

# Import Libraries and Load the dataset.

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

```
In [2]: xls = pd.ExcelFile(r'C:\Users\NEW\Downloads\Data set for BA.xlsx')
sheet_names = xls.sheet_names
dfs = {sheet_name: xls.parse(sheet_name) for sheet_name in sheet_names}
dfs
```

```
Out[2]: {'Report Snapshot': Empty DataFrame
Columns: []
Index: [],
'User Acquisition':   First user default channel group  New users  Engaged sessions \
0                Display                9957            12008
1          Organic Search                7652            18141
2          Paid Search                3025             4408
3                Direct                1903            4975
4          Unassigned                 325             1619
5          Organic Social                 10              13

Engagement rate  Engaged sessions per user  Average engagement time \
0      0.544457                1.206107            58.86209
1      0.813680                2.367041            534.31280
2      0.474284                1.458154            102.23780
3      0.318808                2.261364           1128.88100
4      0.813159                4.981538            798.34150
5      0.722222                1.300000           145.30000

Event count  Conversions  Total revenue
0      204820         37434              0
1      770710        109801              0
2       81997         14770              0
3      227434         31093              0
4       33320          789              0
5         248          27              0 ,
'Traffic Aquisition':   Session default channel group  Users  Sessions  Engaged sessions \
0                Unassigned  20263    13448            1481
1                Display    9613    18292           10613
2          Organic Search    7689    21241           17814
3                Direct    4042    13220            7649
4          Paid Search    2909     6788            3452
5          Organic Social     11      16             12

Average engagement time per session  Engaged sessions per user \
0                34.11704            0.073089
1                28.52198            1.104026
2           195.94340            2.316816
3           177.17060            1.892380
4                36.65321            1.186662
5           60.06250            1.090909

Events per session  Engagement rate  Event count  Conversions \
0      18.023130            0.110128    242375    114161
1       9.069320            0.580199    165896     20031
2      29.302290            0.838661    622410     33612
3      17.135850            0.578593    226536     18496
4       8.989982            0.508544     61024      7595
5      18.000000            0.750000      288       19
```

Total revenue	
0	0
1	0
2	0
3	0
4	0
5	0 ,

'Event Report':	Event name	Event count	Total users \
0	screen_view	694729	23254
1	notification_receive	125146	1700
2	user_engagement	124836	22699
3	notification_dismiss	70128	1369
4	session_start	61163	23226
..	...	...	...
374	Promilo119_myProfile_mediator	1	1
375	Promilo_feeds	1	1
376	feeds	1	1
377	my_interests_screen	1	1
378	(not set)	0	22269

	Event count per user	Total revenue
0	30.865870	0
1	138.896800	0
2	5.622230	0
3	144.000000	0
4	3.121357	0
..	...	...
374	1.000000	0
375	1.000000	0
376	1.000000	0
377	1.000000	0
378	0.000000	0

[379 rows x 5 columns],

'Conversion Report':	Event name	Conversions	Total users	Tota
1 revenue				
0	notification_receive	94890	1311	0
1	session_start	56203	21674	0
2	first_open	22872	23059	0
3	app_remove	12468	12538	0
4	Promilo111 otp_screen	1738	855	0
5	Promilo111_Event_Enter_Feed_Page	1594	969	0
6	Promilo106_login	1458	603	0
7	Promilo106_feeds	683	185	0
8	os_update	672	634	0
9	notification_open	569	308	0
10	Promilo106_feedDetails	195	67	0
11	Promilo106_my_meetings_screen	136	20	0
12	Promilo106 otp_screen	128	90	0
13	Promilo106_resume_builder	127	66	0
14	Promilo106_my_interests_screen	117	23	0
15	Promilo106_dashboard	23	16	0
16	Promilo106_my_profile_learners	21	10	0
17	Promilo106_campaign_interest	20	10	0,

'Pages & Screens Report':	Page path and screen class	Views	Users	Views pe
r user \				
0	Flutter	156708	8726	17.958740
1	MainActivity	44326	8978	4.937180
2	feeds	18514	4358	4.248279
3	login	16883	7291	2.315595
4	my_rewards_screen	15381	2045	7.521271
5	storyboard	8189	5244	1.561594
6	SignInHubActivity	6650	3778	1.760191
7	registration_screen	5501	3566	1.542625
8	feedDetails	3971	1047	3.792741

9	otp_screen	3291	1678	1.961263
10	video_viewer_screem	2880	1521	1.893491
11	FacebookActivity	2299	675	3.405926
12	resume_builder	1781	828	2.150966
13	CustomTabMainActivity	1301	193	6.740933
14	notification_store	1062	648	1.638889
15	dashboard	1058	411	2.574209
16	myProfile_mediator	1056	600	1.760000
17	WebViewActivity	878	490	1.791837
18	video_tutorial_view	835	722	1.156510
19	my_profile_learners	804	321	2.504673
20	FlutterViewController	758	155	4.890323
21	my_meetings_screen	715	273	2.619048
22	my_interests_screen	688	375	1.834667
23	discovery_screen	486	225	2.160000
24	calculator_intro	388	281	1.380783
25	campaign_interest	244	58	4.206897
26	calculator_one	199	151	1.317881
27	UIActivityViewSuccessController	99	18	5.500000
28	UIActivityContentViewController	97	17	5.705882
29	calculator_two	88	73	1.205479
30	SFAuthenticationViewController	76	45	1.688889
31	my_profile_professional	67	23	2.913043
32	CustomTabActivity	35	24	1.458333
33	SFSafariViewController	34	21	1.619048
34	CheckoutActivity	24	3	8.000000
35	my_profile_others	20	12	1.666667
36	PHPickerViewController	13	10	1.300000
37	SLComposeViewController	6	4	1.500000
38	campaign_meeting	4	3	1.333333
39	CAMImagePickerCameraViewController	2	2	1.000000
40	UIAlertController	2	2	1.000000
41	(not set)	0	9145	0.000000

	Average engagement time	Event count	Conversions	Total revenue
0	83.412220	203901	328	0
1	78.292160	53374	101	0
2	61.600050	37628	253	0
3	34.881770	40772	435	0
4	94.179950	32910	5	0
5	5.341152	15676	115	0
6	0.003176	6653	0	0
7	45.075720	13496	136	0
8	69.316140	7820	84	0
9	46.864720	10833	32	0
10	28.120970	5256	31	0
11	0.524444	2310	0	0
12	118.043500	3776	17	0
13	0.005181	1302	0	0
14	10.137350	1971	0	0
15	38.114360	2279	4	0
16	57.020000	2276	4	0
17	105.806100	1321	2	0
18	36.303320	1662	2	0
19	127.722700	1856	4	0
20	15.625810	1060	18	0
21	45.765570	1480	15	0
22	29.189330	1340	7	0
23	70.235560	890	7	0
24	6.391459	680	0	0
25	42.379310	459	0	0
26	41.013250	387	0	0
27	0.111111	101	0	0
28	32.411760	177	0	0
29	9.287671	158	0	0
30	12.200000	161	0	0

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31      184.087000      143      0      0
32      1.666667      36      0      0
33      46.857140      74      0      0
34      304.333300      26      0      0
35      123.833300      49      1      0
36      9.500000      27      0      0
37      15.250000      13      0      0
38      14.333330      7      0      0
39      7.500000      4      0      0
40      1.000000      4      0      0
41      0.001093      109189      88518      0 ,
'Retention Overview': Empty DataFrame
Columns: []
Index: [],
'User Engagement Overview': Empty DataFrame
Columns: []
Index: [],
'Demographics Report':
Country  Users  New users  Engaged sessions  Enga
gement rate \
0      India  23024      22528      41479      0.593626
1      United States  272      213      197      0.491272
2      Canada  37      18      25      0.416667
3      (not set)  36      36      17      0.459459
4      United Kingdom  20      8      13      0.371429
5      Singapore  17      6      13      0.419355
6      Japan  11      6      11      0.550000
7      Australia  10      7      8      0.500000
8      Bangladesh  7      2      10      0.625000
9      Germany  7      2      6      0.500000
10     Malaysia  7      7      7      0.636364
11     Nepal  7      3      5      0.357143
12     Saudi Arabia  7      5      6      0.600000
13     United Arab Emirates  5      3      3      0.300000
14     Kuwait  4      3      5      0.833333
15     Myanmar (Burma)  3      2      8      0.533333
16     Qatar  3      2      3      0.600000
17     China  2      2      4      1.000000
18     Indonesia  2      1      2      1.000000
19     Ireland  2      2      1      0.500000
20     Italy  2      2      2      0.666667
21     Netherlands  2      2      2      0.666667
22     South Korea  2      0      4      0.444444
23     Switzerland  2      0      1      0.500000
24     Afghanistan  1      1      1      1.000000
25     Argentina  1      1      1      1.000000
26     Bahamas  1      1      1      1.000000
27     Dominican Republic  1      1      1      1.000000
28     France  1      0      1      0.333333
29     Guernsey  1      1      1      1.000000
30     Iran  1      0      0      0.000000
31     Kyrgyzstan  1      1      1      1.000000
32     Latvia  1      1      1      1.000000
33     Norway  1      0      0      0.000000
34     Oman  1      1      1      1.000000
35     Panama  1      0      1      1.000000
36     Romania  1      1      1      1.000000
37     Russia  1      1      1      0.500000
38     Serbia  1      1      1      1.000000
39     Sweden  1      1      1      1.000000
40     Czechia  0      0      0      0.000000
41     Hungary  0      0      0      0.000000
42     Kenya  0      0      0      0.000000
43     Maldives  0      0      0      0.000000
44     Pakistan  0      0      0      0.000000
45     Sri Lanka  0      0      0      0.000000
46     Ukraine  0      0      0      0.000000

```

	Engaged sessions per user	Average engagement time	Event count \
0	1.801555	334.81660	1312097
1	0.724265	50.96324	3157
2	0.675676	43.21622	410
3	0.472222	24.80556	241
4	0.650000	61.85000	289
5	0.764706	70.00000	299
6	1.000000	51.45455	283
7	0.800000	26.90000	132
8	1.428571	49.85714	121
9	0.857143	15.42857	82
10	1.000000	536.00000	507
11	0.714286	13.85714	74
12	0.857143	25.28571	74
13	0.600000	7.60000	64
14	1.250000	36.50000	91
15	2.666667	47.00000	142
16	1.000000	20.66667	27
17	2.000000	61.00000	42
18	1.000000	8.50000	21
19	0.500000	72.00000	61
20	1.000000	14.50000	23
21	1.000000	156.50000	58
22	2.000000	45.00000	49
23	0.500000	2.00000	9
24	1.000000	17.00000	6
25	1.000000	12.00000	6
26	1.000000	13.00000	6
27	1.000000	160.00000	25
28	1.000000	14.00000	18
29	1.000000	30.00000	8
30	0.000000	1.00000	3
31	1.000000	20.00000	11
32	1.000000	16.00000	7
33	0.000000	1.00000	7
34	1.000000	2.00000	9
35	1.000000	6.00000	9
36	1.000000	13.00000	6
37	1.000000	152.00000	23
38	1.000000	32.00000	8
39	1.000000	9.00000	8
40	0.000000	0.00000	2
41	0.000000	0.00000	1
42	0.000000	0.00000	1
43	0.000000	0.00000	1
44	0.000000	0.00000	3
45	0.000000	0.00000	1
46	0.000000	0.00000	7

	Conversions	Total revenue
0	192766	0
1	643	0
2	121	0
3	54	0
4	43	0
5	29	0
6	24	0
7	22	0
8	20	0
9	9	0
10	19	0
11	18	0
12	13	0
13	15	0
14	11	0

15	16	0
16	4	0
17	5	0
18	3	0
19	4	0
20	4	0
21	10	0
22	11	0
23	2	0
24	2	0
25	2	0
26	2	0
27	2	0
28	7	0
29	2	0
30	1	0
31	2	0
32	1	0
33	1	0
34	2	0
35	1	0
36	1	0
37	4	0
38	2	0
39	2	0
40	2	0
41	1	0
42	0	0
43	1	0
44	2	0
45	1	0
46	7	0

'Citiwise Report':			Town/City	Users	New users	Engaged sessions	Engagement rate
\							
0	Bengaluru	6097	5685		15013	0.769385	
1	Patna	1594	1467		2127	0.440646	
2	Hyderabad	1038	920		1578	0.569264	
3	Indore	983	915		1241	0.426460	
4	Lucknow	897	839		1125	0.450180	
..	...	...	...		...	...	
569	Titwala	0	0		0	0.000000	
570	Vagator	0	0		0	0.000000	
571	Vellakoil	0	0		0	0.000000	
572	Wardha	0	0		0	0.000000	
573	Washington	0	0		0	0.000000	

	Engaged sessions per user	Average engagement time	Event count	\
0	2.462359	762.20550	607200	
1	1.334379	98.22208	38830	
2	1.520231	243.69080	96826	
3	1.262462	67.89115	21383	
4	1.254181	83.40580	21041	
..	...	...	...	
569	0.000000	0.00000	1	
570	0.000000	0.00000	16	
571	0.000000	0.00000	1	
572	0.000000	0.00000	1	
573	0.000000	0.00000	1	

	Conversions	Total revenue
0	62939	0
1	6980	0
2	34103	0
3	4121	0
4	3650	0
..	...	...

569	1	0
570	16	0
571	0	0
572	1	0
573	1	0

[574 rows x 10 columns],

'Gender Report':			Gender	Users	New users	Engaged sessions	Engagement rate \
0	unknown	13142	12691		23161	0.564077	
1	male	7218	5877		10467	0.543091	
2	female	4944	4304		7877	0.637710	

	Engaged sessions per user	Average engagement time	Event count \
0	1.762365	439.5776	761771
1	1.450125	128.2319	282504
2	1.593244	208.7407	274254

	Conversions	Total revenue
0	93180	0
1	65651	0
2	35083	0 ,

'User By Interest':		Interests	Users	New use
rs \				
0	Shoppers	10950	9256	
1	Media & Entertainment/Comics & Animation Fans	10946	9247	
2	Technology/Mobile Enthusiasts	10934	9239	
3	Food & Dining/Cooking Enthusiasts	8410	6970	
4	Sports & Fitness/Health & Fitness Buffs	5844	4580	
..	...	...	...	
84	Food & Dining	15	4	
85	Home & Garden	15	5	
86	Sports & Fitness/Sports Fans/Racquetball Enthu...	11	11	
87	Vehicles & Transportation	11	3	
88	Sports & Fitness/Sports Fans/Fans of American ...	10	4	

	Engaged sessions	Engagement rate	Engaged sessions per user \
0	15652	0.581534	1.429406
1	15680	0.583008	1.432487
2	15619	0.582451	1.428480
3	12332	0.602325	1.466350
4	8226	0.588328	1.407598
..	...	...	...
84	24	0.489796	1.600000
85	12	0.631579	0.800000
86	21	0.840000	1.909091
87	9	0.450000	0.818182
88	18	0.782609	1.800000

	Average engagement time	Event count	Conversions	Total revenue
0	162.83470	490664	86846	0
1	165.17720	491025	86845	0
2	162.69450	489353	86742	0
3	176.95670	409713	73814	0
4	155.14510	257831	43074	0
..	...	...	...	...
84	70.86667	460	58	0
85	133.86670	453	107	0
86	487.45450	736	39	0
87	71.54545	161	27	0
88	201.40000	375	81	0

[89 rows x 10 columns],

'User by Language':			Language	Users	New users	Engaged sessions	Engagement rate \
0	English	22495	21990		40639	0.595147	
1	Hindi	586	552		798	0.406314	
2	Marathi	85	84		98	0.426087	

3	Gujarati	78	77	100	0.448430
4	Telugu	43	42	56	0.455285
5	Tamil	36	36	43	0.518072
6	Malayalam	17	15	36	0.654545
7	Bengali	14	11	18	0.600000
8	Chinese	13	13	13	1.000000
9	Kannada	13	12	31	0.500000
10	Panjabi	9	9	17	0.708333
11	Persian	8	8	6	0.400000
12	Spanish	6	6	8	0.470588
13	Finnish	4	3	4	0.571429
14	Japanese	4	4	3	0.428571
15	Oriya	4	4	2	0.666667
16	Afrikaans	1	1	1	1.000000
17	Assamese	1	0	1	1.000000
18	German	1	1	0	0.000000
19	Malay	1	1	1	1.000000
20	Nepali	1	1	1	1.000000
21	Russian	1	1	0	0.000000
22	Urdu	1	1	0	0.000000
23	Sanskrit	0	0	0	0.000000

	Engaged sessions per user	Average engagement time	Event count \
0	1.806579	341.36350	1297970
1	1.361775	60.03413	13523
2	1.152941	38.48235	1589
3	1.282051	46.53846	1794
4	1.302326	36.65116	812
5	1.194444	45.86111	615
6	2.117647	161.94120	548
7	1.285714	50.07143	217
8	1.000000	136.76920	138
9	2.384615	249.07690	680
10	1.888889	92.44444	229
11	0.750000	28.25000	99
12	1.333333	22.16667	113
13	1.000000	89.25000	64
14	0.750000	9.25000	49
15	0.500000	7.50000	29
16	1.000000	37.00000	12
17	1.000000	42.00000	6
18	0.000000	0.00000	5
19	1.000000	2.00000	7
20	1.000000	5.00000	7
21	0.000000	70.00000	12
22	0.000000	1.00000	7
23	0.000000	0.00000	4

	Conversions	Total revenue
0	189946	0
1	2699	0
2	323	0
3	327	0
4	170	0
5	115	0
6	71	0
7	39	0
8	20	0
9	75	0
10	35	0
11	23	0
12	22	0
13	11	0
14	12	0
15	10	0
16	2	0



17	1	0
18	2	0
19	2	0
20	2	0
21	1	0
22	4	0
23	2	0

'User By Age':			Age	Users	New users	Engaged sessions	Engagement rate \
0	unknown	14303		13636		24976	0.569098
1	18-24	4282		3678		7291	0.695308
2	25-34	2920		2161		3749	0.504780
3	65+	1422		1081		1640	0.539829
4	55-64	1403		979		1552	0.519411
5	35-44	1202		785		1420	0.510424
6	45-54	810		552		881	0.561862

	Engaged sessions per user	Average engagement time	Event count \
0	1.746207	422.22330	817501
1	1.702709	251.16300	309328
2	1.283904	97.24144	90074
3	1.153305	52.30661	24780
4	1.106201	55.37063	25169
5	1.181364	96.08236	33016
6	1.087654	84.54321	18661

	Conversions	Total revenue
0	99310	0
1	53661	0
2	20172	0
3	4891	0
4	4823	0
5	8111	0
6	2946	0

'Google Ads Report':			Session	Google Ads campaign	Users	Sessions \
0	App Installation for May --Shahid		5429		10936	
1	App Install-States-A200Inst-20Jun22		842		1655	
2	App Install-States-B100Installs-22Jun22		742		1332	
3	App Install for April -- Shahid		473		976	
4	Video-AppInstall-PS-Internships-11Jul22		510		966	
5	App promotion-App-3		732		945	
6	App Instal-States-B200 &A100Inst-22Jun22		373		742	
7	App Install-1to5NC-StateA200-07Jul22		370		610	
8	App Instal-6to10NC-States-A200Inst-07Jul22		242		432	
9	Video-AppInstall-PS-Browsing-11Jul22		91		188	
10	Video-AppInstall-PS-Webinar-11Jul22		78		124	
11	Video-AppInstall-PS-Colleges-11Jul22		46		77	
12	Video-AppInstall-PS-Videos-11Jul22		38		75	
13	Video-AppInstall-PS-Jobs-11Jul22		25		49	
14	App installation for May 06-05-2022 T1		2		5	

	Engaged sessions	Google Ads clicks	Google Ads cost \
0	6276	147100	179175.00000
1	968	28742	24309.13000
2	780	17809	22374.58000
3	546	19302	20525.18000
4	515	9831	6377.83300
5	763	5793	12084.04000
6	425	10595	11993.01000
7	462	3659	8839.72300
8	296	4475	9204.69600
9	112	1899	1535.27000
10	81	893	1903.41800
11	50	1851	1263.62500
12	39	1706	935.72270
13	32	1649	800.09510
14	5	14	16.62396

	Google Ads cost per click	Conversions	Cost per conversion	Event count	\
0	1.218049	12257	14.618180	97802	
1	0.845770	1794	13.550240	15311	
2	1.256363	1422	15.734580	11640	
3	1.063370	1115	18.408230	8001	
4	0.648747	1032	6.180071	10323	
5	2.085972	922	13.106330	10825	
6	1.131950	851	14.092850	7504	
7	2.415885	709	12.467870	10864	
8	2.056915	630	14.610630	6092	
9	0.808462	206	7.452768	2172	
10	2.131487	150	12.689450	1600	
11	0.682671	101	12.511140	1268	
12	0.548489	76	12.312140	728	
13	0.485200	53	15.096130	731	
14	1.187426	5	3.324793	163	

	Total revenue	Return on ad spend
0	0	0
1	0	0
2	0	0
3	0	0
4	0	0
5	0	0
6	0	0
7	0	0
8	0	0
9	0	0
10	0	0
11	0	0
12	0	0
13	0	0
14	0	0 }

## Lets Check the sheets & their names.

In [3]: `sheet_names`

Out[3]:

```
['Report Snapshot',
 'User Acquisition',
 'Traffic Aquisition',
 'Event Report',
 'Conversion Report',
 'Pages & Screens Report',
 'Retention Overview',
 'User Engagement Overview',
 'Demographics Report',
 'Citiwise Report',
 'Gender Report',
 'User By Interest',
 'User by Language',
 'User By Age',
 'Google Ads Report']
```

In [4]:

```
df1 = dfs['User Acquisition']
df2 = dfs['Traffic Aquisition']
df3 = dfs['Event Report']
df4 = dfs['Conversion Report']
df5 = dfs['Pages & Screens Report']
df6 = dfs['Demographics Report']
df7 = dfs['Citiwise Report']
df8 = dfs['Gender Report']
df9 = dfs['User By Interest']
```

```
df10 = dfs['User by Language']
df11 = dfs['User By Age']
df12 = dfs['Google Ads Report']
```

## Data cleaning.

### Let's find if there is any null values in the dataframe

```
In [5]: dfs_list = [df1, df2, df3, df4, df5, df6, df7, df8, df9, df10, df11, df12]
```

```
for i, df in enumerate(dfs_list, start=1):
    print(f"DataFrame {i}:")
    print(df.isnull().sum())
    print("\n---\n")
```

```
DataFrame 1:
First user default channel group    0
New users                          0
Engaged sessions                    0
Engagement rate                     0
Engaged sessions per user           0
Average engagement time              0
Event count                         0
Conversions                         0
Total revenue                       0
dtype: int64
```

---

```
DataFrame 2:
Session default channel group    0
Users                            0
Sessions                         0
Engaged sessions                 0
Average engagement time per session 0
Engaged sessions per user        0
Events per session               0
Engagement rate                  0
Event count                      0
Conversions                      0
Total revenue                    0
dtype: int64
```

---

```
DataFrame 3:
Event name          1
Event count         0
Total users         0
Event count per user 0
Total revenue       0
dtype: int64
```

---

```
DataFrame 4:
Event name    0
Conversions   0
Total users   0
Total revenue 0
dtype: int64
```

---

```
DataFrame 5:
Page path and screen class    0
Views                        0
Users                        0
Views per user                0
Average engagement time      0
Event count                  0
Conversions                  0
Total revenue                 0
dtype: int64
```

---

```
DataFrame 6:
Country                      0
Users                        0
New users                    0
Engaged sessions             0
Engagement rate              0
Engaged sessions per user    0
Average engagement time      0
Event count                  0
Conversions                  0
Total revenue                 0
dtype: int64
```

---

```
DataFrame 7:
Town/City                    0
Users                        0
New users                    0
Engaged sessions             0
Engagement rate              0
Engaged sessions per user    0
Average engagement time      0
Event count                  0
Conversions                  0
Total revenue                 0
dtype: int64
```

---

```
DataFrame 8:
Gender                       0
Users                        0
New users                    0
Engaged sessions             0
Engagement rate              0
Engaged sessions per user    0
Average engagement time      0
Event count                  0
Conversions                  0
Total revenue                 0
dtype: int64
```

---

```
DataFrame 9:
Interests                    0
Users                        0
New users                    0
Engaged sessions             0
Engagement rate              0
Engaged sessions per user    0
```

```
Average engagement time    0
Event count                  0
Conversions                  0
Total revenue                0
dtype: int64
```

---

```
DataFrame 10:
Language                     0
Users                       0
New users                   0
Engaged sessions            0
Engagement rate             0
Engaged sessions per user   0
Average engagement time     0
Event count                  0
Conversions                  0
Total revenue                0
dtype: int64
```

---

```
DataFrame 11:
Age                         0
Users                       0
New users                   0
Engaged sessions            0
Engagement rate             0
Engaged sessions per user   0
Average engagement time     0
Event count                  0
Conversions                  0
Total revenue                0
dtype: int64
```

---

```
DataFrame 12:
Session Google Ads campaign  0
Users                       0
Sessions                     0
Engaged sessions            0
Google Ads clicks            0
Google Ads cost              0
Google Ads cost per click    0
Conversions                  0
Cost per conversion          0
Event count                  0
Total revenue                0
Return on ad spend           0
dtype: int64
```

---

```
In [6]: df3 = df3.dropna()
df3.isnull().sum()
```

```
Out[6]: Event name          0
Event count                0
Total users                0
Event count per user       0
Total revenue              0
dtype: int64
```

(Now, there is no null value & seems fine)

## For a Data Visualization, I choose the 'Seaborn' & 'matplotlib' library...

### User Acquisition

Let's analyse in the User Acquisition Data & let's which has to be improve...

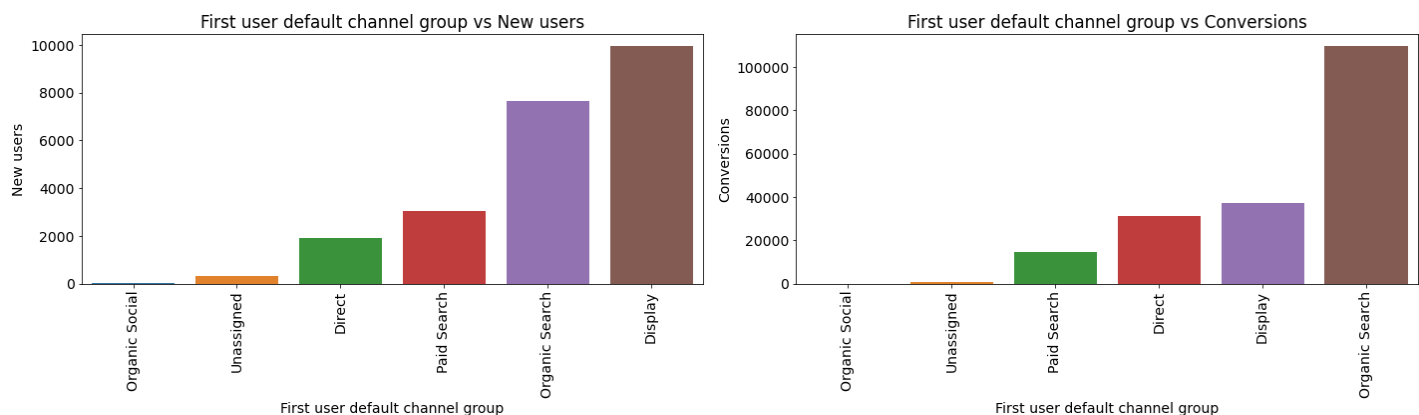
```
In [7]: plt.rcParams.update({'font.size': 14})

fig, axs = plt.subplots(nrows=1, ncols=2, figsize=(20, 6))

# channel group vs New users...
df1 = df1.sort_values('New users')
sns.barplot(x='First user default channel group', y='New users', data=df1, ax=axs[0])
axs[0].set_title('First user default channel group vs New users')
axs[0].set_xlabel('First user default channel group')
axs[0].set_ylabel('New users')
axs[0].tick_params(axis='x', rotation=90)

# channel group vs Conversions...
df1 = df1.sort_values('Conversions')
sns.barplot(x='First user default channel group', y='Conversions', data=df1, ax=axs[1])
axs[1].set_title('First user default channel group vs Conversions')
axs[1].set_xlabel('First user default channel group')
axs[1].set_ylabel('Conversions')
axs[1].tick_params(axis='x', rotation=90)

plt.tight_layout()
plt.show()
```



from the data, the Organic search is play a vital role for a conversion.

### Traffic Acquisition

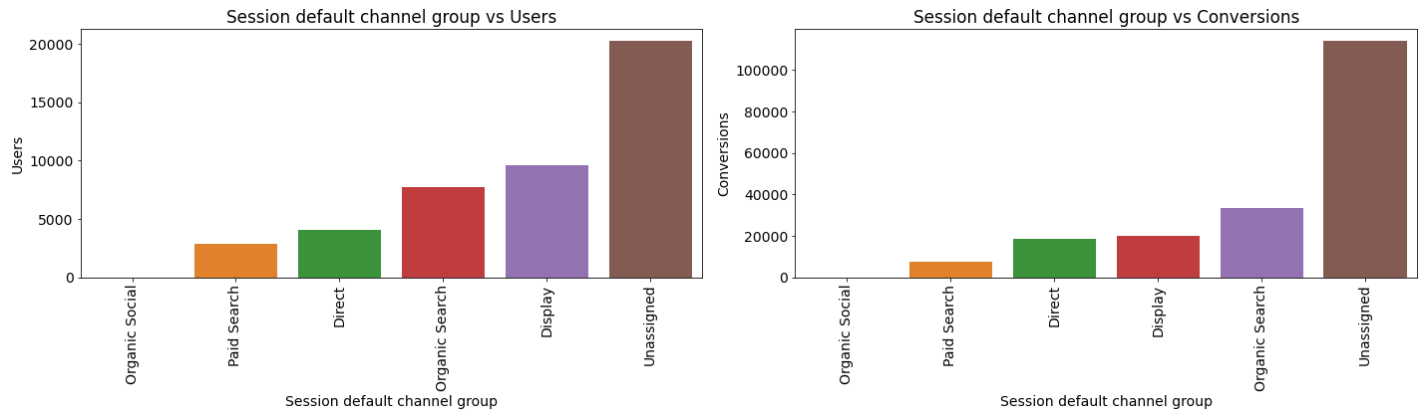
```
In [8]: plt.rcParams.update({'font.size': 14})

fig, axs = plt.subplots(ncols=2, figsize=(20, 6))

df2 = df2.sort_values('Users')
sns.barplot(x='Session default channel group', y='Users', data=df2, ax=axs[0])
axs[0].set_title('Session default channel group vs Users')
axs[0].set_xlabel('Session default channel group')
axs[0].set_ylabel('Users')
axs[0].tick_params(axis='x', rotation=90)
```

```
df2 = df2.sort_values('Conversions')
sns.barplot(x='Session default channel group', y='Conversions', data=df2, ax=axes[1])
axes[1].set_title('Session default channel group vs Conversions')
axes[1].set_xlabel('Session default channel group')
axes[1].set_ylabel('Conversions')
axes[1].tick_params(axis='x', rotation=90)

plt.tight_layout()
plt.show()
```



"Here, the Unassigned category beats organic search when it comes to conversions."

"Paid search should be improved for the conversion rate."

"The conversion rate is low in paid search."

## Event Report

In [9]:

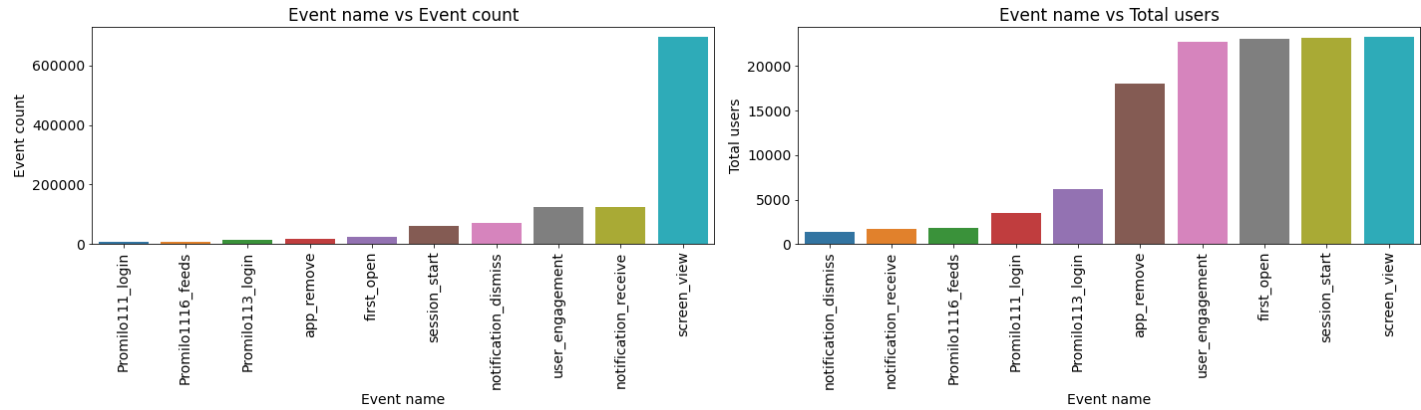
```
top_events = df3.nlargest(10, 'Event count')['Event name']

fig, axes = plt.subplots(ncols=2, figsize=(20, 6))

df_top = df3[df3['Event name'].isin(top_events)].sort_values('Event count')
sns.barplot(x='Event name', y='Event count', data=df_top, ax=axes[0])
axes[0].set_title('Event name vs Event count')
axes[0].set_xlabel('Event name')
axes[0].set_ylabel('Event count')
axes[0].tick_params(axis='x', rotation=90)

df_top = df3[df3['Event name'].isin(top_events)].sort_values('Total users')
sns.barplot(x='Event name', y='Total users', data=df_top, ax=axes[1])
axes[1].set_title('Event name vs Total users')
axes[1].set_xlabel('Event name')
axes[1].set_ylabel('Total users')
axes[1].tick_params(axis='x', rotation=90)

plt.tight_layout()
plt.show()
```

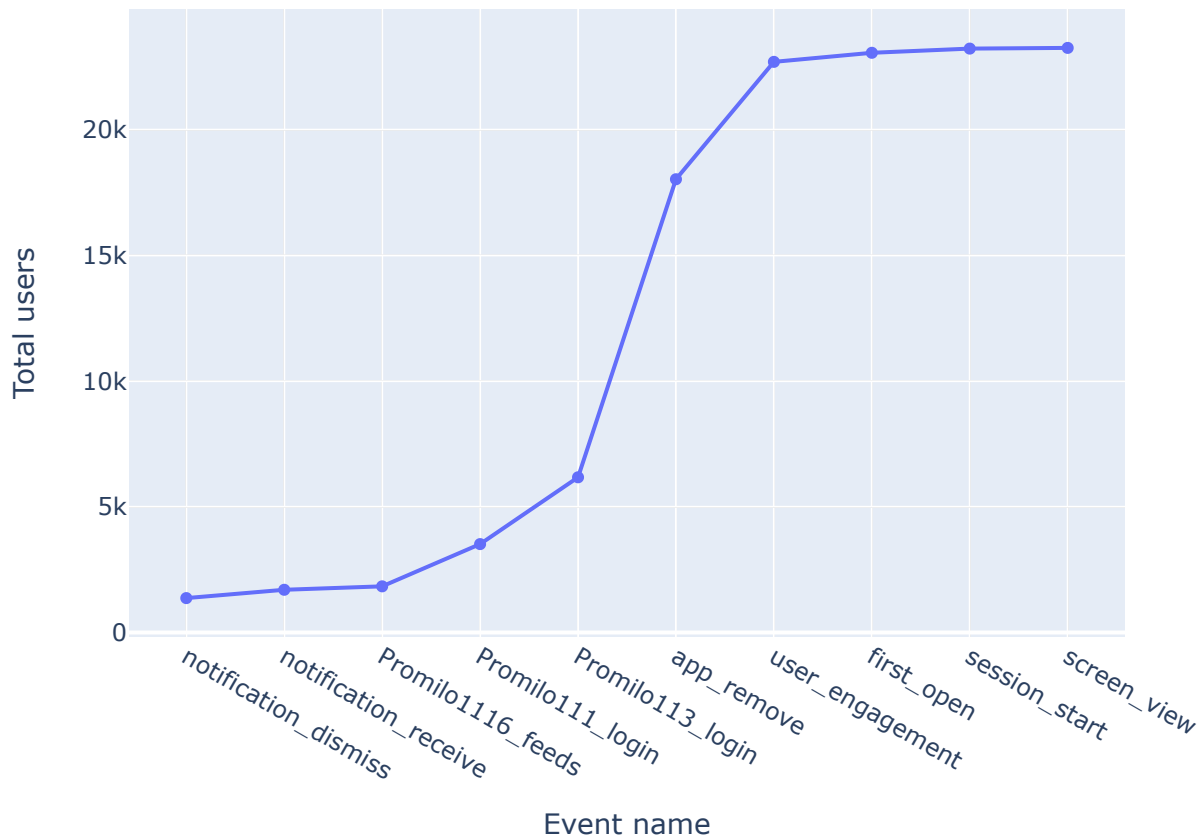


In [10]:

```
import plotly.express as px

# For Better Understanding... (Event name vs Total users)
df_top = df3[df3['Event name'].isin(top_events)].sort_values('Total users')
fig1 = px.line(df_top, x='Event name', y='Total users', title='Event name vs Total users')
fig1.update_traces(mode='markers+lines')
fig1.show()
```

## Event name vs Total users



When it comes to Promilo, the user's time spent on the app is minimal, and screen views predominantly take precedence, with the user login rate also being very low.

## Conversion Report

In [11]:

```
df4 = dfs['Conversion Report']
```



```
df4.head()
```

Out[11]:

	Event name	Conversions	Total users	Total revenue
0	notification_receive	94890	1311	0
1	session_start	56203	21674	0
2	first_open	22872	23059	0
3	app_remove	12468	12538	0
4	Promilo111_otp_screen	1738	855	0

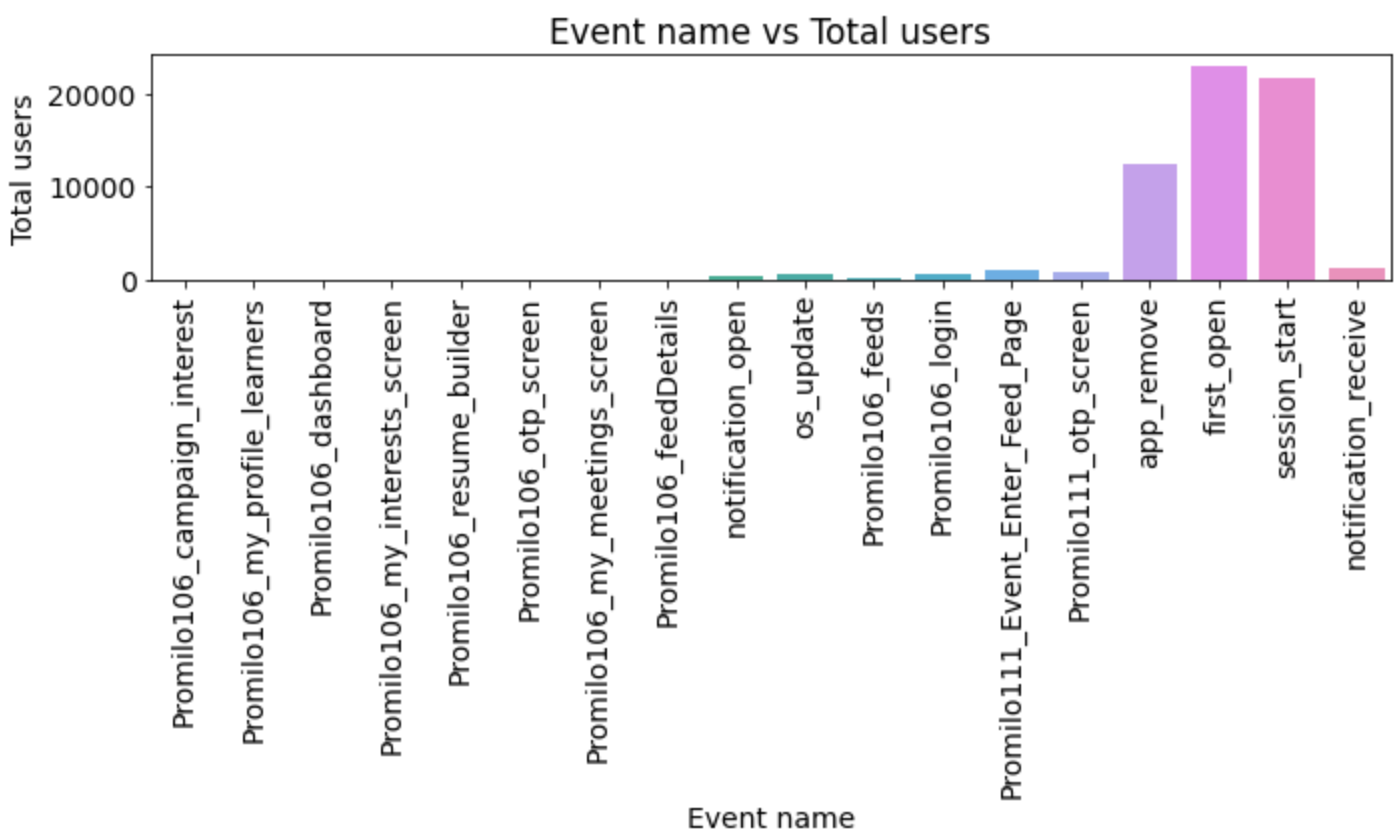
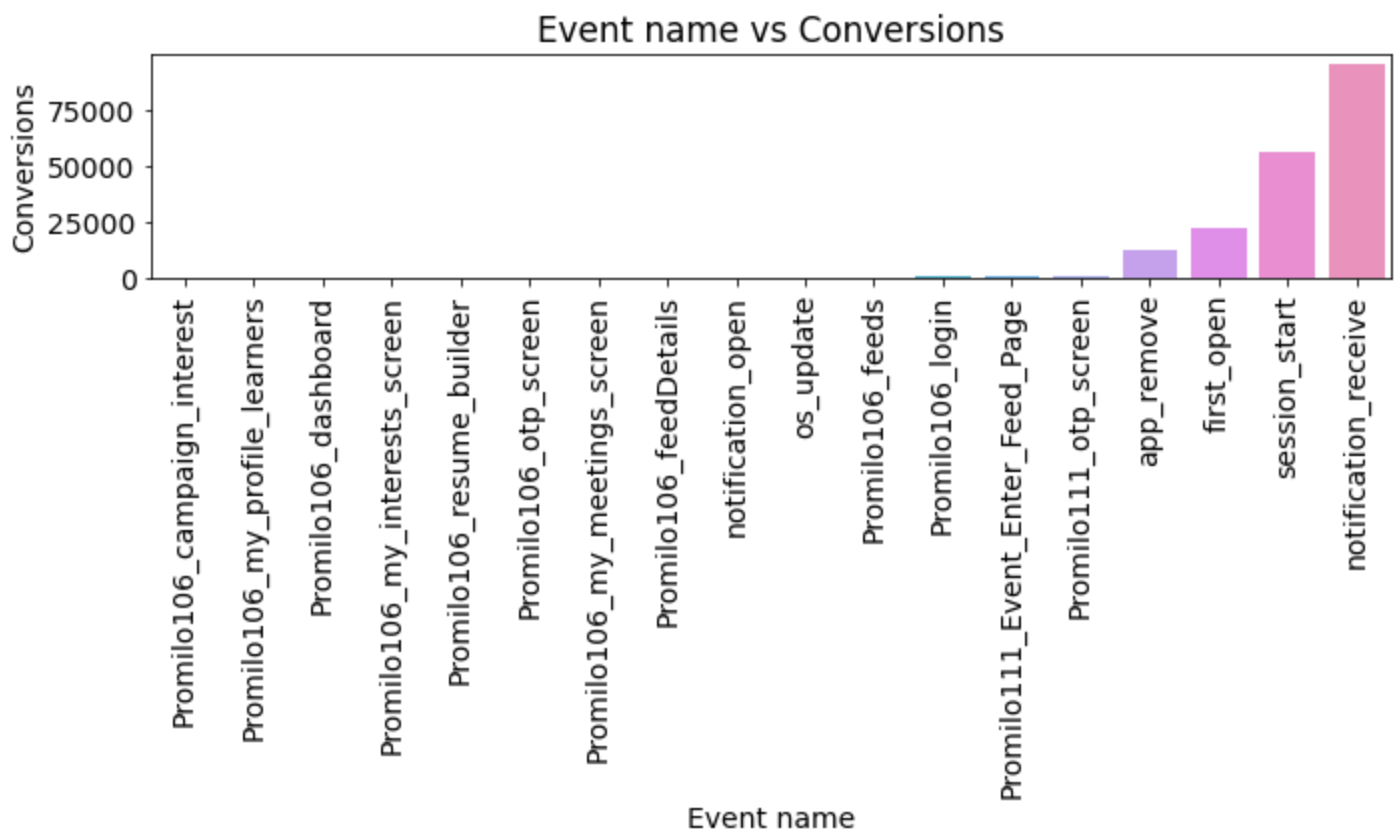
In [12]:

```
fig, axs = plt.subplots(nrows=2, figsize=(10, 12))

# Event name Vs Conversion...
df4_sorted = df4.sort_values('Conversions')
sns.barplot(x='Event name', y='Conversions', data=df4_sorted, ax=axs[0])
axs[0].set_title('Event name vs Conversions')
axs[0].set_xlabel('Event name')
axs[0].set_ylabel('Conversions')
axs[0].tick_params(axis='x', rotation=90)

# Event name vs Total Users...
sns.barplot(x='Event name', y='Total users', data=df4_sorted, ax=axs[1])
axs[1].set_title('Event name vs Total users')
axs[1].set_xlabel('Event name')
axs[1].set_ylabel('Total users')
axs[1].tick_params(axis='x', rotation=90)

plt.tight_layout()
plt.show()
```



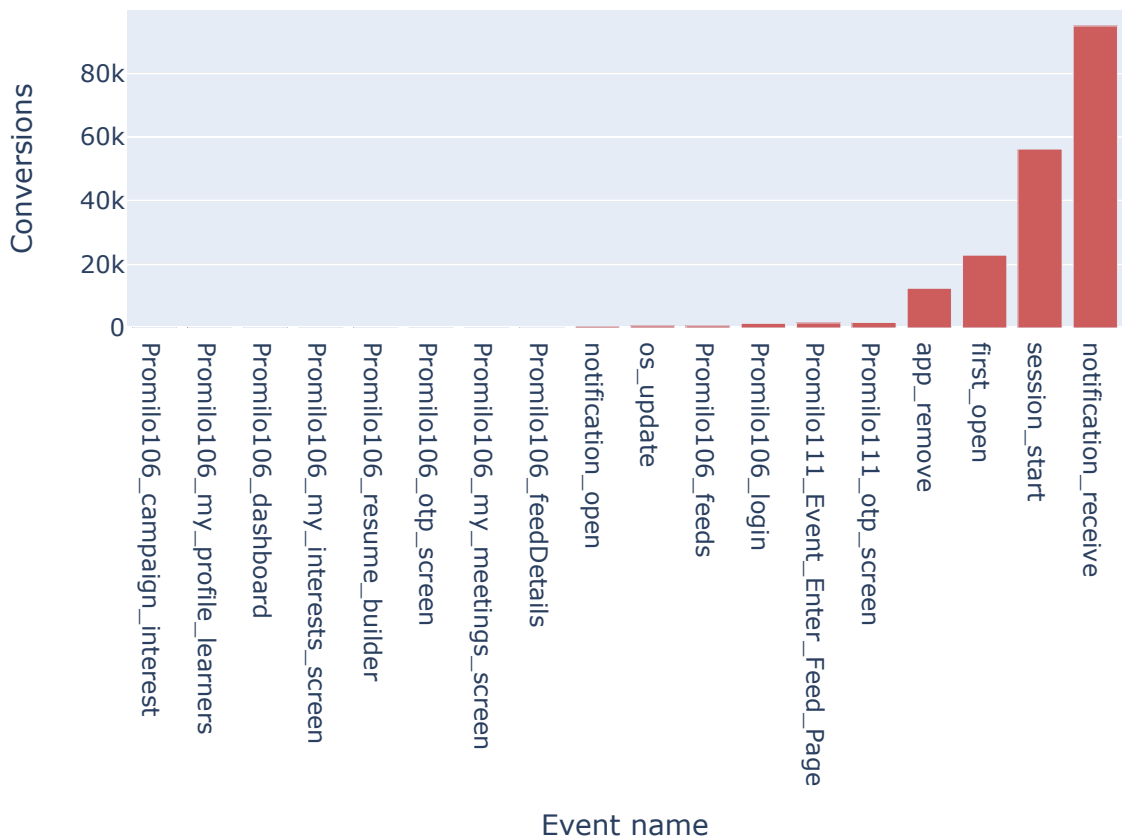
```
In [13]: import plotly.express as px

# Event name Vs Conversions...
df4_sorted = df4.sort_values('Conversions')
fig2 = px.bar(df4_sorted, x='Event name', y='Conversions', title='Event name vs Conversions')
fig2.update_traces(marker_color='indianred')

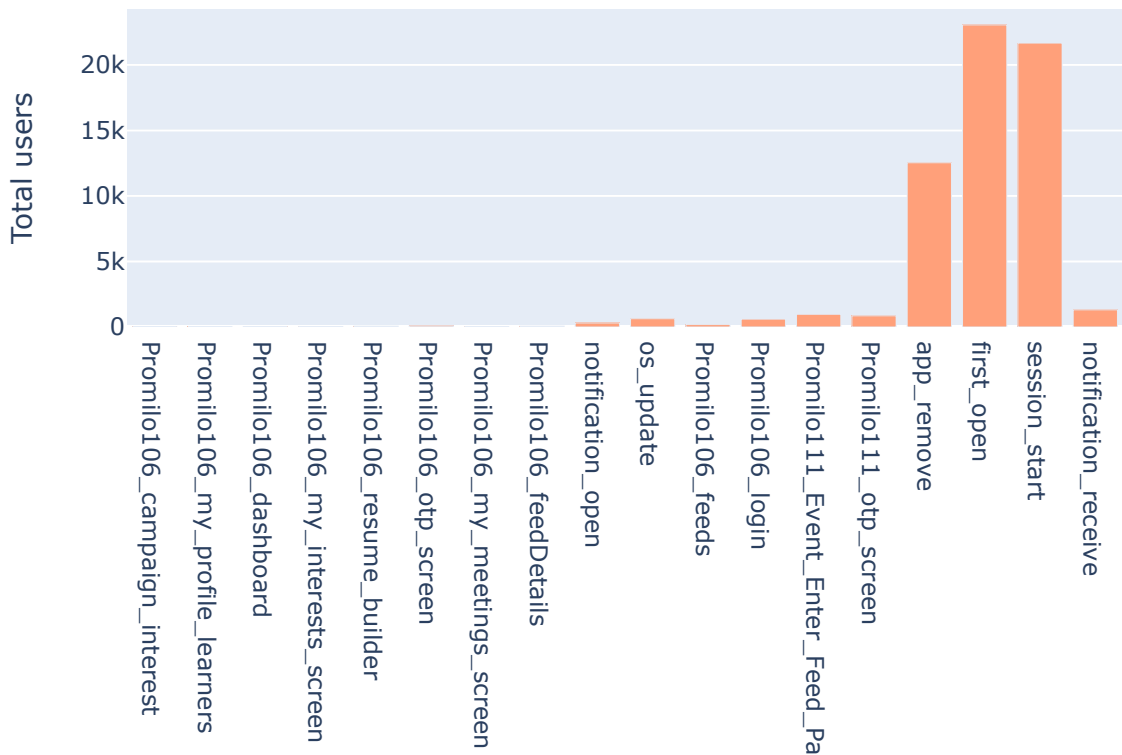
# Event name vs Total Users...
fig3 = px.bar(df4_sorted, x='Event name', y='Total users', title='Event name vs Total users')
fig3.update_traces(marker_color='lightsalmon')
```

```
fig2.show()  
fig3.show()
```

Event name vs Conversions



Event name vs Total users



In the Conversion Vs Event Name chart, notifications play a vital role in conversions. Perhaps, when a customer sees a notification, it triggers the user to engage with it.

## Page path and screen class

```
In [14]: print(df5.columns)
df5.head(2)
```

```
Index(['Page path and screen class', 'Views', 'Users', 'Views per user',
      'Average engagement time', 'Event count', 'Conversions',
      'Total revenue'],
      dtype='object')
```

Out[14]:

	Page path and screen class	Views	Users	Views per user	Average engagement time	Event count	Conversions	Total revenue
0	Flutter	156708	8726	17.95874	83.41222	203901	328	0
1	MainActivity	44326	8978	4.93718	78.29216	53374	101	0

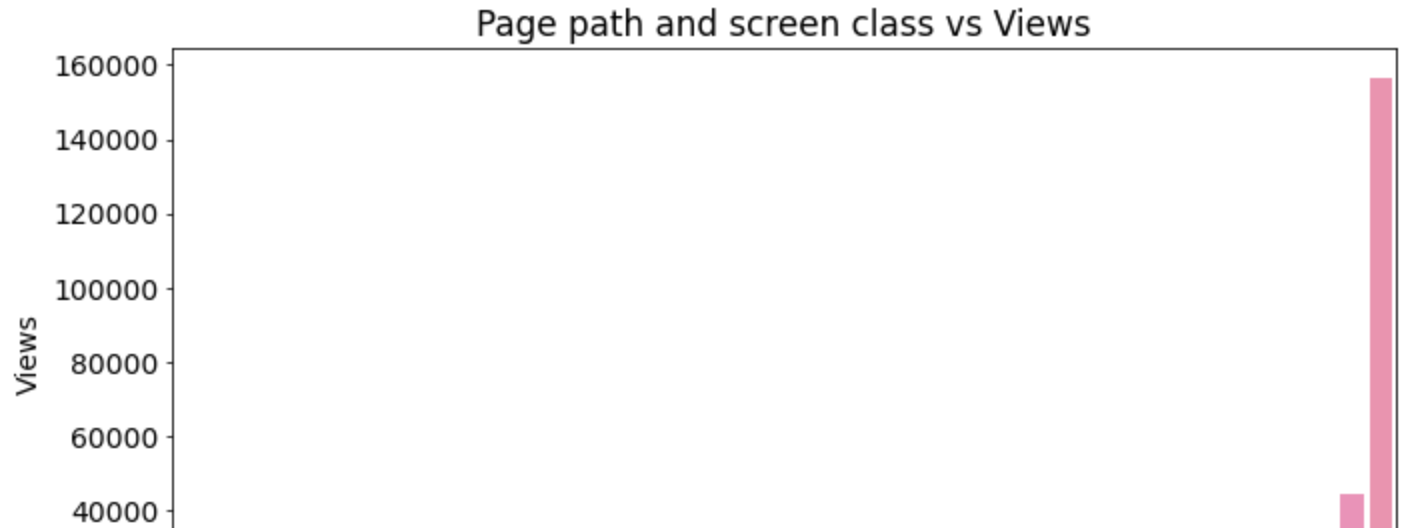
```
In [15]: fig, axs = plt.subplots(nrows=2, figsize=(10, 18))

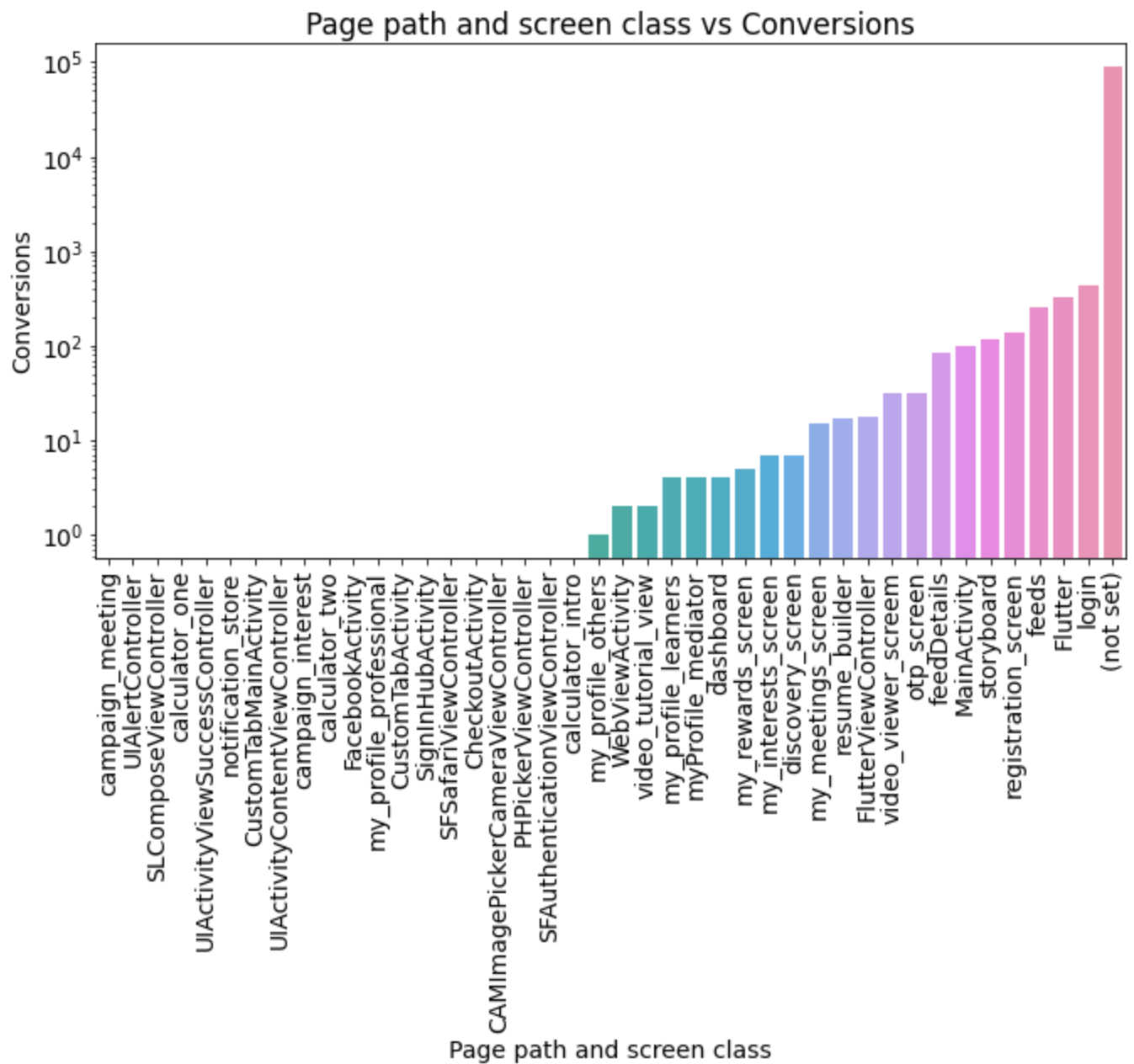
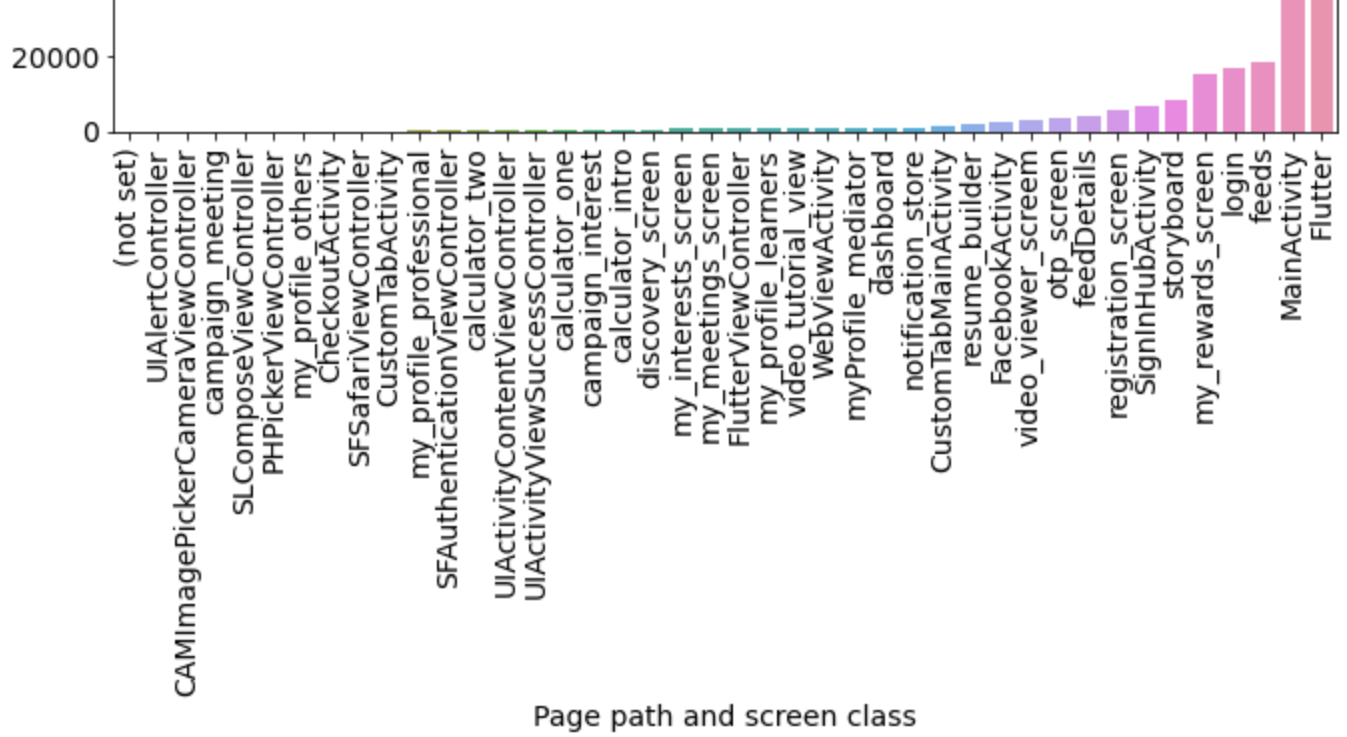
df5_sorted = df5.sort_values('Views')
sns.barplot(x='Page path and screen class', y='Views', data=df5_sorted, ax=axs[0])
axs[0].set_title('Page path and screen class vs Views')
axs[0].set_xlabel('Page path and screen class')
axs[0].set_ylabel('Views')
axs[0].tick_params(axis='x', rotation=90)

df5_sorted = df5.sort_values('Conversions')
sns.barplot(x='Page path and screen class', y='Conversions', data=df5_sorted, ax=axs[1])
axs[1].set_title('Page path and screen class vs Conversions')
axs[1].set_xlabel('Page path and screen class')
axs[1].set_ylabel('Conversions')
axs[1].tick_params(axis='x', rotation=90)

# Set y-axis to logarithmic scale
axs[1].set_yscale('log')

plt.tight_layout()
plt.show()
```





In these pages, such as the main activity, Flutter, login, and storyboard, are highly viewed by the users.

# Demographics Report

In [16]:

```
df6 = dfs['Demographics Report']
print(df6.columns)
df6.head()
```

Index(['Country', 'Users', 'New users', 'Engaged sessions', 'Engagement rate',  
 'Engaged sessions per user', 'Average engagement time', 'Event count',  
 'Conversions', 'Total revenue'],  
 dtype='object')

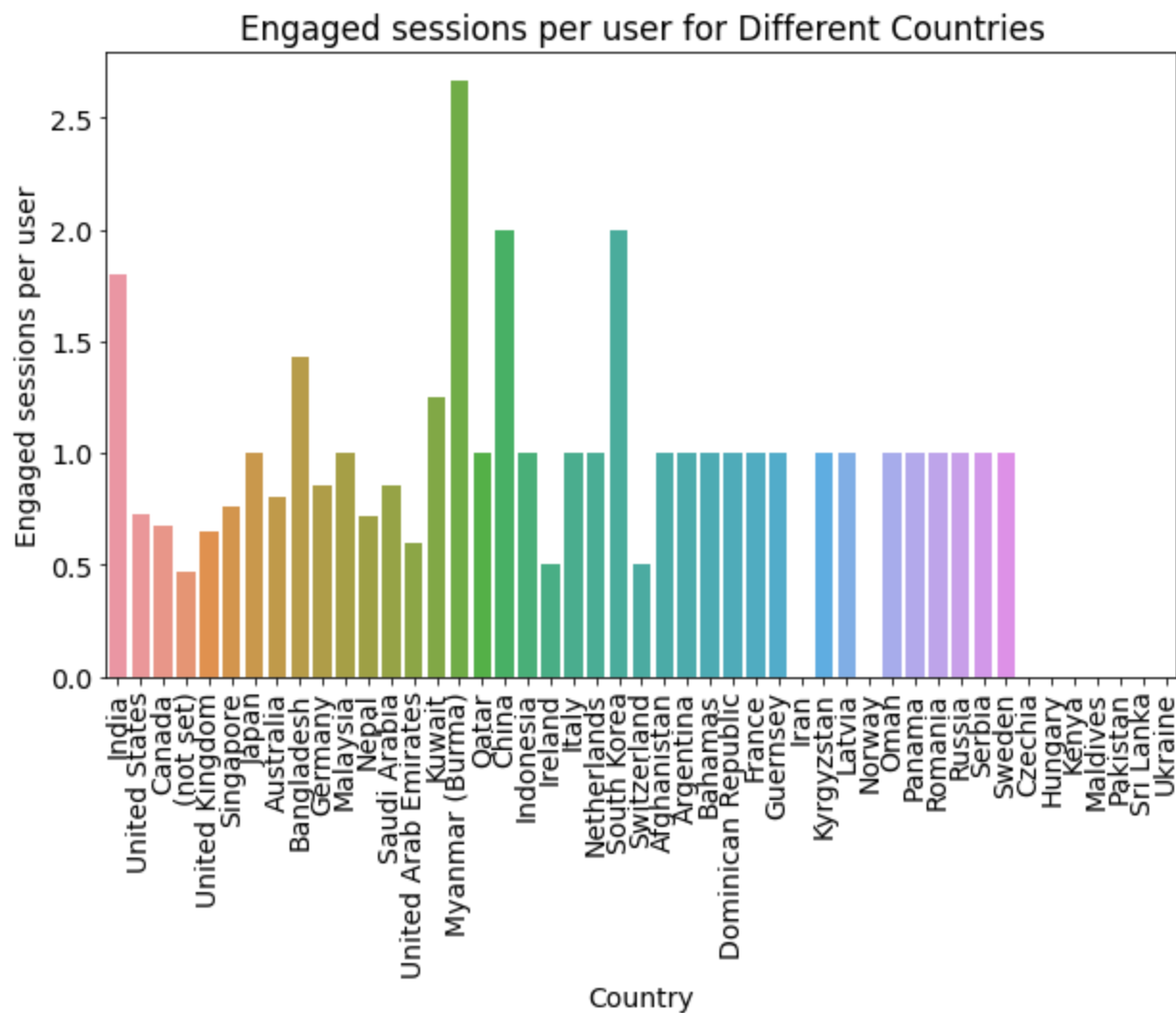
Out[16]:

	Country	Users	New users	Engaged sessions	Engagement rate	Engaged sessions per user	Average engagement time	Event count	Conversions	Total revenue
0	India	23024	22528	41479	0.593626	1.801555	334.81660	1312097	192766	0
1	United States	272	213	197	0.491272	0.724265	50.96324	3157	643	0
2	Canada	37	18	25	0.416667	0.675676	43.21622	410	121	0
3	(not set)	36	36	17	0.459459	0.472222	24.80556	241	54	0
4	United Kingdom	20	8	13	0.371429	0.650000	61.85000	289	43	0

In [17]:

```
df6 = dfs['Demographics Report']

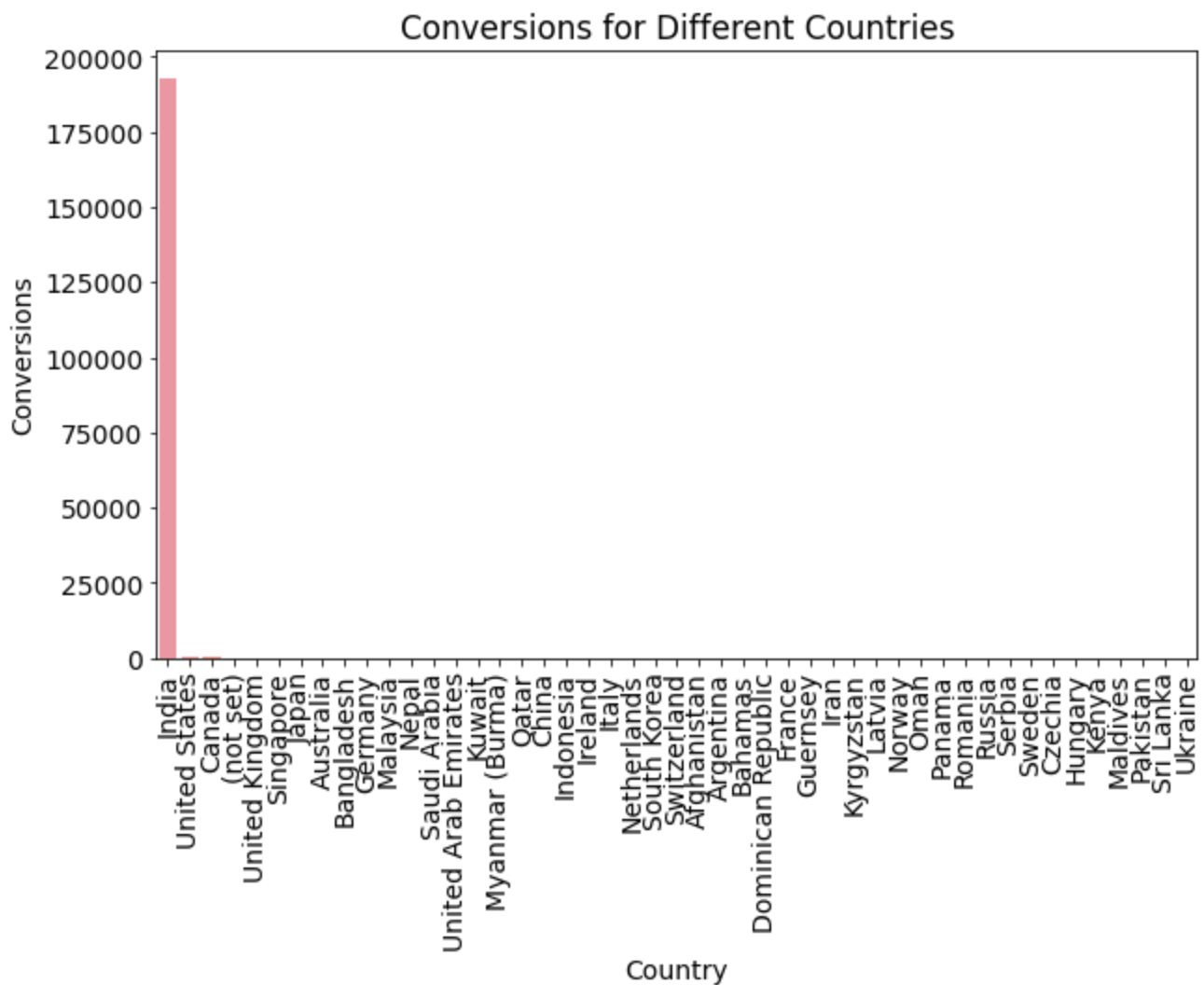
plt.figure(figsize=(10, 6))
sns.barplot(x=df6['Country'], y=df6['Engaged sessions per user'])
plt.title('Engaged sessions per user for Different Countries')
plt.xlabel('Country')
plt.ylabel('Engaged sessions per user')
plt.xticks(rotation=90)
plt.show()
```



The Users from the Myanmar(Burma) are more engaging in this session...

In [18]:

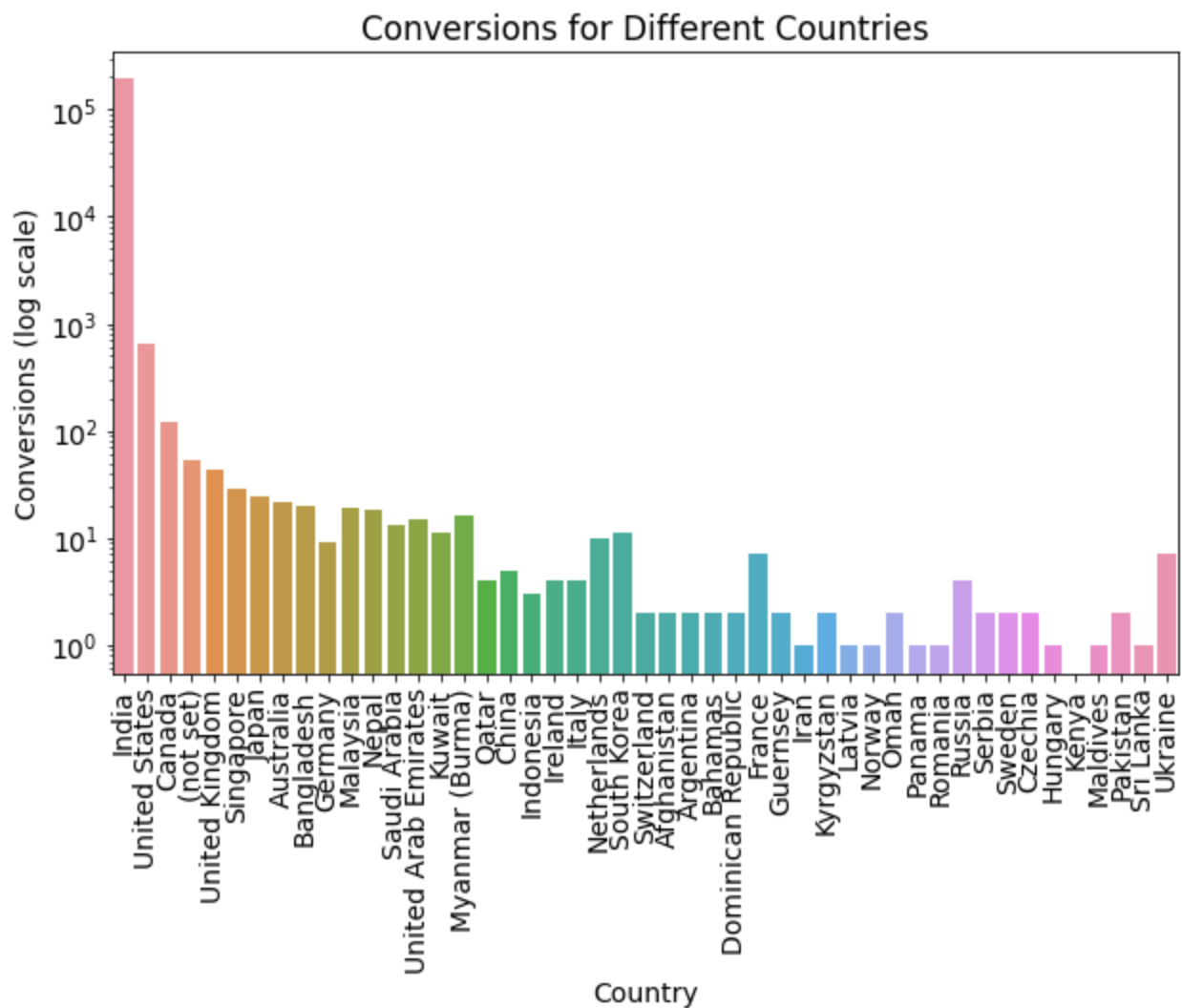
```
plt.figure(figsize=(10, 6))
sns.barplot(x=df6['Country'], y=df6['Conversions'])
plt.title('Conversions for Different Countries')
plt.xlabel('Country')
plt.ylabel('Conversions')
plt.xticks(rotation=90)
plt.show()
```



India has a higher conversion rate

```
In [19]: plt.figure(figsize=(10, 6))
sns.barplot(x=df6['Country'], y=df6['Conversions'])
plt.title('Conversions for Different Countries')
plt.xlabel('Country')
plt.ylabel('Conversions (log scale)')
plt.yscale('log') # Set the y-axis to log scale
plt.xticks(rotation=90)
plt.show()
```





when it comes to country, India plays a vital role in conversion.

## Citywise Report

```
In [20]: df7 = dfs['Citiwise Report']
df7.head()
```

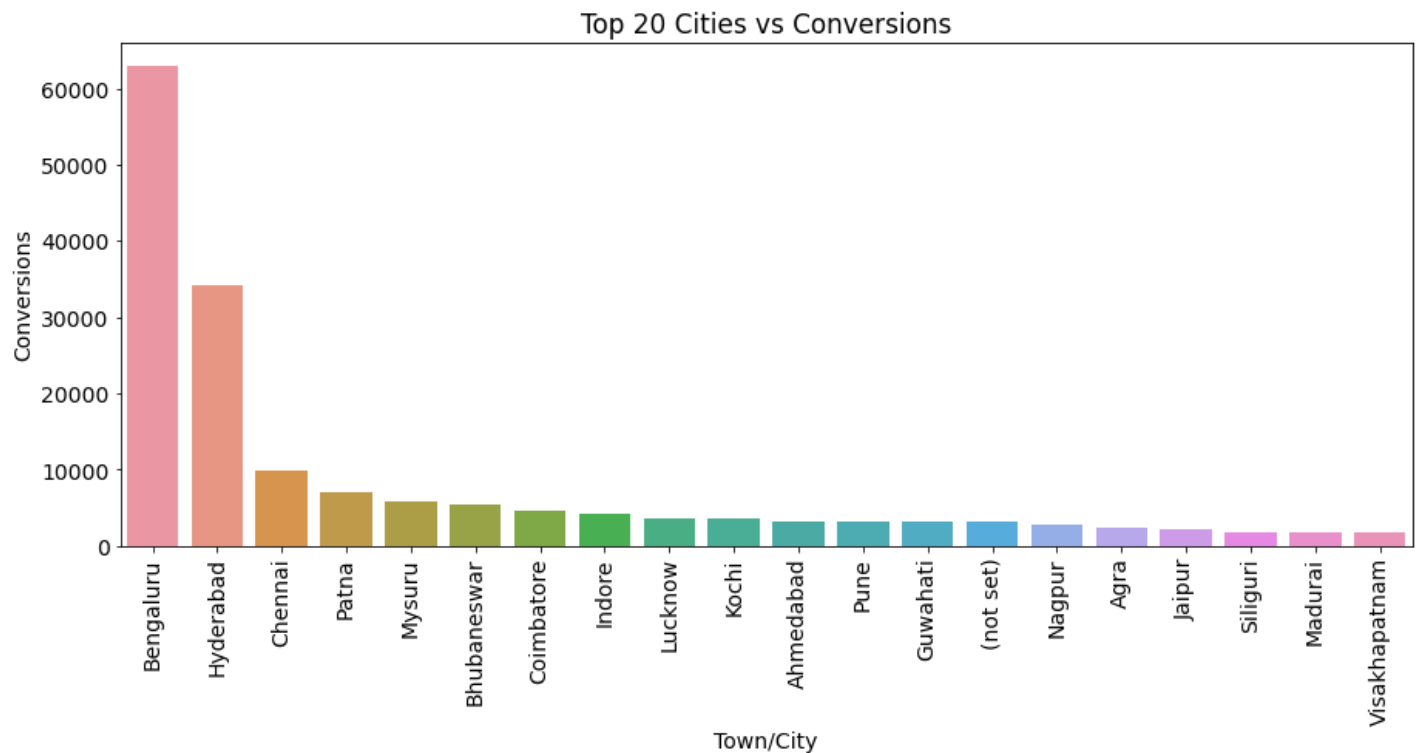
Out[20]:

	Town/City	Users	New users	Engaged sessions	Engagement rate	Engaged sessions per user	Average engagement time	Event count	Conversions	Total revenue
0	Bengaluru	6097	5685	15013	0.769385	2.462359	762.20550	607200	62939	0
1	Patna	1594	1467	2127	0.440646	1.334379	98.22208	38830	6980	0
2	Hyderabad	1038	920	1578	0.569264	1.520231	243.69080	96826	34103	0
3	Indore	983	915	1241	0.426460	1.262462	67.89115	21383	4121	0
4	Lucknow	897	839	1125	0.450180	1.254181	83.40580	21041	3650	0

```
In [21]: top_cities = df7.nlargest(20, 'Conversions')

plt.figure(figsize=(15, 6))
sns.barplot(x='Town/City', y='Conversions', data=top_cities)
plt.title('Top 20 Cities vs Conversions')
plt.xlabel('Town/City')
plt.ylabel('Conversions')
```

```
plt.xticks(rotation=90) # Rotate x-axis labels by 90 degrees
plt.show()
```



In citiwise, the bengaluru has the highest conversion rate, second highest is hyderabad.

## Gender Report

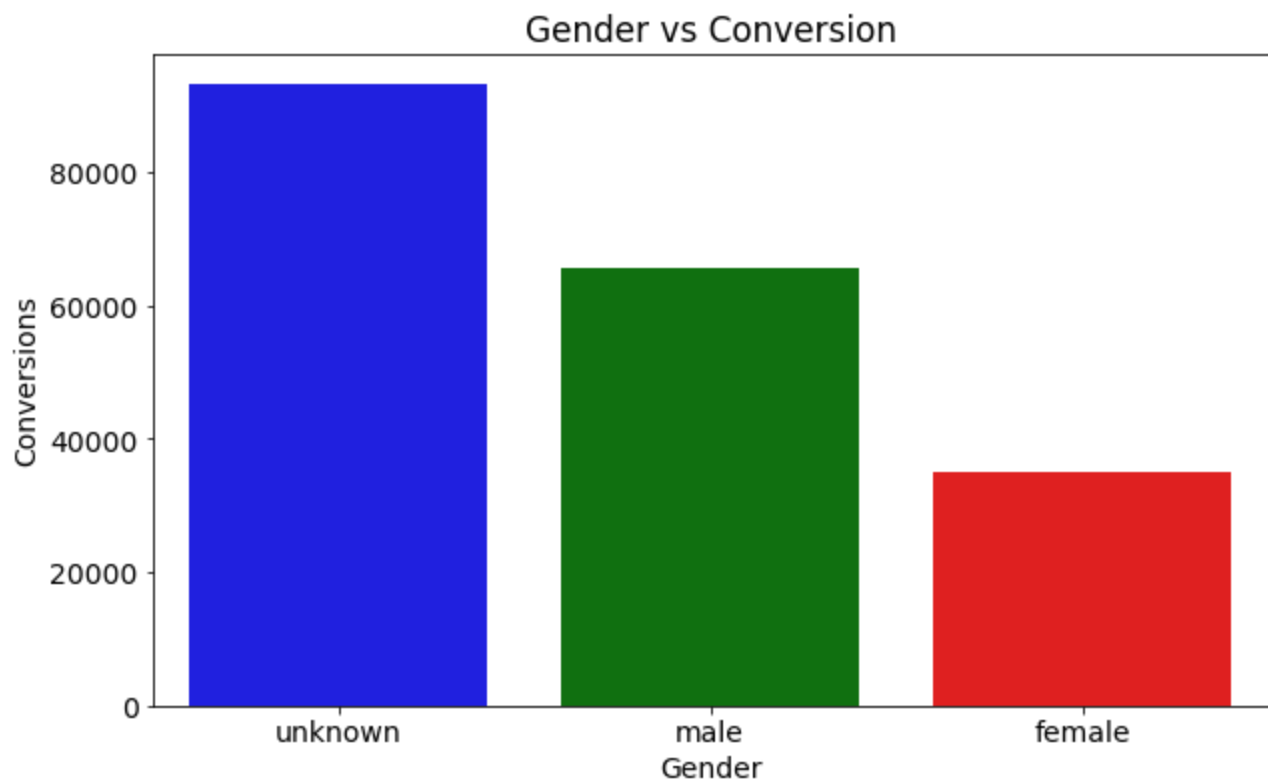
```
In [22]: df8 = dfs['Gender Report']
df8.head()
```

```
Out[22]:
```

	Gender	Users	New users	Engaged sessions	Engagement rate	Engaged sessions per user	Average engagement time	Event count	Conversions	Total revenue
0	unknown	13142	12691	23161	0.564077	1.762365	439.5776	761771	93180	0
1	male	7218	5877	10467	0.543091	1.450125	128.2319	282504	65651	0
2	female	4944	4304	7877	0.637710	1.593244	208.7407	274254	35083	0

```
In [23]: genders = df8['Gender']
conversions = df8['Conversions']

plt.figure(figsize=(10,6))
sns.barplot(x=genders, y=conversions, palette=['blue', 'green', 'red'])
plt.title('Gender vs Conversion')
plt.xlabel('Gender')
plt.ylabel('Conversions')
plt.show()
```



Users hide their gender when logging in, so it may be considered as unknown. Apart from that, males have a higher conversion rate.

## User By Instrest

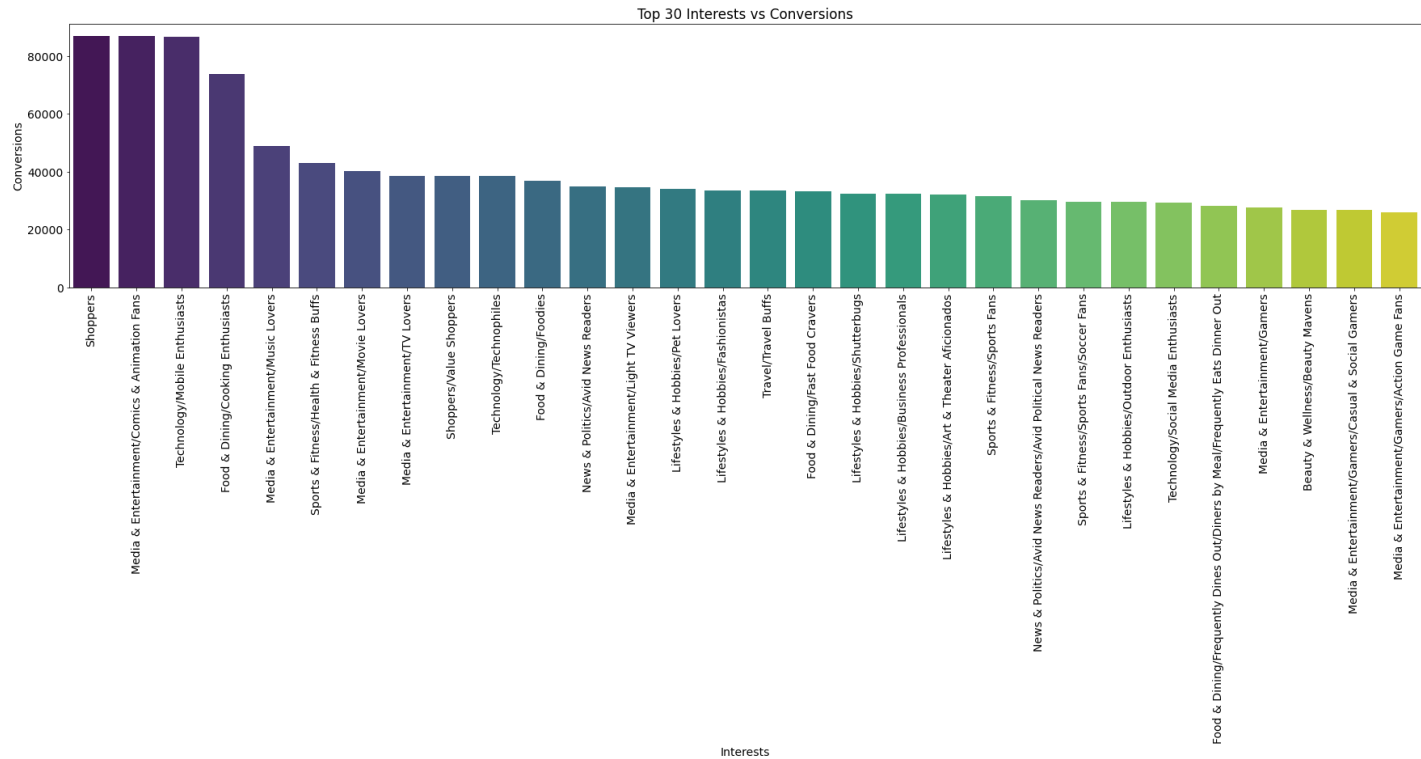
```
In [24]: df9 = dfs['User By Interest']  
  
df9.head()
```

Out[24]:

	Interests	Users	New users	Engaged sessions	Engagement rate	Engaged sessions per user	Average engagement time	Event count	Conversions	Total revenue
0	Shoppers	10950	9256	15652	0.581534	1.429406	162.8347	490664	86846	0
1	Media & Entertainment/Comics & Animation Fans	10946	9247	15680	0.583008	1.432487	165.1772	491025	86845	0
2	Technology/Mobile Enthusiasts	10934	9239	15619	0.582451	1.428480	162.6945	489353	86742	0
3	Food & Dining/Cooking Enthusiasts	8410	6970	12332	0.602325	1.466350	176.9567	409713	73814	0
4	Sports & Fitness/Health & Fitness Buffs	5844	4580	8226	0.588328	1.407598	155.1451	257831	43074	0

```
In [25]: df_top30 = df9.sort_values('Conversions', ascending=False).head(30)  
  
interests = df_top30['Interests']  
conversions = df_top30['Conversions']  
  
plt.figure(figsize=(30,6))  
sns.barplot(x=interests, y=conversions, palette='viridis')
```

```
plt.title('Top 30 Interests vs Conversions')
plt.xlabel('Interests')
plt.ylabel('Conversions')
plt.xticks(rotation=90) # Rotate x-axis labels by 90 degrees
plt.show()
```

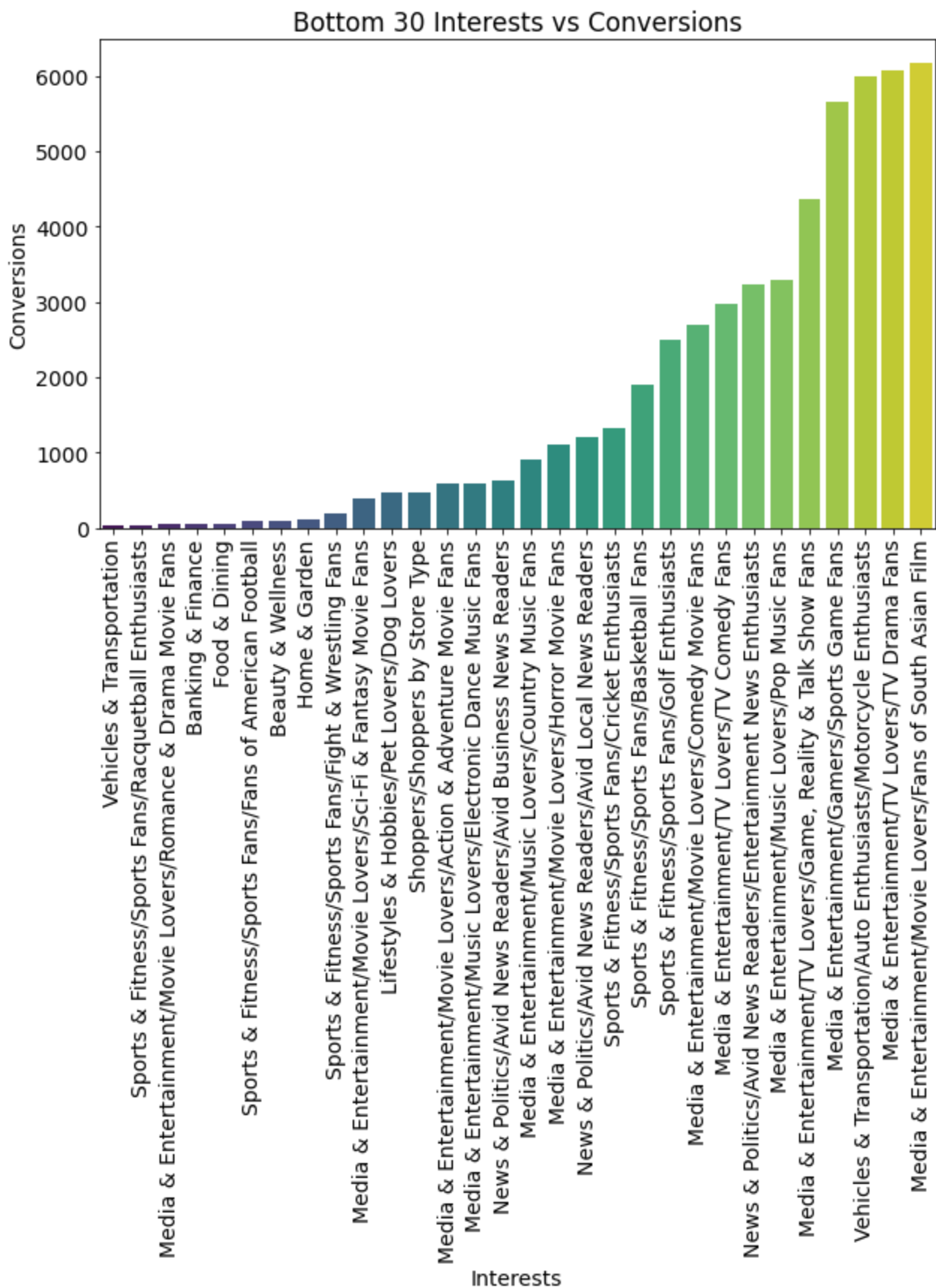


In [26]:

```
# Sort by 'Conversions' and take bottom 30
df_bottom30 = df9.sort_values('Conversions', ascending=True).head(30)

interests = df_bottom30['Interests']
conversions = df_bottom30['Conversions']

plt.figure(figsize=(10,6))
sns.barplot(x=interests, y=conversions, palette='viridis')
plt.title('Bottom 30 Interests vs Conversions')
plt.xlabel('Interests')
plt.ylabel('Conversions')
plt.xticks(rotation=90) # Rotate x-axis labels by 90 degrees
plt.show()
```



from the top & bottom analysis of User's Interest, User highly focusing the shopping, media entertainment and technology and mobile enthusiasts.

and users are not focusing on vehicles, fitness, & banking category

## User by Language.

```
In [27]: df10 = dfs['User by Language']
```

```
df10.head(3)
```

Out[27]:

	Language	Users	New users	Engaged sessions	Engagement rate	Engaged sessions per user	Average engagement time	Event count	Conversions	Total revenue
0	English	22495	21990	40639	0.595147	1.806579	341.36350	1297970	189946	0
1	Hindi	586	552	798	0.406314	1.361775	60.03413	13523	2699	0
2	Marathi	85	84	98	0.426087	1.152941	38.48235	1589	323	0

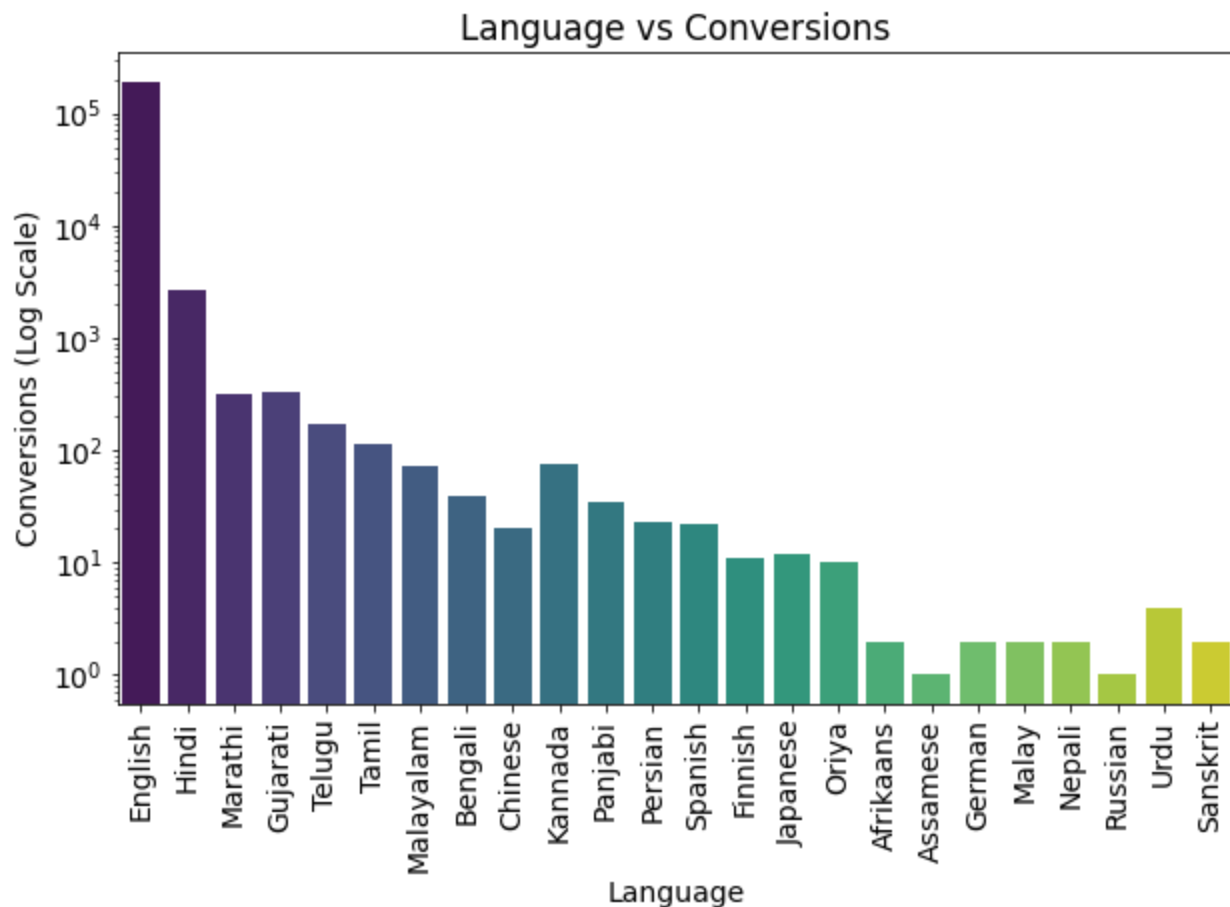
In [28]:

```
languages = df10['Language']
conversions = df10['Conversions']

plt.figure(figsize=(10,6))
sns.barplot(x=languages, y=conversions, palette='viridis')
plt.title('Language vs Conversions')
plt.xlabel('Language')
plt.ylabel('Conversions (Log Scale)')
plt.xticks(rotation=90) # Rotate x-axis labels by 90 degrees

# Set y-axis to logarithmic scale
plt.yscale('log')

plt.show()
```

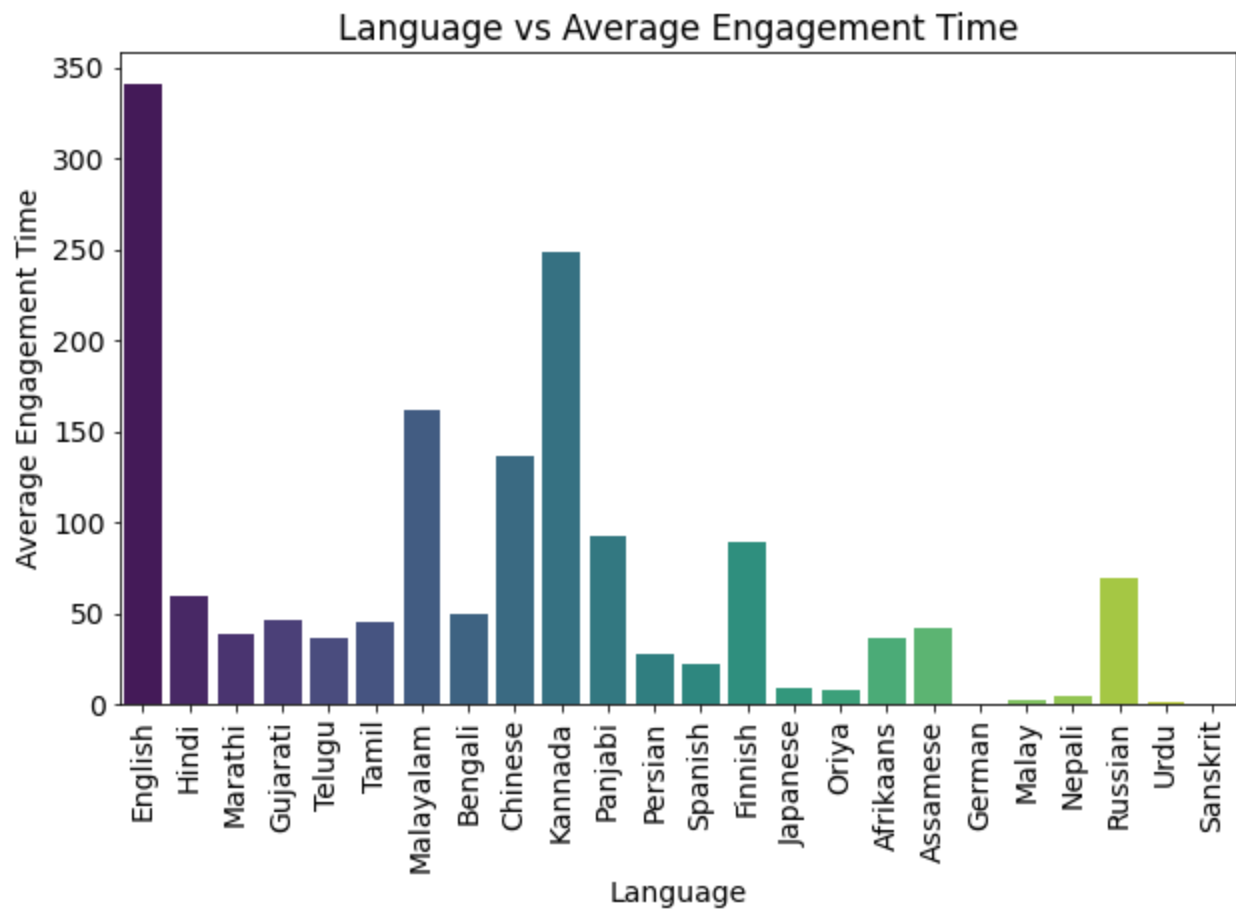


In [29]:

```
languages = df10['Language']
avg_engagement_time = df10['Average engagement time']

plt.figure(figsize=(10,6))
sns.barplot(x=languages, y=avg_engagement_time, palette='viridis')
plt.title('Language vs Average Engagement Time')
```

```
plt.xlabel('Language')
plt.ylabel('Average Engagement Time')
plt.xticks(rotation=90) # Rotate x-axis labels by 90 degrees
plt.show()
```



English Language is play mojour role when it comes to conversion and engagement time.

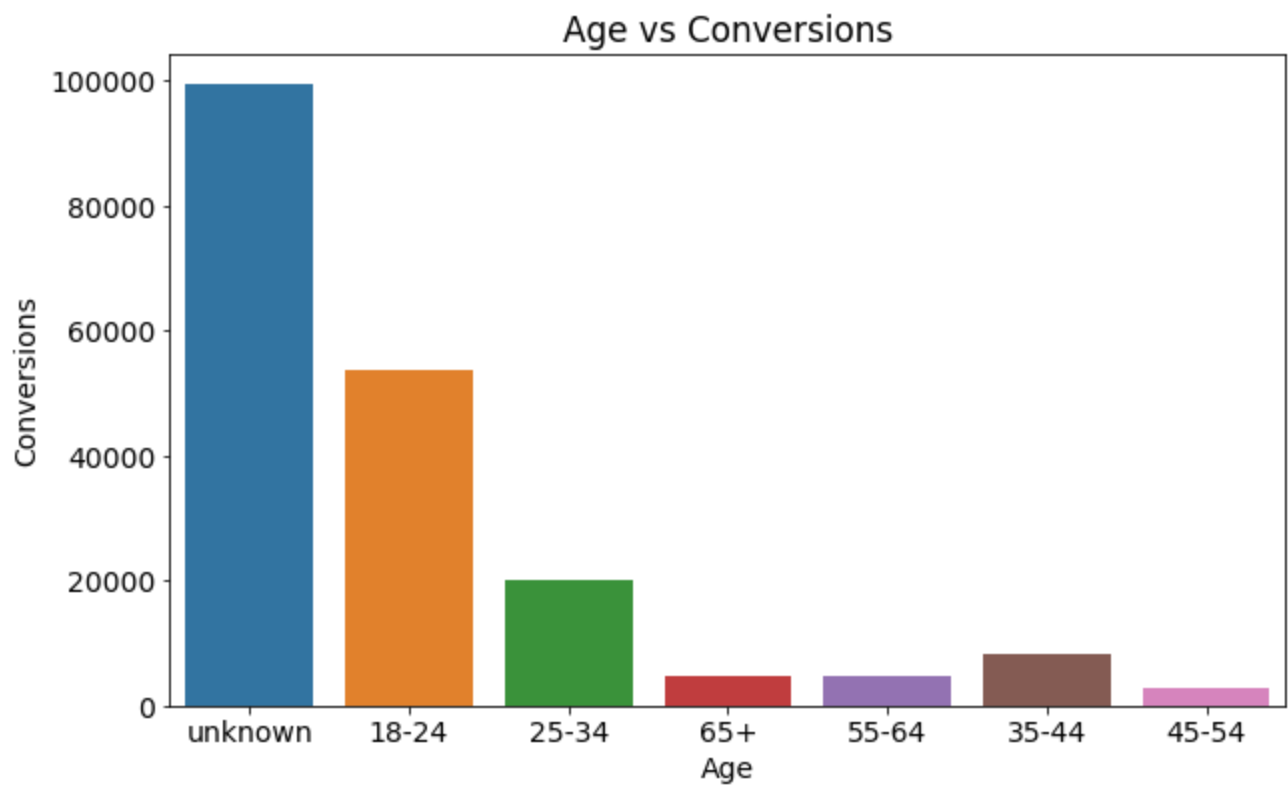
## User by Age

```
In [30]: df11 = dfs['User By Age']
df11.head(2)
```

Out[30]:

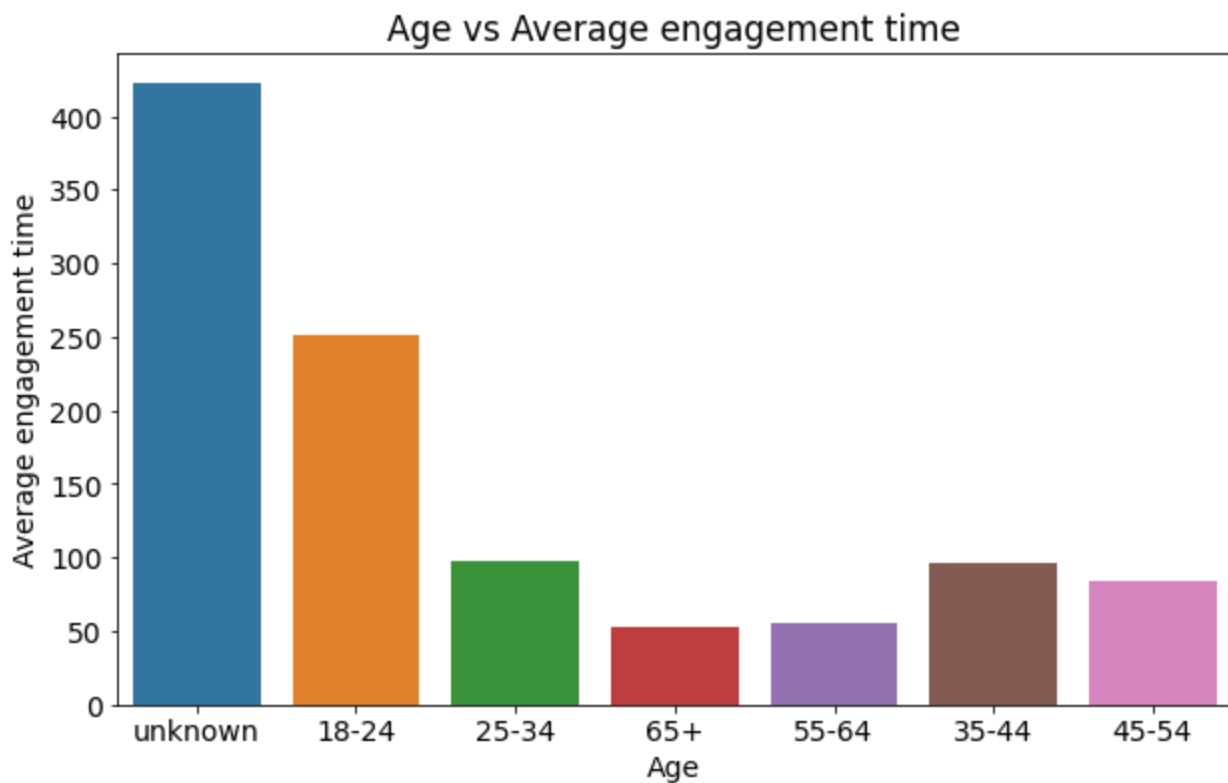
	Age	Users	New users	Engaged sessions	Engagement rate	Engaged sessions per user	Average engagement time	Event count	Conversions	Total revenue
0	unknown	14303	13636	24976	0.569098	1.746207	422.2233	817501	99310	0
1	18-24	4282	3678	7291	0.695308	1.702709	251.1630	309328	53661	0

```
In [31]: plt.figure(figsize=(10, 6))
sns.barplot(x='Age', y='Conversions', data=df11)
plt.title('Age vs Conversions')
plt.show()
```



In [32]:

```
plt.figure(figsize=(10, 6))
sns.barplot(x='Age', y='Average engagement time', data=df11)
plt.title('Age vs Average engagement time')
plt.show()
```



people hide their age, while log in. so its consider as a unknown. The unknown category are in higher conversion.

18-24 age people are second highest in conversion

45-54 age people are less when comes to conversion



65+ age people are heving less engagement time

# Google Ads Report

We need to more focus paid marketing campaign.

Let see what are the thing working and not working for a converison the Users

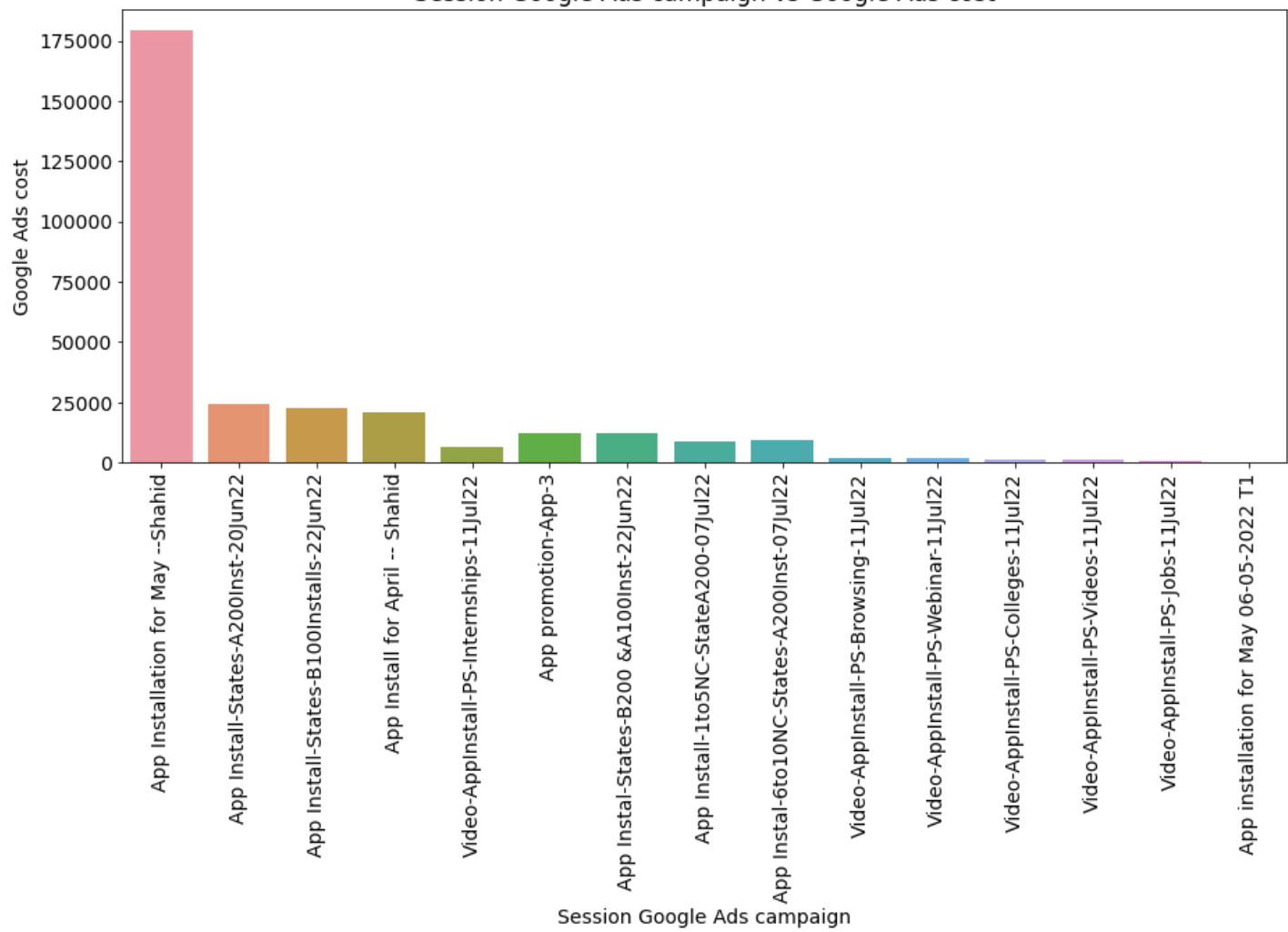
```
In [33]: df12 = dfs['Google Ads Report']
df12.head(2)
```

Out[33]:

	Session Google Ads campaign	Users	Sessions	Engaged sessions	Google Ads clicks	Google Ads cost	Google Ads cost per click	Conversions	Cost per conversion	Event count	Total revenue	Re o sj
0	App Installation for May -- Shahid	5429	10936	6276	147100	179175.00	1.218049	12257	14.61818	97802	0	
1	App Install- States- A200Inst- 20Jun22	842	1655	968	28742	24309.13	0.845770	1794	13.55024	15311	0	

```
In [34]: plt.figure(figsize=(15, 6))
sns.barplot(x='Session Google Ads campaign', y='Google Ads cost', data=df12)
plt.title('Session Google Ads campaign vs Google Ads cost')
plt.xticks(rotation=90)
plt.show()
```

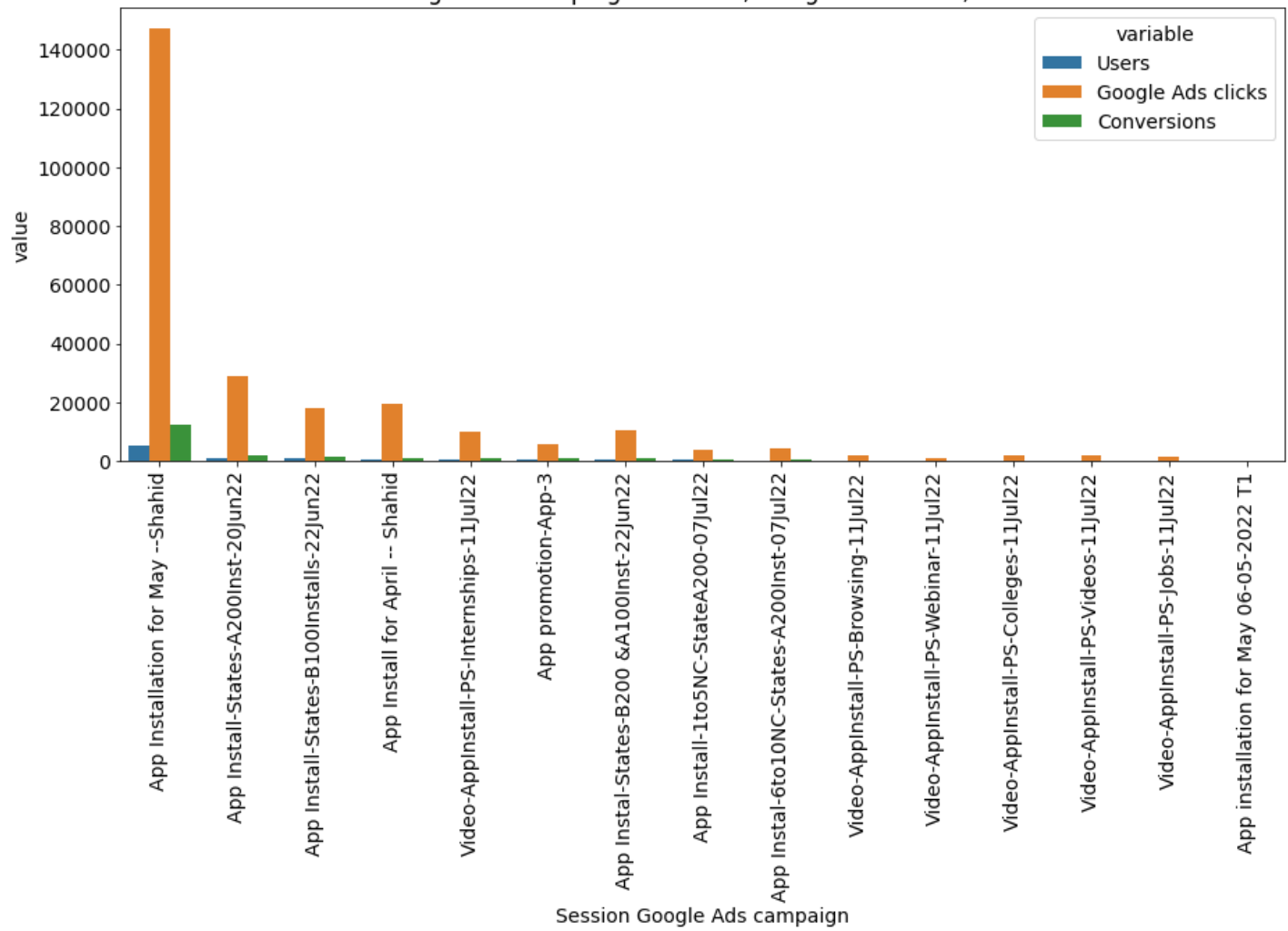
Session Google Ads campaign vs Google Ads cost



In [35]:

```
df_melted = df12.melt(id_vars='Session Google Ads campaign', value_vars=['Users', 'Google
plt.figure(figsize=(15, 6))
sns.barplot(x='Session Google Ads campaign', y='value', hue='variable', data=df_melted)
plt.title('Session Google Ads campaign vs Users, Google Ads clicks, Conversions')
plt.xticks(rotation=90)
plt.show()
```

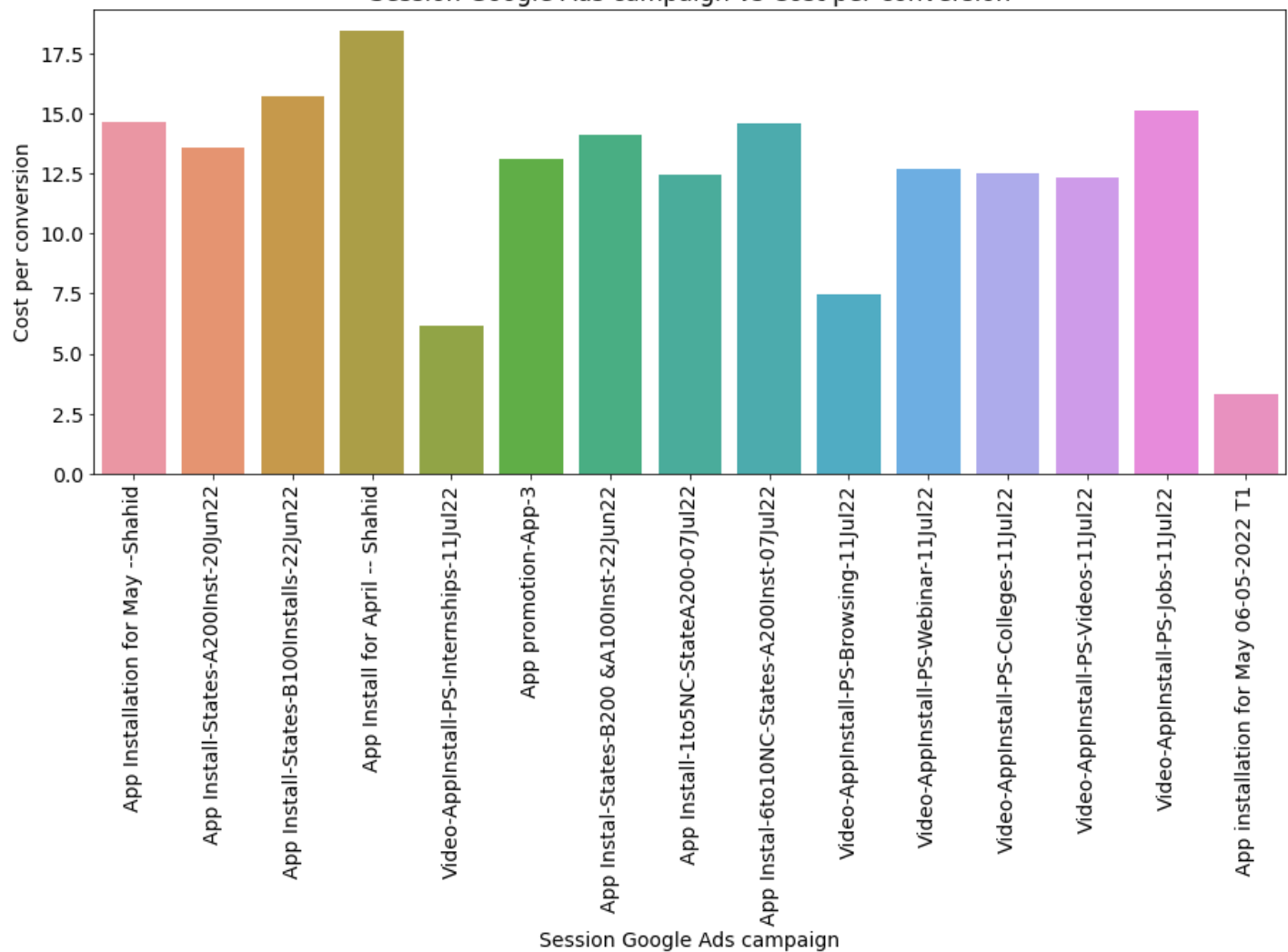
Session Google Ads campaign vs Users, Google Ads clicks, Conversions



In [36]:

```
plt.figure(figsize=(15, 6))
sns.barplot(x='Session Google Ads campaign', y='Cost per conversion', data=df12)
plt.title('Session Google Ads campaign vs Cost per conversion')
plt.xticks(rotation=90)
plt.show()
```

Session Google Ads campaign vs Cost per conversion



## ----- IMPORTANT THINGS FROM THE ANALYSIS -----

1. In User Acquisition: Organic Search is the most important factor for user conversion.
2. In Traffic Acquisition: The Unassigned Category is crucial for conversions.
3. In Event Report: Screen Views, Session Starts, First Opens, and User Engagements occur more frequently during user usage.
4. In Conversion Report: Notifications play a vital role in conversions. When a customer sees a notification, it triggers user engagement.
5. In Page Path and Screen Class: If a conversion occurs, the Login Page appears more frequently. Login Flutter, Feeds, Storyboard, etc., are major components here.
6. In Demographics Report: India has the highest conversion rate, while Kenya, Panama, Maldives, Sri Lanka, and Romania have the lowest conversions.
7. In Citywise Report: Bengaluru has the highest conversion rate, followed by Hyderabad.
8. In Gender Report: Users often do not specify their gender during login, resulting in it being considered unknown. Additionally, males have a higher conversion rate.
9. In User by Interest: Users are interested in shopping, media entertainment, technology, and mobile usage.

10. In User by Language: Users primarily use English, followed by Hindi, which has the highest conversion rate.
11. In User by Age: People aged 18-24 have the second-highest conversion rate, while those aged 45-54 have lower conversion rates. Those aged 65+ have less engagement time.
12. In Google Ads Report: The cost per conversion is higher for April app installations. Despite the amount paid for Google Ads, the conversion rate is very low.

## ----- NEEDS TO IMPROVRE -----

1. Users are required to provide their correct age and gender during login; only then can we analyze which categories people interact with our products most.
1. In the Indian region, we need to focus on tier-2 and tier-3 cities.
1. In the 25-34 age category, the conversion rate is very low. Despite users being independent in that category, conversions are low. We need to focus more on that age group.
1. Vehicle, Food & Dining, Finance & Banking, Home & Garden, and Wellness are the least interesting to users. Perhaps we should focus on improving or discontinuing them.
1. We must be cautious about investing in Google Ads because the conversion rate is very low considering the money we allocate to the ads.

In [ ]:

In [ ]:

In [ ]: