import pandas as pd pwd df = pd.read_csv(r'/content/dataset_Facebook.csv', ';'); <ipython-input-8-629e9ce8c81f>:1: FutureWarning: In a future version of pandas all arguments of read_csv except for the argument 'd
 df = pd.read_csv(r'/content/dataset_Facebook.csv', ';'); • Post Lifetime Lifetime Post Post Total Engaged Total Impressions Users 139441 Photo 0.0 Photo 0.0 139441 Photo 0.0 85093 Photo 0.0 81370 Photo 0.0 81370 Photo 4 NaN df.shape (500, 19) df.describe()

df.head()

Page total likes	Туре	Category	Post Month	Post Weekday	Post Hour	Paid	Lifetime Post Total	Lifetime Post Total Impressions		Lifetime Post Consumers	Cons
TIKES							Reach	Tillbi.essTolls	users	Consumers	COII:

0	139441	Photo	2	12	4	3	0.0	2752	5091	178	109
									19057		
2	139441	Photo	3	12	3	3	0.0	2413	4373	177	113
3	139441	Photo	2	12	2	10	1.0	50128	87991	2211	790
4	139441	Photo	2	12	2	3	0.0	7244	13594	671	410



df.tail()

Page total likes	Туре	Category	Post Month	Post Weekday	Post Hour	Paid	Lifetime Post Total	Lifetime Post Total Impressions			Con
likes							Reach	impressions	users	consumers	Con

495	85093	Photo	3	1	7	2	0.0	4684	7536	733	708
496	81370								6229		508
497	81370	Photo	1	1	5	2	0.0	3778	7216	625	572
498	81370	Photo	3	1	4	11	0.0	4156	7564	626	574
499	81370	Photo	2	1	4	4	NaN	4188	7292	564	524



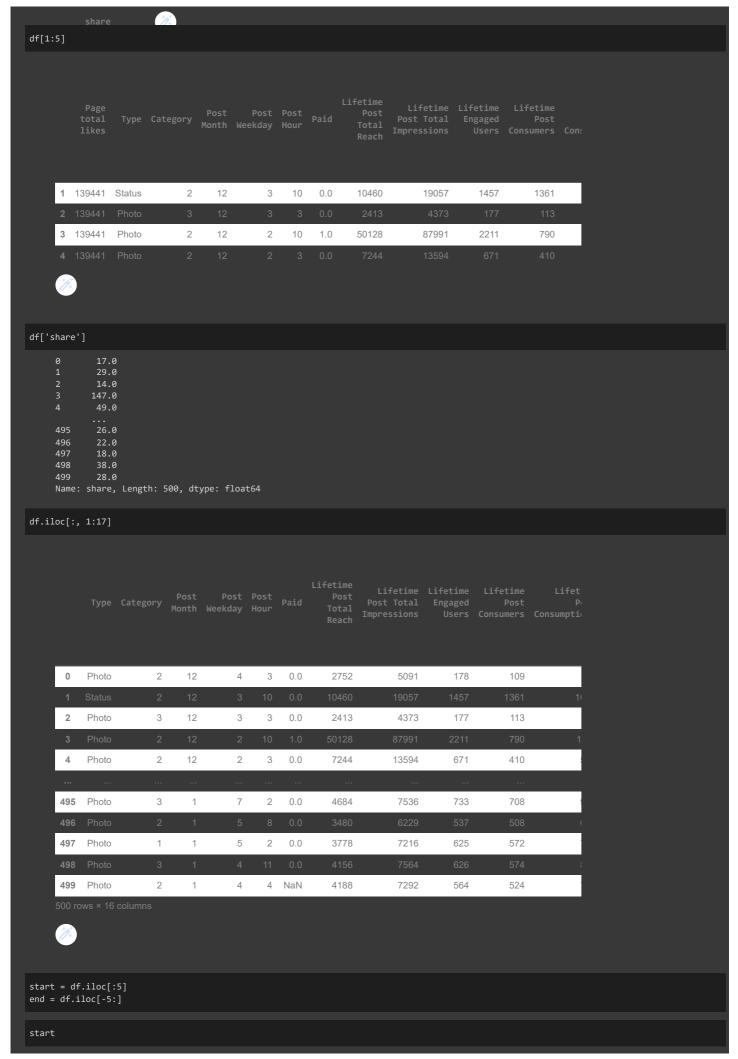
df.dtypes

Page total likes Category Post Month Post Weekday Post Hour int64 Paid Lifetime Post Total Reach Lifetime Post Total Impressions Lifetime Engaged Users Lifetime Post Consumptions Lifetime Post Impressions by people who have liked your Page Lifetime Post reach by people who like your Page Lifetime People who have liked your Page and engaged with your post comment like float64 float64 share Total Interactions

df.columns

- 'Lifetime Post Impressions by people who have liked your Page',

```
'Lifetime Post reach by people who like your Page',
'Lifetime People who have liked your Page and engaged with your post',
'comment', 'like', 'share', 'Total Interactions'],
dtype='object')
df['share']
                 ...
26.0
                 22.0
                 18.0
                 38.0
                 28.0
      Name: share, Length: 500, dtype: float64
df.share
                 29.0
                 14.0
                 49.0
                 ...
26.0
                 22.0
                 38.0
                 28.0
      Name: share, Length: 500, dtype: float64
df[['share', 'like']]
         0
                17.0
                          79.0
         2
                14.0
                          66.0
         4
                         325.0
                49.0
        495
                26.0
                          53.0
        497
                18.0
                          93.0
        499
                          91.0
                28.0
df[['share']]
```



	Page total likes	Туре	Category	Post Month	Post Weekday	Post Hour	Paid	Lifetime Post Total Reach	Lifetime Post Total Impressions	Lifetime Engaged Users	Lifetime Post Consumers	Con:
0	139441	Photo	2	12	4	3	0.0	2752	5091	178	109	
1	139441	Status	2	12	3	10	0.0	10460	19057	1457	1361	
2	139441	Photo	3	12	3	3	0.0	2413	4373	177	113	
3	139441	Photo	2	12	2	10	1.0	50128	87991	2211	790	
4	139441	Photo	2	12	2	3	0.0	7244	13594	671	410	



end

	Page total likes	Туре	Category	Post Month	Post Weekday	Post Hour	Paid	Lifetime Post Total Reach	Lifetime Post Total Impressions	Lifetime Engaged Users	Lifetime Post Consumers	Cons
495	85093	Photo	3	1	7	2	0.0	4684	7536	733	708	
496	81370								6229		508	
497	81370	Photo	1	1	5	2	0.0	3778	7216	625	572	
498	81370	Photo	3	1	4	11	0.0	4156	7564	626	574	
499	81370	Photo	2	1	4	4	NaN	4188	7292	564	524	·



merged_data = pd.concat([start, end])

merged_data

```
data = df[df['like'] > 100]
data
                                                  Post Post Paid
                                                                          fetime Lifetime Lifetime Lifetime
Post Post Total Engaged Post
Total Impressions Users Consumers Co
Reach
            139441 Status
                                                                 0.0
                                                                          10460
                                                                                        19057
                                                                                                    1457
                                                                                                                1361
        1
                                    2
                                           12
                                                      3
                                                            10
                                                      2
            139441
                     Photo
                                    2
                                           12
                                                            3
                                                                 0.0
                                                                           7244
                                                                                        13594
                                                                                                     671
                                                                                                                  410
        4
                                                      1
        6
            139441
                     Photo
                                    3
                                           12
                                                            3
                                                                 1.0
                                                                          11692
                                                                                        19479
                                                                                                      481
                                                                                                                  265
       488
             85979
                     Photo
                                     3
                                            1
                                                      7
                                                            10
                                                                 0.0
                                                                           9700
                                                                                        17442
                                                                                                     1407
                                                                                                                 1271
       492
             85979
                                            1
                                                      5
                                                            11
                                                                          45920
                      Link
                                     1
                                                                 0.0
                                                                                         5808
                                                                                                      753
                                                                                                                  655
       494
             85093 Photo
                                     3
                                            1
                                                      7
                                                            10
                                                                 0.0
                                                                           5400
                                                                                         9218
                                                                                                      810
                                                                                                                  756
new_data = []
for index, row in df.iterrows():
  flag = False
  for val in row.values:
      flag = True
      break
  if flag:
    new_data.append(row)
new_df = pd.DataFrame(new_data);
temp1 = df[df['Post Month'] > 10]
temp2 = df[df['like'] > 120]
temp1
```

		Page total likes	Туре	Category	Post Month	Post Weekday	Post Hour	Paid	Lifetime Post Total Reach	Lifetime Post Total Impressions	Engaged	Lifetime Post Consumers	Соі	
	0	139441	Photo	2	12	4	3	0.0	2752	5091	178	109		
temp2														

Engaged Post Users Consumers Co Status 0.0 Photo 0.0 139441 Photo 1.0 Photo 0.0 0.0 Link 85093 Photo 0.0

217 rows × 19 columns



temp3 = pd.concat([temp1, temp2])

temp3

temp4 = pd.merge(temp1, temp2, left_index=True, right_index=True); temp4

. – - –	-,						_		- ,	,	
	O	133441	FIIUIU	J	IZ	1	J	1.0	11092	19479	4(
									13720		50
	8	139441	Status	2	12	7	3	0.0	11844	22538	150
	10	139441	Status	2	12	5	10	0.0	21744	42334	428
	13	139441	Photo	2	12	5	3	0.0	2549	4896	24
	14	138414	Photo	2	12	4	5	1.0	22784	39941	88
	17	138414	Photo	1	12	2	12	1.0	53264	111785	170
	26	138458	Status	2	12	5	11	0.0	19552	34143	280
emp5	= df		st Month'l > :	10) & (df['like'l	> 120)]					

temp5

3	133441	Otatus	2	12	,	9	0.0	10472	20049	1191	1073	
6	139441	Photo	3	12	1	3	1.0	11692	19479	481	265	
7	139441	Photo	3	12	7	9	1.0	13720	24137	537	232	
8	139441	Status	2	12	7	3	0.0	11844	22538	1530	1407	•
10	139441	Status	2	12	5	10	0.0	21744	42334	4258	4100	
13	139441	Photo	2	12	5	3	0.0	2549	4896	249	134	
14	138414	Photo	2	12	4	5	1.0	22784	39941	887	337	
17	138414	Photo	1	12	2	12	1.0	53264	111785	1706	1103	
26	138458	Status	2	12	5	11	0.0	19552	34143	2806	2531	
28	138895	Photo	2	12	5	3	0.0	9560	18264	973	559	
20	128805	Vidoo	1	19	Л	11	1 ∩	38308	61262	11/11	1062	

sorted_data = df.sort_values(['Page total likes', 'comment'], ascending=False)

sorted_data.head(20)

	Page total likes	Туре	Category	Post Month	Post Weekday	Post Hour	Paid	Lifetime Post Total Reach	Lifetime Post Total Impressions	Lifetime Engaged Users	Lifetime Post Consumers	Соі
3	139441	Photo	2	12	2	10	1.0	50128	87991	2211	790	
4	139441	Photo	2	12	2	3	0.0	7244	13594	671	410	
1	139441	Status	2	12	3	10	0.0	10460	19057	1457	1361	
13	139441	Photo	2	12	5	3	0.0	2549	4896	249	134	
0	139441	Photo	2	12	4	3	0.0	2752	5091	178	109	
6	139441	Photo	3	12	1	3	1.0	11692	19479	481	265	
9	139441	Photo	3	12	6	10	0.0	4694	8668	280	183	
5	139441	Status	2	12	1	9	0.0	10472	20849	1191	1073	
2	139441	Photo	3	12	3	3	0.0	2413	4373	177	113	
7	139441	Photo	3	12	7	9	1.0	13720	24137	537	232	
8	139441	Status	2	12	7	3	0.0	11844	22538	1530	1407	
10	139441	Status	2	12	5	10	0.0	21744	42334	4258	4100	
11	139441	Photo	2	12	5	10	0.0	3112	5590	208	127	
12	139441	Photo	2	12	5	10	0.0	2847	5133	193	115	
28	138895	Photo	2	12	5	3	0.0	9560	18264	973	559	
30	138895	Photo	2	12	4	2	0.0	4940	9390	385	306	
29	138895	Video	1	12	4	11	1.0	36208	61262	1141	1068	
37	138895	Photo	1	12	7	10	1.0	19800	28663	479	424	
38	138895	Status	2	12	7	9	0.0	17576	33058	5352	5202	
								13280				



transpose_data = df.transpose()

transpose_data

Category 2 2 3 3 2 2 2 3 3 3 Photo Month 12 12 12 12 12 12 12 12 12 12 12 12 12	2023, 20:18								ynb - Co	labo
Post Meekday 4 3 3 2 2 1 1 7 7 Post Hour 3 10 3 10 3 9 3 9 Paid 0.0 0.0 0.0 1.0 0.0 0.0 1.0 1.0 Lifetime Post Total Rasch 2752 10460 2413 50128 7244 10472 11692 13720 1 Lifetime Post Total Impressions Lifetime Post Consumers 109 1361 113 790 410 1073 265 232 1 Lifetime Post Consumers 109 1361 113 790 410 1073 265 232 1 Lifetime Post Consumptions 159 1674 154 1119 580 1389 364 305 Lifetime Post Lifetime Post Inference Post Inference Post Consumptions 159 1674 154 1119 580 1389 364 305 Lifetime Post Lifetime Post Inference Post Inference Post Post Post Post Post Post Post Post										St
Post Weekday 4 3 3 2 2 1 1 7 Post Hour 3 10 3 10 3 9 3 9 Paid 0.0 0.0 0.0 1.0 0.0 0.0 1.0 1.0 Lifetime Post Total So91 19057 4373 87991 13594 20849 19479 24137 2: Impressions Lifetime Post Total Impressions Lifetime Engaged 178 1457 177 2211 671 1191 481 537 Lifetime Post Consumers Lifetime Post Consumptions 159 1674 154 1119 580 1389 384 305 Lifetime Post Consumptions 159 1674 154 1119 580 1389 384 305 Lifetime Post Impressions by people who have liked your Page Lifetime Post reach by people who have liked your Page Lifetime Post reach by people who have liked your Page Lifetime Post reach by people with an angular to the people who have liked your Page Lifetime Post reach by people who have liked 119 1108 132 1386 396 1016 379 422 reach by people with angular to the people who have liked your Page Lifetime Post reach by people with angular to the people who have liked 119 1108 132 1386 396 1016 379 422 reach by people with angular to the people who have liked your Page Lifetime Post reach by people with angular to the people who have liked your Page 110 108 132 1386 396 1016 379 422 reach by people with angular to the people who have liked your Page 300 110 110 110 110 110 110 110 110 110										
Post Hour 3 10 3 10 3 9 3 9 3 9 1										
Lifetime Post Total Reach Lifetime Post Total South 19057										
Total Reach Lifetime Post Total Sone 1 19057	Paid	0.0	0.0	0.0	1.0	0.0	0.0	1.0	1.0	7
Total 5091 19057 4373 87991 13594 20849 19479 24137 2:			10460	2413	50128	7244	10472	11692	13720	1'
Lifetime Post Consumers 109 1361 113 790 410 1073 265 232 1056 1058 1	Total	5091	19057	4373	87991	13594	20849	19479	24137	22
Consumers 109 1361 113 790 410 1073 265 232 Lifetime Post Consumptions 159 1674 154 1119 580 1389 364 305 Lifetime Post Impressions by people who have liked your Page Lifetime Post reach by people who have liked your Page and engaged with your post comment 4 5 0 58 19 1 3 0 Like 79.0 130.0 66.0 1572.0 325.0 152.0 249.0 325.0 1 share 17.0 29.0 14.0 147.0 49.0 33.0 27.0 14.0 Total Interactions 100 164 80 1777 393 186 279 339 19 rows × 500 columns hape_data = pd.pivot_table(df, index=['Category'], values=['comment']) Lifetime Post reach by people who have liked your Page and engaged with your post comment 4 5 0 58 19 1 3 0 Lifetime Post reach by people who have liked 119 1108 132 1386 396 1016 379 422 1900 1000 1000 1000 1000 1000 1000 10	Engaged	178	1457	177	2211	671	1191	481	537	
Consumptions			1361	113	790	410	1073	265	232	
Impressions by people who have liked your Page	Lifetime Po		1674						305	
reach by people who like your Page Lifetime People who have liked your Page and engaged with your post comment	Impression by people w have liked your Page	ho 3078	11710	2812	61027	6228	16034	15432	19728	15
People who have liked your Page and engaged with your post comment	reach by people who				32048	3200		9328		
like 79.0 130.0 66.0 1572.0 325.0 152.0 249.0 325.0 1 share 17.0 29.0 14.0 147.0 49.0 33.0 27.0 14.0 Total	Lifetime People who have liked your Page a engaged wi	o I ₁₁₉ nd th	1108	132	1386	396	1016	379	422	
share 17.0 29.0 14.0 147.0 49.0 33.0 27.0 14.0 Total Interactions 100 164 80 1777 393 186 279 339 19 rows × 500 columns hape_data = pd.pivot_table(df, index=['Category'], values=['comment']) hape_data comment	comment	4	5	0	58	19	1	3	0	
Total Interactions 100 164 80 1777 393 186 279 339 19 rows × 500 columns nape_data = pd.pivot_table(df, index=['Category'], values=['comment']) Category 1										1
Interactions 100 104 80 1777 393 186 279 339 19 rows × 500 columns nape_data = pd.pivot_table(df, index=['Category'], values=['comment']) category 1		17.0	29.0	14.0	147.0	49.0	33.0	27.0	14.0	
nape_data = pd.pivot_table(df, index=['Category'], values=['comment']) nape_data comment Category 1		s 100	164	80	1777	393	186	279	339	
chape_data = pd.pivot_table(df, index=['Category'], values=['comment']) chape_data comment Category 1										
comment Category 1 5.804651 2 11.100000 3 6.774194 Fort numpy as np Chaping_arr= np.array([1,2,3,4,5,6]) Chaping_arr.reshape(3,2) Category Analy ([1,2], [3,4],		d.pivot_tab	le(df, in	dex=['Ca	ategory'	, values	s=['comm	ent'])		
Category 1										
Category 1		commont								
1 5.804651 2 11.100000 3 6.774194 ort numpy as np haping_arr= np.array([1,2,3,4,5,6]) haping_arr.reshape(3,2) array([[1, 2],		comment								
3 6.774194 port numpy as np shaping_arr= np.array([1,2,3,4,5,6]) shaping_arr.reshape(3,2) array([[1, 2],		5.804651								
ort numpy as np haping_arr= np.array([1,2,3,4,5,6]) haping_arr.reshape(3,2) array([[1, 2],	2 1	1.100000								
haping_arr= np.array([1,2,3,4,5,6]) haping_arr.reshape(3,2) array([[1, 2],	3	6.774194								
haping_arr= np.array([1,2,3,4,5,6]) haping_arr.reshape(3,2) array([[1, 2],										
aping_arr.reshape(3,2) array([[1, 2],										
[3, 4],			2,3,4,5,6])						
	[3, 4	4],								