Pre-processing

5 %matplotlib inline

Importing the necessary libraries and the dataset

In [1]:

```
1 # from google.colab import drive
2 # drive.mount('/content/drive', force_remount=True)
```

In [1]:

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

In [2]:

```
1  df = pd.read_csv('combined.csv')
2  df.head()
```

Out[2]:

	Date	Profile URL	First Name	Last Name	Full Name	
0	29/03/2021, 19:22:46	https://www.linkedin.com/in/karteek-pallerla	Karteek	Pallerla (KP)	Karteek Pallerla (KP)	
1	29/03/2021, 19:22:47	https://www.linkedin.com/in/ravitejadupuguntla	Ravi Teja	Dupuguntla	Ravi Teja Dupuguntla	1
2	29/03/2021, 19:22:48	https://www.linkedin.com/in/aroonmathai	Aroon	Mathai	Aroon Mathai	
3	29/03/2021, 19:22:48	https://www.linkedin.com/in/shubhanjan- chakrab	Shubhanjan	Chakrabarty	Shubhanjan Chakrabarty	Е
4	29/03/2021, 19:22:49	https://www.linkedin.com/in/varsha-agarwal- 8a7	Varsha	Agarwal	Varsha Agarwal	

5 rows × 28 columns

In [3]:

```
1 df.info()
```

<class 'pandas.core.frame.DataFrame'> RangeIndex: 4910 entries, 0 to 4909 Data columns (total 28 columns):

#	Column	Non-Null Count	Dtype
	D-4-	4010 11	
0	Date	4910 non-null	object
1	Profile URL	4910 non-null	object
2	First Name	4834 non-null	object
3	Last Name	4833 non-null	object
4	Full Name	4834 non-null	object
5	Location	3476 non-null	object
6	Job Title	4898 non-null	object
7	Company Name	4892 non-null	object
8	Industry	4900 non-null	object
9	Company Location	4910 non-null	object
10	Social Handle	326 non-null	object
11	Social Network	326 non-null	object
12	Websites	676 non-null	object
13	New Job (90 Days)	4760 non-null	object
14	Current Position	4898 non-null	object
15	Job Description	1259 non-null	object
16	Month Started	4760 non-null	float64
17	Year Started	4816 non-null	float64
18	Profile Headline	4892 non-null	object
19	Profile Summary	3465 non-null	object
20	School	4658 non-null	object
21	Degree	4716 non-null	object
22	Education Start	4680 non-null	float64
23	Education End	4672 non-null	float64
24	Field of Study	4532 non-null	object
25	_	4910 non-null	int64
26	Degree of Connection	4910 non-null	int64
27	•	4910 non-null	object
dtyp	es: float64(4), int64(3

memory usage: 1.0+ MB

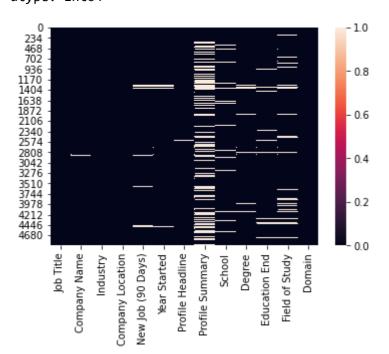
In [4]:

```
1 df1 = df.drop(['Date','Location','Current Position','Job Description','Education Start
  # df1.head()
2
```

In [5]:

1	<pre>sns.heatmap(df1.isnull())</pre>	
2	<pre>print(df1.isnull().sum())</pre>	

Job Title	12
Company Name	18
Industry	10
Company Location	0
New Job (90 Days)	150
Year Started	94
Profile Headline	18
Profile Summary	1445
School	252
Degree	194
Education End	238
Field of Study	378
Domain	0
dtype: int64	



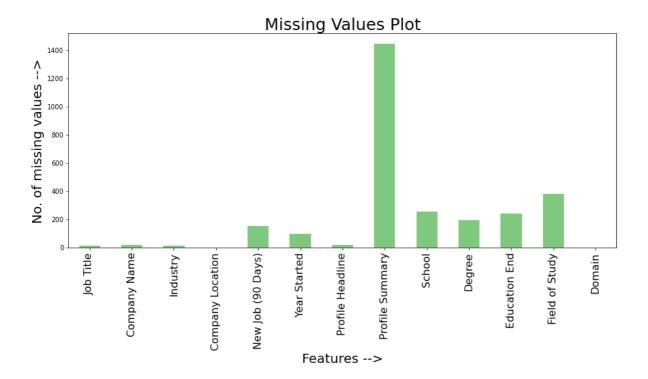
Checking the number of missing values in each coloum before Pre-processing

In [6]:

```
print('Percentage of missing values :')
print(df1.isnull().sum()*100/(df1.notnull().sum()+df1.isnull().sum()))

plt.figure(figsize=(15,6))
fd1.isnull().sum().plot(kind='bar', colormap='Accent')
plt.title('Missing Values Plot', fontsize = 25)
plt.xlabel('Features -->', fontsize = 20)
plt.ylabel('No. of missing values -->', fontsize = 20)
plt.xticks(fontsize=16)
plt.show()
```

```
Percentage of missing values :
Job Title
                       0.244399
Company Name
                       0.366599
Industry
                       0.203666
Company Location
                       0.000000
New Job (90 Days)
                       3.054990
Year Started
                       1.914460
Profile Headline
                       0.366599
Profile Summary
                      29.429735
                       5.132383
School
Degree
                       3.951120
Education End
                       4.847251
Field of Study
                       7.698574
                       0.000000
Domain
dtype: float64
```



Plotting a TreeMap to understand the hierarchy of jobs in the companies

```
In [7]:
```

```
import plotly.express as px
import numpy as np
import plotly as plt
import ipywidgets as widgets

plt.offline.init_notebook_mode(connected=True)
```

In [8]:

```
df1['CompanyName']=df1['Company Name']
df1['JobTitle']=df1['Job Title']
df1.JobTitle = df1['JobTitle'].fillna('not_given')
df1.CompanyName = df1['CompanyName'].fillna('not_given')
df1.Industry = df1.Industry.fillna('not_given')
df1['My Network']='network'
```

In [9]:

```
1 #!pip install --upgrade plotly
```

In [10]:

```
1 # fig1 = px.treemap(df1, path=['My Network', 'Domain', 'Industry'], width=1000, height=
2 # fig1.show()
3 # # renderer = "colab"
```

In [11]:

```
1 # fig2 = px.treemap(df1, path=['My Network', 'Domain', 'JobTitle'], width=1000, height=
2 # fig2.show()
```

In [12]:

```
1 # fig2 = px.treemap(df1, path=['My Network', 'Domain', 'JobTitle'], width=1000, height=
2 # fig2.show()
3
4 # fig3 = px.treemap(df1, path=['My Network', 'Domain', 'CompanyName'], width=1000, height=
5 # fig3.show()
```

Checking the number of unique values in each coloum before Pre-processing

In [13]:

```
import matplotlib.pyplot as plt

print(df1.nunique())

plt.figure(figsize=(20,4))

plt.plot(df1.nunique())

plt.grid()

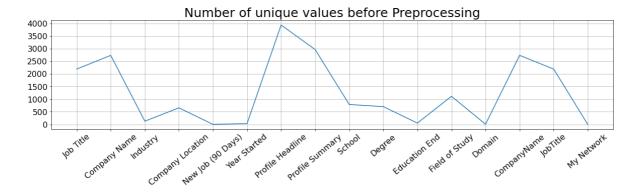
plt.title('Number of unique values before Preprocessing', fontsize = 25)

plt.xticks(fontsize=16, rotation=40)

plt.yticks(fontsize=16)

plt.show()
```

Job Title	2194
Company Name	2738
Industry	128
Company Location	657
New Job (90 Days)	2
Year Started	32
Profile Headline	3941
Profile Summary	2968
School	789
Degree	706
Education End	53
Field of Study	1115
Domain	7
CompanyName	2739
JobTitle	2195
My Network	1
dtype: int64	



Preprocessing the Company Locations column

Two steps:

- 1. Adding a country column using the location column
- 2. Adding a continent column using the country column created above

STEP - 1

```
In [14]:
```

```
1 df1['Country'] = df1['Company Location']
```

In [15]:

```
df1['Country'].replace({'Bengaluru, Karnataka' : 'Bengaluru, India',
 2
                  'Greater Bengaluru Area' : 'Greater Bengaluru Area, India',
 3
                 'Canada | Netherlands' : 'Canada',
 4
                 'Mumbai, Maharashtra' : 'Mumbai, Maharashtra, India',
                 'Pune': 'Pune, India',
 5
 6
                'Woburn, MA' : 'Woburn, MA, United States',
                'New York City Metropolitan Area': 'New York, United States',
 7
                'Greater Paris Metropolitan Region' : 'Greater Paris Metropolitan Region,
 8
 9
                'Vellore' : 'Vellore, India',
                'Near Kakinada, A.P.': 'Near Kakinada, A.P, India',
10
                'Greater Leicester Area': 'Greater Leicester Area, England',
11
12
                'VIT Vellore' : 'VIT Vellore, India',
               'Greater San Diego Area' : 'Greater San Diego Area, India',
13
                'Jaipur' : 'Jaipur, India',
14
                'CHENNAI' : 'CHENNAI, India',
15
16
                'Houston, Texas Area' : 'Houston, Texas Area, United States',
                'Raleigh-Durham, North Carolina Area' : 'Raleigh-Durham, North Carolina Ar
17
                'Vellore, Tamil Nadu' : 'Vellore, Tamil Nadu, India',
18
                'Bangalore - India' : 'Bangalore, India',
19
                'pune' : 'pune, India',
20
21
                'Ottawa, Canada Area' : 'Ottawa, Canada',
                'Bangalore' : 'Bangalore, India',
22
                'Chennai, Tamil Nadu' : 'Chennai, Tamil Nadu, India',
23
                'Russelsheim' : 'Russelsheim, Germany',
24
25
                'Seattle, Washington' : 'Seattle, Washington, United States',
               'Tamil Nadu, India and Doha, Qatar' : 'Doha, Qatar',
26
                'Kochi, India' : 'Kochi, India',
27
               'Ghandhinagar' : 'Ghandhinagar, India',
28
                'Raipur' : 'Raipur, India',
29
                'Tamilnadu' : 'Tamilnadu, India',
30
31
                'VIT Vellore ' : 'VIT Vellore , India',
                'Hyderabad ': 'Hyderabad , India',
32
                'Hosur': 'Hosur, India',
33
                'Greater New York City Area' : 'Greater New York City Area, United States'
34
35
                'New Delhi, Delhi' : 'New Delhi, Delhi, India',
36
                'Chennai' : 'Chennai, India',
                'Hyderabad' : 'Hyderabad, India',
37
38
                'Greater Noida' : 'Greater Noida, India',
                'Richardson, Texas': 'Richardson, Texas, United States',
39
                'Chittoor, Andhra Pradesh' : 'Chittoor, Andhra Pradesh, India',
40
                'MN, Minnesota' : 'MN, Minnesota, United States',
41
42
                'Pantnagar, Uttarakhand' : 'Pantnagar, Uttarakhand, India',
43
                'Mumbai' : 'Mumbai, India',
                'Lyndhurst new jersey': 'Lyndhurst new jersey, United States',
44
               'Kalpakkam, Chennai' : 'Kalpakkam, Chennai, India',
45
               'Surajpur noida' : 'Surajpur noida, India',
46
47
                'Wixom, MI' : 'Wixom, MI, United States',
                'Jacksonville, Florida Area' : 'Jacksonville, Florida Area, United States'
48
                'San Francisco Bay Area' : 'San Francisco Bay Area, United States',
49
50
                'Greater Seattle Area' : 'Greater Seattle Area, United States',
51
                'Greater Sydney Area' : 'Greater Sydney Area, United States',
                'Greater Houston': 'Greater Houston, United States',
52
53
                'Greater Chicago Area' : 'Greater Chicago Area, United States',
                'Dallas-Fort Worth Metroplex' : 'Dallas-Fort Worth Metroplex, United State
54
                'Charlotte Metro' : 'Charlotte Metro, United States',
55
56
                'Greater Cambridge Area' : 'Greater Cambridge Area, England',
57
                'Other': 'India',
58
                'Greater Delhi Area' : 'Greater Delhi Area, India',
                'Greater Chennai Area' : 'Greater Chennai Area, India',
59
```

```
'Greater Hyderabad Area' : 'Greater Hyderabad Area, India',
 60
 61
                 'San Francisco Bay Area' : 'San Francisco Bay Area, United States',
                 'Greater Reading Area' : 'Greater Reading Area, United Kingdom',
 62
                 'Boise Metropolitan Area' : 'Boise Metropolitan Area, United States',
 63
                 'Berlin Metropolitan Area' : 'Berlin Metropolitan Area, Germany',
 64
                 'Brabantine City Row' : 'Brabantine City Row, Netherlands',
 65
                 'Pune/Pimpri-Chinchwad Area' : 'Pune/Pimpri-Chinchwad Area, India',
 66
                 'Greater Coventry Area' : 'Greater Coventry Area, England',
 67
                 'Greater Sacramento' : 'Greater Sacramento, United States',
 68
                 'Hong Kong SAR' : 'Hong Kong',
 69
                 'Geneva Metropolitan Area' : 'Geneva Metropolitan Area, Switzerland',
 70
                 'Greater Boston' : 'Greater Boston, United States',
 71
 72
                 'Greater Indore Area' : 'Greater Indore Area, India',
                 'Detroit Metropolitan Area' : 'Detroit Metropolitan Area, United States',
 73
 74
                 'Greater Montreal Metropolitan Area' : 'Greater Montreal Metropolitan Area
 75
                 'Greater Tuscaloosa Area' : 'Greater Tuscaloosa Area, United States',
                 'Greater Melbourne Area' : 'Greater Melbourne Area, Australia',
 76
                 'Gothenburg Metropolitan Area' : 'Gothenburg Metropolitan Area, Sweden',
 77
                 'Greater Brisbane Area' : 'Greater Brisbane Area, Australia',
 78
 79
                 'Greater Dublin' : 'Greater Dublin, Ireland',
                 'Greater Allahabad Area' : 'Greater Allahabad Area, India',
 80
 81
                 'Greater Perth Area' : 'Greater Perth Area, Australia',
                 'Greater Hamburg Area' : 'Greater Hamburg Area, Germany',
 82
                 'Mumbai Metropolitan Region' : 'Mumbai Metropolitan Region, India',
 83
                 'Los Angeles Metropolitan Area' : 'Los Angeles Metropolitan Area, India',
 84
                 'Greater Vancouver Metropolitan Area' : 'Greater Vancouver Metropolitan Ar
 85
 86
                 'Greater Adelaide Area' : 'Greater Adelaide Area, Australia',
                 'Greater Kassel Area' : 'Greater Kassel Area, Germany',
 87
                 'Greater Barcelona Metropolitan Area' : 'Greater Barcelona Metropolitan Ar
 88
                 'Da Nang Metropolitan Area' : 'Da Nang Metropolitan Area, Central Vietnam'
 89
                 'Greater Lille Metropolitan Area' : 'Greater Lille Metropolitan Area, Fran
 90
                 'Texas Metropolitan Area' : ' Texas Metropolitan Area, United States',
 91
                 'Greater Toulouse Metropolitan Area' : 'Greater Toulouse Metropolitan Area
 92
                 'Johannesburg Metropolitan Area' : 'Johannesburg Metropolitan Area, South
 93
                 'Cincinnati Metropolitan Area' : 'Cincinnati Metropolitan Area, United Sta
 94
                 ' South Carolina Area' : 'South Carolina Area, United states',
 95
 96
                 'Stockholm Metropolitan Area' : 'Stockholm Metropolitan Area, Sweden',
                 'Oregon Metropolitan Area' : ' Oregon Metropolitan Area, United States',
 97
                 'Greater Kolkata Area' : 'Greater Kolkata Area, India',
 98
                 'Greater Syracuse-Auburn Area' : 'Greater Syracuse-Auburn Area, United Sta
 99
                 'Greater Madrid Metropolitan Area' : 'Greater Madrid Metropolitan Area, Sp
100
                 'Greater Newcastle Area' : 'Greater Newcastle Area, England',
101
                 'Cork Metropolitan Area' : 'Cork Metropolitan Area, Ireland',
102
103
                 'Helsinki Metropolitan Area' : 'Helsinki Metropolitan Area, Finland',
                 'Atlanta Metropolitan Area' : 'Atlanta Metropolitan Area, United States',
104
                 'Ghent Metropolitan Area' : 'Ghent Metropolitan Area, Belgium',
105
                 'Washington DC-Baltimore Area' : 'Washington DC-Baltimore Area, United Sta
106
                 'Greater Tampa Bay Area' : 'Greater Tampa Bay Area, United States',
107
                 'New York Metropolitan Area' : ' New York Metropolitan Area, United States
108
                 'Greater Orlando' : 'Greater Orlando, Florida',
109
110
                 'Greater Minneapolis-St. Paul Area' : 'Greater Minneapolis-St. Paul Area,
                'Greater Munich Metropolitan Area' : 'Greater Munich Metropolitan Area, Ger
111
                 'Stuttgart Region' : 'Stuttgart Region, Germany',
112
                 'Brussels Metropolitan Area' : 'Brussels Metropolitan Area, Belgium',
113
                'Greater St. Louis' : 'Greater St. Louis, United States',
114
                 'Greater Hartford' : 'Greater Hartford, United States',
115
                 'Greater Edmonton Metropolitan Area' : 'Greater Edmonton Metropolitan Area
116
                 'Frankfurt Rhine-Main Metropolitan Area' : 'Frankfurt Rhine-Main Metropoli
117
118
                'Greater Indianapolis' : 'Greater Indianapolis, United States',
                 'Denver Metropolitan Area' : 'Denver Metropolitan Area, United States',
119
                 'Greater Bordeaux Metropolitan Area' : 'Greater Bordeaux Metropolitan Area
120
```

'Rochester, New York Metropolitan Area' : 'Rochester, New York Metropolitan 127 'Greenville-Spartanburg-Anderson, South Carolina Area' : 'Greenville-Spartan 128 'Portland, Oregon Metropolitan Area': 'Portland, Oregon Metropolitan Area, 129 'San Antonio, Texas Metropolitan Area' : 'San Antonio, Texas Metropolitan Ar 130 131

'kanchipuram': 'kanchipuram, India'}, inplace = True)

In [16]:

```
countries = []
 1
 2
   for index, loc in enumerate(df1['Country']):
3
        if type(loc) != float:
4
            country = loc.strip().split(',')[-1]
 5
            country = country.rstrip().lstrip()
            countries.append(country)
 6
7
        else:
            countries.append("not-given")
8
9
  df1['Country'] = countries
10
```

In [17]:

```
1 df1['Country'].unique()
```

Out[17]:

```
array(['United States', 'India', 'Denmark', 'England', 'France',
        'United Arab Emirates', 'United Kingdom', 'Switzerland', 'Germany',
        'Australia', 'Nepal', 'Netherlands', 'Canada', 'Bangladesh',
        'Taiwan', 'Kuwait', 'Singapore', 'Ireland', 'Sweden', 'China',
        'Saudi Arabia', 'Brazil', 'Fiji', 'Hong Kong', 'Uganda',
        'Indonesia', 'Qatar', 'Finland', 'Oman', 'Nigeria', 'New Zealand',
        'Bahrain', 'Belgium', 'Italy', 'Spain', 'Central Vietnam', 'Japan',
        'Malaysia', 'Kenya', 'Thailand', 'South Africa', 'Florida', 'Norway', 'Luxembourg', 'Mexico', 'Unknown',
        'Republic of the Congo'], dtype=object)
```

In [18]:

```
1 print("Number of unique countries of country locations: ", df1['Country'].nunique())
```

Number of unique countries of country locations: 47

```
In [19]:
```

```
1 df1['Country'].value_counts(dropna = False)
```

Out[19]:

India 3913	3
United States 420	
Canada 77	
United Arab Emirates 75	
Australia 75	
Germany 46	
United Kingdom 46	5
France 26	5
Netherlands 25	5
China 23	3
Singapore 22	
Sweden 17	
Qatar 15	
Saudi Arabia	
New Zealand 16	
Ireland 16	
•	3
<u> </u>	7
	5
Denmark	5
Malaysia	5
Finland	5
Kuwait	5
Spain	5
England	1
	1
	1
T+oly	י ר
Italy	,
Japan	3
Hong Kong	3
Norway	3 3 3 3 3
Central Vietnam	3
3	
Republic of the Congo 2	2
South Africa	2
Luxembourg	2 2 2 2
Bangladesh 2)
Thailand)
	- L
	L
	L L
,	
	1
0	L
	L
	L
Name: Country, dtype: int64	

STEP - 2

In [20]:

```
df1['Continent'] = df1['Country']
```

```
In [21]:
```

```
1 # !pip install pycountry_convert
```

In [22]:

```
import pycountry_convert as pc
import pycountry
import pycountry_convert as pc
import pycountry
import pycountry_convert as pc
import pycountry
import p
```

In [23]:

```
1 countries = {}
   for country in pycountry.countries:
        countries[country.name] = country.alpha_2
 3
4 | continents = []
   codes = [countries.get(country, 'Unknown code') for country in input_countries]
 6
   for code in codes:
7
        if code != 'Unknown code':
            continents.append(pc.country_alpha2_to_continent_code(code))
8
9
       else:
            continents.append('unknown')
10
11
   # print(continents)
12
   df1['Continent'] = continents
```

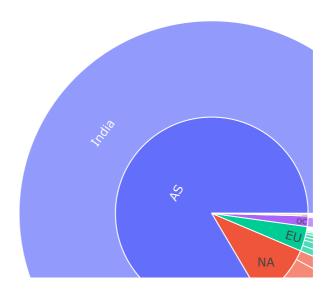
In [24]:

```
1 df1['Continent'].unique()
Out[24]:
array(['NA', 'AS', 'EU', 'unknown', 'OC', 'SA', 'AF'], dtype=object)
```

Plotting sunbursts to visualise the distribution of companies distributed over continents

In [25]:

```
fig4 = px.sunburst(df1, path=['Continent', 'Country'])
fig4.show()
```



Pre-processing the Job Title coloum

```
In [26]:
```

```
1 df1['Job Title'].value_counts(dropna=False)
```

Out[26]:

Project Manager	348
Product Manager	217
Analyst	213
Program Manager	113
Account Manager	66
Trainee decision scientist	1
entertainers and event management	1
Internet of Things Intern	1
Operations Executive	1
Associate Video Producer	1
Name: Job Title, Length: 2195, dtype	: int64

```
In [27]:
 1 df['Job Title'].describe()
Out[27]:
                     4898
count
                     2194
unique
top
          Project Manager
                      348
freq
Name: Job Title, dtype: object
In [28]:
 1 | df['Job Title'].value_counts()
Out[28]:
Project Manager
                                      348
Product Manager
                                      217
Analyst
                                      213
Program Manager
                                      113
Account Manager
                                       66
Trainee decision scientist
                                        1
entertainers and event management
                                        1
Internet of Things Intern
                                        1
Operations Executive
                                        1
Junior Analyst 2
                                        1
Name: Job Title, Length: 2194, dtype: int64
In [29]:
 1 df1['Job Title'].unique()
Out[29]:
array(['Battery Designer', 'Digital DevOps Engineer', 'Product Designer',
       ..., 'Product Manager Haematology', 'Principal Product Manager',
       'Product Technical Manager at Infosys'], dtype=object)
In [30]:
    df1['JobTitle']=df1['Job Title']
    df1['JobTitle'] = df1['JobTitle'].str.replace('[^\w\s]','')
    df1['JobTitle']=df1['JobTitle'].str.lower()
    df1['JobTitle']=df1['JobTitle'].str.replace(' ','')
 5
 6
    df1.JobTitle = df1.JobTitle.fillna('not_given')
```

In [31]:

```
df1.loc[df1.JobTitle.str.contains('manage'), 'JobTitle'] = 'Manager'
   df1.loc[df1.JobTitle.str.contains('analys'), 'JobTitle'] = 'Analyst'
   df1.loc[df1.JobTitle.str.contains('developer'), 'JobTitle'] = 'Developer'
   df1.loc[df1.JobTitle.str.contains('web'), 'JobTitle'] = 'Developer'
 5
    df1.loc[df1.JobTitle.str.contains('designer'), 'JobTitle'] = 'Designer'
    df1.loc[df1.JobTitle.str.contains('dev'), 'JobTitle'] = 'developer'
    df1.loc[df1.JobTitle.str.contains('founder'), 'JobTitle'] = 'Founder'
 7
   df1.loc[df1.JobTitle.str.contains('owner'), 'JobTitle'] = 'Founder'
df1.loc[df1.JobTitle.str.contains('intern'), 'JobTitle'] = 'Intern'
 9
    df1.loc[df1.JobTitle.str.contains('freelance'), 'JobTitle'] = 'freelancer'
10
   df1.loc[df1.JobTitle.str.contains('associate'), 'JobTitle'] = 'Associate'
11
   df1.loc[df1.JobTitle.str.contains('president'), 'JobTitle'] = 'Board Member'
12
   df1.loc[df1.JobTitle.str.contains('vice'), 'JobTitle'] = 'Board Member'
13
   df1.loc[df1.JobTitle.str.contains('ceo'), 'JobTitle'] = 'Board Member'
   df1.loc[df1.JobTitle.str.contains('director'), 'JobTitle'] = 'Board Member'
df1.loc[df1.JobTitle.str.contains('board'), 'JobTitle'] = 'Board Member'
15
16
    df1.loc[df1.JobTitle.str.contains('engineer'), 'JobTitle'] = 'Engineer'
17
   df1.loc[df1.JobTitle.str.contains('chair'), 'JobTitle'] = 'Board Member'
18
   df1.loc[df1.JobTitle.str.contains('professor'), 'JobTitle'] = 'Professor'
19
   df1.loc[df1.JobTitle.str.contains('officer'), 'JobTitle'] = 'Officer'
20
   df1.loc[df1.JobTitle.str.contains('blogger'), 'JobTitle' = 'Blogger'
21
   df1.loc[df1.JobTitle.str.contains('scientist'), 'JobTitle'] = 'Research Scientist'
22
   df1.loc[df1.JobTitle.str.contains('research'), 'JobTitle'] = 'PhD Student'
23
24
   df1.loc[df1.JobTitle.str.contains('phd'), 'JobTitle'] = 'PhD Student'
   df1.loc[df1.JobTitle.str.contains('thesis'), 'JobTitle'] = 'PhD Student'
25
   df1.loc[df1.JobTitle.str.contains('core'), 'JobTitle'] = 'Commitee Member'
26
27
    df1.loc[df1.JobTitle.str.contains('member'), 'JobTitle'] = 'Commitee Member'
   df1.loc[df1.JobTitle.str.contains('head'), 'JobTitle'] = 'Team Leader'
df1.loc[df1.JobTitle.str.contains('lead'), 'JobTitle'] = 'Team Leader'
28
   df1.loc[df1.JobTitle.str.contains('admin'), 'JobTitle'] = 'Team Leader'
df1.loc[df1.JobTitle.str.contains('sales'), 'JobTitle'] = 'Sales Representative'
30
31
32
   df1.loc[df1.JobTitle.str.contains('market'), 'JobTitle'] = 'Marketing'
33
   df1.loc[df1.JobTitle.str.contains('specialist'), 'JobTitle'] = 'Specialist'
34 df1.loc[df1.JobTitle.str.contains('edit'), 'JobTitle'] = 'Editor'
35
   df1.loc[df1.JobTitle.str.contains('ambassador'), 'JobTitle'] = 'Product ambassador'
36
    df1.loc[df1.JobTitle.str.contains('campus'), 'JobTitle'] = 'Product ambassador'
    df1.loc[df1.JobTitle.str.contains('grapher'), 'JobTitle'] = 'Photographer'
37
    df1.loc[df1.JobTitle.str.contains('assistant'), 'JobTitle'] = 'Assistant'
38
39
    df1.loc[df1.JobTitle.str.contains('archi'), 'JobTitle'] = 'Architect'
    df1.loc[df1.JobTitle.str.contains('creator'), 'JobTitle'] = 'Content Creator'
40
   df1.loc[df1.JobTitle.str.contains('content'), 'JobTitle'] = 'Content Creator'
41
   df1.loc[df1.JobTitle.str.contains('desi'), 'JobTitle'] = 'Product Designer'
42
   df1.loc[df1.JobTitle.str.contains('test'), 'JobTitle'] = 'Sofware tester'
43
   df1.loc[df1.JobTitle.str.contains('rep'), 'JobTitle'] = 'Product Representative'
df1.loc[df1.JobTitle.str.contains('art'), 'JobTitle'] = 'Artist'
44
45
46
    df1.loc[df1.JobTitle.str.contains('volunteer'), 'JobTitle'] = 'Volunteer'
47
    df1.loc[df1.JobTitle.str.contains('coordinator'), 'JobTitle'] = 'Coordinator'
   df1.loc[df1.JobTitle.str.contains('consult'), 'JobTitle'] = 'Consultant'
48
   df1.loc[df1.JobTitle.str.contains('advi'), 'JobTitle'] = 'Advisor'
49
    df1.loc[df1.JobTitle.str.contains('writ'), 'JobTitle'] = 'Writer'
50
   df1.loc[df1.JobTitle.str.contains('read'), 'JobTitle'] = 'Writer'
51
   df1.loc[df1.JobTitle.str.contains('grow'), 'JobTitle'] = 'Growth Hacker'
52
   df1.loc[df1.JobTitle.str.contains('youtube'), 'JobTitle'] = 'Youtuber'
53
   df1.loc[df1.JobTitle.str.contains('compos'), 'JobTitle'] = 'Composer'
55 df1.loc[df1.JobTitle.str.contains('executive'), 'JobTitle'] = 'Executive'
56 df1.loc[df1.JobTitle.str.contains('solution'), 'JobTitle'] = 'Solution Expert'
57 df1.loc[df1.JobTitle.str.contains('expert'), 'JobTitle'] = 'Product Expert'
58 df1.loc[df1.JobTitle.str.contains('actor'), 'JobTitle'] = 'Actor'
59 df1.loc[df1.JobTitle.str.contains('review'), 'JobTitle'] = 'Reviewer'
```

```
df1.loc[df1.JobTitle.str.contains('organize'), 'JobTitle'] = 'Organizer'
df1.loc[df1.JobTitle.str.contains('agent'), 'JobTitle'] = 'Product Agent'
df1.loc[df1.JobTitle.str.contains('promot'), 'JobTitle'] = 'Promoter'
df1.loc[df1.JobTitle.str.contains('student'), 'JobTitle'] = 'Student'
df1.loc[df1.JobTitle.str.contains('communicat'), 'JobTitle'] = 'Communicator'
df1.loc[df1.JobTitle.str.contains('business'), 'JobTitle'] = 'Business Stratergy'
df1.loc[df1.JobTitle.str.contains('retail'), 'JobTitle'] = 'Retailer'
df1.loc[df1.JobTitle.str.contains('public'), 'JobTitle'] = 'Public Relations'
df1.loc[df1.JobTitle.str.contains('secret'), 'JobTitle'] = 'Secretary'
df1.loc[df1.JobTitle.str.contains('trainee'), 'JobTitle'] = 'Trainee'
df1.loc[df1.JobTitle.str.contains('control'), 'JobTitle'] = 'Project Control'
df1.loc[df1.JobTitle.str.contains('operat'), 'JobTitle'] = 'Project Control'
df1.loc[df1.JobTitle.str.contains('transla'), 'JobTitle'] = 'Translator'
df1.loc[df1.JobTitle.str.contains('tutor'), 'JobTitle'] = 'Trainer'
df1.loc[df1.JobTitle.str.contains('tutor'), 'JobTitle'] = 'Trainer'
df1.loc[df1.JobTitle.str.contains('instruct'), 'JobTitle'] = 'Trainer'
```

In [32]:

```
1 print("Number of unique Job Titles: ", df1['JobTitle'].nunique())
```

Number of unique Job Titles: 177

Pre-processing the Industry coloum

```
In [33]:
```

```
1 df1['Industry'].value_counts(dropna=False)
```

Out[33]:

Information Technology and Services	785
Computer Software	554
Internet	181
Financial Services	179
Management Consulting	178
Fine Art	1
Wireless	1
Executive Office	1
Wine and Spirits	1
Warehousing	1
Name: Industry, Length: 128, dtype:	int64

In [34]:

```
df1['Industry'].replace({'Higher Education':'Primary/Secondary Education',
                              'Medical Practice': 'Hospital & Health Care',
 2
 3
                              'Mental Health Care': 'Hospital & Health Care',
                              'Health, Wellness and Fitness': 'Hospital & Health Care',
 4
 5
                              'Medical Devices': 'Hospital & Health Care',
 6
                              'Pharmaceuticals': 'Hospital & Health Care',
 7
                              'Veterinary': 'Hospital & Health Care',
                              'Computer Software': 'Information Technology and Services',
 8
 9
                              'Wireless': 'Information Technology and Services',
                              'Computer Games': 'Information Technology and Services',
10
                              'Information Services': 'Information Technology and Services',
11
                              'Computer & Network Security':'Information Technology and Serv
12
                              'Computer Networking': 'Information Technology and Services',
13
14
                              'Internet': 'Information Technology and Services',
15
                              'Automotive': 'Mechanical or Industrial Engineering',
16
                              'Construction':'Civil Engineering',
                              'Building Materials':'Civil Engineering',
17
                              'Railroad Manufacture': 'Civil Engineering',
18
                              'Investment Banking': 'Banking',
19
20
                              'Online Media': 'Media Production',
21
                              'Broadcast Media': 'Media Production',
22
                              'Food & Beverages':'Consumer Goods',
23
                              'Restaurants':'Consumer Goods',
24
                              'Wine and Spirits':'Consumer Goods',
25
                              'Oil & Energy': 'Consumer Goods',
26
                              'Chemicals':'Consumer Goods',
27
                              'Insurance': 'Consumer Services',
28
                              'Hospitality': 'Consumer Services',
29
                              'Telecommunications':'Consumer Services',
                              'Arts and Crafts':'Consumer Goods',
30
31
                              'Newspapers':'Consumer Goods',
                              'Plastics':'Consumer Goods',
32
                              'Fine Art':'Consumer Goods',
33
                              'Wholesale': 'Consumer Goods',
34
35
                              'Paper & Forest Products': 'Consumer Goods',
36
                              'Textiles':'Consumer Goods',
37
                              'Food Production': 'Consumer Goods',
                              'Luxury Goods & Jewelry':'Consumer Goods',
38
                              'Consumer Electronics':'Consumer Goods',
39
40
                              'Staffing and Recruiting': 'Consumer Services',
                              'Environmental Services': 'Consumer Services',
41
42
                              'Financial Services': 'Consumer Services',
43
                              'Events Services': 'Consumer Services',
                              'Legal Services':'Consumer Services',
44
45
                              'Individual & Family Services': 'Consumer Services',
                              'Facilities Services': 'Consumer Services',
46
47
                              'Supermarkets':'Consumer Goods',
                              'Retail':'Consumer Goods',
48
49
                              'Package/Freight Delivery':'Consumer Services',
50
                              'Import and Export':'International Trade and Development',
51
                              'International Affairs':'International Trade and Development'
                              'Outsourcing/Offshoring':'International Trade and Development
52
53
                              'Transportation/Trucking/Railroad':'Logistics and Supply Chair
                              'Leisure, Travel & Tourism': 'Logistics and Supply Chain',
54
55
                              'Writing and Editing': 'Publishing',
                              'Music': 'Entertainment',
56
57
                              'Animation': 'Entertainment',
58
                              'Media Production': 'Entertainment',
                              'Photography': 'Entertainment',
```

```
'Philanthropy': 'Nonprofit Organization Management',
60
61
                              'Market Research': 'Research',
                              'Think Tanks': 'Research',
62
                              'Computer Hardware':'Consumer Goods',
63
                              'Apparel & Fashion':'Consumer Goods',
64
                              'Semiconductors':'Consumer Goods',
65
                              'Cosmetics':'Consumer Goods',
66
                              'Packaging and Containers': 'Consumer Services',
67
                              'Public Relations and Communications': 'Consumer Services',
68
                              'Printing':'Consumer Services',
69
                              'Glass, Ceramics & Concrete':'Consumer Goods',
70
                              'Machinery':'Consumer Goods',
71
                              'Utilities': 'Consumer Goods',
72
73
                              'Shipbuilding':'Consumer Services',
74
                              'Banking':'Consumer Services',
                              'Motion Pictures and Film': 'Entertainment',
75
76
                              'Accounting':'Consumer Services',
                              'Human Resources':'Consumer Services',
77
                              'Translation and Localization':'Consumer Services',
78
79
                              'Publishing':'Consumer Services',
                              'Professional Training & Coaching': 'Consumer Services',
80
                              'Program Development':'Consumer Services'
81
82
                             },inplace = True)
83
```

In [35]:

```
1 df1['Industry'].unique()
```

Out[35]:

```
array(['Mechanical or Industrial Engineering',
       'Information Technology and Services', 'Consumer Goods',
       'Consumer Services', 'Hospital & Health Care', 'Design',
       'Management Consulting', 'Entertainment',
       'Electrical/Electronic Manufacturing', 'Civil Engineering',
       'Education Management', 'Industrial Automation',
       'Primary/Secondary Education', 'Aviation & Aerospace', 'Research',
       'Nonprofit Organization Management', 'Publishing',
       'Marketing and Advertising', 'Real Estate', 'Graphic Design',
       'International Trade and Development', 'Renewables & Environment',
       'Law Practice', 'Biotechnology', 'E-Learning',
       'Architecture & Planning', 'Sports',
'Venture Capital & Private Equity', 'Logistics and Supply Chain',
       'not_given', 'Public Safety', 'Mining & Metals',
       'Civic & Social Organization', 'Airlines/Aviation',
       'Security and Investigations', 'Media Production',
       'Business Supplies and Equipment', 'Capital Markets',
       'Commercial Real Estate', '-1', 'Warehousing', 'Defense & Space',
       'Investment Management', 'Judiciary', 'Public Policy',
       'Government Administration', 'Farming', 'Banking',
       'Executive Office', 'Nanotechnology'], dtype=object)
```

In [36]:

1 df1['Industry'].value_counts()[0:50]

Out[36]:

Information Technology and Services	1617
Consumer Services	481
Consumer Goods	433
Mechanical or Industrial Engineering	296
Management Consulting	178
Nonprofit Organization Management	178
Marketing and Advertising	175
Hospital & Health Care	159
Electrical/Electronic Manufacturing	136
Entertainment	124
Civil Engineering	124
Education Management	122
Biotechnology	102
E-Learning	93
Research	92
Design	85
Primary/Secondary Education	74
Publishing	57
Logistics and Supply Chain	54
Aviation & Aerospace	41
Industrial Automation	36
Media Production	28
Graphic Design	26
Renewables & Environment	24
International Trade and Development	20
Real Estate	17
Venture Capital & Private Equity	15
Sports	13
Business Supplies and Equipment	12
Architecture & Planning	12
Capital Markets	11
not_given	10
Mining & Metals	9
-1	9
Civic & Social Organization	7
Airlines/Aviation	7
Investment Management	6
Commercial Real Estate	3
Public Safety	3
Security and Investigations	3
Banking	3
Defense & Space	
Nanotechnology	3 3
Law Practice	2
Government Administration	2
Public Policy	1
Judiciary	1
Farming	1
Executive Office	1
Warehousing	1
Name: Industry, dtype: int64	

Preprocessing the Degree Column

```
In [39]:
```

```
df1['Degree'] = df1['Degree'].str.replace('[^\w\s]','')
df1['Degree']=df1['Degree'].str.lower()
df1['Degree']=df1['Degree'].str.replace(' ','')

df1.Degree = df1.Degree.fillna('not_given')
```

In [40]:

```
df1.loc[df1.Degree.str.contains('null'), 'Degree'] = 'null'
   df1.loc[df1.Degree.str.contains('bachelor'), 'Degree'] = 'ug'
   df1.loc[df1.Degree.str.contains('under'), 'Degree'] = 'ug'
   df1.loc[df1.Degree.str.contains('be'), 'Degree'] = 'ug'
   df1.loc[df1.Degree.str.contains('btec'), 'Degree'] = 'ug'
 5
   df1.loc[df1.Degree.str.contains('bachlors'), 'Degree'] = 'ug'
   df1.loc[df1.Degree.str.contains('engineer'), 'Degree'] = 'ug'
 7
   df1.loc[df1.Degree.str.contains('bc'), 'Degree'] = 'ug'
9
   df1.loc[df1.Degree.str.contains('mtech'), 'Degree'] = 'pg'
   df1.loc[df1.Degree.str.contains('master'), 'Degree'] = 'pg'
10
   df1.loc[df1.Degree.str.contains('mca'), 'Degree'] = 'pg'
df1.loc[df1.Degree.str.contains('mba'), 'Degree'] = 'pg'
11
12
   df1.loc[df1.Degree.str.contains('business'), 'Degree'] = 'pg'
13
   df1.loc[df1.Degree.str.contains('management'), 'Degree'] = 'pg'
   df1.loc[df1.Degree.str.contains('post'), 'Degree'] = 'pg'
15
16
   df1.loc[df1.Degree.str.contains('ms'), 'Degree'] = 'pg'
   df1.loc[df1.Degree.str.contains('ma'), 'Degree'] = 'pg'
17
   df1.loc[df1.Degree.str.contains('pg'), 'Degree'] = 'pg'
18
   df1.loc[df1.Degree.str.contains('mdes'), 'Degree'] = 'pg'
19
20
   df1.loc[df1.Degree.str.contains('doctor'), 'Degree'] = 'phd'
21
   df1.loc[df1.Degree.str.contains('research'), 'Degree'] = 'phd'
22
   df1.loc[df1.Degree.str.contains('phd'), 'Degree'] = 'phd'
   df1.loc[df1.Degree.str.contains('bs'), 'Degree'] = 'ug'
23
24
   df1.loc[df1.Degree.str.contains('school'), 'Degree'] = 'school'
   df1.loc[df1.Degree.str.contains('high'), 'Degree'] = 'school'
26
   df1.loc[df1.Degree.str.contains('ba'), 'Degree'] = 'ug'
27
   df1.loc[df1.Degree.str.contains('graduate'), 'Degree'] = 'ug'
    df1.loc[df1.Degree.str.contains('cs'), 'Degree'] = 'ug'
28
29
    df1.loc[df1.Degree.str.contains('mechanical'), 'Degree'] = 'ug'
30
31
    df1['Degree'] = df1['Degree'].str.replace('ugprofessionalyearug', 'ug')
32
    df1['Degree'].replace({'byech':'ug','executiveeducation':'pg','me':'pg','sslc':'school
33
                            'preuniversity':'school','10thboards':'school','intermediate':'
34
35
                            'student':'school','12th':'school','preuniversitycource':'schoo
                            'ece':'ug','grduate':'ug','thesis':'phd','llicenciatura':'other
36
                            'pncandkrdegreecollege':'ug', 'arts':'ug',
37
38
           'onlinesaleinindia':'other',
           'certificateprogramme':'other',
39
           'financialmodellingandvaluation':'other',
40
           '2styear':'other', 'vitvellore':'ug',
41
           'professionaldegreecertificate':'other',
42
43
            'valuenagotiation':'other', 'idp':'other',
           'contractlaw':'ug', 'projectstudent':'other',
44
           'projectplanninganalysisandcontrol': 'other',
45
            'ee':'other', 'dme':'other', 'pmpusa':'other', 'aws':'other',
46
           'ocw':'other', 'accountingfundamental':'other', 'extensionstudies':'other', 'mi
47
           'gniit':'ug', '学士':'other',
48
            'bt':'other', 'dece':'ug', 'diplomo':'ug', 'amie':'other', 'biotech':'ug', 'de
49
           'presidencycollegeofchennai':'ug', 'commerce':'ug', 'fellowshipprogram':'phd',
50
51
           'innovationandentrepreneurship':'other', 'onlinecourse':'other', 'no':'other',
           'associateofscienceas':'ug', 'inprocess':'other',
52
53
           'mobileapplicationdevelopment':'other',
           'chineseenglishtranslatorinchennai09910713101':'other', 'grandeécole':'other',
54
           'craftingcreativecommunication':'other', 'documentaryproduction':'other',
55
           'computerscience':'ug', 'nanodegree':'pg', 'diplômeetudiantentrepreneur':'ug'
56
            'professionalyear''ug',
57
           'advancedjavaframeworks':'other', 'chefdeprojetmultimedia':'other',
58
           'intermediatescience':'pg', 'mphil':'pg', 'cqfcertification':'other', 'dhmct':'
59
```

```
60
            'certificationingamedesigining':'other', 'licencedinsuranceagent':'other', 'aecp
61
                            'professionalyear':'pg','ugprofessionalyearug':'ug'},inplace=Tr
In [41]:
 1 df1['Degree'].unique()
Out[41]:
array(['pg', 'ug', 'not_given', 'phd', 'other', 'school',
       'ugprofessionalyearug'], dtype=object)
In [42]:
 1 df1['Degree'].value_counts(dropna=False)
Out[42]:
                        2660
ug
                        1918
pg
                         194
not_given
phd
                          78
                          39
other
                          20
school
ugprofessionalyearug
Name: Degree, dtype: int64
In [43]:
 1 print("Number of unique Degree: ", df1['Degree'].nunique())
```

Number of unique Degree: 7

Preprocessing the Field of Study Column

```
In [44]:
```

```
1 df1['FieldOfStudy']=df1['Field of Study']
 2 df1['FieldOfStudy'].value_counts()
Out[44]:
Computer Science
Mechanical Engineering
371
Information Technology
206
```

Electrical and Electronics Engineering

Digital Game Design

Mechanical engineering with specialization in energy engineering

Electrical, Electronics and Communications Engineering

Computer Science Engineering with Specialization in Information Security

Electronics and Communication with specialisation in IOT and Sensors

Applied Biology

Name: FieldOfStudy, Length: 1115, dtype: int64

In [45]:

```
1 df1['FieldOfStudy'] = df1['FieldOfStudy'].str.replace('[^\w\s]','')
  df1['FieldOfStudy']=df1['FieldOfStudy'].str.lower()
  df1['FieldOfStudy']=df1['FieldOfStudy'].str.replace(' ','')
4
5
  df1.FieldOfStudy = df1.FieldOfStudy.fillna('not_given')
```

In [46]:

```
df1.loc[df1.FieldOfStudy.str.contains('computer'), 'FieldOfStudy'] = 'Computer Science'
    df1.loc[df1.FieldOfStudy.str.contains('security'), 'FieldOfStudy'] = 'Computer Science'
 2
    df1.loc[df1.FieldOfStudy.str.contains('data'), 'FieldOfStudy'] = 'Computer Science'
    df1.loc[df1.FieldOfStudy.str.contains('artificial'), 'FieldOfStudy'] = 'Computer Science
 5
    df1.loc[df1.FieldOfStudy.str.contains('cse'), 'FieldOfStudy'] = 'Computer Science'
    df1.loc[df1.FieldOfStudy.str.contains('information'), 'FieldOfStudy'] = 'Computer Scient'
    df1.loc[df1.FieldOfStudy.str.contains('network'), 'FieldOfStudy'] = 'Computer Science'
 7
    df1.loc[df1.FieldOfStudy.str.contains('electric'), 'FieldOfStudy'] = 'Electrical, Electrical df1.loc[df1.FieldOfStudy.str.contains('software'), 'FieldOfStudy'] = 'Computer Science'
 9
    df1.loc[df1.FieldOfStudy.str.contains('electronics'), 'FieldOfStudy'] = 'Electrical, El
10
    df1.loc[df1.FieldOfStudy.str.contains('communica'), 'FieldOfStudy'] = 'Electrical, Electrical')
11
    df1.loc[df1.FieldOfStudy.str.contains('ece'), 'FieldOfStudy'] = 'Electrical, Electronic
12
    df1.loc[df1.FieldOfStudy.str.contains('instrument'), 'FieldOfStudy'] = 'Electrical, Ele
13
    df1.loc[df1.FieldOfStudy.str.contains('sensor'), 'FieldOfStudy'] = 'Electrical, Electro
df1.loc[df1.FieldOfStudy.str.contains('biotech'), 'FieldOfStudy'] = 'Biotechnology'
15
    df1.loc[df1.FieldOfStudy.str.contains('mech'), 'FieldOfStudy'] = 'Mechanical Engineeric
16
    df1.loc[df1.FieldOfStudy.str.contains('auto'), 'FieldOfStudy'] = 'Mechanical Engineeric
17
    df1.loc[df1.FieldOfStudy.str.contains('market'), 'FieldOfStudy'] = 'Marketing and Final
18
    df1.loc[df1.FieldOfStudy.str.contains('financ'), 'FieldOfStudy'] = 'Marketing and Finand df1.loc[df1.FieldOfStudy.str.contains('manage'), 'FieldOfStudy'] = 'Management'
19
20
    df1.loc[df1.FieldOfStudy.str.contains('chem'), 'FieldOfStudy'] = 'Chemical Engineering
21
    df1.loc[df1.FieldOfStudy.str.contains('civil'), 'FieldOfStudy'] = 'Civil Engineering'
22
    df1.loc[df1.FieldOfStudy.str.contains('material'), 'FieldOfStudy'] = 'Civil Engineering
23
    df1.loc[df1.FieldOfStudy.str.contains('struct'), 'FieldOfStudy'] = 'Civil Engineering'
df1.loc[df1.FieldOfStudy.str.contains('busines'), 'FieldOfStudy'] = 'Business Studies'
24
25
    df1.loc[df1.FieldOfStudy.str.contains('innovat'), 'FieldOfStudy'] = 'Business Studies'
26
    df1.loc[df1.FieldOfStudy.str.contains('entrep'), 'FieldOfStudy'] = 'Business Studies'
27
    df1.loc[df1.FieldOfStudy.str.contains('mba'), 'FieldOfStudy'] = 'Business Studies'
28
    df1.loc[df1.FieldOfStudy.str.contains('media'), 'FieldOfStudy'] = 'Media'
    df1.loc[df1.FieldOfStudy.str.contains('manufact'), 'FieldOfStudy'] = 'Manufacturing and
30
    df1.loc[df1.FieldOfStudy.str.contains('production'), 'FieldOfStudy'] = 'Manufacturing a
31
32
    df1.loc[df1.FieldOfStudy.str.contains('design'), 'FieldOfStudy'] = 'Design'
33
    df1.loc[df1.FieldOfStudy.str.contains('commerce'), 'FieldOfStudy'] = 'Commerce'
    df1.loc[df1.FieldOfStudy.str.contains('arts'), 'FieldOfStudy'] = 'Arts'
34
35
    df1.loc[df1.FieldOfStudy.str.contains('english'), 'FieldOfStudy'] = 'Arts'
    df1.loc[df1.FieldOfStudy.str.contains('operat'), 'FieldOfStudy'] = 'Operations'
36
    df1.loc[df1.FieldOfStudy.str.contains('research'), 'FieldOfStudy'] = 'Research'
37
    df1.loc[df1.FieldOfStudy.str.contains('energy'), 'FieldOfStudy'] = 'Energy Engineering
df1.loc[df1.FieldOfStudy.str.contains('visual'), 'FieldOfStudy'] = 'Visual Communication'
38
39
    df1.loc[df1.FieldOfStudy.str.contains('system'), 'FieldOfStudy'] = 'System Engineering'
40
    df1.loc[df1.FieldOfStudy.str.contains('envi'), 'FieldOfStudy'] = 'Environmental Engine
41
    df1.loc[df1.FieldOfStudy.str.contains('food'), 'FieldOfStudy'] = 'Food Science'
42
43
    df1.loc[df1.FieldOfStudy.str.contains('archi'), 'FieldOfStudy'] = 'Architecture'
    df1.loc[df1.FieldOfStudy.str.contains('psycho'), 'FieldOfStudy'] = 'Psychology'
44
    df1.loc[df1.FieldOfStudy.str.contains('math'), 'FieldOfStudy'] = 'Mathematics'
df1.loc[df1.FieldOfStudy.str.contains('econo'), 'FieldOfStudy'] = 'Economics'
45
46
    df1.loc[df1.FieldOfStudy.str.contains('journal'), 'FieldOfStudy'] = 'Journalism'
    df1.loc[df1.FieldOfStudy.str.contains('literat'), 'FieldOfStudy'] = 'Arts'
48
```

In [47]:

```
1
   df1['FieldOfStudy'].replace({'backendwebdevelopment':'Computer Science','compter':'Com
 2
                                  'imageprocessing':'Computer Science', 'urbanplanning':'Civ
 3
                                 'remotesensing':'Electrical, Electronics and Communication
 4
           'hr': 'Management',
           'entrepreneurshipandnewventurecreation':'Business Studies', 'generalmangement':
 5
           'executiveeducation': 'Business Studies', 'socialwork': 'not_given', 'strategylead
 6
           'liberalstudies':'Law', 'bomedicalengneering':'Biotechnology', 'bkfs':'not_give
 7
           'strategyandorganization':'Business Studies', 'masteroftechnologymtechcadcam':'
 8
9
           'departmentofhaematologyhaemostasisoncologyandstemcelltransplantation':'Biotech
           'foundationfielbus': 'not given',
10
           'robotics':'Mechanical Engineering','animationinteractivetechnologyvideographic
11
            'bba':'Law','law':'Law','basiclaw':'Law','highschoolsecondarycertificateprogram
12
           'technologyandpolicy':'not_given', 'smartcities':'not_given', 'leanengineering'
13
           'prourementinventory':'not_given','petroleumengineering':'Energy Engineering',
14
            'semiconductordevicefabrication':'Electrical, Electronics and Communication Eng
15
16
            'strategyandleadership':'Business Studies','strategy':'Business Studies','':'no
            'cs':'Computer Science','broadcastengineering':'Media','pcmb':'School','mca':'(
17
            'bachelorofengineeringbe':'not_given','btech':'not_given', 'cloudcomputing':'Co
18
                                }, inplace=True)
19
```

In [48]:

```
1 df1['FieldOfStudy'].value_counts()
```

Out[48]:

```
Computer Science
                                                           1731
Electrical, Electronics and Communication Engineering
                                                            525
Mechanical Engineering
                                                            492
not_given
                                                            393
Marketing and Finance
                                                            350
technical
                                                              1
pcminformatics
                                                              1
cellbiology
                                                              1
                                                              1
regulatoryaffairsofdrugsbiologicsandmedicaldevices
Name: FieldOfStudy, Length: 157, dtype: int64
```

Preprocessing the New Job Column

1. Filling Missing Values

In [49]:

```
print("Number of unique values: ", df1['New Job (90 Days)'].nunique())
print("Number of rows: ", df1['New Job (90 Days)'].shape[0])
print("Number of null values: ", df1['New Job (90 Days)'].isnull().sum())
```

```
Number of unique values: 2
Number of rows: 4910
Number of null values: 150
```

```
In [50]:
 1 print(df['New Job (90 Days)'].describe())
           4760
count
unique
              2
top
          False
freq
           3984
Name: New Job (90 Days), dtype: object
In [51]:
 1 print(df1.index[(df1['Year Started'] == 2021) & (df1['New Job (90 Days)'] != True)].to]
 2 print(df1.index[(df1['Year Started'] == 2021) & (df1['New Job (90 Days)'] == False)].te
[830, 1527, 2184]
In [52]:
 1 df1['New Job (90 Days)'].fillna(False, inplace = True)
In [53]:
 1 df1['New Job (90 Days)'].isnull().sum()
Out[53]:
0
```

Preprocessing the Profile Headline column

1. Removing Null Values

```
In [54]:
```

```
print("Number of unique values: ", df1['Profile Headline'].nunique())
print("Number of rows: ", df1['Profile Headline'].shape[0])
print("Number of null values: ", df1['Profile Headline'].isnull().sum())
Number of unique values: 3941
```

Number of rows: 4910
Number of null values: 18

In [55]:

```
nan_profileheadline = df1.index[df1['Profile Headline'].isnull()].tolist()
nan_profileheadline
```

Out[55]:

```
[66,
 275,
 552,
 710,
 904,
 1653,
 1876,
 1878,
 1908,
 2146,
 2334,
 2551,
 2628.
 2647,
 2699,
 2999,
 3175,
 3676]
```

In [56]:

```
1 df1.at[66, 'Profile Headline'] = 'Computer Science and Engineering Student'
   df1.at[275, 'Profile Headline'] = 'Electrical, Electronics and Communications Engineer:
   df1.at[66, 'Profile Headline'] = 'Computer Science and Engineering Student'
   df1.at[552, 'Profile Headline'] = 'Mechanical Engineering Student'
   df1.at[710, 'Profile Headline'] = 'Mechanical Engineering Student'
 5
   df1.at[904, 'Profile Headline'] = 'Computer Science and Engineering Student'
   df1.at[1653, 'Profile Headline'] = 'Computer Science and Engineering Student'
7
   df1.at[1876, 'Profile Headline'] = 'Computer Science and Engineering Student'
 8
   df1.at[1878, 'Profile Headline'] = 'Mechanical Engineering Student | Business Developme
9
   df1.at[1908, 'Profile Headline'] = 'Automotive Engineering Student'
10
   df1.at[2146, 'Profile Headline'] = 'Computer Science and Engineering Student | Cyber S€
11
   df1.at[2334, 'Profile Headline'] = 'Computer Science and Engineering Student | Web Deve
12
   df1.at[2551, 'Profile Headline'] = 'Mechanical Engineering Student'
13
   df1.at[2551, 'Education End'] = 2021.0
14
15
   df1.at[2628, 'Profile Headline'] = 'Computer Science and Engineering Student'
   df1.at[2647, 'Profile Headline'] = 'Computer Science and Engineering Student'
16
   df1.at[2699, 'Profile Headline'] = 'Automotive Engineering student'
17
   df1.at[2999, 'Profile Headline'] = 'Electrical and Electronics Engineering Student | A
18
   df1.at[2999, 'Education End'] = 2022.0
19
   df1.at[3175, 'Profile Headline'] = 'Automotive Engineering Student'
20
   df1.at[3676, 'Profile Headline'] = 'Electrical, Electronics and Communications Engineer
21
```

```
In [57]:
 1 i = nan profileheadline[15]
 2 print(i)
 3 df1.iloc[i]
 4 # df1.iloc[i]['Profile Summary']
 5 # df1.iloc[i]['Company Name']
2999
Out[57]:
Job Title
                                                  Java developer intern
Company Name
                                             Tata Consultancy Services
Industry
                                   Electrical/Electronic Manufacturing
Company Location
                                              Faridabad, Haryana, India
New Job (90 Days)
                                                                   True
Year Started
Profile Headline
                     Electrical and Electronics Engineering Student...
Profile Summary
                     Aerial robotics enthusiast with a keen interes...
                                       Vellore Institute of Technology
School
Degree
                                                                     ug
Education End
                                                                   2022
                                Electrical and Electronics Engineering
Field of Study
Domain
                                                  Business Development
CompanyName
                                             Tata Consultancy Services
JobTitle
                                                              Developer
```

Name: 2999, dtype: object

```
In [58]:
```

My Network

Country

Continent

FieldOfStudy

```
df1['Profile Headline'].isnull().sum()
```

network

Computer Science

India

AS

Out[58]:

0

2. Named Entity Recognition

Preprocessing the Education End column

```
In [59]:
```

```
1 df1['Education End'].value_counts(dropna = False)
Out[59]:
2022.0
          533
2020.0
          479
2021.0
          448
2023.0
          422
2019.0
          408
2018.0
          367
2017.0
          299
2016.0
          259
2015.0
          246
NaN
          236
2014.0
          155
2013.0
          129
2024.0
          106
2012.0
          101
2011.0
           96
            85
2010.0
2008.0
           72
           69
2009.0
           67
2005.0
2007.0
           59
2006.0
           49
2003.0
           33
2004.0
           31
1999.0
           21
1997.0
           17
           14
2000.0
2025.0
           12
2002.0
           12
1993.0
           11
            9
1998.0
            9
2001.0
            7
1996.0
            4
1995.0
1988.0
            4
             4
1994.0
1989.0
            3
            3
1981.0
            3
1992.0
            3
1990.0
            3
1980.0
             3
1987.0
            3
1977.0
            2
1985.0
            2
1969.0
             2
1976.0
            2
1979.0
1973.0
            1
            1
1966.0
            1
1982.0
1986.0
            1
            1
1972.0
             1
1965.0
1960.0
            1
1991.0
            1
```

Name: Education End, dtype: int64

```
In [60]:
```

```
1 # df1['Education End'].fillna(0, inplace = True)
```

In [61]:

```
1 arr = df1['Education End'].unique()
2 arr = np.sort(arr)[::-1]
```

In [62]:

```
1 arr
```

Out[62]:

```
array([ nan, 2025., 2024., 2023., 2022., 2021., 2020., 2019., 2018., 2017., 2016., 2015., 2014., 2013., 2012., 2011., 2010., 2009., 2008., 2007., 2006., 2005., 2004., 2003., 2002., 2001., 2000., 1999., 1998., 1997., 1996., 1995., 1994., 1993., 1992., 1991., 1990., 1989., 1988., 1987., 1986., 1985., 1982., 1981., 1980., 1979., 1977., 1976., 1973., 1972., 1969., 1966., 1965., 1960.])
```

In [63]:

```
df1['Education End'].replace({2025 : 'group1', 2024 : 'group1', 2023 : 'group1', 2022
 2
                                  2020 : 'group2', 2019 : 'group2', 2018 : 'group2', 2017
                                          'group3', 2013 : 'group3', 2012 :
                                                                            'group3', 2011
 3
 4
                                  2009 : 'group3', 2008 : 'group3', 2007 : 'group3', 2006
 5
                                  2004 : 'group3', 2003 : 'group3', 2002 : 'group3', 2001
                                  1999 : 'group4', 1998 : 'group4', 1997 :
                                                                            'group4', 1996
 6
 7
                                  1994 : 'group4', 1993 : 'group4', 1992 : 'group4', 1991
                                  1989 : 'group4', 1988 : 'group4', 1987 : 'group4', 1986
 8
                                  1982 : 'group4', 1981 : 'group4', 1980 : 'group4', 1979
 9
                                  1976 : 'group4', 1973 : 'group4', 1972 : 'group4', 1969
10
                                  1965 : 'group4', 1960 : 'group4', np.nan: 'not-given'},
11
12
13
14
   group 1: 2021 - 2015
15
   group 2: 2020 - 2015
   group 3: 2000 - 2015
16
    group 4: before 2000
17
18
    4
```

Out[63]:

```
'\ngroup 1: 2021 - 2015\ngroup 2: 2020 - 2015\ngroup 3: 2000 - 2015\ngroup 4: before 2000\n'
```

In [64]:

```
1 df1['Education End'].value_counts(dropna = False)
```

Out[64]:

```
group2 1812
group1 1521
group3 1227
not-given 236
group4 114
Name: Education End, dtype: int64
```

Final Preprocessing before applying BERT and converting to Embeddings

```
In [65]:
```

```
1 df1.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4910 entries, 0 to 4909
Data columns (total 19 columns):
    Column
                      Non-Null Count
                                     Dtype
    -----
                      -----
                                     ----
    Job Title
0
                      4898 non-null
                                     object
1
    Company Name
                      4892 non-null
                                     object
2
    Industry
                      4910 non-null
                                     object
3
    Company Location 4910 non-null
                                     object
4
    New Job (90 Days) 4910 non-null
                                     bool
5
    Year Started
                      4816 non-null
                                     float64
6
    Profile Headline
                      4910 non-null
                                     object
7
    Profile Summary
                      3465 non-null
                                     object
8
    School
                      4658 non-null
                                     object
9
    Degree
                      4910 non-null
                                     object
10 Education End
                      4910 non-null
                                     object
11 Field of Study
                      4532 non-null
                                     object
12 Domain
                      4910 non-null
                                     object
13 CompanyName
                      4910 non-null
                                     object
14 JobTitle
                      4910 non-null
                                     object
15 My Network
                      4910 non-null
                                     object
16 Country
                      4910 non-null
                                     object
17 Continent
                      4910 non-null
                                     object
18 FieldOfStudy
                    4910 non-null
                                      object
dtypes: bool(1), float64(1), object(17)
memory usage: 695.4+ KB
```

In [66]:

```
1 df1.isnull().count()
```

Out[66]:

Job Title	4910
Company Name	4910
Industry	4910
Company Location	4910
New Job (90 Days)	4910
Year Started	4910
Profile Headline	4910
Profile Summary	4910
School	4910
Degree	4910
Education End	4910
Field of Study	4910
Domain	4910
CompanyName	4910
JobTitle	4910
My Network	4910
Country	4910
Continent	4910
FieldOfStudy	4910
dtype: int64	

In [67]:

```
# df1['Job Title'].fillna("", inplace = True)
# df1['Company Name'].fillna("", inplace = True)
# df1['Industry'].fillna("", inplace = True)
# df1['Year Started'].fillna(0, inplace = True)
# df1['Profile Summary'].fillna("", inplace = True)
# df1['School'].fillna("", inplace = True)
# df1['Degree'].fillna("", inplace = True)
# df1['Education End'].fillna(0, inplace = True)
# df1['Field of Study'].fillna("", inplace = True)
```

In [68]:

```
1 # sns.heatmap(df1.isnull())
```

In [69]:

```
1 # df1.to_csv('Before_BERT.csv')
```

In [70]:

```
import matplotlib.pyplot as plt

print(df1.nunique())

plt.figure(figsize=(20,4))

plt.plot(df1.nunique())

plt.grid()

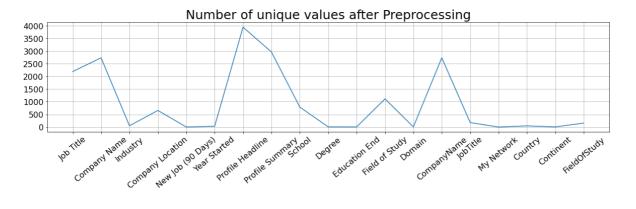
plt.title('Number of unique values after Preprocessing', fontsize = 25)

plt.xticks(fontsize=16, rotation=40)

plt.yticks(fontsize=16)

plt.show()
```

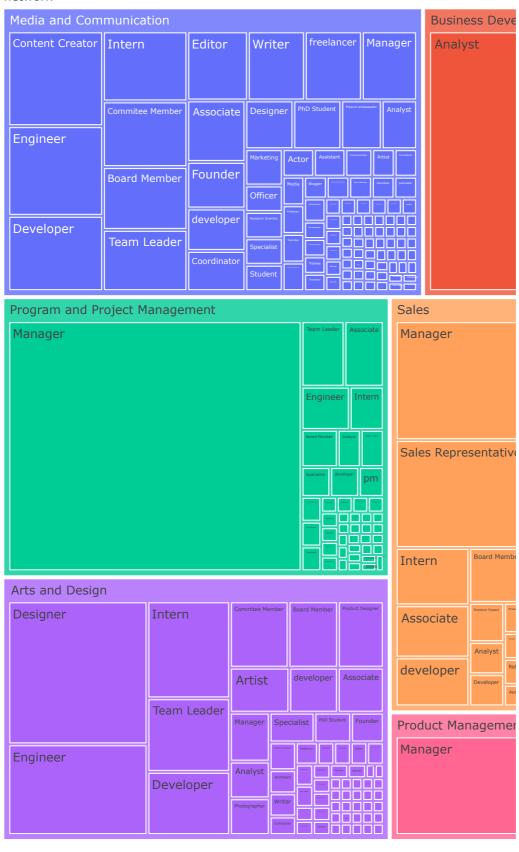
Job Title	2194
Company Name	2738
· •	50
Industry	
Company Location	657
New Job (90 Days)	2
Year Started	32
Profile Headline	3950
Profile Summary	2968
School	789
Degree	7
Education End	5
Field of Study	1115
Domain	7
CompanyName	2739
JobTitle	177
My Network	1
Country	47
Continent	7
FieldOfStudy	157
dtype: int64	



In [72]:

fig2 = px.treemap(df1, path=['My Network', 'Domain', 'JobTitle'], width=1000, height=1000
fig2.show()

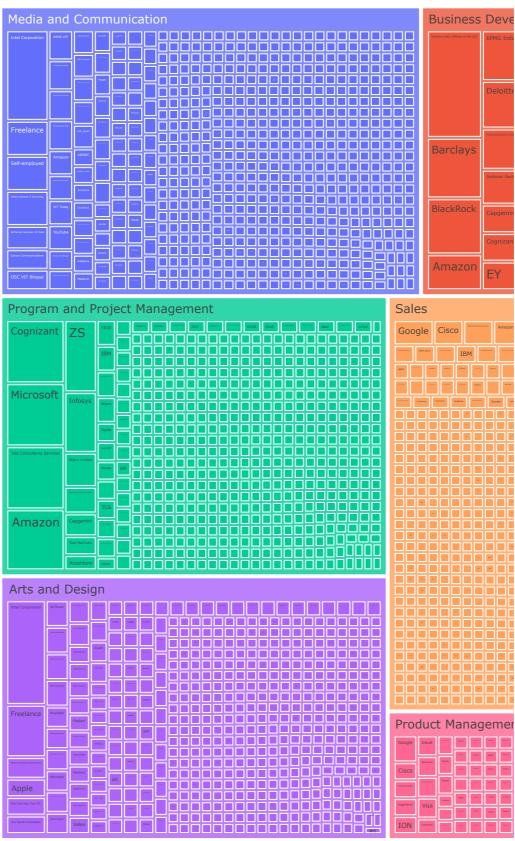
network



In [73]:

fig3 = px.treemap(df1, path=['My Network', 'Domain', 'CompanyName'], width=1000, height
fig3.show()

network



4	>
In []:	
1	
In []:	
1	
In []:	
1	
In []:	
1	