

Lead Score

Problem Statement

- X Education is an Edtech company which provides online courses for industry professionals.
- If any person shows interest in their courses they provide X Education with their personal information, it is termed as leads.
- Company collects leads from different online platforms and then its employees try to connect with these leads.
- In this process about 30% of the leads opt for any of the courses available and get converted into a paying customer.
- Identify those leads who have high conversion probability.
- Provide lead score to them such that person with high lead score have higher chances to convert into a paying customer. while person with lower score have less chances to convert.

Goals and Objectives

There are quite a few goals for this case study.

Build a logistic regression model to assign a lead score between 0 and 100 to each of the leads which can be used by the company to target potential leads. A higher score would mean that the lead is hot, i.e. is most likely to convert whereas a lower score would mean that the lead is cold and will mostly not get converted. There are some more problems presented by the company which your model should be able to adjust to if the company's requirement changes in the future so you will need to handle these as well. These problems are provided in a separate doc file. Please fill it based on the logistic regression model you got in the first step. Also, make sure you include this in your final PPT where you'll make recommendations.

#Importing required packages

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

from sklearn.model_selection import train_test_split
from sklearn.preprocessing import MinMaxScaler
import statsmodels.api as sm
from statsmodels.stats.outliers_influence import variance_inflation_factor
from sklearn.linear_model import LogisticRegression
from sklearn.feature_selection import RFE
```

```

from sklearn import metrics
from sklearn.metrics import confusion_matrix
from sklearn.metrics import precision_score
from sklearn.metrics import precision_recall_curve

```

```
# Ignoring Warnings
```

```

import warnings
warnings.filterwarnings('ignore')

```

Importing and Cleaning Data

```

df_leads = pd.read_csv('Leads.csv')
df_leads.head()

```

	Prospect ID	Lead Number	Lead
Origin \			
0 7927b2df-8bba-4d29-b9a2-b6e0beafe620		660737	
API			
1 2a272436-5132-4136-86fa-dcc88c88f482		660728	
API			
2 8cc8c611-a219-4f35-ad23-fdfd2656bd8a		660727	Landing Page
Submission			
3 0cc2df48-7cf4-4e39-9de9-19797f9b38cc		660719	Landing Page
Submission			
4 3256f628-e534-4826-9d63-4a8b88782852		660681	Landing Page
Submission			

	Lead Source	Do Not Email	Do Not Call	Converted	TotalVisits	\
0	Olark Chat	No	No	0	0.0	
1	Organic Search	No	No	0	5.0	
2	Direct Traffic	No	No	1	2.0	
3	Direct Traffic	No	No	0	1.0	
4	Google	No	No	1	2.0	

	Total Time Spent on Website	Page Views	Per Visit	...	\
0	0		0.0	...	
1	674		2.5	...	
2	1532		2.0	...	
3	305		1.0	...	
4	1428		1.0	...	

	Get updates on DM Content	Lead Profile	City	\
0	No	Select	Select	
1	No	Select	Select	
2	No	Potential Lead	Mumbai	
3	No	Select	Mumbai	
4	No	Select	Mumbai	

	Asymmetrique Activity Index	Asymmetrique Profile Index	\
0	02.Medium	02.Medium	

1	02.Medium	02.Medium
2	02.Medium	01.High
3	02.Medium	01.High
4	02.Medium	01.High

	Asymmetrique Activity Score	Asymmetrique Profile Score \
0	15.0	15.0
1	15.0	15.0
2	14.0	20.0
3	13.0	17.0
4	15.0	18.0

	I agree to pay the amount through cheque \
0	No
1	No
2	No
3	No
4	No

	A free copy of Mastering The Interview	Last Notable Activity
0	No	Modified
1	No	Email Opened
2	Yes	Email Opened
3	No	Modified
4	No	Modified

[5 rows x 37 columns]

df_leads.shape

(9240, 37)

df_leads.info()

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 9240 entries, 0 to 9239

Data columns (total 37 columns):

#	Column	Non-Null Count
0	Prospect ID	9240 non-null
1	Lead Number	9240 non-null
2	Lead Origin	9240 non-null
3	Lead Source	9204 non-null
4	Do Not Email	9240 non-null

5	Do Not Call	9240 non-null
object		
6	Converted	9240 non-null
int64		
7	TotalVisits	9103 non-null
float64		
8	Total Time Spent on Website	9240 non-null
int64		
9	Page Views Per Visit	9103 non-null
float64		
10	Last Activity	9137 non-null
object		
11	Country	6779 non-null
object		
12	Specialization	7802 non-null
object		
13	How did you hear about X Education	7033 non-null
object		
14	What is your current occupation	6550 non-null
object		
15	What matters most to you in choosing a course	6531 non-null
object		
16	Search	9240 non-null
object		
17	Magazine	9240 non-null
object		
18	Newspaper Article	9240 non-null
object		
19	X Education Forums	9240 non-null
object		
20	Newspaper	9240 non-null
object		
21	Digital Advertisement	9240 non-null
object		
22	Through Recommendations	9240 non-null
object		
23	Receive More Updates About Our Courses	9240 non-null
object		
24	Tags	5887 non-null
object		
25	Lead Quality	4473 non-null
object		
26	Update me on Supply Chain Content	9240 non-null
object		
27	Get updates on DM Content	9240 non-null
object		
28	Lead Profile	6531 non-null
object		
29	City	7820 non-null
object		

```

30 Asymmetrique Activity Index          5022 non-null
object
31 Asymmetrique Profile Index          5022 non-null
object
32 Asymmetrique Activity Score          5022 non-null
float64
33 Asymmetrique Profile Score          5022 non-null
float64
34 I agree to pay the amount through cheque 9240 non-null
object
35 A free copy of Mastering The Interview 9240 non-null
object
36 Last Notable Activity                9240 non-null
object
dtypes: float64(4), int64(3), object(30)
memory usage: 2.6+ MB

```

```
df_leads.describe()
```

	Lead Number	Converted	TotalVisits	Total Time Spent on Website \
count	9240.000000	9240.000000	9103.000000	9240.000000
mean	617188.435606	0.385390	3.445238	487.698268
std	23405.995698	0.486714	4.854853	548.021466
min	579533.000000	0.000000	0.000000	0.000000
25%	596484.500000	0.000000	1.000000	12.000000
50%	615479.000000	0.000000	3.000000	248.000000
75%	637387.250000	1.000000	5.000000	936.000000
max	660737.000000	1.000000	251.000000	2272.000000

	Page Views Per Visit	Asymmetrique Activity Score \
count	9103.000000	5022.000000
mean	2.362820	14.306252
std	2.161418	1.386694
min	0.000000	7.000000
25%	1.000000	14.000000
50%	2.000000	14.000000
75%	3.000000	15.000000
max	55.000000	18.000000

	Asymmetrique Profile Score
count	5022.000000

mean	16.344883
std	1.811395
min	11.000000
25%	15.000000
50%	16.000000
75%	18.000000
max	20.000000

Data cleaning

Identifying features with missing terms

```
round(100*(df_leads.isnull().sum()/len(df_leads.index)), 2)
```

Prospect ID	0.00
Lead Number	0.00
Lead Origin	0.00
Lead Source	0.39
Do Not Email	0.00
Do Not Call	0.00
Converted	0.00
TotalVisits	1.48
Total Time Spent on Website	0.00
Page Views Per Visit	1.48
Last Activity	1.11
Country	26.63
Specialization	15.56
How did you hear about X Education	23.89
What is your current occupation	29.11
What matters most to you in choosing a course	29.32
Search	0.00
Magazine	0.00
Newspaper Article	0.00
X Education Forums	0.00
Newspaper	0.00
Digital Advertisement	0.00
Through Recommendations	0.00
Receive More Updates About Our Courses	0.00
Tags	36.29
Lead Quality	51.59
Update me on Supply Chain Content	0.00
Get updates on DM Content	0.00
Lead Profile	29.32
City	15.37
Asymmetrique Activity Index	45.65
Asymmetrique Profile Index	45.65
Asymmetrique Activity Score	45.65
Asymmetrique Profile Score	45.65
I agree to pay the amount through cheque	0.00
A free copy of Mastering The Interview	0.00

Last Notable Activity 0.00
dtype: float64

```
df_leads.select_dtypes(include = 'object').info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 9240 entries, 0 to 9239
```

```
Data columns (total 30 columns):
```

#	Column	Non-Null Count
Dtype		
---	-----	-----

0	Prospect ID	9240 non-null
object		
1	Lead Origin	9240 non-null
object		
2	Lead Source	9204 non-null
object		
3	Do Not Email	9240 non-null
object		
4	Do Not Call	9240 non-null
object		
5	Last Activity	9137 non-null
object		
6	Country	6779 non-null
object		
7	Specialization	7802 non-null
object		
8	How did you hear about X Education	7033 non-null
object		
9	What is your current occupation	6550 non-null
object		
10	What matters most to you in choosing a course	6531 non-null
object		
11	Search	9240 non-null
object		
12	Magazine	9240 non-null
object		
13	Newspaper Article	9240 non-null
object		
14	X Education Forums	9240 non-null
object		
15	Newspaper	9240 non-null
object		
16	Digital Advertisement	9240 non-null
object		
17	Through Recommendations	9240 non-null
object		
18	Receive More Updates About Our Courses	9240 non-null
object		
19	Tags	5887 non-null

```

object
  20 Lead Quality 4473 non-null
object
  21 Update me on Supply Chain Content 9240 non-null
object
  22 Get updates on DM Content 9240 non-null
object
  23 Lead Profile 6531 non-null
object
  24 City 7820 non-null
object
  25 Asymetrique Activity Index 5022 non-null
object
  26 Asymetrique Profile Index 5022 non-null
object
  27 I agree to pay the amount through cheque 9240 non-null
object
  28 A free copy of Mastering The Interview 9240 non-null
object
  29 Last Notable Activity 9240 non-null
object
dtypes: object(30)
memory usage: 2.1+ MB

```

Replacing Select with nul values

```
df_leads = df_leads.replace('Select', np.nan)
```

```
df_leads.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 9240 entries, 0 to 9239
```

```
Data columns (total 37 columns):
```

#	Column	Non-Null Count
0	Prospect ID	9240 non-null
1	Lead Number	9240 non-null
2	Lead Origin	9240 non-null
3	Lead Source	9204 non-null
4	Do Not Email	9240 non-null
5	Do Not Call	9240 non-null
6	Converted	9240 non-null
7	TotalVisits	9103 non-null

float64		
8	Total Time Spent on Website	9240 non-null
int64		
9	Page Views Per Visit	9103 non-null
float64		
10	Last Activity	9137 non-null
object		
11	Country	6779 non-null
object		
12	Specialization	5860 non-null
object		
13	How did you hear about X Education	1990 non-null
object		
14	What is your current occupation	6550 non-null
object		
15	What matters most to you in choosing a course	6531 non-null
object		
16	Search	9240 non-null
object		
17	Magazine	9240 non-null
object		
18	Newspaper Article	9240 non-null
object		
19	X Education Forums	9240 non-null
object		
20	Newspaper	9240 non-null
object		
21	Digital Advertisement	9240 non-null
object		
22	Through Recommendations	9240 non-null
object		
23	Receive More Updates About Our Courses	9240 non-null
object		
24	Tags	5887 non-null
object		
25	Lead Quality	4473 non-null
object		
26	Update me on Supply Chain Content	9240 non-null
object		
27	Get updates on DM Content	9240 non-null
object		
28	Lead Profile	2385 non-null
object		
29	City	5571 non-null
object		
30	Asymmetrique Activity Index	5022 non-null
object		
31	Asymmetrique Profile Index	5022 non-null
object		
32	Asymmetrique Activity Score	5022 non-null

```

float64
 33 Asymmetrique Profile Score                    5022 non-null
float64
 34 I agree to pay the amount through cheque      9240 non-null
object
 35 A free copy of Mastering The Interview        9240 non-null
object
 36 Last Notable Activity                        9240 non-null
dtypes: float64(4), int64(3), object(30)
memory usage: 2.6+ MB

# From the data converting categorical features into binary values
for features in ['Do Not Email', 'Do Not
Call', 'Search', 'Magazine', 'Newspaper Article', 'X Education
Forums', 'Newspaper',
                  'Digital Advertisement', 'Through
Recommendations', 'Receive More Updates About Our Courses', 'Update me
on Supply Chain Content',
                  'Get updates on DM Content', 'I agree to pay the
amount through cheque', 'A free copy of Mastering The Interview']:
    df_leads[features] = df_leads[features].apply(lambda x : 1 if x ==
'Yes' else 0)

```

```
df_leads.head()
```

	Prospect ID	Lead Number	Lead
Origin \			
0 7927b2df-8bba-4d29-b9a2-b6e0beafe620		660737	
API			
1 2a272436-5132-4136-86fa-dcc88c88f482		660728	
API			
2 8cc8c611-a219-4f35-ad23-fdfd2656bd8a		660727	Landing Page
Submission			
3 0cc2df48-7cf4-4e39-9de9-19797f9b38cc		660719	Landing Page
Submission			
4 3256f628-e534-4826-9d63-4a8b88782852		660681	Landing Page
Submission			

	Lead Source	Do Not Email	Do Not Call	Converted	
TotalVisits \					
0	0lark Chat	0	0	0	0.0
1	Organic Search	0	0	0	5.0
2	Direct Traffic	0	0	1	2.0
3	Direct Traffic	0	0	0	1.0

4	Google	0	0	1	2.0
---	--------	---	---	---	-----

	Total Time Spent on Website	Page Views	Per Visit	...	\
0	0		0.0	...	
1	674		2.5	...	
2	1532		2.0	...	
3	305		1.0	...	
4	1428		1.0	...	

	Get updates on DM Content	Lead Profile	City	\
0	0	NaN	NaN	
1	0	NaN	NaN	
2	0	Potential Lead	Mumbai	
3	0	NaN	Mumbai	
4	0	NaN	Mumbai	

	Asymmetrique Activity Index	Asymmetrique Profile Index	\
0	02.Medium	02.Medium	
1	02.Medium	02.Medium	
2	02.Medium	01.High	
3	02.Medium	01.High	
4	02.Medium	01.High	

	Asymmetrique Activity Score	Asymmetrique Profile Score	\
0	15.0	15.0	
1	15.0	15.0	
2	14.0	20.0	
3	13.0	17.0	
4	15.0	18.0	

	I agree to pay the amount through cheque	\
0	0	
1	0	
2	0	
3	0	
4	0	

	A free copy of Mastering The Interview	Last Notable Activity
0	0	Modified
1	0	Email Opened
2	1	Email Opened
3	0	Modified
4	0	Modified

[5 rows x 37 columns]

Checking labels of other categorical features

for column **in** df_leads:

print(df_leads[column].astype('category').value_counts())

```
print('-----')
-----')
```

```
000104b9-23e4-4ddc-8caa-8629fe8ad7f4    1
a7a319ea-b6ae-4c6b-afc5-183b933d10b5    1
aa27a0af-eeab-4007-a770-fa8a93fa53c8    1
aa30ebb2-8476-41ce-9258-37cc025110d3    1
aa405742-17ac-4c65-b19e-ab91c241cc53    1
..
539eb309-df36-4a89-ac58-6d3651393910    1
539ffa32-1be7-4fe1-b04c-faf1bab763cf    1
53aabdb84-5dcc-4299-bbe3-62f3764b07b1    1
53ac14bd-2bb2-4315-a21c-94562d1b6b2d    1
ffffb0e5e-9f92-4017-9f42-781a69da4154    1
Name: Prospect ID, Length: 9240, dtype: int64
-----
```

```
-----
579533    1
629593    1
630390    1
630403    1
630405    1
..
602534    1
602540    1
602557    1
602561    1
660737    1
Name: Lead Number, Length: 9240, dtype: int64
-----
```

```
-----
Landing Page Submission    4886
API                        3580
Lead Add Form              718
Lead Import                55
Quick Add Form             1
Name: Lead Origin, dtype: int64
-----
```

```
-----
Google                2868
Direct Traffic        2543
Olark Chat            1755
Organic Search        1154
Reference              534
Welingak Website      142
Referral Sites        125
Facebook              55
bing                   6
google                 5
```

Click2call	4
Press_Release	2
Social_Media	2
Live Chat	2
WeLearn	1
Pay per Click Ads	1
NC_EDM	1
blog	1
testone	1
welearnblog_Home	1
youtubechannel	1

Name: Lead Source, dtype: int64

0	8506
1	734

Name: Do Not Email, dtype: int64

0	9238
1	2

Name: Do Not Call, dtype: int64

0	5679
1	3561

Name: Converted, dtype: int64

0.0	2189
2.0	1680
3.0	1306
4.0	1120
5.0	783
6.0	466
1.0	395
7.0	309
8.0	224
9.0	164
10.0	114
11.0	86
13.0	48
12.0	45
14.0	36
16.0	21
15.0	18
17.0	16
18.0	15
20.0	12
19.0	9

23.0	6
21.0	6
24.0	5
25.0	5
27.0	5
22.0	3
26.0	2
28.0	2
29.0	2
54.0	1
141.0	1
115.0	1
74.0	1
55.0	1
30.0	1
43.0	1
42.0	1
41.0	1
32.0	1
251.0	1

Name: TotalVisits, dtype: int64

0	2193
60	19
75	18
74	18
127	18

...	
1091	1
1088	1
1085	1
1084	1
2272	1

Name: Total Time Spent on Website, Length: 1731, dtype: int64

0.0	2189
2.0	1795
3.0	1196
4.0	896
1.0	651

...	
3.57	1
3.8	1
3.82	1
3.83	1
55.0	1

Name: Page Views Per Visit, Length: 114, dtype: int64

```

-----
Email Opened                3437
SMS Sent                    2745
Olark Chat Conversation     973
Page Visited on Website    640
Converted to Lead          428
Email Bounced              326
Email Link Clicked         267
Form Submitted on Website  116
Unreachable                 93
Unsubscribed                61
Had a Phone Conversation    30
Approached upfront          9
View in browser link Clicked 6
Email Received              2
Email Marked Spam           2
Resubscribed to emails      1
Visited Booth in Tradeshow  1
Name: Last Activity, dtype: int64
-----

```

```

-----
India                        6492
United States                69
United Arab Emirates        53
Singapore                   24
Saudi Arabia                 21
United Kingdom              15
Australia                   13
Qatar                       10
Bahrain                      7
Hong Kong                   7
France                       6
Oman                        6
unknown                     5
Kuwait                      4
Nigeria                     4
South Africa                 4
Germany                     4
Canada                      4
Sweden                      3
Uganda                      2
Philippines                  2
Asia/Pacific Region         2
Italy                       2
Ghana                       2
China                       2
Belgium                     2
Bangladesh                   2
Netherlands                  2
Malaysia                     1

```

Liberia	1
Russia	1
Kenya	1
Indonesia	1
Sri Lanka	1
Switzerland	1
Tanzania	1
Denmark	1
Vietnam	1

Name: Country, dtype: int64

Finance Management	976
Human Resource Management	848
Marketing Management	838
Operations Management	503
Business Administration	403
IT Projects Management	366
Supply Chain Management	349
Banking, Investment And Insurance	338
Media and Advertising	203
Travel and Tourism	203
International Business	178
Healthcare Management	159
Hospitality Management	114
E-COMMERCE	112
Retail Management	100
Rural and Agribusiness	73
E-Business	57
Services Excellence	40

Name: Specialization, dtype: int64

Online Search	808
Word Of Mouth	348
Student of SomeSchool	310
Other	186
Multiple Sources	152
Advertisements	70
Social Media	67
Email	26
SMS	23

Name: How did you hear about X Education, dtype: int64

Unemployed	5600
Working Professional	706
Student	210
Other	16
Housewife	10


```

Businessman          8
Name: What is your current occupation, dtype: int64
-----
-----
Better Career Prospects      6528
Flexibility & Convenience    2
Other                        1
Name: What matters most to you in choosing a course, dtype: int64
-----
-----
0      9226
1      14
Name: Search, dtype: int64
-----
-----
0      9240
Name: Magazine, dtype: int64
-----
-----
0      9238
1      2
Name: Newspaper Article, dtype: int64
-----
-----
0      9239
1      1
Name: X Education Forums, dtype: int64
-----
-----
0      9239
1      1
Name: Newspaper, dtype: int64
-----
-----
0      9236
1      4
Name: Digital Advertisement, dtype: int64
-----
-----
0      9233
1      7
Name: Through Recommendations, dtype: int64
-----
-----
0      9240
Name: Receive More Updates About Our Courses, dtype: int64
-----
-----
Will revert after reading the email      2072
Ringing                                  1203

```

Interested in other courses	513
Already a student	465
Closed by Horizzon	358
switched off	240
Busy	186
Lost to EINS	175
Not doing further education	145
Interested in full time MBA	117
Graduation in progress	111
invalid number	83
Diploma holder (Not Eligible)	63
wrong number given	47
opp hangup	33
number not provided	27
in touch with EINS	12
Lost to Others	7
Still Thinking	6
Want to take admission but has financial problems	6
Interested in Next batch	5
In confusion whether part time or DLP	5
Lateral student	3
Shall take in the next coming month	2
University not recognized	2
Recognition issue (DEC approval)	1

Name: Tags, dtype: int64

Might be	1560
Not Sure	1092
High in Relevance	637
Worst	601
Low in Relevance	583

Name: Lead Quality, dtype: int64

0 9240

Name: Update me on Supply Chain Content, dtype: int64

0 9240

Name: Get updates on DM Content, dtype: int64

Potential Lead	1613
Other Leads	487
Student of SomeSchool	241
Lateral Student	24
Dual Specialization Student	20

Name: Lead Profile, dtype: int64

Mumbai 3222
Thane & Outskirts 752
Other Cities 686
Other Cities of Maharashtra 457
Other Metro Cities 380
Tier II Cities 74
Name: City, dtype: int64

02.Medium 3839
01.High 821
03.Low 362
Name: Asymmetrique Activity Index, dtype: int64

02.Medium 2788
01.High 2203
03.Low 31
Name: Asymmetrique Profile Index, dtype: int64

14.0 1771
15.0 1293
13.0 775
16.0 467
17.0 349
12.0 196
11.0 95
10.0 57
9.0 9
18.0 5
8.0 4
7.0 1
Name: Asymmetrique Activity Score, dtype: int64

15.0 1759
18.0 1071
16.0 599
17.0 579
20.0 308
19.0 245
14.0 226
13.0 204
12.0 22
11.0 9
Name: Asymmetrique Profile Score, dtype: int64


```

0      9240
Name: I agree to pay the amount through cheque, dtype: int64
-----
0      6352
1      2888
Name: A free copy of Mastering The Interview, dtype: int64
-----
Modified                3407
Email Opened            2827
SMS Sent                2172
Page Visited on Website  318
Olark Chat Conversation  183
Email Link Clicked       173
Email Bounced           60
Unsubscribed             47
Unreachable              32
Had a Phone Conversation  14
Email Marked Spam         2
Approached upfront        1
Email Received            1
Form Submitted on Website  1
Resubscribed to emails    1
View in browser link Clicked 1
Name: Last Notable Activity, dtype: int64
-----
-----

```

Handling data and the missing values

```

# Converting all the values to lower case as google is mentioned in
bot upper and lower case
df_leads = df_leads.applymap(lambda s:s.lower() if type(s) == str else
s)

```

```

# Checking if there are columns with one unique value since it won't
affect our analysis
df_leads.nunique()

```

```

Prospect ID                9240
Lead Number                9240
Lead Origin                 5
Lead Source                 20
Do Not Email                2
Do Not Call                 2
Converted                  2
TotalVisits                 41
Total Time Spent on Website 1731
Page Views Per Visit        114

```

Last Activity	17
Country	38
Specialization	18
How did you hear about X Education	9
What is your current occupation	6
What matters most to you in choosing a course	3
Search	2
Magazine	1
Newspaper Article	2
X Education Forums	2
Newspaper	2
Digital Advertisement	2
Through Recommendations	2
Receive More Updates About Our Courses	1
Tags	26
Lead Quality	5
Update me on Supply Chain Content	1
Get updates on DM Content	1
Lead Profile	5
City	6
Asymmetrique Activity Index	3
Asymmetrique Profile Index	3
Asymmetrique Activity Score	12
Asymmetrique Profile Score	10
I agree to pay the amount through cheque	1
A free copy of Mastering The Interview	2
Last Notable Activity	16
dtype: int64	

Dropping unique valued columns

```
df_leads= df_leads.drop(['Magazine','Receive More Updates About Our
Courses','Update me on Supply Chain Content',
                        'Get updates on DM Content','I agree to pay
the amount through cheque'], axis =1)
```

Checking the percentage of missing values

```
round(100*(df_leads.isnull().sum()/len(df_leads.index)), 2)
```

Prospect ID	0.00
Lead Number	0.00
Lead Origin	0.00
Lead Source	0.39
Do Not Email	0.00
Do Not Call	0.00
Converted	0.00
TotalVisits	1.48
Total Time Spent on Website	0.00
Page Views Per Visit	1.48
Last Activity	1.11
Country	26.63
Specialization	36.58

How did you hear about X Education	78.46
What is your current occupation	29.11
What matters most to you in choosing a course	29.32
Search	0.00
Newspaper Article	0.00
X Education Forums	0.00
Newspaper	0.00
Digital Advertisement	0.00
Through Recommendations	0.00
Tags	36.29
Lead Quality	51.59
Lead Profile	74.19
City	39.71
Asymmetrique Activity Index	45.65
Asymmetrique Profile Index	45.65
Asymmetrique Activity Score	45.65
Asymmetrique Profile Score	45.65
A free copy of Mastering The Interview	0.00
Last Notable Activity	0.00
dtype: float64	

Removing all the columns that are no required and have 35% null values

```
df_2 = df_leads.drop(['Asymmetrique Profile Index','Asymmetrique
Activity Index','Asymmetrique Activity Score',
                    'Asymmetrique Profile Score','Lead
Profile','Tags','Lead Quality','How did you hear about X Education',
                    'City','Lead Number'],axis=1)
df_2.head()
```

	Prospect ID	Lead Origin \
0	7927b2df-8bba-4d29-b9a2-b6e0beafe620	api
1	2a272436-5132-4136-86fa-dcc88c88f482	api
2	8cc8c611-a219-4f35-ad23-fdfd2656bd8a	landing page submission
3	0cc2df48-7cf4-4e39-9de9-19797f9b38cc	landing page submission
4	3256f628-e534-4826-9d63-4a8b88782852	landing page submission

	Lead Source	Do Not Email	Do Not Call	Converted	TotalVisits \
0	olark chat	0	0	0	0.0
1	organic search	0	0	0	5.0
2	direct traffic	0	0	1	2.0
3	direct traffic	0	0	0	1.0
4	google	0	0	1	2.0

	Total Time Spent on Website	Page Views Per Visit	Last
Activity \			
0	0	0.0	page visited on
website			
1	674	2.5	email
opened			
2	1532	2.0	email
opened			
3	305	1.0	
unreachable			
4	1428	1.0	converted
to lead			

	... What is your current occupation \
0	...
1	unemployed
2	unemployed
3	student
4	unemployed

	What matters most to you in choosing a course Search Newspaper
Article \	
0	better career prospects
0	
1	better career prospects
0	
2	better career prospects
0	
3	better career prospects
0	
4	better career prospects
0	

	X Education Forums	Newspaper	Digital Advertisement \
0	0	0	0
1	0	0	0
2	0	0	0
3	0	0	0
4	0	0	0

	Through Recommendations	A free copy of Mastering The Interview \
0	0	0
1	0	0
2	0	1
3	0	0
4	0	0

	Last Notable Activity
0	modified
1	email opened

```
2          email opened
3          modified
4          modified
```

```
[5 rows x 22 columns]
```

```
# Rechecking the percentage of missing values
```

```
round(100*(df_2.isnull().sum()/len(df_2.index)), 2)
```

```
Prospect ID          0.00
Lead Origin          0.00
Lead Source          0.39
Do Not Email         0.00
Do Not Call          0.00
Converted            0.00
TotalVisits          1.48
Total Time Spent on Website 0.00
Page Views Per Visit 1.48
Last Activity        1.11
Country              26.63
Specialization       36.58
What is your current occupation 29.11
What matters most to you in choosing a course 29.32
Search              0.00
Newspaper Article   0.00
X Education Forums  0.00
Newspaper           0.00
Digital Advertisement 0.00
Through Recommendations 0.00
A free copy of Mastering The Interview 0.00
Last Notable Activity 0.00
dtype: float64
```

```
# Replacing the remaing null values with not provided as removing
these values will result in huge data loss
```

```
df_2['Specialization'] = df_2['Specialization'].fillna('not provided')
```

```
df_2['What matters most to you in choosing a course'] = df_2['What
matters most to you in choosing a course'].fillna('not provided')
```

```
df_2['Country'] = df_2['Country'].fillna('not provided')
```

```
df_2['What is your current occupation'] = df_2['What is your current
occupation'].fillna('not provided')
```

```
df_2.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 9240 entries, 0 to 9239
```

```
Data columns (total 22 columns):
```

```
#    Column                                     Non-Null Count
Dtype
```

```
---
```

```
-----
```



```

-----
 0 Prospect ID 9240 non-null
object
 1 Lead Origin 9240 non-null
object
 2 Lead Source 9204 non-null
object
 3 Do Not Email 9240 non-null
int64
 4 Do Not Call 9240 non-null
int64
 5 Converted 9240 non-null
int64
 6 TotalVisits 9103 non-null
float64
 7 Total Time Spent on Website 9240 non-null
int64
 8 Page Views Per Visit 9103 non-null
float64
 9 Last Activity 9137 non-null
object
10 Country 9240 non-null
object
11 Specialization 9240 non-null
object
12 What is your current occupation 9240 non-null
object
13 What matters most to you in choosing a course 9240 non-null
object
14 Search 9240 non-null
int64
15 Newspaper Article 9240 non-null
int64
16 X Education Forums 9240 non-null
int64
17 Newspaper 9240 non-null
int64
18 Digital Advertisement 9240 non-null
int64
19 Through Recommendations 9240 non-null
int64
20 A free copy of Mastering The Interview 9240 non-null
int64
21 Last Notable Activity 9240 non-null
object
dtypes: float64(2), int64(11), object(9)
memory usage: 1.6+ MB

df_2["Country"].value_counts()

```

india	6492
not provided	2461
united states	69
united arab emirates	53
singapore	24
saudi arabia	21
united kingdom	15
australia	13
qatar	10
bahrain	7
hong kong	7
oman	6
france	6
unknown	5
kuwait	4
south africa	4
canada	4
nigeria	4
germany	4
sweden	3
philippines	2
uganda	2
italy	2
bangladesh	2
netherlands	2
asia/pacific region	2
china	2
belgium	2
ghana	2
kenya	1
sri lanka	1
tanzania	1
malaysia	1
liberia	1
switzerland	1
denmark	1
russia	1
vietnam	1
indonesia	1

Name: Country, dtype: int64

Function to replace feature country into only 3 values

```
def slots(x):
    category = ""
    if x == "india":
        category = "india"
    elif x == "not provided":
        category = "not provided"
    else:
        category = "outside india"
```

return category

```
df_2['Country'] = df_2.apply(lambda x:slots(x['Country']), axis = 1)
df_2['Country'].value_counts()
```

```
india          6492
not provided    2461
outside india    287
Name: Country, dtype: int64
```

```
# Rechecking the percentage of missing values
round(100*(df_2.isnull().sum()/len(df_2.index)), 2)
```

```
Prospect ID          0.00
Lead Origin          0.00
Lead Source          0.39
Do Not Email         0.00
Do Not Call          0.00
Converted            0.00
TotalVisits          1.48
Total Time Spent on Website 0.00
Page Views Per Visit 1.48
Last Activity        1.11
Country              0.00
Specialization        0.00
What is your current occupation 0.00
What matters most to you in choosing a course 0.00
Search               0.00
Newspaper Article    0.00
X Education Forums   0.00
Newspaper            0.00
Digital Advertisement 0.00
Through Recommendations 0.00
A free copy of Mastering The Interview 0.00
Last Notable Activity 0.00
dtype: float64
```

```
df3 = df_2[df_2.isnull().sum(axis=1) <1]
```

```
# Rechecking the percentage of missing values
round(100*(df3.isnull().sum()/len(df3.index)), 2)
```

```
Prospect ID          0.0
Lead Origin          0.0
Lead Source          0.0
Do Not Email         0.0
Do Not Call          0.0
Converted            0.0
TotalVisits          0.0
Total Time Spent on Website 0.0
Page Views Per Visit 0.0
```

```

Last Activity                                0.0
Country                                    0.0
Specialization                              0.0
What is your current occupation              0.0
What matters most to you in choosing a course 0.0
Search                                      0.0
Newspaper Article                           0.0
X Education Forums                          0.0
Newspaper                                   0.0
Digital Advertisement                       0.0
Through Recommendations                     0.0
A free copy of Mastering The Interview      0.0
Last Notable Activity                       0.0
dtype: float64

```

Removing Id values since they are unique for everyone

```

df_final = df3.drop('Prospect ID',1)
df_final.shape

```

```

(9074, 21)

```

Univariate Analysis

```

plt.figure(figsize = (20,40))

```

```

plt.subplot(6,2,1)
sns.countplot(df_final['Lead Origin'])
plt.title('Lead Origin')

```

```

plt.subplot(6,2,2)
sns.countplot(df_final['Do Not Email'])
plt.title('Do Not Email')

```

```

plt.subplot(6,2,3)
sns.countplot(df_final['Do Not Call'])
plt.title('Do Not Call')

```

```

plt.subplot(6,2,4)
sns.countplot(df_final['Country'])
plt.title('Country')

```

```

plt.subplot(6,2,5)
sns.countplot(df_final['Search'])
plt.title('Search')

```

```

plt.subplot(6,2,6)
sns.countplot(df_final['Newspaper Article'])
plt.title('Newspaper Article')

```

```
plt.subplot(6,2,7)
sns.countplot(df_final['X Education Forums'])
plt.title('X Education Forums')

plt.subplot(6,2,8)
sns.countplot(df_final['Newspaper'])
plt.title('Newspaper')

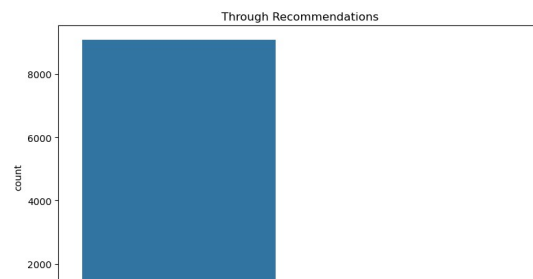
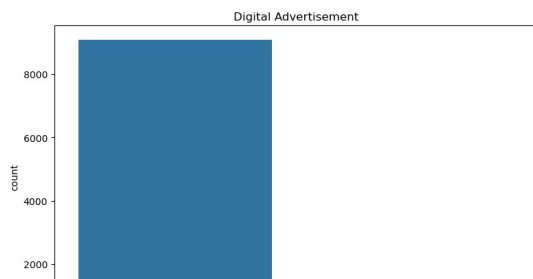
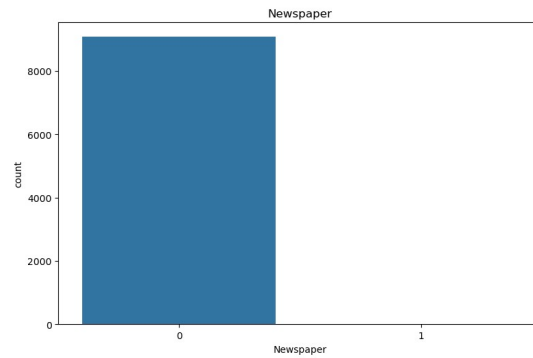
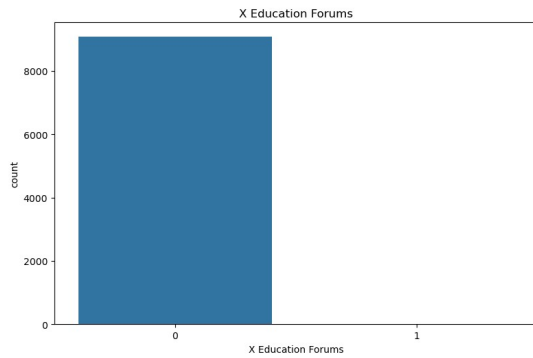
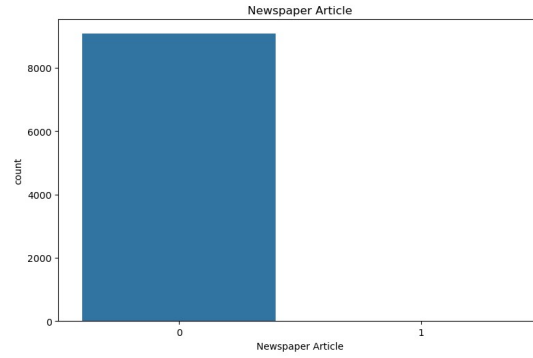
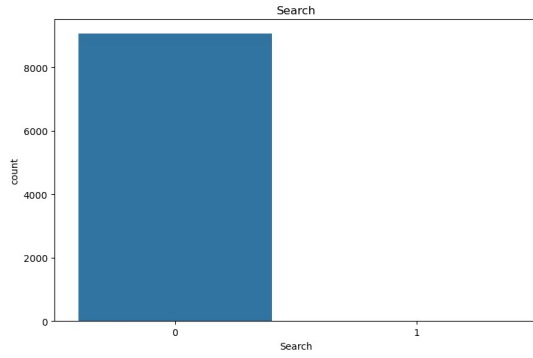
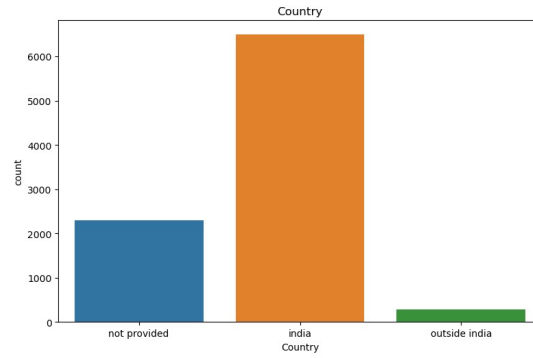
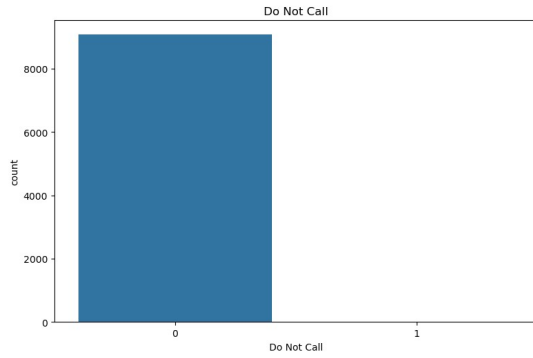
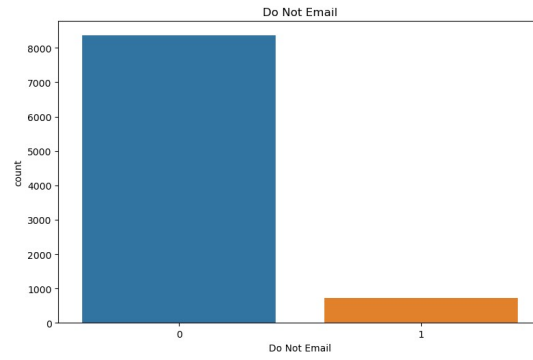
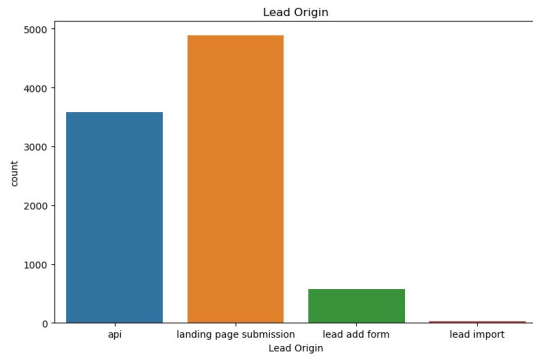
plt.subplot(6,2,9)
sns.countplot(df_final['Digital Advertisement'])
plt.title('Digital Advertisement')

plt.subplot(6,2,10)
sns.countplot(df_final['Through Recommendations'])
plt.title('Through Recommendations')

plt.subplot(6,2,11)
sns.countplot(df_final['A free copy of Mastering The Interview'])
plt.title('A free copy of Mastering The Interview')

plt.subplot(6,2,12)
sns.countplot(df_final['Last Notable Activity']).tick_params(axis='x',
rotation = 90)
plt.title('Last Notable Activity')

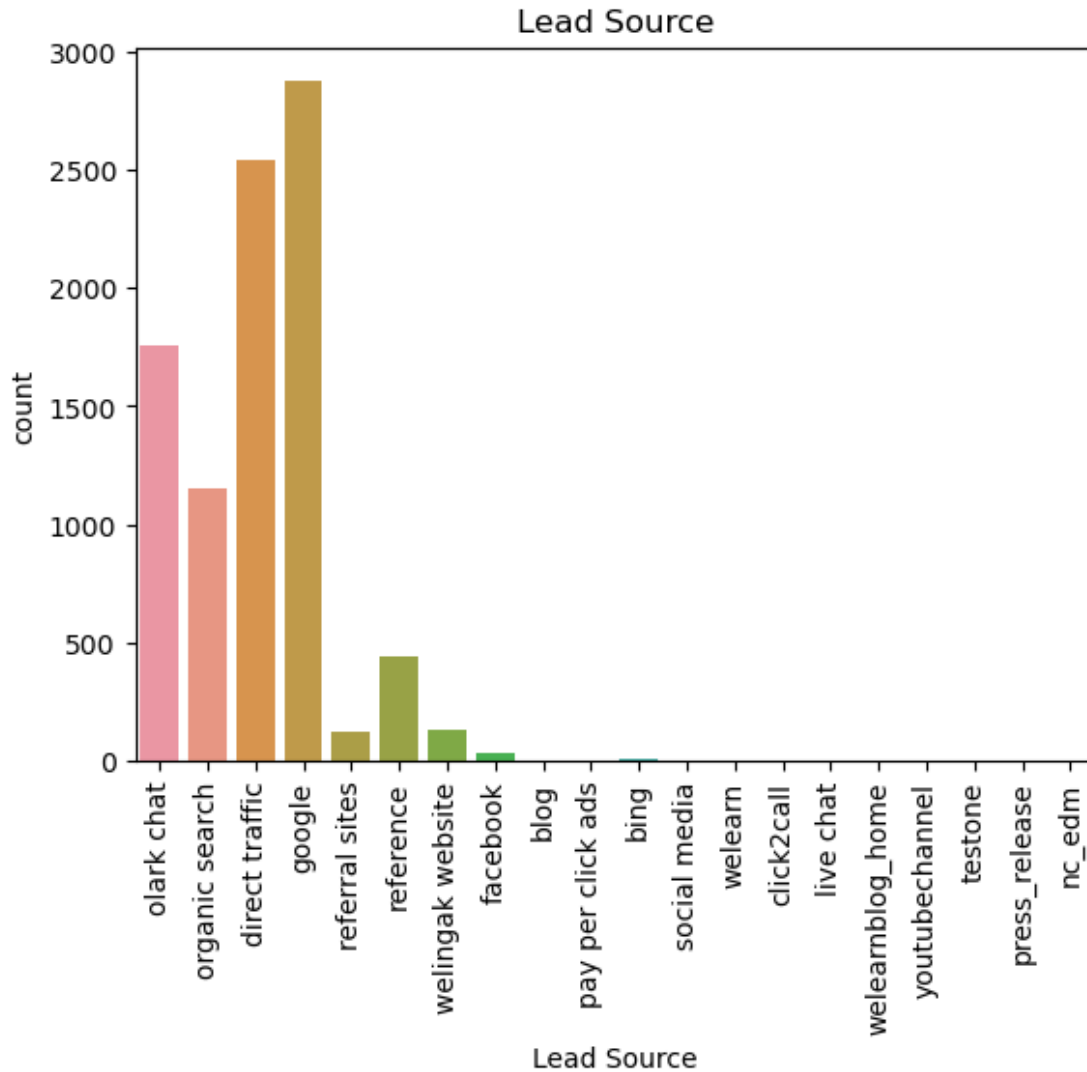
Text(0.5, 1.0, 'Last Notable Activity')
```



```

sns.countplot(df_final['Lead Source']).tick_params(axis='x', rotation
= 90)
plt.title('Lead Source')
plt.show()

```

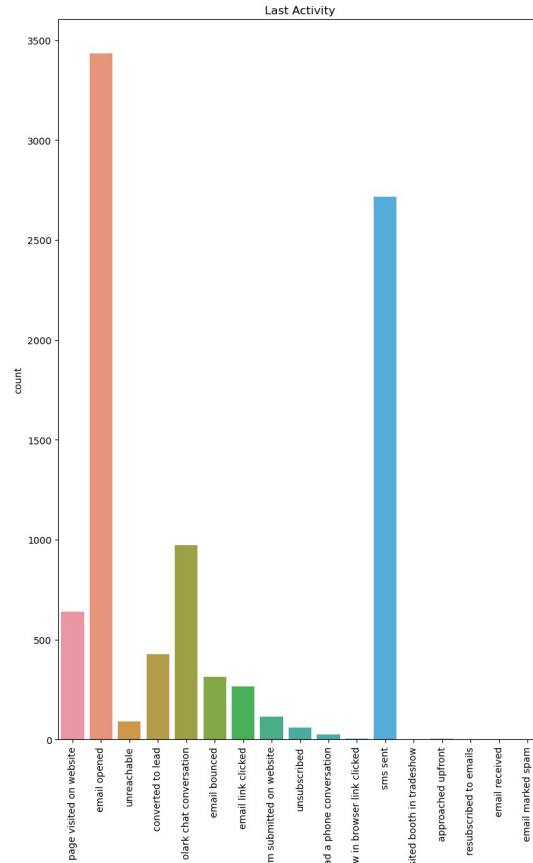
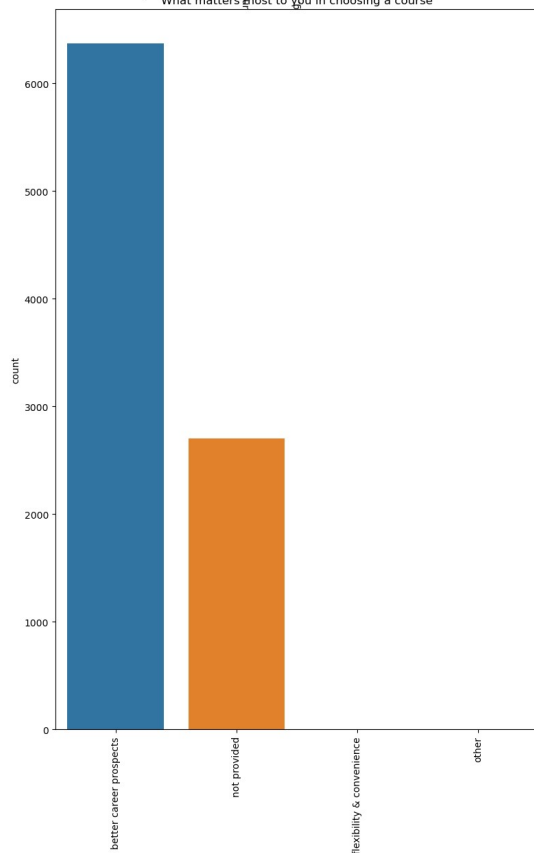
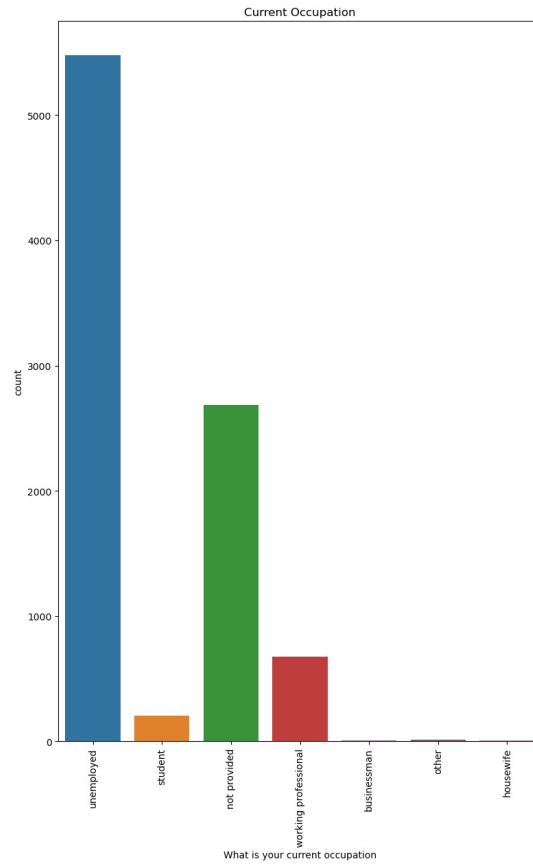
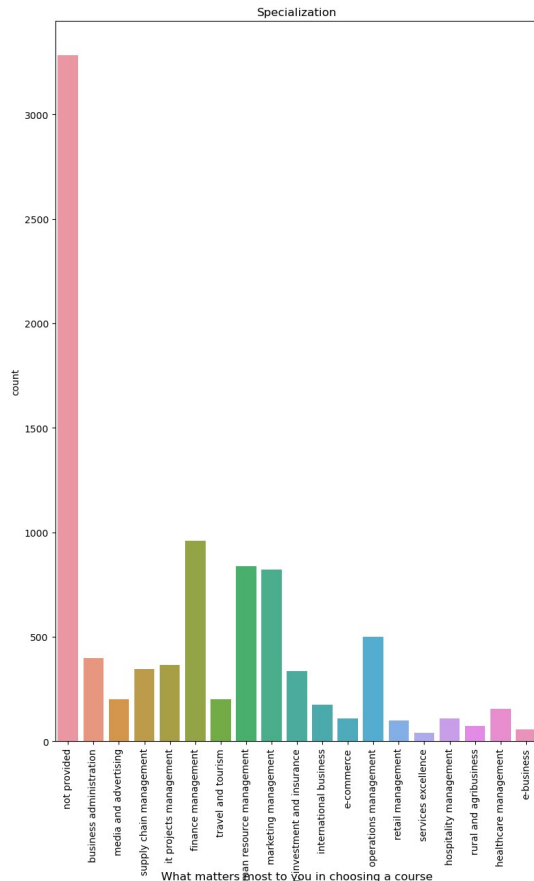


```

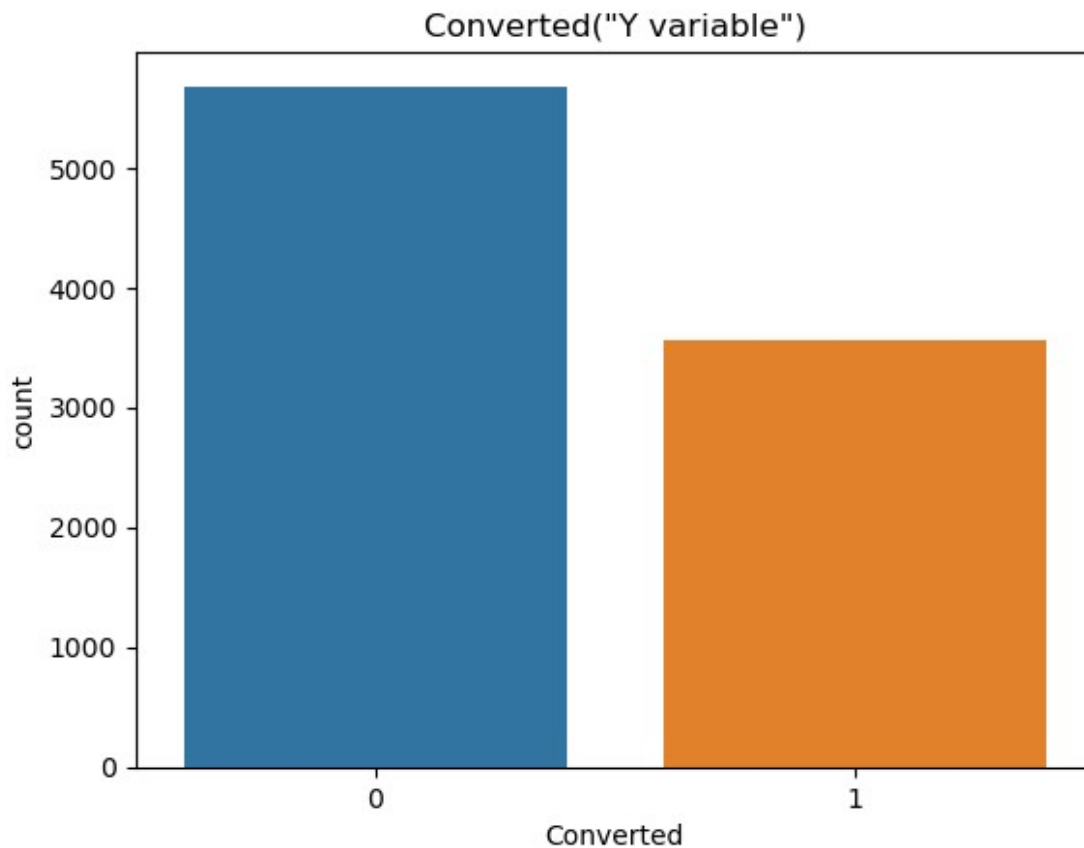
plt.figure(figsize = (20,30))
plt.subplot(2,2,1)
sns.countplot(df_final['Specialization']).tick_params(axis='x',
rotation = 90)
plt.title('Specialization')
plt.subplot(2,2,2)
sns.countplot(df_final['What is your current
occupation']).tick_params(axis='x', rotation = 90)
plt.title('Current Occupation')
plt.subplot(2,2,3)
sns.countplot(df_final['What matters most to you in choosing a
course']).tick_params(axis='x', rotation = 90)

```

```
plt.title('What matters most to you in choosing a course')
plt.subplot(2,2,4)
sns.countplot(df_final['Last Activity']).tick_params(axis='x',
rotation = 90)
plt.title('Last Activity')
plt.show()
```

```
sns.countplot(df_leads['Converted'])
plt.title('Converted("Y variable")')
plt.show()
```



```
# Numerical variables
```

```
plt.figure(figsize = (10,10))
```

```
plt.subplot(221)
```

```
plt.hist(df_final['TotalVisits'], bins = 200)
```

```
plt.title('Total Visits')
```

```
plt.xlim(0,25)
```

```
plt.subplot(222)
```

```
plt.hist(df_final['Total Time Spent on Website'], bins = 10)
```

```
plt.title('Total Time Spent on Website')
```

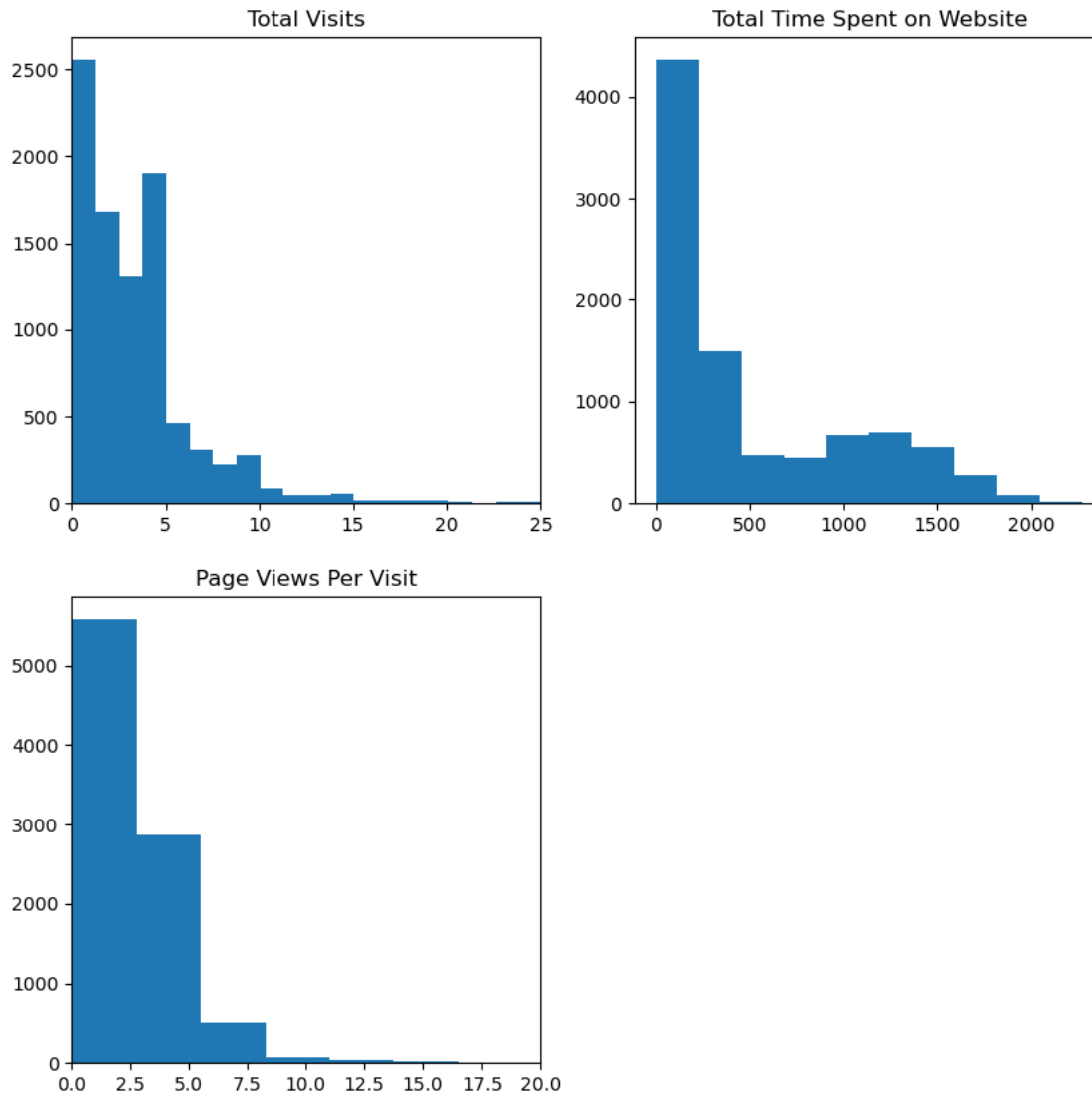
```
plt.subplot(223)
```

```
plt.hist(df_final['Page Views Per Visit'], bins = 20)
```

```
plt.title('Page Views Per Visit')
```

```
plt.xlim(0,20)
```

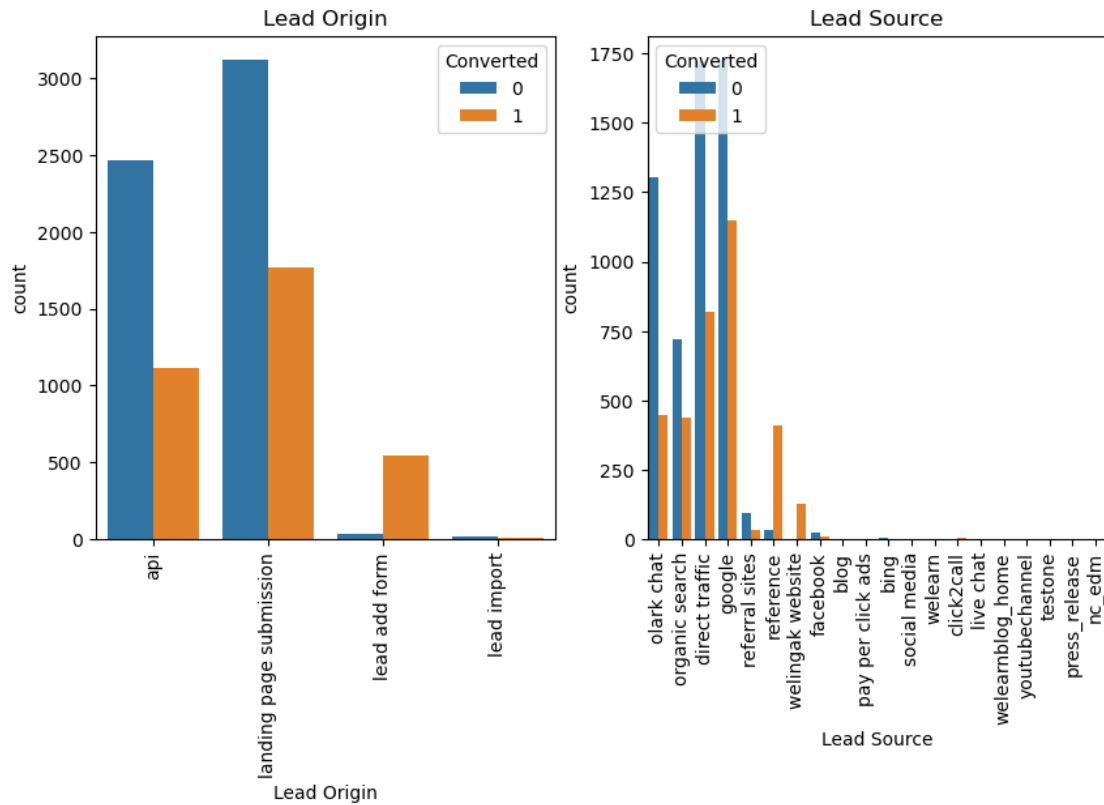
```
plt.show()
```



```
plt.figure(figsize = (10,5))
```

```
plt.subplot(1,2,1)
sns.countplot(x='Lead Origin', hue='Converted', data=
df_final).tick_params(axis='x', rotation = 90)
plt.title('Lead Origin')
```

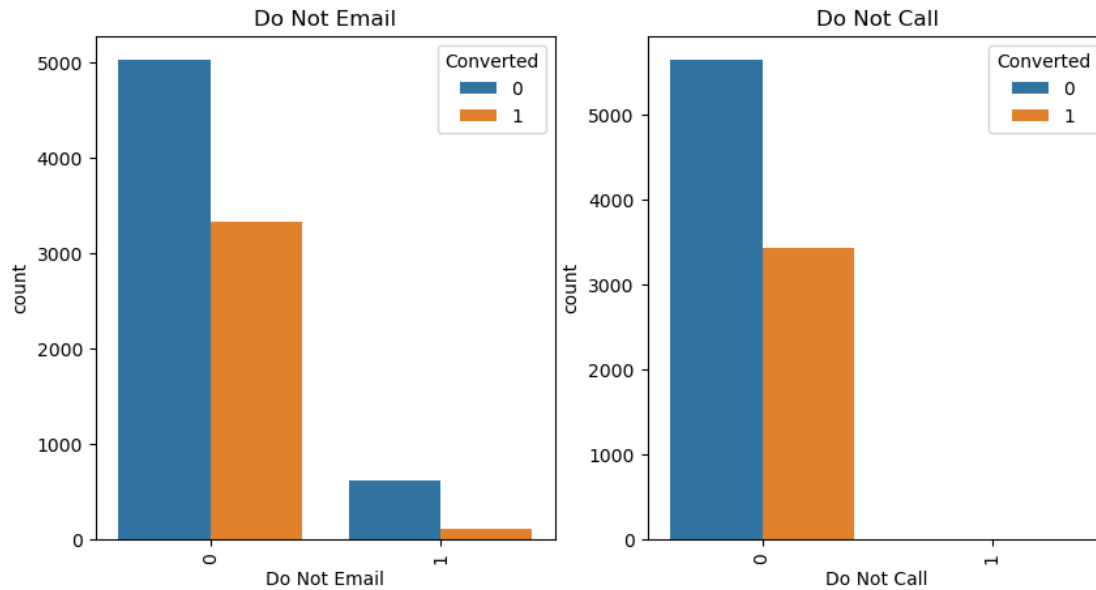
```
plt.subplot(1,2,2)
sns.countplot(x='Lead Source', hue='Converted', data=
df_final).tick_params(axis='x', rotation = 90)
plt.title('Lead Source')
plt.show()
```



```
plt.figure(figsize = (10,5))
```

```
plt.subplot(1,2,1)
sns.countplot(x='Do Not Email', hue='Converted', data=
df_final).tick_params(axis='x', rotation = 90)
plt.title('Do Not Email')
```

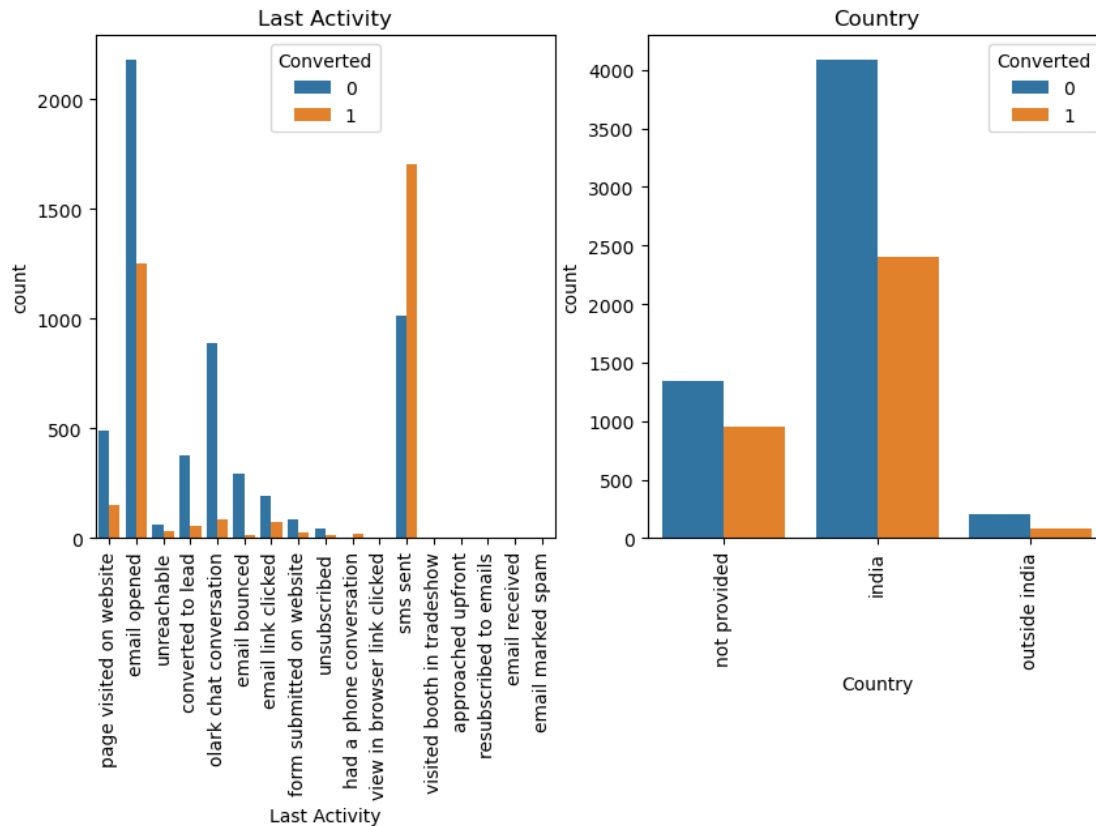
```
plt.subplot(1,2,2)
sns.countplot(x='Do Not Call', hue='Converted', data=
df_final).tick_params(axis='x', rotation = 90)
plt.title('Do Not Call')
plt.show()
```



```
plt.figure(figsize = (10,5))
```

```
plt.subplot(1,2,1)
sns.countplot(x='Last Activity', hue='Converted', data=
df_final).tick_params(axis='x', rotation = 90)
plt.title('Last Activity')
```

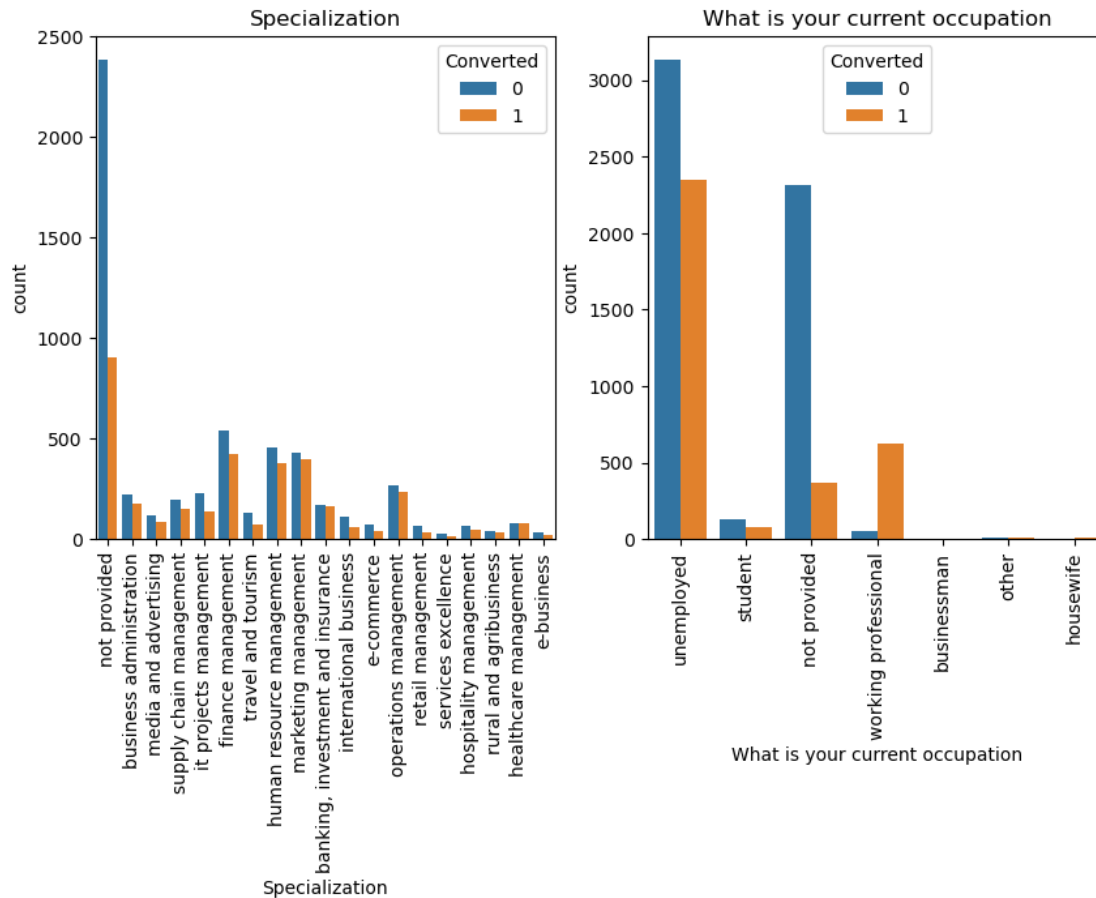
```
plt.subplot(1,2,2)
sns.countplot(x='Country', hue='Converted', data=
df_final).tick_params(axis='x', rotation = 90)
plt.title('Country')
plt.show()
```



```
plt.figure(figsize = (10,5))
```

```
plt.subplot(1,2,1)
sns.countplot(x='Specialization', hue='Converted', data=
df_final).tick_params(axis='x', rotation = 90)
plt.title('Specialization')
```

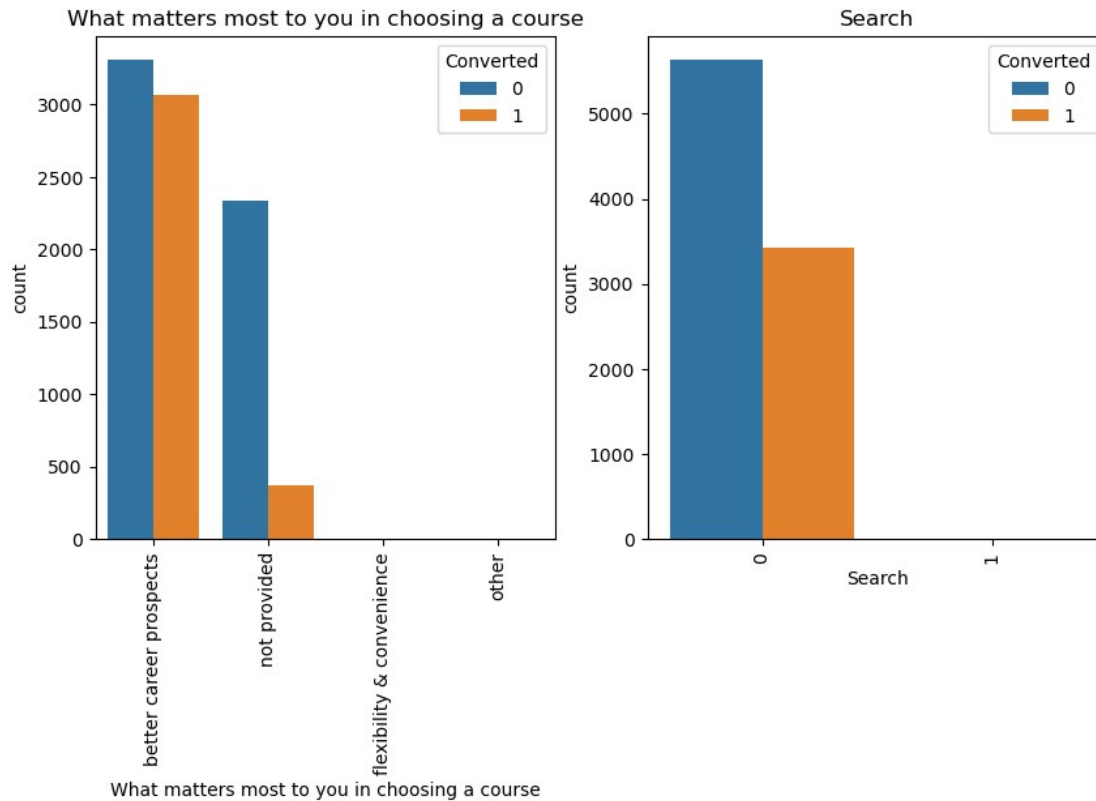
```
plt.subplot(1,2,2)
sns.countplot(x='What is your current occupation', hue='Converted',
data= df_final).tick_params(axis='x', rotation = 90)
plt.title('What is your current occupation')
plt.show()
```



```
plt.figure(figsize = (10,5))
```

```
plt.subplot(1,2,1)
sns.countplot(x='What matters most to you in choosing a course',
hue='Converted', data= df_final).tick_params(axis='x', rotation = 90)
plt.title('What matters most to you in choosing a course')
```

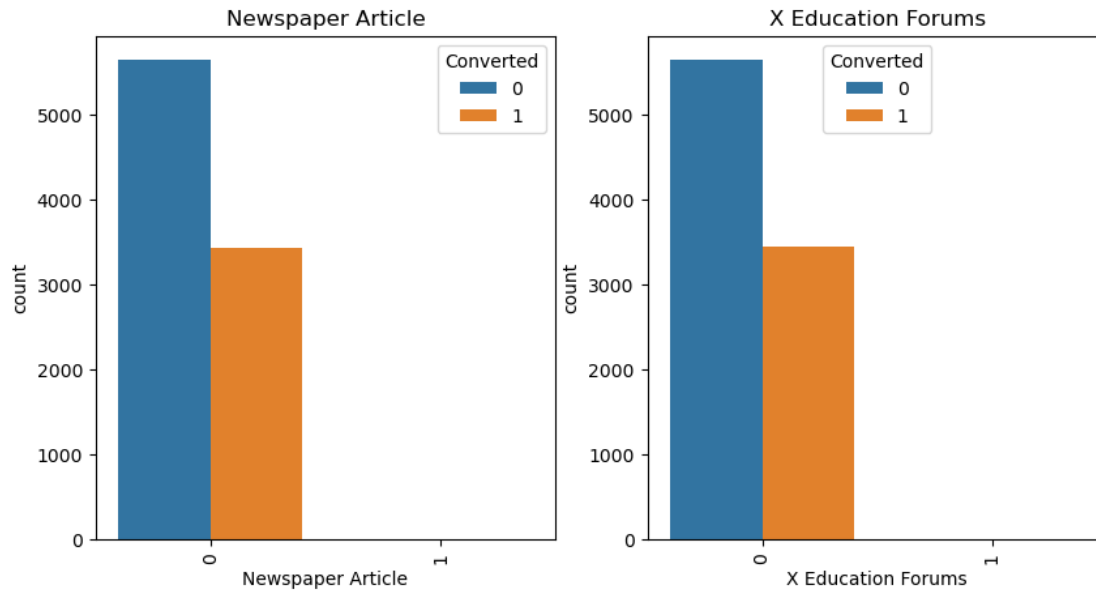
```
plt.subplot(1,2,2)
sns.countplot(x='Search', hue='Converted', data=
df_final).tick_params(axis='x', rotation = 90)
plt.title('Search')
plt.show()
```



```
plt.figure(figsize = (10,5))
```

```
plt.subplot(1,2,1)
sns.countplot(x='Newspaper Article', hue='Converted', data=
df_final).tick_params(axis='x', rotation = 90)
plt.title('Newspaper Article')
```

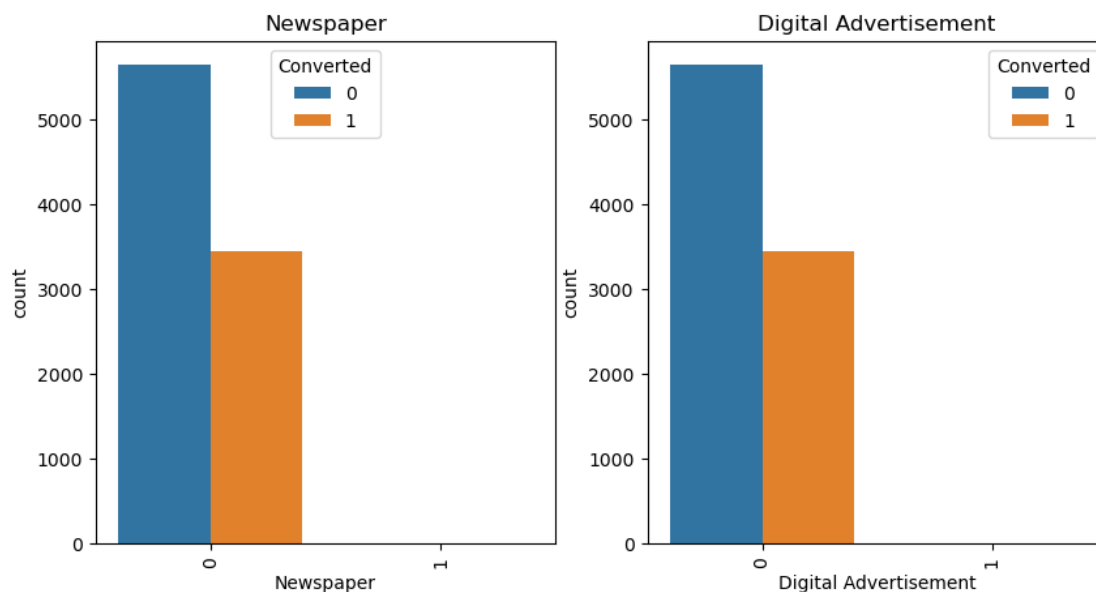
```
plt.subplot(1,2,2)
sns.countplot(x='X Education Forums', hue='Converted', data=
df_final).tick_params(axis='x', rotation = 90)
plt.title('X Education Forums')
plt.show()
```

```
plt.figure(figsize = (10,5))
```

```
plt.subplot(1,2,1)
sns.countplot(x='Newspaper', hue='Converted', data=
df_final).tick_params(axis='x', rotation = 90)
plt.title('Newspaper')
```

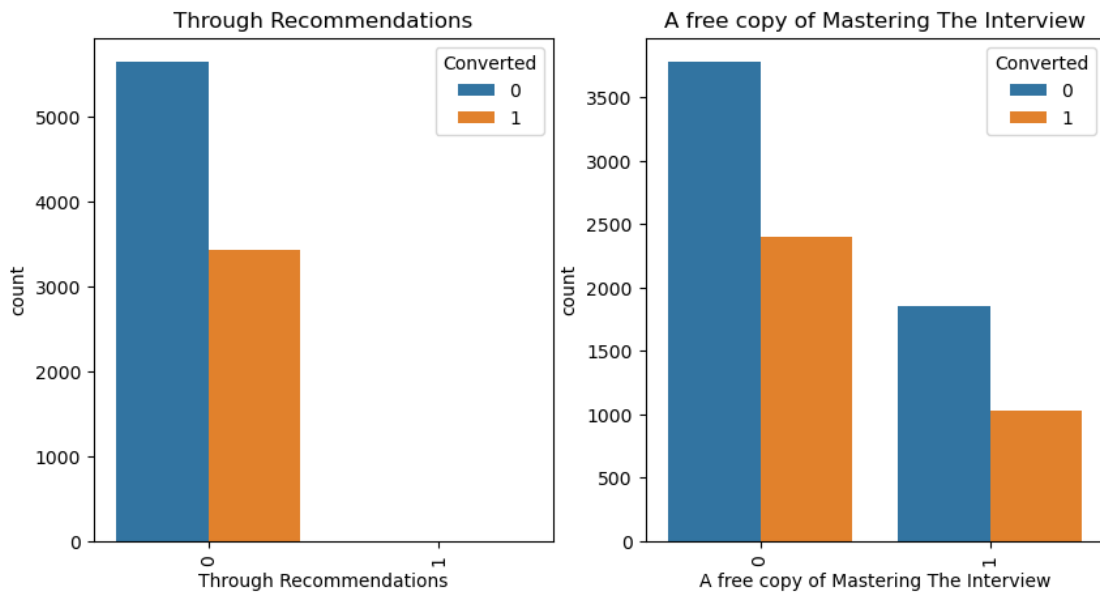
```
plt.subplot(1,2,2)
sns.countplot(x='Digital Advertisement', hue='Converted', data=
df_final).tick_params(axis='x', rotation = 90)
plt.title('Digital Advertisement')
plt.show()
```



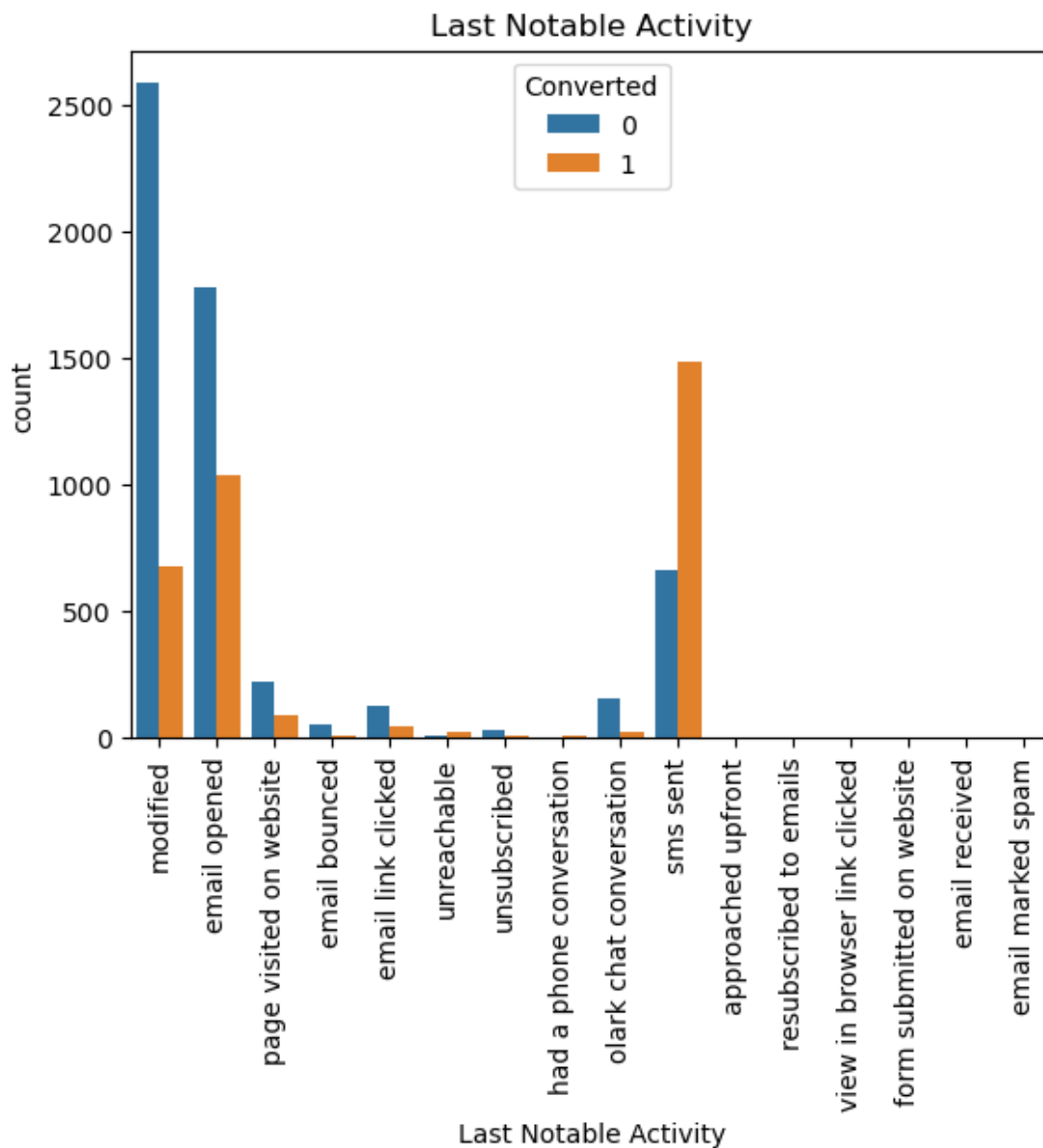
```
plt.figure(figsize = (10,5))

plt.subplot(1,2,1)
sns.countplot(x='Through Recommendations', hue='Converted', data=
df_final).tick_params(axis='x', rotation = 90)
plt.title('Through Recommendations')

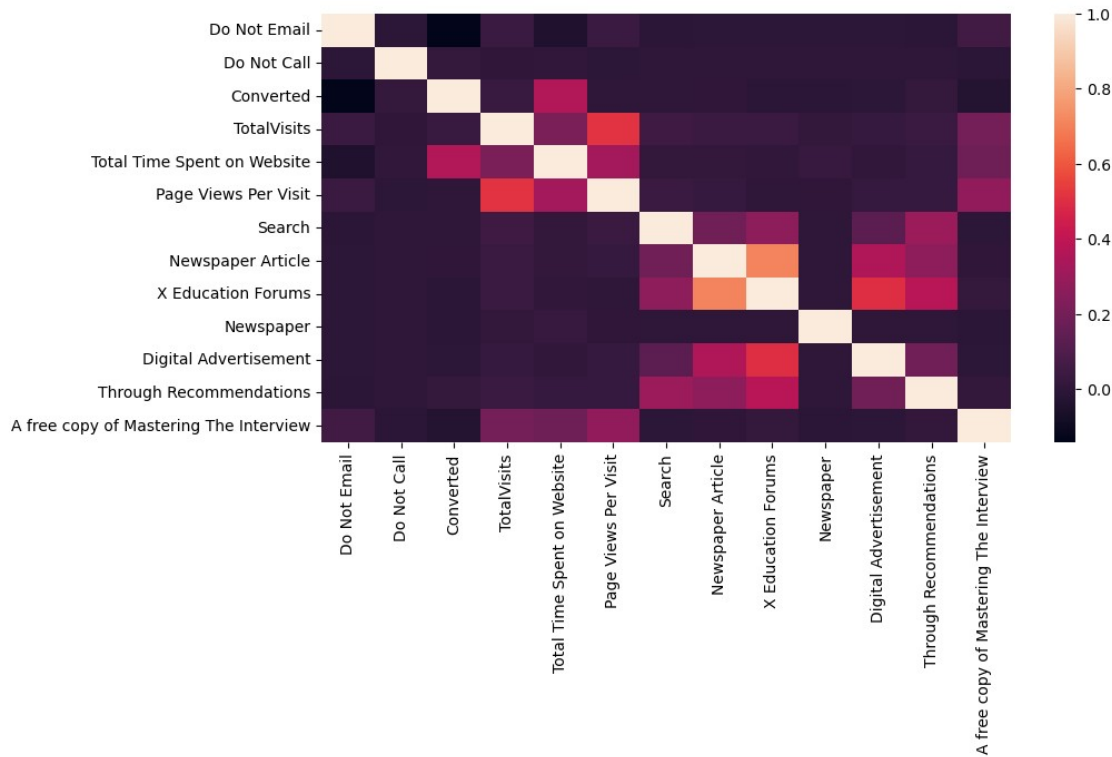
plt.subplot(1,2,2)
sns.countplot(x='A free copy of Mastering The Interview',
hue='Converted', data= df_final).tick_params(axis='x', rotation = 90)
plt.title('A free copy of Mastering The Interview')
plt.show()
```



```
sns.countplot(x='Last Notable Activity', hue='Converted', data=
df_final).tick_params(axis='x', rotation = 90)
plt.title('Last Notable Activity')
plt.show()
```



```
# To check the correlation among variables
plt.figure(figsize=(10,5))
sns.heatmap(df_final.corr())
plt.show()
```



```
numeric = df_final[['TotalVisits','Total Time Spent on Website','Page Views Per Visit']]
numeric.describe(percentiles=[0.25,0.5,0.75,0.9,0.99])
```

	TotalVisits	Total Time Spent on Website	Page Views Per Visit
count	9074.000000	9074.000000	9074.000000
mean	3.456028	482.887481	2.370151
std	4.858802	545.256560	2.160871
min	0.000000	0.000000	0.000000
25%	1.000000	11.000000	1.000000
50%	3.000000	246.000000	2.000000
75%	5.000000	922.750000	3.200000
90%	7.000000	1373.000000	5.000000
99%	17.000000	1839.000000	9.000000
max	251.000000	2272.000000	55.000000

There are no major outliers in the data

Dummy variables

```
df_final.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 9074 entries, 0 to 9239
Data columns (total 21 columns):
 #   Column
Dtype
```

Non-Null Count

```

--- -----
-----
0    Lead Origin          9074 non-null
object
1    Lead Source          9074 non-null
object
2    Do Not Email        9074 non-null
int64
3    Do Not Call          9074 non-null
int64
4    Converted            9074 non-null
int64
5    TotalVisits          9074 non-null
float64
6    Total Time Spent on Website 9074 non-null
int64
7    Page Views Per Visit 9074 non-null
float64
8    Last Activity        9074 non-null
object
9    Country              9074 non-null
object
10   Specialization       9074 non-null
object
11   What is your current occupation 9074 non-null
object
12   What matters most to you in choosing a course 9074 non-null
object
13   Search                9074 non-null
int64
14   Newspaper Article     9074 non-null
int64
15   X Education Forums    9074 non-null
int64
16   Newspaper             9074 non-null
int64
17   Digital Advertisement 9074 non-null
int64
18   Through Recommendations 9074 non-null
int64
19   A free copy of Mastering The Interview 9074 non-null
int64
20   Last Notable Activity 9074 non-null
object
dtypes: float64(2), int64(11), object(8)
memory usage: 1.5+ MB

df_final.loc[:, df_final.dtypes == 'object'].columns

Index(['Lead Origin', 'Lead Source', 'Last Activity', 'Country',
      'Specialization', 'What is your current occupation',

```

```

    'What matters most to you in choosing a course',
    'Last Notable Activity'],
    dtype='object')

```

```

# Create dummy variables using the 'get_dummies'
dummy = pd.get_dummies(df_final[['Lead Origin', 'Specialization', 'Lead
Source', 'Do Not Email', 'Last Activity', 'What is your current
occupation', 'A free copy of Mastering The Interview', 'Last Notable
Activity']], drop_first=True)
# Add the results to the master dataframe
df_final_dum = pd.concat([df_final, dummy], axis=1)
df_final_dum

```

Call \	Lead Origin	Lead Source	Do Not Email	Do Not
0	api	olark chat	0	
0				
1	api	organic search	0	
0				
2	landing page submission	direct traffic	0	
0				
3	landing page submission	direct traffic	0	
0				
4	landing page submission	google	0	
0				
...
..				
9235	landing page submission	direct traffic	1	
0				
9236	landing page submission	direct traffic	0	
0				
9237	landing page submission	direct traffic	1	
0				
9238	landing page submission	google	0	
0				
9239	landing page submission	direct traffic	0	
0				

	Converted	TotalVisits	Total Time Spent on Website \
0	0	0.0	0
1	0	5.0	674
2	1	2.0	1532
3	0	1.0	305
4	1	2.0	1428
...
9235	1	8.0	1845
9236	0	2.0	238
9237	0	2.0	199
9238	1	3.0	499
9239	1	6.0	1279

	Page Views Per Visit	Last Activity	
Country ... \			
0	0.00	page visited on website	not
provided ...			
1	2.50	email opened	
india ...			
2	2.00	email opened	
india ...			
3	1.00	unreachable	
india ...			
4	1.00	converted to lead	
india ...			
...
.			
9235	2.67	email marked spam	outside
india ...			
9236	2.00	sms sent	
india ...			
9237	2.00	sms sent	
india ...			
9238	3.00	sms sent	
india ...			
9239	3.00	sms sent	outside
india ...			

	Last Notable Activity_form submitted on website \
0	0
1	0
2	0
3	0
4	0
...	...
9235	0
9236	0
9237	0
9238	0
9239	0

	Last Notable Activity_had a phone conversation \
0	0
1	0
2	0
3	0
4	0
...	...
9235	0
9236	0
9237	0
9238	0

9239 0

	Last Notable Activity_modified \
0	1
1	0
2	0
3	1
4	1
...	...
9235	0
9236	0
9237	0
9238	0
9239	1

	Last Notable Activity_olark chat conversation \
0	0
1	0
2	0
3	0
4	0
...	...
9235	0
9236	0
9237	0
9238	0
9239	0

	Last Notable Activity_page visited on website \
0	0
1	0
2	0
3	0
4	0
...	...
9235	0
9236	0
9237	0
9238	0
9239	0

	Last Notable Activity_resubscribed to emails \
0	0
1	0
2	0
3	0
4	0
...	...
9235	0
9236	0

9237	0
9238	0
9239	0

	Last Notable Activity_sms sent	Last Notable Activity_unreachable \
--	--------------------------------	-------------------------------------

0	0	
0		
1	0	
0		
2	0	
0		
3	0	
0		
4	0	
0		
...
.		
9235	0	
0		
9236	1	
0		
9237	1	
0		
9238	1	
0		
9239	0	
0		

	Last Notable Activity_unsubscribed \
--	--------------------------------------

0	0
1	0
2	0
3	0
4	0
...	...
9235	0
9236	0
9237	0
9238	0
9239	0

	Last Notable Activity_view in browser link clicked
--	--

0	0
1	0
2	0
3	0
4	0
...	...
9235	0

9236	0
9237	0
9238	0
9239	0

[9074 rows x 100 columns]

```
df_final_dum = df_final_dum.drop(['What is your current occupation_not
provided','Lead Origin', 'Lead Source', 'Do Not Email', 'Do Not
Call','Last Activity', 'Country', 'Specialization',
'Specialization_not provided','What is your current occupation','What
matters most to you in choosing a course', 'Search','Newspaper
Article', 'X Education Forums', 'Newspaper','Digital Advertisement',
'Through Recommendations','A free copy of Mastering The Interview',
'Last Notable Activity'], 1)
df_final_dum
```

	Converted	TotalVisits	Total Time Spent on Website \
0	0	0.0	0
1	0	5.0	674
2	1	2.0	1532
3	0	1.0	305
4	1	2.0	1428
...
9235	1	8.0	1845
9236	0	2.0	238
9237	0	2.0	199
9238	1	3.0	499
9239	1	6.0	1279

	Page Views Per Visit	Lead Origin_landing page submission \
0	0.00	0
1	2.50	0
2	2.00	1
3	1.00	1
4	1.00	1
...
9235	2.67	1
9236	2.00	1
9237	2.00	1
9238	3.00	1
9239	3.00	1

	Lead Origin_lead add form	Lead Origin_lead import \
0	0	0
1	0	0
2	0	0
3	0	0
4	0	0
...

9235	0	0
9236	0	0
9237	0	0
9238	0	0
9239	0	0

Specialization_business administration	Specialization_e-
business \	
0	0
0	
1	0
0	
2	1
0	
3	0
0	
4	0
0	
...	...
.	..
9235	0
0	
9236	0
0	
9237	1
0	
9238	0
0	
9239	0
0	

Specialization_e-commerce	...	\
0	0	...
1	0	...
2	0	...
3	0	...
4	0	...
...
9235	0	...
9236	0	...
9237	0	...
9238	0	...
9239	0	...

Last Notable Activity_form submitted on website	\
0	0
1	0
2	0
3	0
4	0

...	...
9235	0
9236	0
9237	0
9238	0
9239	0

	Last Notable Activity_had a phone conversation \
--	--

0	0
1	0
2	0
3	0
4	0

...	...
9235	0
9236	0
9237	0
9238	0
9239	0

	Last Notable Activity_modified \
--	----------------------------------

0	1
1	0
2	0
3	1
4	1

...	...
9235	0
9236	0
9237	0
9238	0
9239	1

	Last Notable Activity_olark chat conversation \
--	---

0	0
1	0
2	0
3	0
4	0

...	...
9235	0
9236	0
9237	0
9238	0
9239	0

	Last Notable Activity_page visited on website \
--	---

0	0
1	0
2	0

3	0
4	0
...	...
9235	0
9236	0
9237	0
9238	0
9239	0

Last Notable Activity_resubscribed to emails \	
0	0
1	0
2	0
3	0
4	0
...	...
9235	0
9236	0
9237	0
9238	0
9239	0

Last Notable Activity_sms sent		Last Notable Activity_unreachable \
0	0	
0		
1	0	
0		
2	0	
0		
3	0	
0		
4	0	
0		
...
.		
9235	0	
0		
9236	1	
0		
9237	1	
0		
9238	1	
0		
9239	0	
0		

Last Notable Activity_unsubscribed \	
0	0
1	0

2	0
3	0
4	0
...	...
9235	0
9236	0
9237	0
9238	0
9239	0

	Last Notable Activity_view in browser link clicked
0	0
1	0
2	0
3	0
4	0
...	...
9235	0
9236	0
9237	0
9238	0
9239	0

[9074 rows x 79 columns]

Train test and split

```
X = df_final_dum.drop(['Converted'], 1)
X.head()
```

	TotalVisits	Total Time Spent on Website	Page Views Per Visit	\
0	0.0	0	0.0	
1	5.0	674	2.5	
2	2.0	1532	2.0	
3	1.0	305	1.0	
4	2.0	1428	1.0	

	Lead Origin_landing page submission	Lead Origin_lead add form	\
0	0	0	
1	0	0	
2	1	0	
3	1	0	
4	1	0	

	Lead Origin_lead import	Specialization_business administration	\
0	0	0	
1	0	0	
2	0	1	
3	0	0	
4	0	0	

	Specialization_e-business	Specialization_e-commerce	\
0	0	0	
1	0	0	
2	0	0	
3	0	0	
4	0	0	

	Specialization_finance management	...	\
0	0	...	
1	0	...	
2	0	...	
3	0	...	
4	0	...	

	Last Notable Activity_form submitted on website	\
0	0	
1	0	
2	0	
3	0	
4	0	

	Last Notable Activity_had a phone conversation	\
0	0	
1	0	
2	0	
3	0	
4	0	

	Last Notable Activity_modified	\
0	1	
1	0	
2	0	
3	1	
4	1	

	Last Notable Activity_olark chat conversation	\
0	0	
1	0	
2	0	
3	0	
4	0	

	Last Notable Activity_page visited on website	\
0	0	
1	0	
2	0	
3	0	
4	0	

	Last Notable Activity_resubscribed to emails \
0	0
1	0
2	0
3	0
4	0

	Last Notable Activity_sms sent	Last Notable Activity_unreachable \
0	0	0
1	0	0
2	0	0
3	0	0
4	0	0

	Last Notable Activity_unsubscribed \
0	0
1	0
2	0
3	0
4	0

	Last Notable Activity_view in browser link clicked
0	0
1	0
2	0
3	0
4	0

[5 rows x 78 columns]

```
# Putting the target variable in y
y = df_final_dum['Converted']
y.head()
```

0	0
1	0
2	1
3	0
4	1

Name: Converted, dtype: int64


```
# Split the dataset into 70% and 30% for train and test respectively
X_train, X_test, y_train, y_test = train_test_split(X, y,
train_size=0.7, test_size=0.3, random_state=10)
```

```
# Scale the three numeric features
```

```
scaler = MinMaxScaler()
X_train[['TotalVisits', 'Page Views Per Visit', 'Total Time Spent on
Website']] = scaler.fit_transform(X_train[['TotalVisits', 'Page Views
Per Visit', 'Total Time Spent on Website']])
X_train.head()
```

	TotalVisits	Total Time Spent on Website	Page Views Per
Visit \			
1289	0.014184	0.612676	0.083333
3604	0.000000	0.000000	0.000000
5584	0.042553	0.751761	0.250000
7679	0.000000	0.000000	0.000000
7563	0.014184	0.787852	0.083333

	Lead Origin_landing page submission	Lead Origin_lead add
form \		
1289	1	0
3604	0	0
5584	1	0
7679	0	0
7563	1	0

	Lead Origin_lead import	Specialization_business administration
\		
1289	0	0
3604	0	0
5584	0	0
7679	0	0
7563	0	0

	Specialization_e-business	Specialization_e-commerce	\
1289	0	0	
3604	0	0	
5584	0	0	
7679	0	0	
7563	0	0	

	Specialization_finance management	...	\
1289	1	...	
3604	0	...	
5584	0	...	
7679	0	...	
7563	0	...	

	Last Notable Activity_form submitted on website	\
1289	0	
3604	0	
5584	0	
7679	0	
7563	0	

	Last Notable Activity_had a phone conversation	\
1289	0	
3604	0	
5584	0	
7679	0	
7563	0	

	Last Notable Activity_modified	\
1289	0	
3604	0	
5584	0	
7679	0	
7563	1	

	Last Notable Activity_olark chat conversation	\
1289	0	
3604	0	
5584	0	
7679	0	
7563	0	

	Last Notable Activity_page visited on website	\
1289	0	
3604	1	
5584	0	
7679	0	
7563	0	

	Last Notable Activity_resubscribed to emails \
1289	0
3604	0
5584	0
7679	0
7563	0

	Last Notable Activity_sms sent	Last Notable Activity_unreachable \
1289	0	
0		
3604	0	
0		
5584	0	
0		
7679	0	
0		
7563	0	
0		

	Last Notable Activity_unsubscribed \
1289	0
3604	0
5584	0
7679	0
7563	0

	Last Notable Activity_view in browser link clicked
1289	0
3604	0
5584	0
7679	0
7563	0

[5 rows x 78 columns]

Building the model

```
logreg = LogisticRegression()
# Running RFE with 15 variables as output
rfe = RFE(logreg,n_features_to_select=15)
rfe = rfe.fit(X_train, y_train)

# Features that have been selected by RFE
list(zip(X_train.columns, rfe.support_, rfe.ranking_))

[('TotalVisits', True, 1),
 ('Total Time Spent on Website', True, 1),
 ('Page Views Per Visit', False, 4),
```

('Lead Origin_landing page submission', False, 25),
('Lead Origin_lead add form', True, 1),
('Lead Origin_lead import', False, 38),
('Specialization_business administration', False, 33),
('Specialization_e-business', False, 32),
('Specialization_e-commerce', False, 24),
('Specialization_finance management', False, 30),
('Specialization_healthcare management', False, 27),
('Specialization_hospitality management', False, 43),
('Specialization_human resource management', False, 31),
('Specialization_international business', False, 36),
('Specialization_it projects management', False, 29),
('Specialization_marketing management', False, 21),
('Specialization_media and advertising', False, 40),
('Specialization_operations management', False, 26),
('Specialization_retail management', False, 60),
('Specialization_rural and agribusiness', False, 23),
('Specialization_services excellence', False, 22),
('Specialization_supply chain management', False, 28),
('Specialization_travel and tourism', False, 35),
('Lead Source_blog', False, 41),
('Lead Source_click2call', False, 61),
('Lead Source_direct traffic', False, 16),
('Lead Source_facebook', False, 39),
('Lead Source_google', False, 18),
('Lead Source_live chat', False, 44),
('Lead Source_nc_edm', False, 63),
('Lead Source_olark chat', True, 1),
('Lead Source_organic search', False, 17),
('Lead Source_pay per click ads', False, 62),
('Lead Source_press_release', False, 34),
('Lead Source_reference', False, 2),
('Lead Source_referral sites', False, 19),
('Lead Source_social media', False, 20),
('Lead Source_testone', False, 42),
('Lead Source_welearn', False, 45),
('Lead Source_welearnblog_home', False, 46),
('Lead Source_welingak website', True, 1),
('Lead Source_youtubechannel', False, 48),
('Last Activity_converted to lead', False, 10),
('Last Activity_email bounced', True, 1),
('Last Activity_email link clicked', False, 56),
('Last Activity_email marked spam', False, 49),
('Last Activity_email opened', False, 37),
('Last Activity_email received', False, 52),
('Last Activity_form submitted on website', False, 51),
('Last Activity_had a phone conversation', False, 3),
('Last Activity_olark chat conversation', True, 1),
('Last Activity_page visited on website', False, 13),
('Last Activity_resubscribed to emails', False, 5),

```

('Last Activity_sms sent', True, 1),
('Last Activity_unreachable', False, 15),
('Last Activity_unsubscribed', False, 11),
('Last Activity_view in browser link clicked', False, 59),
('Last Activity_visited booth in tradeshow', False, 55),
('What is your current occupation_housewife', True, 1),
('What is your current occupation_other', True, 1),
('What is your current occupation_student', True, 1),
('What is your current occupation_unemployed', True, 1),
('What is your current occupation_working professional', True, 1),
('Last Notable Activity_email bounced', False, 53),
('Last Notable Activity_email link clicked', False, 9),
('Last Notable Activity_email marked spam', False, 47),
('Last Notable Activity_email opened', False, 12),
('Last Notable Activity_email received', False, 57),
('Last Notable Activity_form submitted on website', False, 58),
('Last Notable Activity_had a phone conversation', True, 1),
('Last Notable Activity_modified', False, 6),
('Last Notable Activity_olark chat conversation', False, 7),
('Last Notable Activity_page visited on website', False, 8),
('Last Notable Activity_resubscribed to emails', False, 14),
('Last Notable Activity_sms sent', False, 50),
('Last Notable Activity_unreachable', True, 1),
('Last Notable Activity_unsubscribed', False, 54),
('Last Notable Activity_view in browser link clicked', False, 64)]

# Put all the columns selected by RFE in the variable 'col'
col = X_train.columns[rfe.support_]

# Selecting columns selected by RFE
X_train = X_train[col]

X_train_sm = sm.add_constant(X_train)
logml = sm.GLM(y_train, X_train_sm, family = sm.families.Binomial())
res = logml.fit()
res.summary()

<class 'statsmodels.iolib.summary.Summary'>
"""

```

Generalized Linear Model Regression Results

```

=====
=====
Dep. Variable:          Converted    No. Observations:
6351
Model:                  GLM         Df Residuals:
6335
Model Family:          Binomial     Df Model:
15
Link Function:          Logit        Scale:
1.0000

```

Method: IRLS Log-Likelihood:
-2654.3
Date: Thu, 15 Jun 2023 Deviance:
5308.6
Time: 19:23:58 Pearson chi2:
6.59e+03
No. Iterations: 22 Pseudo R-squ. (CS):
0.3926
Covariance Type: nonrobust

=====					
=====					
err	z	P> z	[0.025	0.975]	coef std

const					-3.4921
0.114	-30.632	0.000	-3.716	-3.269	4.4247
TotalVisits					4.4247
1.467	3.016	0.003	1.549	7.300	4.6634
Total Time Spent on Website					4.6634
0.166	28.038	0.000	4.337	4.989	3.6800
Lead Origin_lead add form					3.6800
0.225	16.385	0.000	3.240	4.120	1.6015
Lead Source_olark chat					1.6015
0.112	14.344	0.000	1.383	1.820	2.6284
Lead Source_welingak website					2.6284
1.036	2.537	0.011	0.597	4.659	-1.8713
Last Activity_email bounced					-1.8713
0.337	-5.559	0.000	-2.531	-1.212	-1.4071
Last Activity_olark chat conversation					-1.4071
0.167	-8.405	0.000	-1.735	-1.079	1.2137
Last Activity_sms sent					1.2137
0.074	16.472	0.000	1.069	1.358	25.4295
What is your current occupation_housewife					25.4295
3.09e+04	0.001	0.999	-6.05e+04	6.06e+04	2.2360
What is your current occupation_other					2.2360
0.756	2.959	0.003	0.755	3.717	1.3091
What is your current occupation_student					1.3091
0.226	5.798	0.000	0.867	1.752	1.1793
What is your current occupation_unemployed					1.1793
0.086	13.747	0.000	1.011	1.347	3.7384
What is your current occupation_working professional					3.7384
0.205	18.224	0.000	3.336	4.141	24.0520
Last Notable Activity_had a phone conversation					24.0520
2.16e+04	0.001	0.999	-4.23e+04	4.24e+04	1.8612
Last Notable Activity_unreachable					1.8612
0.602	3.092	0.002	0.681	3.041	
=====					

```

=====
"""

# Make a VIF dataframe for all the variables present
vif = pd.DataFrame()
vif['Features'] = X_train.columns
vif['VIF'] = [variance_inflation_factor(X_train.values, i) for i in
range(X_train.shape[1])]
vif['VIF'] = round(vif['VIF'], 2)
vif = vif.sort_values(by = "VIF", ascending = False)
vif

11          Features  VIF
1          Total Time Spent on Website 2.07
0          TotalVisits 1.82
2          Lead Origin_lead add form 1.59
7          Last Activity_sms sent 1.55
3          Lead Source_olark chat 1.51
6          Last Activity_olark chat conversation 1.37
12 What is your current occupation_working profes... 1.32
4          Lead Source_welingak website 1.31
10         What is your current occupation_student 1.05
5          Last Activity_email bounced 1.03
9          What is your current occupation_other 1.01
14         Last Notable Activity_unreachable 1.01
8          What is your current occupation_housewife 1.00
13         Last Notable Activity_had a phone conversation 1.00

# Revoming Last Notable Activity had a phone conversation as it has
high p value
X_train.drop('Last Notable Activity_had a phone conversation', axis =
1, inplace = True)

# Refit the model with the new set of features
X_train_sm = sm.add_constant(X_train)
logm2 = sm.GLM(y_train, X_train_sm, family = sm.families.Binomial())
res = logm2.fit()
res.summary()

<class 'statsmodels.iolib.summary.Summary'>
"""

```

Generalized Linear Model Regression Results

```

=====
=====
Dep. Variable:          Converted    No. Observations:
6351
Model:                  GLM    Df Residuals:
6336
Model Family:          Binomial    Df Model:

```

14

Link Function: Logit Scale:
 1.0000
 Method: IRLS Log-Likelihood:
 -2662.3
 Date: Thu, 15 Jun 2023 Deviance:
 5324.5
 Time: 19:26:29 Pearson chi2:
 6.59e+03
 No. Iterations: 20 Pseudo R-squ. (CS):
 0.3911
 Covariance Type: nonrobust

=====					
=====					
err	z	P> z	[0.025	0.975]	coef std

const					-3.4879
0.114	-30.606	0.000	-3.711	-3.265	4.6014
TotalVisits					4.6014
1.477	3.115	0.002	1.707	7.496	4.6490
Total Time Spent on Website					4.6490
0.166	28.026	0.000	4.324	4.974	3.6740
Lead Origin_lead add form					3.6740
0.225	16.361	0.000	3.234	4.114	1.5975
Lead Source_olark chat					1.5975
0.112	14.310	0.000	1.379	1.816	2.6282
Lead Source_welingak website					2.6282
1.036	2.536	0.011	0.597	4.659	-1.8760
Last Activity_email bounced					-1.8760
0.336	-5.576	0.000	-2.535	-1.217	-1.4115
Last Activity_olark chat conversation					-1.4115
0.167	-8.434	0.000	-1.740	-1.083	1.2055
Last Activity_sms sent					1.2055
0.074	16.383	0.000	1.061	1.350	23.4238
What is your current occupation_housewife					23.4238
1.14e+04	0.002	0.998	-2.22e+04	2.23e+04	2.2289
What is your current occupation_other					2.2289
0.755	2.950	0.003	0.748	3.710	1.3076
What is your current occupation_student					1.3076
0.226	5.795	0.000	0.865	1.750	1.1845
What is your current occupation_unemployed					1.1845
0.086	13.821	0.000	1.017	1.352	3.7363
What is your current occupation_working professional					3.7363
0.205	18.225	0.000	3.334	4.138	1.8518
Last Notable Activity_unreachable					1.8518
0.602	3.078	0.002	0.673	3.031	
=====					


```

=====
"""

# Make a VIF dataframe for all the variables present
vif = pd.DataFrame()
vif['Features'] = X_train.columns
vif['VIF'] = [variance_inflation_factor(X_train.values, i) for i in
range(X_train.shape[1])]
vif['VIF'] = round(vif['VIF'], 2)
vif = vif.sort_values(by = "VIF", ascending = False)
vif

11          Features  VIF
1          Total Time Spent on Website 2.07
0          TotalVisits 1.82
2          Lead Origin_lead add form 1.59
7          Last Activity_sms sent 1.55
3          Lead Source_olark chat 1.51
6          Last Activity_olark chat conversation 1.37
12 What is your current occupation_working profes... 1.32
4          Lead Source_welingak website 1.31
10         What is your current occupation_student 1.05
5          Last Activity_email bounced 1.03
9          What is your current occupation_other 1.01
13         Last Notable Activity_unreachable 1.01
8          What is your current occupation_housewife 1.00

# Removing What is your current occupation _housewife as it has high p
value
X_train.drop('What is your current occupation_housewife', axis = 1,
inplace = True)

# Refit the model with the new set of features
X_train_sm = sm.add_constant(X_train)
logm3 = sm.GLM(y_train, X_train_sm, family = sm.families.Binomial())
res = logm3.fit()
res.summary()

<class 'statsmodels.iolib.summary.Summary'>
"""

                Generalized Linear Model Regression Results

=====
=====
Dep. Variable:                Converted    No. Observations:
6351
Model:                        GLM        Df Residuals:
6337
Model Family:                Binomial    Df Model:
13

```

Link Function: Logit Scale:
 1.0000
 Method: IRLS Log-Likelihood:
 -2670.9
 Date: Thu, 15 Jun 2023 Deviance:
 5341.7
 Time: 19:27:54 Pearson chi2:
 6.61e+03
 No. Iterations: 7 Pseudo R-squ. (CS):
 0.3895
 Covariance Type: nonrobust

=====					
=====					
err	z	P> z	[0.025	0.975]	coef std

const					-3.4579
0.113	-30.555	0.000	-3.680	-3.236	4.5335
TotalVisits					4.5335
1.472	3.080	0.002	1.649	7.418	4.6435
Total Time Spent on Website					4.6435
0.166	28.042	0.000	4.319	4.968	3.6867
Lead Origin_lead add form					3.6867
0.225	16.419	0.000	3.247	4.127	1.5866
Lead Source_olark chat					1.5866
0.111	14.247	0.000	1.368	1.805	2.6112
Lead Source_welingak website					2.6112
1.036	2.520	0.012	0.580	4.642	-1.8831
Last Activity_email bounced					-1.8831
0.336	-5.600	0.000	-2.542	-1.224	-1.4171
Last Activity_olark chat conversation					-1.4171
0.167	-8.474	0.000	-1.745	-1.089	1.1971
Last Activity_sms sent					1.1971
0.073	16.298	0.000	1.053	1.341	2.2060
What is your current occupation_other					2.2060
0.755	2.920	0.003	0.725	3.686	1.2845
What is your current occupation_student					1.2845
0.225	5.697	0.000	0.843	1.726	1.1625
What is your current occupation_unemployed					1.1625
0.085	13.650	0.000	0.996	1.329	3.7125
What is your current occupation_working professional					3.7125
0.205	18.134	0.000	3.311	4.114	1.8421
Last Notable Activity_unreachable					1.8421
0.601	3.063	0.002	0.663	3.021	
=====					
=====					
" " "					

```
# Make a VIF dataframe for all the variables present
vif = pd.DataFrame()
vif['Features'] = X_train.columns
vif['VIF'] = [variance_inflation_factor(X_train.values, i) for i in
range(X_train.shape[1])]
vif['VIF'] = round(vif['VIF'], 2)
vif = vif.sort_values(by = "VIF", ascending = False)
vif
```

	Features	VIF
10	What is your current occupation_unemployed	2.30
1	Total Time Spent on Website	2.06
0	TotalVisits	1.82
2	Lead Origin_lead add form	1.58
7	Last Activity_sms sent	1.55
3	Lead Source_olark chat	1.51
6	Last Activity_olark chat conversation	1.37
11	What is your current occupation_working profes...	1.32
4	Lead Source_welingak website	1.31
9	What is your current occupation_student	1.05
5	Last Activity_email bounced	1.03
8	What is your current occupation_other	1.01
12	Last Notable Activity_unreachable	1.01

```
# p value of what is your current occupation_other is not correct
X_train.drop('What is your current occupation_other', axis = 1,
inplace = True)
```

```
# Refit the model with the new set of features
X_train_sm = sm.add_constant(X_train)
logm4 = sm.GLM(y_train, X_train_sm, family = sm.families.Binomial())
res = logm4.fit()
res.summary()
```

```
<class 'statsmodels.iolib.summary.Summary'>
"""
```

Generalized Linear Model Regression Results

```
=====
=====
Dep. Variable:          Converted    No. Observations:
6351
Model:                  GLM         Df Residuals:
6338
Model Family:          Binomial     Df Model:
12
Link Function:          Logit        Scale:
1.0000
Method:                 IRLS         Log-Likelihood:
-2675.6
Date:                   Thu, 15 Jun 2023    Deviance:
```

```

5351.2
Time:                19:31:22    Pearson chi2:
6.61e+03
No. Iterations:      7    Pseudo R-squ. (CS):
0.3886
Covariance Type:     nonrobust

```

```

=====
=====
err          z      P>|z|      [0.025      0.975]      coef      std
-----
const                                -3.4394
0.113    -30.490      0.000      -3.660      -3.218
TotalVisits                                4.7279
1.483      3.187      0.001      1.820      7.635
Total Time Spent on Website                                4.6530
0.166     28.107      0.000      4.328      4.977
Lead Origin_lead add form                                3.6934
0.225     16.446      0.000      3.253      4.134
Lead Source_olark chat                                1.5847
0.111     14.225      0.000      1.366      1.803
Lead Source_welingak website                                2.6117
1.036      2.520      0.012      0.581      4.643
Last Activity_email bounced                                -1.8882
0.336     -5.617      0.000      -2.547      -1.229
Last Activity_olark chat conversation                                -1.4128
0.167     -8.456      0.000      -1.740      -1.085
Last Activity_sms sent                                1.1913
0.073     16.238      0.000      1.047      1.335
What is your current occupation_student                                1.2606
0.225      5.596      0.000      0.819      1.702
What is your current occupation_unemployed                                1.1385
0.085     13.472      0.000      0.973      1.304
What is your current occupation_working professional                                3.6882
0.204     18.039      0.000      3.287      4.089
Last Notable Activity_unreachable                                1.8333
0.601      3.049      0.002      0.655      3.012
=====
=====
"""

```

```

# Make a VIF dataframe for all the variables present
vif = pd.DataFrame()
vif['Features'] = X_train.columns
vif['VIF'] = [variance_inflation_factor(X_train.values, i) for i in
range(X_train.shape[1])]
vif['VIF'] = round(vif['VIF'], 2)

```

```
vif = vif.sort_values(by = "VIF", ascending = False)
vif
```

	Features	VIF
9	What is your current occupation_unemployed	2.30
1	Total Time Spent on Website	2.06
0	TotalVisits	1.82
2	Lead Origin_lead add form	1.58
7	Last Activity_sms sent	1.55
3	Lead Source_olark chat	1.51
6	Last Activity_olark chat conversation	1.37
10	What is your current occupation_working profes...	1.32
4	Lead Source_welingak website	1.31
8	What is your current occupation_student	1.05
5	Last Activity_email bounced	1.03
11	Last Notable Activity_unreachable	1.01

Now all the vifs and p values are good

Prediction

Predicting the probabilities on the train set

```
y_train_pred = res.predict(X_train_sm)
y_train_pred[:10]
```

```
1289    0.649527
3604    0.135329
5584    0.164040
7679    0.135329
7563    0.387899
7978    0.758862
7780    0.155930
7863    0.982089
838     0.776544
708     0.146284
dtype: float64
```

Reshaping to an array

```
y_train_pred = y_train_pred.values.reshape(-1)
y_train_pred[:10]
```

```
array([0.64952699, 0.13532885, 0.16403992, 0.13532885, 0.38789903,
       0.75886225, 0.15593025, 0.98208925, 0.77654367, 0.14628394])
```

Data frame with given conversion rate and probability of predicted ones

```
y_train_pred_final = pd.DataFrame({'Converted':y_train.values,
'Conversion_Prob':y_train_pred})
y_train_pred_final.head()
```

	Converted	Conversion_Prob
0	1	0.649527
1	0	0.135329
2	0	0.164040
3	0	0.135329
4	0	0.387899

Substituting 0 or 1 with the cut off as 0.5

```
y_train_pred_final['Predicted'] =
y_train_pred_final.Conversion_Prob.map(lambda x: 1 if x > 0.5 else 0)
y_train_pred_final.head()
```

	Converted	Conversion_Prob	Predicted
0	1	0.649527	1
1	0	0.135329	0
2	0	0.164040	0
3	0	0.135329	0
4	0	0.387899	0

Evaluating the model

Creating confusion matrix

```
confusion = metrics.confusion_matrix(y_train_pred_final.Converted,
y_train_pred_final.Predicted )
confusion
```

```
array([[3442, 453],
       [ 752, 1704]], dtype=int64)
```

Check the overall accuracy

```
metrics.accuracy_score(y_train_pred_final.Converted,
y_train_pred_final.Predicted)
```

```
0.810266099826799
```

Substituting the value of true positive

```
TP = confusion[1,1]
```

Substituting the value of true negatives

```
TN = confusion[0,0]
```

Substituting the value of false positives

```
FP = confusion[0,1]
```

Substituting the value of false negatives

```
FN = confusion[1,0]
```

Calculating the sensitivity

```
TP/(TP+FN)
```

```
0.6938110749185668
```

```
# Calculating the specificity  
TN/(TN+FP)
```

```
0.8836970474967908
```

ROC curve

```
# ROC function
```

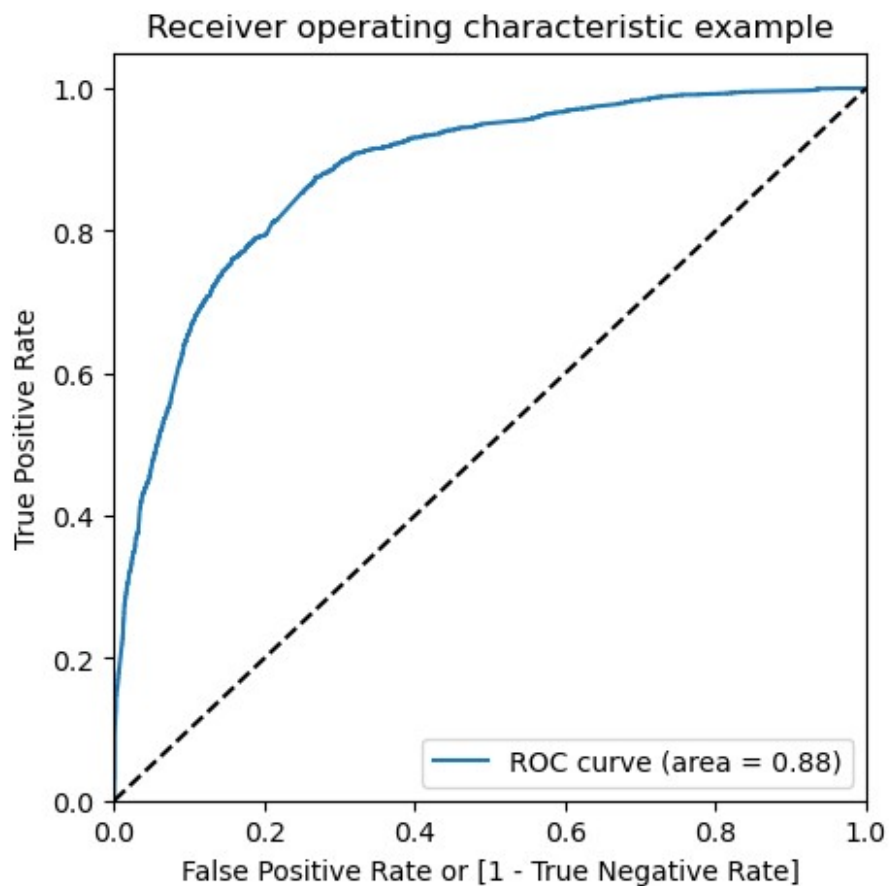
```
def draw_roc( actual, probs ):  
    fpr, tpr, thresholds = metrics.roc_curve( actual, probs,  
                                              drop_intermediate =  
False )  
    auc_score = metrics.roc_auc_score( actual, probs )  
    plt.figure(figsize=(5, 5))  
    plt.plot( fpr, tpr, label='ROC curve (area = %0.2f)' % auc_score )  
    plt.plot([0, 1], [0, 1], 'k--')  
    plt.xlim([0.0, 1.0])  
    plt.ylim([0.0, 1.05])  
    plt.xlabel('False Positive Rate or [1 - True Negative Rate]')  
    plt.ylabel('True Positive Rate')  
    plt.title('Receiver operating characteristic example')  
    plt.legend(loc="lower right")  
    plt.show()
```

```
return None
```

```
fpr, tpr, thresholds =  
metrics.roc_curve( y_train_pred_final.Converted,  
y_train_pred_final.Conversion_Prob)
```

```
# Call the ROC function
```

```
draw_roc(y_train_pred_final.Converted,  
y_train_pred_final.Conversion_Prob)
```



Creating columns with different probability cutoffs

```
numbers = [float(x)/10 for x in range(10)]
```

```
for i in numbers:
```

```
    y_train_pred_final[i]=
```

```
y_train_pred_final.Conversion_Prob.map(lambda x: 1 if x > i else 0)
```

```
y_train_pred_final.head()
```

	Converted	Conversion_Prob	Predicted	0.0	0.1	0.2	0.3	0.4	0.5
0.6 \									
0	1	0.649527	1	1	1	1	1	1	1
1									
1	0	0.135329	0	1	1	0	0	0	0
0									
2	0	0.164040	0	1	1	0	0	0	0
0									
3	0	0.135329	0	1	1	0	0	0	0
0									
4	0	0.387899	0	1	1	1	1	0	0
0									
	0.7	0.8	0.9						
0	0	0	0						
1	0	0	0						


```

2      0      0      0
3      0      0      0
4      0      0      0

```

Creating a dataframe to see the values of accuracy, sensitivity, and specificity at different values of probabiity cutoffs

```

cutoff_df = pd.DataFrame( columns =
['prob','accuracy','sensi','speci'])
# Making confusing matrix to find values of sensitivity, accurace and
specificity for each level of probability

```

```

from sklearn.metrics import confusion_matrix
num = [0.0,0.1,0.2,0.3,0.4,0.5,0.6,0.7,0.8,0.9]
for i in num:
    cm1 = metrics.confusion_matrix(y_train_pred_final.Converted,
y_train_pred_final[i] )
    total1=sum(sum(cm1))
    accuracy = (cm1[0,0]+cm1[1,1])/total1

    speci = cm1[0,0]/(cm1[0,0]+cm1[0,1])
    sensi = cm1[1,1]/(cm1[1,0]+cm1[1,1])
    cutoff_df.loc[i] =[ i ,accuracy,sensi,speci]
cutoff_df

```

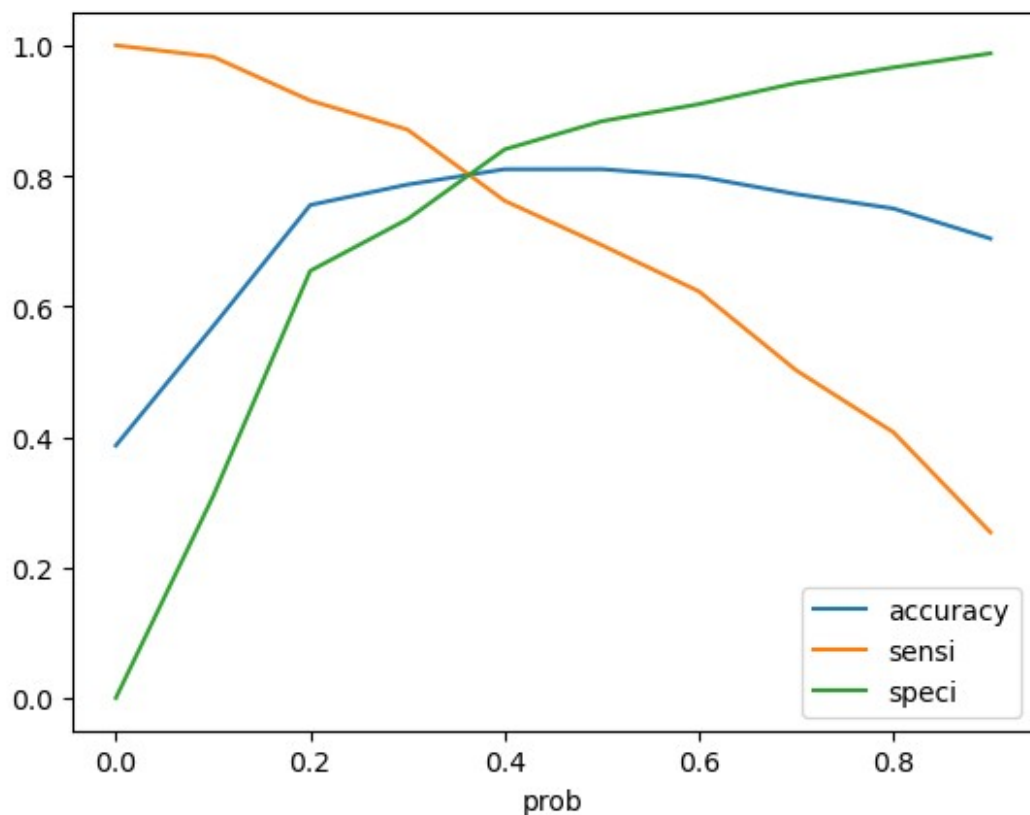
	prob	accuracy	sensi	speci
0.0	0.0	0.386711	1.000000	0.000000
0.1	0.1	0.569359	0.982492	0.308858
0.2	0.2	0.755314	0.915309	0.654429
0.3	0.3	0.786648	0.870928	0.733504
0.4	0.4	0.810109	0.761808	0.840565
0.5	0.5	0.810266	0.693811	0.883697
0.6	0.6	0.798929	0.622964	0.909884
0.7	0.7	0.772004	0.502036	0.942234
0.8	0.8	0.749961	0.407166	0.966110
0.9	0.9	0.703826	0.253664	0.987677

Plotting it

```

cutoff_df.plot.line(x='prob', y=['accuracy','sensi','speci'])
plt.show()

```



```

y_train_pred_final['final_predicted'] =
y_train_pred_final.Conversion_Prob.map( lambda x: 1 if x > 0.35 else
0)
y_train_pred_final.head()

```

	Converted	Conversion_Prob	Predicted	0.0	0.1	0.2	0.3	0.4	0.5
0.6 \									
0	1	0.649527	1	1	1	1	1	1	1
1	0	0.135329	0	1	1	0	0	0	0
2	0	0.164040	0	1	1	0	0	0	0
3	0	0.135329	0	1	1	0	0	0	0
4	0	0.387899	0	1	1	1	1	0	0

	0.7	0.8	0.9	final_predicted
0	0	0	0	1
1	0	0	0	0
2	0	0	0	0
3	0	0	0	0
4	0	0	0	1

```

# Check the overall accuracy
metrics.accuracy_score(y_train_pred_final.Converted,
y_train_pred_final.final_predicted)

0.8001889466225791

# Creating confusion matrix
confusion2 = metrics.confusion_matrix(y_train_pred_final.Converted,
y_train_pred_final.final_predicted )
confusion2

array([[3135, 760],
       [ 509, 1947]], dtype=int64)

# Substituting the value of true positive
TP = confusion2[1,1]
# Substituting the value of true negatives
TN = confusion2[0,0]
# Substituting the value of false positives
FP = confusion2[0,1]
# Substituting the value of false negatives
FN = confusion2[1,0]

# Calculating the sensitivity
TP/(TP+FN)

0.7927524429967426

# Calculating the specificity
TN/(TN+FP)

0.8048780487804879

```

Prediction on Test set

```

# Scaling numeric values
X_test[['TotalVisits', 'Page Views Per Visit', 'Total Time Spent on
Website']] = scaler.transform(X_test[['TotalVisits', 'Page Views Per
Visit', 'Total Time Spent on Website']])

# Substituting all the columns in the final train model
col = X_train.columns

# Select the columns in X_train for X_test as well
X_test = X_test[col]
# Add a constant to X_test
X_test_sm = sm.add_constant(X_test[col])
X_test_sm
X_test_sm

      const  TotalVisits  Total Time Spent on Website \
8308      1.0         0.035461                    0.416813

```

7212	1.0	0.028369	0.001320
2085	1.0	0.000000	0.000000
4048	1.0	0.028369	0.617077
4790	1.0	0.028369	0.005282
...
3261	1.0	0.000000	0.000000
8179	1.0	0.170213	0.148768
6236	1.0	0.000000	0.000000
5240	1.0	0.078014	0.458627
7243	1.0	0.035461	0.499560

	Lead Origin_lead add form	Lead Source_olark chat \
8308	0	0
7212	0	0
2085	1	0
4048	0	0
4790	0	0
...
3261	0	1
8179	0	0
6236	0	1
5240	0	0
7243	0	0

	Lead Source_welingak website	Last Activity_email bounced \
8308	0	0
7212	0	0
2085	1	0
4048	0	0
4790	0	0
...
3261	0	0
8179	0	0
6236	0	0
5240	0	0
7243	0	0

	Last Activity_olark chat conversation	Last Activity_sms sent \
8308	0	0
7212	0	1
2085	0	0
4048	0	1
4790	0	0
...
3261	1	0
8179	0	1
6236	0	0
5240	0	1
7243	0	0

	What is your current occupation_student \
8308	0
7212	0
2085	0
4048	0
4790	0
...	...
3261	0
8179	0
6236	0
5240	0
7243	0

	What is your current occupation_unemployed \
8308	1
7212	0
2085	1
4048	1
4790	1
...	...
3261	1
8179	0
6236	0
5240	1
7243	1

	What is your current occupation_working professional \
8308	0
7212	1
2085	0
4048	0
4790	0
...	...
3261	0
8179	0
6236	0
5240	0
7243	0

	Last Notable Activity_unreachable
8308	0
7212	0
2085	0
4048	0
4790	0
...	...
3261	0
8179	0
6236	0
5240	0

7243

0

[2723 rows x 13 columns]

```

# Storing prediction of test set in the variable 'y_test_pred'
y_test_pred = res.predict(X_test_sm)
# Converting it to df
y_pred_df = pd.DataFrame(y_test_pred)
# Converting y_test to dataframe
y_test_df = pd.DataFrame(y_test)
# Remove index for both dataframes to append them side by side
y_pred_df.reset_index(drop=True, inplace=True)
y_test_df.reset_index(drop=True, inplace=True)
# Append y_test_df and y_pred_df
y_pred_final = pd.concat([y_test_df, y_pred_df], axis=1)
# Renaming column
y_pred_final = y_pred_final.rename(columns = {0 : 'Conversion_Prob'})
y_pred_final.head()

```

	Converted	Conversion_Prob
0	0	0.451705
1	1	0.829251
2	1	0.982089
3	1	0.869411
4	0	0.105066

```

# Making prediction using cut off 0.35
y_pred_final['final_predicted'] =
y_pred_final.Conversion_Prob.map(lambda x: 1 if x > 0.35 else 0)
y_pred_final

```

	Converted	Conversion_Prob	final_predicted
0	0	0.451705	1
1	1	0.829251	1
2	1	0.982089	1
3	1	0.869411	1
4	0	0.105066	0
...
2718	1	0.106317	0
2719	0	0.320571	0
2720	0	0.135329	0
2721	1	0.801105	1
2722	1	0.547662	1

[2723 rows x 3 columns]

```

# Check the overall accuracy
metrics.accuracy_score(y_pred_final['Converted'],
y_pred_final.final_predicted)

```

0.8013220712449505

```

# Creating confusion matrix
confusion2 = metrics.confusion_matrix(y_pred_final['Converted'],
y_pred_final.final_predicted )
confusion2

array([[1392,  352],
       [ 189,  790]], dtype=int64)

# Substituting the value of true positive
TP = confusion2[1,1]
# Substituting the value of true negatives
TN = confusion2[0,0]
# Substituting the value of false positives
FP = confusion2[0,1]
# Substituting the value of false negatives
FN = confusion2[1,0]

# Calculating the sensitivity
TP/(TP+FN)

0.8069458631256384

# Calculating the specificity
TN/(TN+FP)

0.7981651376146789

```

Presion and Recall

```

confusion = metrics.confusion_matrix(y_train_pred_final.Converted,
y_train_pred_final.Predicted )
confusion

array([[3442,  453],
       [ 752, 1704]], dtype=int64)

# Precision = TP / TP + FP
confusion[1,1]/(confusion[0,1]+confusion[1,1])

0.7899860917941586

#Recall = TP / TP + FN
confusion[1,1]/(confusion[1,0]+confusion[1,1])

0.6938110749185668

y_train_pred_final.Converted, y_train_pred_final.Predicted
(0      1
 1      0
 2      0
 3      0
 4      0

```

```

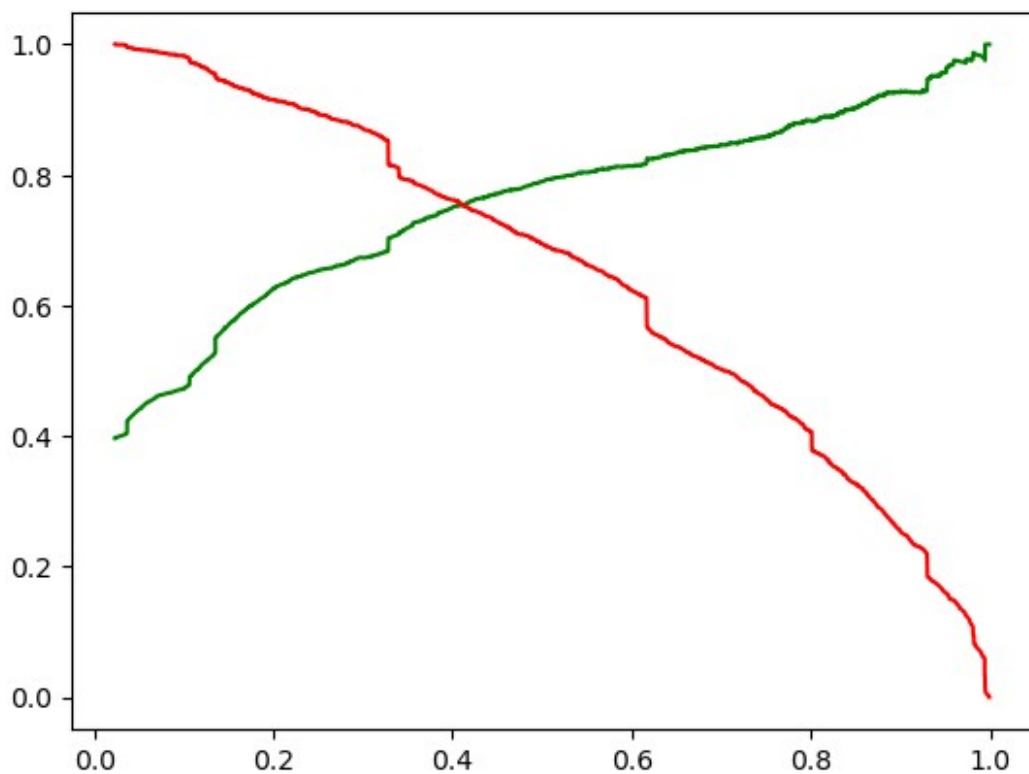
6346    ..
6347    0
6348    0
6349    0
6350    1
Name: Converted, Length: 6351, dtype: int64,
0      1
1      0
2      0
3      0
4      0

6346    ..
6347    0
6348    0
6349    0
6350    0
Name: Predicted, Length: 6351, dtype: int64)

p, r, thresholds =
precision_recall_curve(y_train_pred_final.Converted,
y_train_pred_final.Conversion_Prob)

plt.plot(thresholds, p[:-1], "g-")
plt.plot(thresholds, r[:-1], "r-")
plt.show()

```

```
y_train_pred_final['final_predicted'] =
y_train_pred_final.Conversion_Prob.map(lambda x: 1 if x > 0.41 else 0)
y_train_pred_final.head()
```

	Converted	Conversion_Prob	Predicted	0.0	0.1	0.2	0.3	0.4	0.5
0.6 \									
0	1	0.649527	1	1	1	1	1	1	1
1									
1	0	0.135329	0	1	1	0	0	0	0
0									
2	0	0.164040	0	1	1	0	0	0	0
0									
3	0	0.135329	0	1	1	0	0	0	0
0									
4	0	0.387899	0	1	1	1	1	0	0
0									

	0.7	0.8	0.9	final_predicted
0	0	0	0	1
1	0	0	0	0
2	0	0	0	0
3	0	0	0	0
4	0	0	0	0

```

# Accuracy
metrics.accuracy_score(y_train_pred_final.Converted,
y_train_pred_final.final_predicted)

0.809951188789167

# Creating confusion matrix again
confusion2 = metrics.confusion_matrix(y_train_pred_final.Converted,
y_train_pred_final.final_predicted )
confusion2

array([[3289, 606],
       [ 601, 1855]], dtype=int64)

# Substituting the value of true positive
TP = confusion2[1,1]
# Substituting the value of true negatives
TN = confusion2[0,0]
# Substituting the value of false positives
FP = confusion2[0,1]
# Substituting the value of false negatives
FN = confusion2[1,0]

# Precision = TP / TP + FP
TP / (TP + FP)

0.753758634701341

#Recall = TP / TP + FN
TP / (TP + FN)

0.7552931596091205

```

Prediction on Test set

```

# Storing prediction of test set in the variable 'y_test_pred'
y_test_pred = res.predict(X_test_sm)
# Converting it to df
y_pred_df = pd.DataFrame(y_test_pred)
# Converting y_test to dataframe
y_test_df = pd.DataFrame(y_test)
# Remove index for both dataframes to append them side by side
y_pred_df.reset_index(drop=True, inplace=True)
y_test_df.reset_index(drop=True, inplace=True)
# Append y_test_df and y_pred_df
y_pred_final = pd.concat([y_test_df, y_pred_df],axis=1)
# Renaming column
y_pred_final= y_pred_final.rename(columns = {0 : 'Conversion_Prob'})
y_pred_final.head()

   Converted  Conversion_Prob
0           0           0.451705

```

1	1	0.829251
2	1	0.982089
3	1	0.869411
4	0	0.105066

Making prediction using cut off 0.41

```
y_pred_final['final_predicted'] =
y_pred_final.Conversion_Prob.map(lambda x: 1 if x > 0.41 else 0)
y_pred_final
```

	Converted	Conversion_Prob	final_predicted
0	0	0.451705	1
1	1	0.829251	1
2	1	0.982089	1
3	1	0.869411	1
4	0	0.105066	0
...
2718	1	0.106317	0
2719	0	0.320571	0
2720	0	0.135329	0
2721	1	0.801105	1
2722	1	0.547662	1

[2723 rows x 3 columns]

Check the overall accuracy

```
metrics.accuracy_score(y_pred_final['Converted'],
y_pred_final.final_predicted)
```

0.8138082996694822

Creating confusion matrix

```
confusion2 = metrics.confusion_matrix(y_pred_final['Converted'],
y_pred_final.final_predicted )
confusion2
```

```
array([[1470, 274],
       [ 233, 746]], dtype=int64)
```

Substituting the value of true positive

```
TP = confusion2[1,1]
```

Substituting the value of true negatives

```
TN = confusion2[0,0]
```

Substituting the value of false positives

```
FP = confusion2[0,1]
```

Substituting the value of false negatives

```
FN = confusion2[1,0]
```

Precision = TP / TP + FP

```
TP / (TP + FP)
```

0.7313725490196078

$\#Recall = TP / TP + FN$
 $TP / (TP + FN)$

0.7620020429009193

Conclusion

The variable that matter most are **as** follows **in** decending order

- The total time spend on website
- The total number of visits
- When the source of lead was
 - Google, Direct traffic, Organic search.
- When the last activity was
 - SMS, Olarck chat conversion
- When the Lead origin **is** Lead add **format**
- When their current occupation **is** waorking **as** professionals