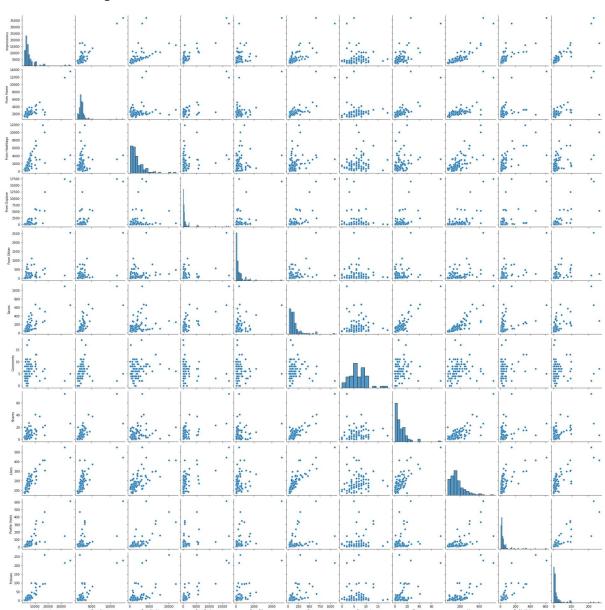
In [6]:

#to display colums heading

EDA and visualization

In [7]: sns.pairplot(sd)

Out[7]: <seaborn.axisgrid.PairGrid at 0x24a532f3250>

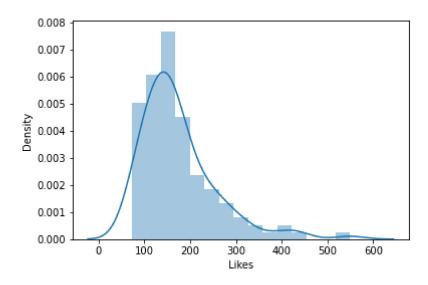


```
In [8]: sns.distplot(sd['Likes'])
```

C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: Fut ureWarning: `distplot` is a deprecated function and will be removed in a futu re version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for hi stograms).

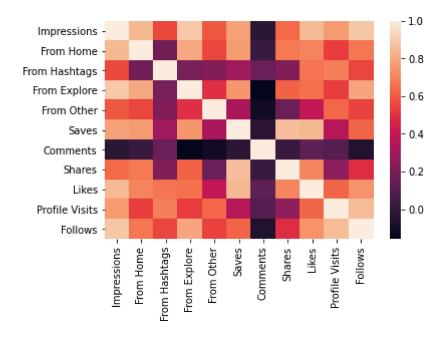
warnings.warn(msg, FutureWarning)

Out[8]: <AxesSubplot:xlabel='Likes', ylabel='Density'>



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

Out[9]: <AxesSubplot:>



TO TRAIN THE MODEL _MODEL BUILDING

we are goint train Liner Regression model; we need to split out the data into two varibles x and y where x is independent on x (output) and y is dependent on x(output) adress coloumn as it is not required our model

```
In [16]: x= ssd[['Impressions', 'From Home', 'From Hashtags', 'From Explore',
                 'From Other', 'Saves', 'Comments', 'Shares']]
         y=ssd['Likes']
In [17]: # To split my dataset into training data and test data
         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 [18]: from sklearn.linear_model import LinearRegression
         lr=LinearRegression()
         lr.fit(x train,y train)
Out[18]: LinearRegression()
In [19]: from sklearn.linear model import LinearRegression
         lr=LinearRegression()
         lr.fit(x_train,y_train)
Out[19]: LinearRegression()
In [20]: |print(lr.intercept_)
          -0.1667758131544872
         coeff= pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
In [21]:
Out[21]:
                        Co-efficient
            Impressions
                          0.012795
             From Home
                          0.028980
          From Hashtags
                          0.004934
            From Explore
                         -0.011974
             From Other
                         -0.031578
                  Saves
                          0.315383
              Comments
                         -0.298466
                 Shares
                         -0.040502
```

```
In [22]: | prediction = lr.predict(x_test)
         plt.scatter(y_test,prediction)
Out[22]: <matplotlib.collections.PathCollection at 0x24a59d26f40>
          450
          400
          350
          300
          250
          200
          150
          100
                   100
                           150
                                   200
                                           250
                                                    300
In [23]: |print(lr.score(x_test,y_test))
         0.730665299562944
In [24]: |lr.score(x_train,y_train)
Out[24]: 0.9309472995791541
In [25]: from sklearn.linear_model import Ridge,Lasso
In [26]: dr=Ridge(alpha=10)
         dr.fit(x_train,y_train)
Out[26]: Ridge(alpha=10)
In [27]: | dr.score(x_test,y_test)
Out[27]: 0.7307134654321605
In [28]: |dr.score(x_train,y_train)
Out[28]: 0.9309472555637033
In [29]: la=Lasso(alpha=10)
         la.fit(x_train,y_train)
Out[29]: Lasso(alpha=10)
In [30]: la.score(x_test,y_test)
Out[30]: 0.7365075747005545
```

```
In [31]: la.score(x_train,y_train)
Out[31]: 0.9306906287698803
In [ ]:
```