EDA with Data Collection, Data Cleaning and Pre- processing with

Instagram dataset

Data cleaning and Pre-procrossing

To import library

In [1]:

import numpy as np
import pandas as pd

To import dataset

```
In [2]:
```

d=pd.read_csv(r"c:\Users\user\Downloads\5_instagram data.csv")
d

Out[2]:

	Impressions	From Home	From Hashtags	From Explore	From Other	Saves	Comments	Shares	Likes	Profile Visits
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	2:
4	2518	1704	255	279	37	96	5	4	123	{
114	13700	5185	3041	5352	77	573	2	38	373	73
115	5731	1923	1368	2266	65	135	4	1	148	2(
116	4139	1133	1538	1367	33	36	0	1	92	34
117	32695	11815	3147	17414	170	1095	2	75	549	148
118	36919	13473	4176	16444	2547	653	5	26	443	611

119 rows × 13 columns

To get top 10 record

In [3]:

d.head(10)

Out[3]:

	Impressions	From Home	From Hashtags	From Explore	From Other	Saves	Comments	Shares	Likes	Profile Visits
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
						_				

To get last 10

In [4]:

d.tail(10)

Out[4]:

	Impressions	From Home	From Hashtags	From Explore	From Other	Saves	Comments	Shares	Likes	Profile Visits
109	17713	2449	2141	12389	561	504	3	23	308	7(
110	5563	3813	362	1135	76	149	5	8	163	22
111	4842	1658	694	2036	310	55	6	4	86	4(
112	11149	4439	747	5762	53	273	4	13	210	6′
113	10206	2371	1624	6000	117	182	10	17	172	237
114	13700	5185	3041	5352	77	573	2	38	373	7;
115	5731	1923	1368	2266	65	135	4	1	148	2(
116	4139	1133	1538	1367	33	36	0	1	92	34
117	32695	11815	3147	17414	170	1095	2	75	549	148
118	36919	13473	4176	16444	2547	653	5	26	443	61 [,]
4										•

To describe statistics Analysis

In [5]:

d.describe()

Out[5]:

	Impressions	From Home	From Hashtags	From Explore	From Other	Saves	Co
count	119.000000	119.000000	119.000000	119.000000	119.000000	119.000000	11
mean	5703.991597	2475.789916	1887.512605	1078.100840	171.092437	153.310924	
std	4843.780105	1489.386348	1884.361443	2613.026132	289.431031	156.317731	
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	
50%	4289.000000	2207.000000	1278.000000	326.000000	74.000000	109.000000	
75%	6138.000000	2602.500000	2363.500000	689.500000	196.000000	169.000000	
max	36919.000000	13473.000000	11817.000000	17414.000000	2547.000000	1095.000000	1
4							

To get rows and columns

In [6]:

np.shape(d)

Out[6]:

(119, 13)

To get number of elements

In [7]:

np.size(d)

Out[7]:

1547

To get the missing value

In [8]:

d.isna()

Out[8]:

	Impressions	From Home	From Hashtags	From Explore	From Other	Saves	Comments	Shares	Likes	Profile Visits
0	False	False	False	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False	False	False	Fals€
3	False	False	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False	False	Fals€
	•••									
114	False	False	False	False	False	False	False	False	False	False
115	False	False	False	False	False	False	False	False	False	Fals€
116	False	False	False	False	False	False	False	False	False	False
117	False	False	False	False	False	False	False	False	False	False
118	False	False	False	False	False	False	False	False	False	False
119 r	ows × 13 colu	ımns								

To drop the missing elements

In [9]:

d.dropna(axis=1,how='any')

Out[9]:

	Impressions	From Home	From Hashtags	From Explore	From Other	Saves	Comments	Shares	Likes	Profile Visits
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	2:
4	2518	1704	255	279	37	96	5	4	123	{
114	13700	5185	3041	5352	77	573	2	38	373	73
115	5731	1923	1368	2266	65	135	4	1	148	2(
116	4139	1133	1538	1367	33	36	0	1	92	34
117	32695	11815	3147	17414	170	1095	2	75	549	148
118	36919	13473	4176	16444	2547	653	5	26	443	611

119 rows × 13 columns

```
In [10]:
```

```
d["Saves"]
Out[10]:
0
          98
        194
1
2
          41
3
        172
          96
114
        573
115
        135
116
          36
117
       1095
        653
118
Name: Saves, Length: 119, dtype: int64
```

Visualization

```
In [11]:
```

```
data=pd.DataFrame(d[['Saves','Shares']][0:500])
data
```

Out[11]:

Saves	Shares
98	5
194	14
41	1
172	7
96	4
573	38
135	1
36	1
1095	75
653	26
	98 194 41 172 96 573 135 36 1095

119 rows × 2 columns

In [12]:

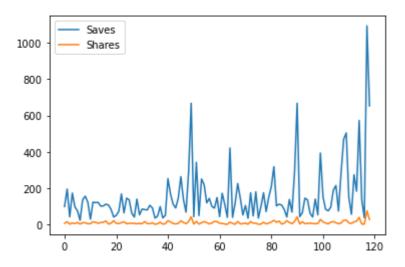
```
import matplotlib.pyplot as pp
```

In [13]:

data.plot.line()

Out[13]:

<AxesSubplot:>

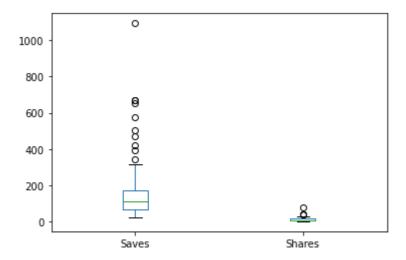


In [14]:

data.plot.box()

Out[14]:

<AxesSubplot:>

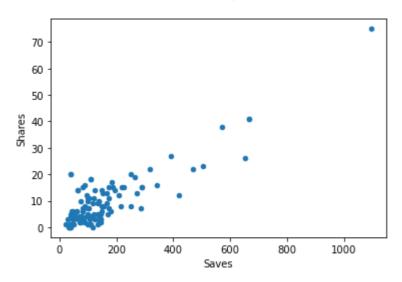


In [15]:

data.plot.scatter(x="Saves",y="Shares")

Out[15]:

<AxesSubplot:xlabel='Saves', ylabel='Shares'>

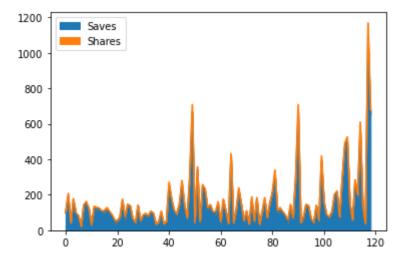


In [16]:

data.plot.area()

Out[16]:

<AxesSubplot:>

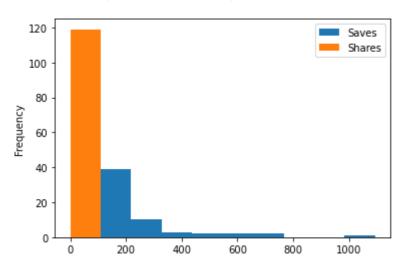


In [17]:

```
data.plot.hist()
```

Out[17]:

<AxesSubplot:ylabel='Frequency'>

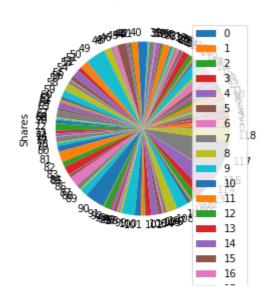


In [18]:

```
data=pd.DataFrame(d[['Saves','Shares']][0:200])
data.plot.pie(y="Shares")
```

Out[18]:

<AxesSubplot:ylabel='Shares'>



Statistics

Mean, median, mode, describe

In [19]:

```
data=pd.DataFrame(d[['Saves','Shares']][0:500])
data
```

Out[19]:

	Saves	Shares
0	98	5
1	194	14
2	41	1
3	172	7
4	96	4
114	573	38
115	135	1
116	36	1
117	1095	75
118	653	26

119 rows × 2 columns

In [20]:

```
print(data.mean())
```

Saves 153.310924 Shares 9.361345

dtype: float64

In [21]:

```
print(data.median())
```

Saves 109.0 Shares 6.0 dtype: float64

In [22]:

print(data.mode())

	Saves	Shares
0	40	3.0
1	135	NaN
2	144	NaN

```
In [23]:
```

```
data.fillna(value=1)
```

Out[23]:

	Saves	Shares
0	98	5
1	194	14
2	41	1
3	172	7
4	96	4
114	573	38
115	135	1
116	36	1
117	1095	75
118	653	26

119 rows × 2 columns

In [24]:

```
print(data.describe())
```

	Saves	Shares
count	119.000000	119.000000
mean	153.310924	9.361345
std	156.317731	10.089205
min	22.000000	0.000000
25%	65.000000	3.000000
50%	109.000000	6.000000
75%	169.000000	13.500000
max	1095.000000	75.000000

Sum,cumsum,count,min,max

In [25]:

print(data.sum())

Saves 18244
Shares 1114
dtype: int64

```
In [26]:
```

```
print(data.cumsum())
     Saves
           Shares
0
        98
                  5
1
       292
                 19
2
       333
                 20
3
       505
                 27
4
       601
                 31
114 16325
              1011
115
     16460
              1012
     16496
              1013
117
     17591
              1088
118
     18244
              1114
[119 rows x 2 columns]
In [27]:
print(data.count())
Saves
          119
Shares
          119
dtype: int64
In [28]:
print(data.min())
Saves
          22
Shares
dtype: int64
In [29]:
print(data.max())
          1095
Saves
            75
Shares
```

covariance and correlation (spearman and pearsons)

dtype: int64

```
In [30]:
data1=data['Saves'][0:10]
data1
Out[30]:
0
      98
1
     194
2
      41
3
     172
4
      96
5
      74
      22
6
7
     135
8
     155
9
     122
Name: Saves, dtype: int64
In [31]:
data2=data['Shares'][0:10]
data2
Out[31]:
      5
0
1
     14
2
      1
3
      7
4
      4
5
     10
6
      1
      9
7
8
      8
9
Name: Shares, dtype: int64
In [32]:
from numpy import cov
print(cov(data1,data2))
[[3091.87777778 171.35555556]
 [ 171.3555556
                   17.51111111]]
In [33]:
from scipy.stats import pearsonr
print(pearsonr(data1,data2))
(0.7364278065804685, 0.015141913073821655)
In [34]:
from scipy.stats import spearmanr
print(spearmanr(data1,data2))
```

SpearmanrResult(correlation=0.6565379871744699, pvalue=0.039204386332556754)

In []: