Fittlyf_Interview_Solution

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BTech Third Year

Part 0: Reading the data:

- **Print all the column names and the data types in each column.**

  Source code-

```
import pandas as pd
#making data frame
data = pd.read_csv("test_DataScience.csv")
# printing the column name and their datatype
DataTypeSeries= data.dtypes
print(DataTypeSeries)
```

OUTPUT:

```
Year                    int64
Month                  object
Laptop/Desktop         object
Type_of_Customers?     object
Coming from            object
Place_in_India         object
Level 1               float64
Level 2               float64
Level 3                 int64
Level 4                 int64
        dtype: object
```

- **Print the cities of India from which the page was accessed.**

  Source code:

```
import pandas as pd
dt=pd.read_csv("test_DataScience.csv")
#fetching the places in india from which the page was accessed.
print(dt[["Place_in_India"]])
```

OUTPUT:

```
     Place_in_India
0         Bengaluru
1         Hyderabad
2          Dehradun
3            Indore
4              Pune
...             ...
2155      Bengaluru
2156      Hyderabad
2157       Dehradun
2158         Indore
2159           Pune

[2160 rows x 1 columns]
```

- Write a brief paragraph about what you think about this dataset along the lines of :
  Q.1] Which geo-location this dataset belongs to?
  Answer: Flipkart is a tech company first then an e-comm firm. The dataset belongs to all famous IT hubs, metro cities in India. It belongs to Bengaluru, Hyderabad, Dehradun, Indore and Pune.
  If we run this query- **data['Place_in_India'].value_counts()**
  We get the output as,

```
Bengaluru    432
Hyderabad    432
Dehradun     432
Indore       432
Pune         432
Name: Place_in_India,
dtype: int64
```

As we can see, number of cities from India from where the page was accessed is 432 each. E-comm is a fast developing segment in India. Hence it targets such geolocations which are metro cities, much ahead of technology. The most interesting aspect of the business is that home-grown companies like Snapdeal and Flipkart are fighting out with global majors like Amazon and eBay. It is no mean feat, considering the Indian companies are still in the nascent stage. So these companies target the tech geo locations for sales and services first.

Q.2] Given that this dataset is for a website like Flipkart, what could be the possible definitions of the columns Level 1, 2, 3, 4 in the given dataset?
Answer: Level 1,2,3,4 are the dependent attributes. They depend on Type_of_customer attribute. So here there is function dependency.
Level 1 and 2 in not applicable for the existing customers. Where as it is applicable for new customers. That means this level 1 and 2 could be steps involved to order a product. It could be subscription, premium, sign in option. Or it could be rank.
Level 3 and 4 is applicable for all. That means it is allowed for both existing customers and new customer.

Part 1: Data cleaning

Write a function called data_cleaning() which, when called, would perform the following :

1] Create a new column, called 'Month_Year', using lambda function. The new column should be at the 3 rd position from the start in the given dataset & its values should be : '01-01-2020' for January, 2020 and '01-02-2020' for February 2020 and so on.

Source code:

```
import pandas as pd
from datetime import timedelta
df=pd.read_csv("test_DataScience.csv")
cols=["Month","Year"]
df['Date'] = df[cols].apply(lambda x: '-'.join(x.values.astype(str)), axis="columns")
df['Date']=pd.to_datetime(df['Date'])
df['Date'] = pd.to_datetime(df['Date']).dt.strftime('%d/%m/%Y')
df = df[['Year', 'Month', 'Date', 'Laptop/Desktop', 'Type_of_Customers?','Coming
from','Place_in_India','Level 1','Level 2','Level 3','Level 4']]
df.head()
```

OUTPUT:

Out[37]:

	Year	Month	Date	Laptop/Desktop	Type_of_Customers?	Coming from	Place_in_India	Level 1	Level 2	Level 3	Level 4
0	2020	Jan	01/01/2020	Desktop_Website	Existing_Customer	Came_From_LinkedIn	Bengaluru	NaN	NaN	56892	17178
1	2020	Jan	01/01/2020	Desktop_Website	Existing_Customer	Came_From_LinkedIn	Hyderabad	NaN	NaN	41460	11916
2	2020	Jan	01/01/2020	Desktop_Website	Existing_Customer	Came_From_LinkedIn	Dehradun	NaN	NaN	55561	19461
3	2020	Jan	01/01/2020	Desktop_Website	Existing_Customer	Came_From_LinkedIn	Indore	NaN	NaN	320923	110667
4	2020	Jan	01/01/2020	Desktop_Website	Existing_Customer	Came_From_LinkedIn	Pune	NaN	NaN	220937	46033

2] Replaces the null values with the average of the respective column in the data.

Source code:

```
import pandas as pd
df=pd.read_csv("test_DataScience.csv")
print(df.isnull().sum())    #no of missing values each column
print(df.isnull().sum().sum())  #no of missing values each column
df["Level 1"]=df["Level 1"].fillna(df["Level 1"].mean())
df["Level 2"]=df["Level 2"].fillna(df["Level 2"].mean())
print(df)
```

OUTPUT:

	Place_in_India	Level 1	Level 2	Level 3	Level 4
0	Bengaluru	7.838702e+05	3.582154e+05	56892	17178
1	Hyderabad	7.838702e+05	3.582154e+05	41460	11916
2	Dehradun	7.838702e+05	3.582154e+05	55561	19461
3	Indore	7.838702e+05	3.582154e+05	320923	110667
4	Pune	7.838702e+05	3.582154e+05	220937	46033
5	Bengaluru	7.838702e+05	3.582154e+05	90241	24229
6	Hyderabad	7.838702e+05	3.582154e+05	77630	18502
7	Dehradun	7.838702e+05	3.582154e+05	91479	24363
8	Indore	7.838702e+05	3.582154e+05	436641	165036
9	Pune	7.838702e+05	3.582154e+05	531446	101317
10	Bengaluru	7.838702e+05	3.582154e+05	32119	6900
11	Hyderabad	7.838702e+05	3.582154e+05	27891	5606
12	Dehradun	7.838702e+05	3.582154e+05	34391	8459
13	Indore	7.838702e+05	3.582154e+05	142422	39296
14	Pune	1.092340e+05	9.810000e+04	120090	20223
15	Bengaluru	1.128690e+05	9.180100e+04	48979	33382
16	Hyderabad	1.103970e+05	8.742900e+04	48899	29031
17	Dehradun	1.564870e+05	1.233240e+05	59084	39804
18	Indore	1.176804e+06	9.601450e+05	604293	373155
19	Pune	3.832600e+05	3.059020e+05	172827	122285
20	Bengaluru	2.682860e+05	1.494760e+05	58622	44999
21	Hyderabad	2.921280e+05	1.318900e+05	50720	32140
22	Dehradun	4.012920e+05	1.713220e+05	51069	36016
23	Indore	1.670668e+06	9.697320e+05	477858	339970
24	Pune	1.748075e+06	8.543770e+05	314289	225823
25	Bengaluru	5.817200e+04	4.409600e+04	18169	8769
26	Hyderabad	8.038400e+04	5.809700e+04	22564	8832
27	Dehradun	1.014530e+05	7.920100e+04	17203	10064
28	Indore	4.944470e+05	3.746380e+05	170498	88331
29	Pune	2.359550e+05	1.628900e+05	54175	30731
30	Bengaluru	7.838702e+05	3.582154e+05	57469	13257
31	Hyderabad	7.838702e+05	3.582154e+05	22092	5405
32	Dehradun	7.838702e+05	3.582154e+05	40947	13345

.....

[2160 rows * 5 columns.]

3.] In column ‘B’ replace Jan with 1, feb with 2, march with 3 and so on.

Source code:

```
import pandas as pd
def GetMonthInInt(month):
MonthInInts=pd.Series([1,2,3,4,5,6,7,8,9,10,11,12],index=['jan','feb','mar','apr','may','jun','jul','aug',
'sep','oct','nov','dec'])
   return MonthInInts[month.lower()]
df=pd.read_csv("test_DataScience.csv")
df['B']= df['Month'].apply(GetMonthInInt)
#print(df)
df.head()
```

OUTPUT:

Out[49]:

	Year	Month	Laptop/Desktop	Type_of_Customers?	Coming from	Place_in_India	Level 1	Level 2	Level 3	Level 4	B
0	2020	Jan	Desktop_Website	Existing_Customer	Came_From_LinkedIn	Bengaluru	NaN	NaN	56892	17178	1
1	2020	Jan	Desktop_Website	Existing_Customer	Came_From_LinkedIn	Hyderabad	NaN	NaN	41460	11916	1
2	2020	Jan	Desktop_Website	Existing_Customer	Came_From_LinkedIn	Dehradun	NaN	NaN	55561	19461	1
3	2020	Jan	Desktop_Website	Existing_Customer	Came_From_LinkedIn	Indore	NaN	NaN	320923	110667	1
4	2020	Jan	Desktop_Website	Existing_Customer	Came_From_LinkedIn	Pune	NaN	NaN	220937	46033	1

	Place_in_India	Level 1	Level 2	Level 3	Level 4	B
0	Bengaluru	NaN	NaN	56892	17178	1
1	Hyderabad	NaN	NaN	41460	11916	1
2	Dehradun	NaN	NaN	55561	19461	1
3	Indore	NaN	NaN	320923	110667	1
4	Pune	NaN	NaN	220937	46033	1
5	Bengaluru	NaN	NaN	90241	24229	1
6	Hyderabad	NaN	NaN	77630	18502	1
7	Dehradun	NaN	NaN	91479	24363	1
8	Indore	NaN	NaN	436641	165036	1
9	Pune	NaN	NaN	531446	101317	1
10	Bengaluru	NaN	NaN	32119	6900	1
11	Hyderabad	NaN	NaN	27891	5606	1
12	Dehradun	NaN	NaN	34391	8459	1
13	Indore	NaN	NaN	142422	39296	1
14	Pune	109234.0	98100.0	120090	20223	1
15	Bengaluru	112869.0	91801.0	48979	33382	1
16	Hyderabad	110397.0	87429.0	48899	29031	1
17	Dehradun	156487.0	123324.0	59084	39804	1
18	Indore	1176804.0	960145.0	604293	373155	1
19	Pune	383260.0	305902.0	172827	122285	1
20	Bengaluru	268286.0	149476.0	58622	44999	1
21	Hyderabad	292128.0	131890.0	50720	32140	1
22	Dehradun	401292.0	171322.0	51069	36016	1
23	Indore	1670668.0	969732.0	477858	339970	1
24	Pune	1748075.0	854377.0	314289	225823	1
25	Bengaluru	58172.0	44096.0	18169	8769	1
26	Hyderabad	80384.0	58097.0	22564	8832	1
27	Dehradun	101453.0	79201.0	17203	10064	1
28	Indore	494447.0	374638.0	170498	88331	1

56	Hyderabad	50889.0	34069.0	10913	5024	1
57	Dehradun	58247.0	41982.0	10229	6334	1
58	Indore	507337.0	365834.0	176165	57549	1
59	Pune	221607.0	127471.0	39922	22624	1
		…….				
		…….				
		.				
		.				
		.				
60	Bengaluru	NaN	NaN	47265	14196	2
61	Hyderabad	NaN	NaN	33702	9671	2
62	Dehradun	NaN	NaN	43417	15551	2
63	Indore	NaN	NaN	245423	80299	2
64	Pune	NaN	NaN	228051	36538	2
65	Bengaluru	NaN	NaN	78378	20261	2
66	Hyderabad	NaN	NaN	64729	15062	2
67	Dehradun	NaN	NaN	75333	19025	2
68	Indore	NaN	NaN	363559	124359	2
69	Pune	NaN	NaN	441298	82153	2
70	Bengaluru	NaN	NaN	26704	5672	2
71	Hyderabad	NaN	NaN	22736	4181	2
72	Dehradun	NaN	NaN	26907	6179	2
73	Indore	NaN	NaN	115601	31589	2
74	Pune	NaN	NaN	103558	16829	2
75	Bengaluru	98574.0	80672.0	39272	26296	2
76	Hyderabad	95163.0	74427.0	39687	23332	2
77	Dehradun	131647.0	103960.0	46826	31260	2
78	Indore	942288.0	760422.0	462969	278204	2
107	Dehradun	163799.0	109105.0	50685	33708	2
108	Indore	1052420.0	733233.0	404716	232852	2
109	Pune	426286.0	304066.0	156062	108686	2
110	Bengaluru	570318.0	245665.0	56844	25016	2
111	Hyderabad	292515.0	128356.0	40287	19591	2
112	Dehradun	466905.0	193933.0	41464	16147	2
113	Indore	2505768.0	1215226.0	397087	182202	2
114	Pune	1376097.0	719141.0	304051	149451	2
115	Bengaluru	63129.0	37419.0	9099	3503	2
116	Hyderabad	43619.0	28122.0	8379	4058	2
		…….				

[2160 rows * 6 columns]

4.] In column ‘E’ Replace “Came_From_LinkedIn” with “LinkedIn” and “Landed_Directly” with “Direct_traffic” .

Source code:

```
import pandas as pd
df=pd.read_csv("test_DataScience.csv")
df['Coming from'] = df['Coming from'].replace(['Came_From_LinkedIn', 'Landed_Directly'],['LinkedIn', 'Direct_Traffic'])
df['E']=df['Coming from']
df
```

OUTPUT:

Out[56]:

	Year	Month	Laptop/Desktop	Type_of_Customers?	Coming from	Place_in_India	Level 1	Level 2	Level 3	Level 4	E
0	2020	Jan	Desktop_Website	Existing_Customer	LinkedIn	Bengaluru	NaN	NaN	56892	17178	LinkedIn
1	2020	Jan	Desktop_Website	Existing_Customer	LinkedIn	Hyderabad	NaN	NaN	41460	11916	LinkedIn
2	2020	Jan	Desktop_Website	Existing_Customer	LinkedIn	Dehradun	NaN	NaN	55561	19461	LinkedIn
3	2020	Jan	Desktop_Website	Existing_Customer	LinkedIn	Indore	NaN	NaN	320923	110667	LinkedIn
4	2020	Jan	Desktop_Website	Existing_Customer	LinkedIn	Pune	NaN	NaN	220937	46033	LinkedIn
5	2020	Jan	Desktop_Website	Existing_Customer	Direct_Traffic	Bengaluru	NaN	NaN	90241	24229	Direct_Traffic
6	2020	Jan	Desktop_Website	Existing_Customer	Direct_Traffic	Hyderabad	NaN	NaN	77630	18502	Direct_Traffic
7	2020	Jan	Desktop_Website	Existing_Customer	Direct_Traffic	Dehradun	NaN	NaN	91479	24363	Direct_Traffic
8	2020	Jan	Desktop_Website	Existing_Customer	Direct_Traffic	Indore	NaN	NaN	436641	165036	Direct_Traffic
9	2020	Jan	Desktop_Website	Existing_Customer	Direct_Traffic	Pune	NaN	NaN	531446	101317	Direct_Traffic
10	2020	Jan	Desktop_Website	Existing_Customer	Unidentified_Sources	Bengaluru	NaN	NaN	32119	6900	Unidentified_Sources

Part 2: Descriptive statistics

Write a function called descriptive_stats(‘Year’, ‘Month’ , ‘Laptop/Desktop’ , ‘Type_of_Customers?’ , ‘Coming from’ , ‘Place_in_India’) which, when called, would perform the following activity:

Q.1] Would filter the dataframe with the given parameters; if any parameter is missed, then consider a default value to that parameter (e.g., default: ‘year’ – 2020, ‘month’-Jan, & so on) . Let’s call this new dataframe ‘df’.

Source code:

```
import pandas as pd
df=pd.read_csv('test_DataScience.csv')
df['LaptopDesktop']=df['Laptop/Desktop']
df.rename(columns = {'Laptop/Desktop':'LaptopDesktop'}, inplace = True)
df.rename(columns = {'Type_of_Customers?':'Type_of_Customers'}, inplace = True)
df.rename(columns = {'Coming from':'Coming_from'}, inplace = True)
 def descriptive_stats(Year='2020',Month='Jan',LaptopDesktop='Laptop',Type_of_Customers='New',
Coming_from='Socialmedia',Place_in_India='Pune'):
      return df.predict_future
df.Year='2020'
df.Month='jan'
df.LaptopDesktop='Laptop'
df.Type_of_Customers='New'
df.Coming_from='Socialmedia'
df.Place_in_India='Pune'
df
```

OUTPUT:

9]:

	Year	Month	LaptopDesktop	Type_of_Customers	Coming_from	Place_in_India	Level 1	Level 2	Level 3	Level 4	LaptopDesktop
0	2020	jan	Laptop	New	Socialmedia	Pune	NaN	NaN	56892	17178	Laptop
1	2020	jan	Laptop	New	Socialmedia	Pune	NaN	NaN	41460	11916	Laptop
2	2020	jan	Laptop	New	Socialmedia	Pune	NaN	NaN	55561	19461	Laptop
3	2020	jan	Laptop	New	Socialmedia	Pune	NaN	NaN	320923	110667	Laptop
4	2020	jan	Laptop	New	Socialmedia	Pune	NaN	NaN	220937	46033	Laptop
...	...	...	...	...	...	...	...	...	...	...	...
2155	2022	jan	Laptop	New	Socialmedia	Pune	67299.0	21255.0	6984	1882	Laptop
2156	2022	jan	Laptop	New	Socialmedia	Pune	430294.0	156510.0	46676	16703	Laptop
2157	2022	jan	Laptop	New	Socialmedia	Pune	48713.0	27770.0	7515	2089	Laptop
2158	2022	jan	Laptop	New	Socialmedia	Pune	593021.0	310836.0	161575	78465	Laptop
2159	2022	jan	Laptop	New	Socialmedia	Pune	372897.0	123057.0	48802	19441	Laptop

2160 rows × 11 columns

Q.2] Generates the summary statistics (Mean, Median, Quartile, standard deviation) of all the numerical columns of the new dataframe, df.

Source Code:

```
import pandas as pd
df=pd.read_csv('test_DataScience.csv')
df['LaptopDesktop']=df['Laptop/Desktop']
df.rename(columns = {'Laptop/Desktop':'LaptopDesktop'}, inplace = True)
df.rename(columns = {'Type_of_Customers?':'Type_of_Customers'}, inplace = True)
df.rename(columns = {'Coming from':'Coming_from'}, inplace = True)

def descriptive_stats(Year='2020',Month='Jan',LaptopDesktop='Laptop',Type_of_Customers='New',
Coming_from='Socialmedia',Place_in_India='Pune'):
      return df.predict_future
df.Year='2020'
df.Month='jan'
df.LaptopDesktop='Laptop'
df.Type_of_Customers='New'
df.Coming_from='Socialmedia'
df.Place_in_India='Pune'
df
#index_labels=['r1','r2','r3','r4','r5','r6']
print("The shape of the dataframe is: ", df.shape)
#df.describe()
dfnew = pd.DataFrame(df,index=index_labels)
df_mean = dfnew["Year"].mean()
print(df_mean) #calculating mean
print(dfnew.median()) #calculating median
dfnew.std(axis = 1, skipna = True)
# Removing the outliers
def removeOutliers(dfnew, Year):
   Q3 = np.quantile(dfnew[Year], 0.75)
   Q1 = np.quantile(dfnew[Year], 0.25)
   IQR = Q3 - Q1

print("IQR value for column %s is: %s" % (Year, IQR))
global outlier_free_list
global filtered_data
lower_range = Q1 - 1.5 * IQR
upper_range = Q3 + 1.5 * IQR
outlier_free_list = [x for x in data[Year] if (
(x > lower_range) & (x < upper_range))]
filtered_data = df.loc[data[Year].isin(outlier_free_list)]
for i in dfnew.columns:
   if i == dfnew.columns[0]:
      removeOutliers(df, i)
   else:
      removeOutliers(filtered_data, i)
```

```
# Assigning filtered data back to our original variable
dfnew = filtered_data
print("Shape of data after outlier removal is: ", dfnew.shape)
```

OUTPUT:

```
The shape of the dataframe is:  (6, 6)
3.367002366936702e+22
Year    2019.0
dtype: float64
```

Q.3] Produce a list of all the unique values & data types present in the non-numeric columns in df.

Source code:

```
import pandas as pd
import numpy as np
df=pd.read_csv('test_DataScience.csv')
df['LaptopDesktop']=df['Laptop/Desktop']
df.rename(columns = {'Laptop/Desktop':'LaptopDesktop'}, inplace = True)
df.rename(columns = {'Type_of_Customers?':'Type_of_Customers'}, inplace = True)
df.rename(columns = {'Coming from':'Coming_from'}, inplace = True)

def predict_future(Year='2020',Month='Jan',LaptopDesktop='Laptop',Type_of_Customers='New',
Coming_from='Socialmedia',Place_in_India='Pune'):
    return df.predict_future
df.Year='2020'
df.Month='jan'
df.LaptopDesktop='Laptop'
df.Type_of_Customers='New'
df.Coming_from='Socialmedia'
df.Place_in_India='Pune'
df
df.applymap(np.isreal).all(1)  #if all values are false then they are non-numeric.
df[~df.applymap(np.isreal).all(1)]
print(df.Place_in_India.unique())
print(df.LaptopDesktop.unique())
print(df.Coming_from.unique())
print(df.Type_of_Customers.unique())
print(pd.unique(df['Year']))
```

OUTPUT:

```
print(df.Coming_from.unique())
print(df.Type_of_Customer.unique())
print(pd.unique(df['Year']))
```

```
['Pune' 'Gujarat' 'Delhi' 'Mumbai' 'Solapur' 'Kolkata']
['Laptop' 'Desktop' 'PC']
['LinkedIN' 'Sources']
['Existing' 'New']
['2020' '2014' '2016' '2021' '2022' '2018']
```

Part 3: Prescriptive statistics

The marketing manager has asked you the following questions, please provide the answers along with summarized data supporting your answer.

1] What are the top 3 "Place_in_India" on the basis of column "Level 1" for the year 2021 and 2022 separately ?

Source code:

```
import pandas as pd
import numpy as np
df=pd.read_csv("test_DataScience.csv")
df.sort_values(['Level 1','Place_in_India'],ascending = False).groupby('Level 1').head(5)
```

OUTPUT:

	Year	Month	Laptop/Desktop	Type_of_Customers?	Coming from	Place_in_India	Level 1	Level 2	Level 3	Level 4
984	2021	May	Desktop_Website	New_Customer	Landed_Directly	Pune	11274131.0	2544078.0	658397	389191
1764	2022	Jun	Desktop_Website	New_Customer	Landed_Directly	Pune	9083552.0	4079301.0	1942557	923720
2064	2022	Nov	Desktop_Website	New_Customer	Landed_Directly	Pune	9036434.0	3881092.0	1573991	119167
924	2021	Apr	Desktop_Website	New_Customer	Landed_Directly	Pune	8949571.0	1932569.0	600182	400768
1284	2021	Oct	Desktop_Website	New_Customer	Landed_Directly	Pune	8188402.0	3435272.0	862600	558073

Q.2 Please, provide the data for all the cities & for all the years, the following format as shown in the below snippet:

Source code:

```
df2 = df.groupby('Place_in_India').sum()
df2
df2['Sum of level 2/Sum of level 1'] = df2['Level 2']/df2['Level 1']
df2['Sum of level 3/Sum of level 1'] = df2['Level 3']/df2['Level 1']
df2['Sum of level 4/Sum of level 1'] = df2['Level 4']/df2['Level 1']
df2
```

OUTPUT:

Place_in_India	Year	Level 1	Level 2	Level 3	Level 4	Sum	Sum of level 2/Sum of level 1	Sum of level 3/Sum of level 1	Sum of level 4/Sum of level 1
Bengaluru	873072	51255804.0	24113122.0	22121810	10124260	51255804.0	0.470447	0.431596	0.197524
Dehradun	873072	62484684.0	25943314.0	22056792	8804705	62484684.0	0.415195	0.352995	0.140910
Hyderabad	873072	132052059.0	62128893.0	50639098	21204313	132052059.0	0.470488	0.383478	0.160575
Indore	873072	282329031.0	153724091.0	134367335	52730177	282329031.0	0.544486	0.475925	0.186769
Pune	873072	319242132.0	121321445.0	97131570	35054534	319242132.0	0.380030	0.304257	0.109805

Q.3] What are the bottom 3 "Place_in_India" on the basis of column "Level 4" for the year 2021 and 2022 separately ?

Source code:

```
import pandas as pd
import numpy as np
df=pd.read_csv("test_DataScience.csv")
df.sort_values(['Level 4','Place_in_India'],ascending = True).groupby('Level 4').head(5)
```

OUTPUT:

	Year	Month	Laptop/Desktop	Type_of_Customers?	Coming from	Place_in_India	Level 1	Level 2	Level 3	Level 4
1482	2022	Jan	Laptop_Website	Existing_Customer	Unidentified_Sources	Dehradun	NaN	NaN	8901	766
1422	2021	Dec	Laptop_Website	Existing_Customer	Unidentified_Sources	Dehradun	NaN	NaN	10240	860
1480	2022	Jan	Laptop_Website	Existing_Customer	Unidentified_Sources	Bengaluru	NaN	NaN	6711	1040
1870	2022	Aug	Desktop_Website	Existing_Customer	Unidentified_Sources	Bengaluru	NaN	NaN	4422	1070
1930	2022	Sep	Desktop_Website	Existing_Customer	Unidentified_Sources	Bengaluru	NaN	NaN	4927	1077

Part 4: Simple Machine learning question:

Write a function called predict_future('Year', 'Month' , 'Laptop/Desktop' , 'Type_of_Customers?' , 'Coming from' , 'Place_in_India') which, when called, would perform the following activity:

Q1.]Predict "Level 4" for the 12 months of 2023 given the parameters of the function. (Please make sure the parameters have default values in place)

Source code:

```
import pandas as pd
import numpy
df=pd.read_csv('test_DataScience.csv')
df['LaptopDesktop']=df['Laptop/Desktop']
df.rename(columns = {'Laptop/Desktop':'LaptopDesktop'}, inplace = True)
df.rename(columns = {'Type_of_Customers?':'Type_of_Customers'}, inplace = True)
df.rename(columns = {'Coming from':'Coming_from'}, inplace = True)

def predict_future(Year='2020',Month='Jan',LaptopDesktop='Laptop',Type_of_Customers='New',
Coming_from='Socialmedia',Place_in_India='Pune'):
    return df.predict_future
df.Year='2020'
df.Month='jan'
df.LaptopDesktop='Laptop'
df.Type_of_Customers='New'
df.Coming_from='Socialmedia'
df.Place_in_India='Pune'
df
#df[(df['Level 4'].dt.month == 1) & (df['Level 4'].dt.day == 1)].mean()
df = df.groupby(by=[df.index.Year, df.index.Level4]).mean()
```

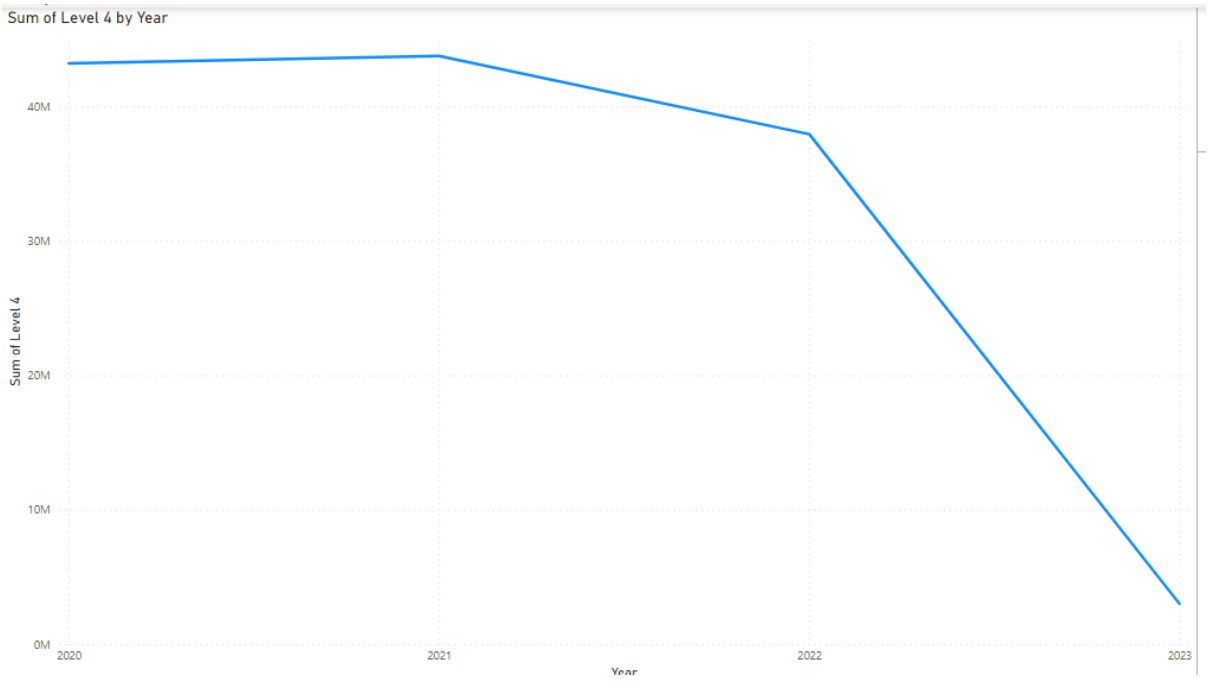
Q.2]Generates the overall Forecast error, MAPE and RMSE of your prediction of the year 2022, 2021 & 2020 for the given parameters.

Source code:

```
import numpy as np
from sklearn.model_selection import train_test_split
import pandas as pd
df=pd.read_csv('test_DataScience.csv')
df['LaptopDesktop']=df['Laptop/Desktop']
df.rename(columns = {'Laptop/Desktop':'LaptopDesktop'}, inplace = True)
df.rename(columns = {'Type_of_Customers?':'Type_of_Customers'}, inplace = True)
df.rename(columns = {'Coming from':'Coming_from'}, inplace = True)
  def predict_future(Year='2020',Month='Jan',LaptopDesktop='Laptop',Type_of_Customers='New',
Coming_from='Socialmedia',Place_in_India='Pune'):
    return df.predict_future
df.Year='2020'
df.Month='jan'
```

```
df.LaptopDesktop='Laptop'
df.Type_of_Customers='New'
df.Coming_from='Socialmedia'
df.Place_in_India='Pune'
df
 #Separating the dependent and independent data variables into two data frames.
X = df.drop(['Year'],axis=1)
Y = df['Year']
 # Splitting the dataset into 80% training data and 20% testing data.
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=.20, random_state=0)
 #Defining MAPE function
def MAPE(Y_actual,Y_Predicted):
    mape = np.mean(np.abs((Y_actual - Y_Predicted)/Y_actual))*100
    return mape
#Building the Linear Regression Model
from sklearn.linear_model import LinearRegression
linear_model = LinearRegression().fit(X_train , Y_train)
 #Predictions on Testing data
LR_Test_predict = linear_model.predict(X_test)
 # Using MAPE error metrics to check for the error rate and accuracy level
LR_MAPE= MAPE(Y_test,LR_Test_predict)
print("MAPE: ",LR_MAPE)
```

Q.3] Plot a line graph of the level 4 actual numbers from 2020-2022 & in the same graph, there should be the predicted numbers for 2023. The x-axis should be the timeline from 2020 Jan to 2023 Dec and the y-axis should be the value of the level 4 column, The below graph is just an example of how your plot should look like.



Part 5: Visualization:

- Please write a code to display :

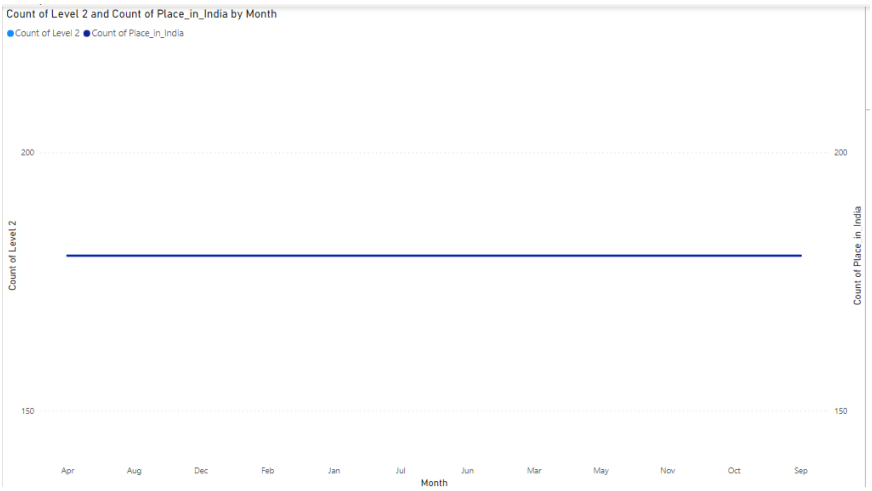
  Q1] A line graph for "Level 2" for the different "Place_in_India?" over the months of the year 2020 & 2021. (Hint: On x-axis, there should be months for 2020 & 2021 and Y axis should be "Level 2" and there should be different lines depicting different regions of "Place_in_India?") Plot a neat graph.

  Source code:

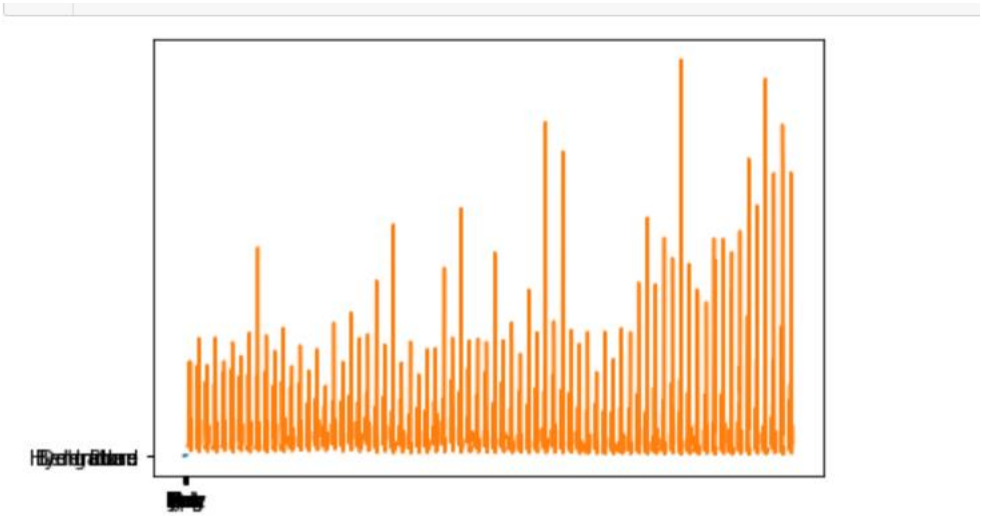
```
import pandas as pd
import matplotlib.pyplot as plt
df=pd.read_csv('test_DataScience.csv')
df = df.head()
data = pd.DataFrame(df, columns=["Month", "Level 2", "Place_in_India"])
data=data.loc["2020":"2021"]
# plot the dataframe
data.plot(x="Month", y=["Level 2", "Place_in_India"], kind="bar", figsize=(9, 8))
# print bar graph
plt.show()
```

  OUTPUT:

  From PowerBI



From Jupyter notebook:

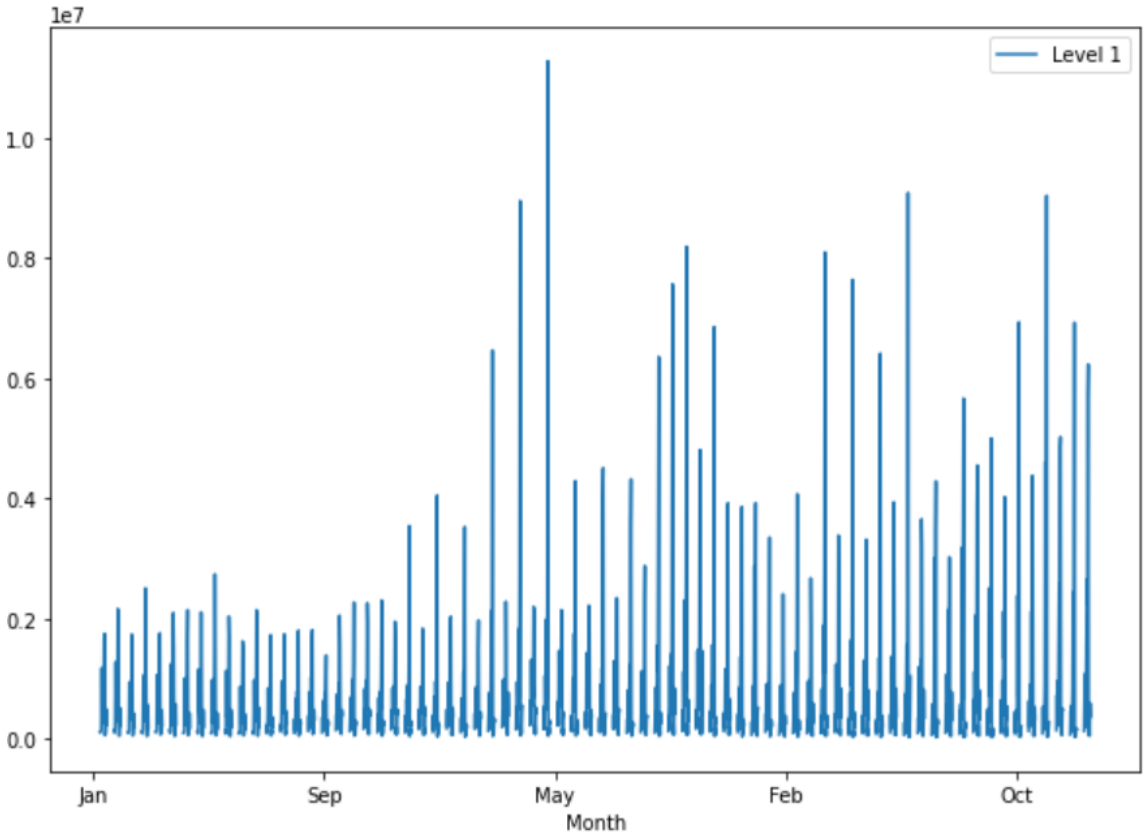


Q2] A line graph for “Level 1” for the different “Laptop/Desktop” over the months of the year 2020 & 2021. (Hint : On x axis there should be months from jan- 2020 to dec- 2021 and Y axis should be the sum of “Level 1” and there should be different lines depicting different devices used.)

Source code:

```
import pandas as pd
import matplotlib.pyplot as plt
df=pd.read_csv('test_DataScience.csv')
#df = df.head()
data = pd.DataFrame(df, columns=["Month", "Level 1", "Laptop/Desktop"])
data=data.loc["2020":"2021"]
data=data.loc[,:"Jan":"Dec"]
# plot the dataframe
data.plot(x="Month", y=["Level 1", "Laptop/Desktop"], kind="line", figsize=(10, 7))
# print bar graph
plt.show()
```

OUTPUT:

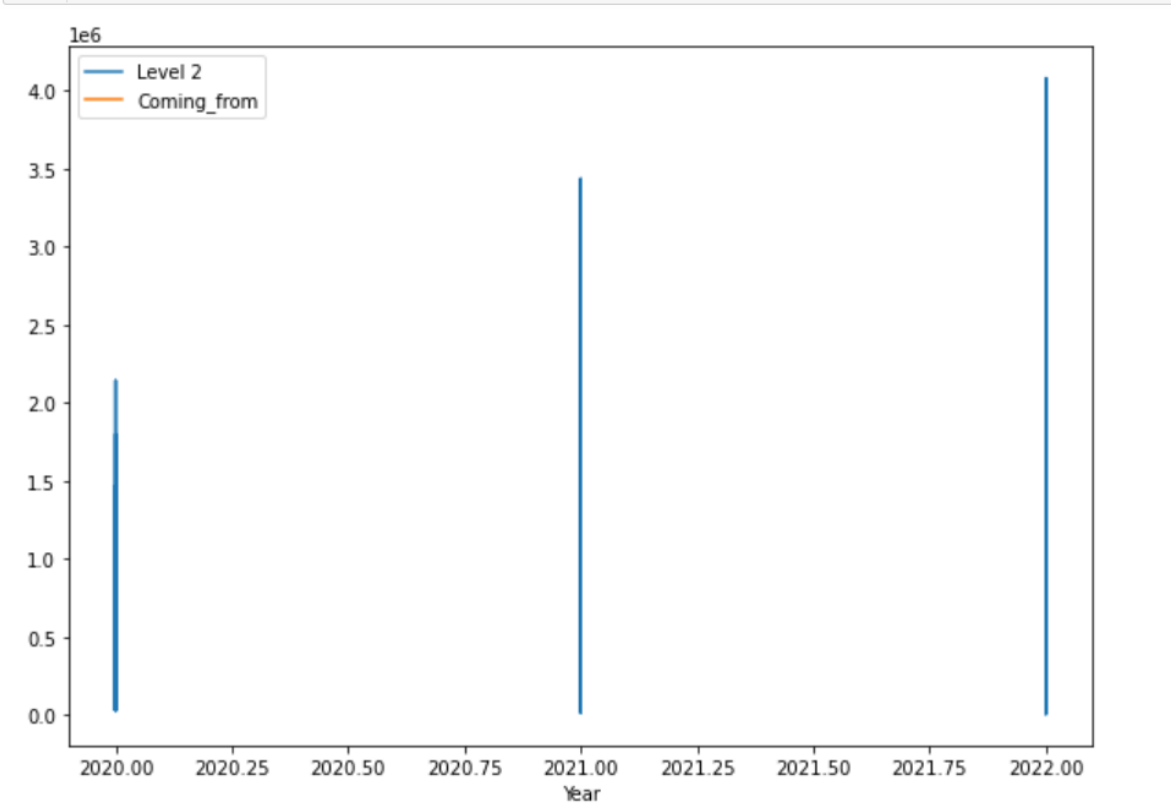


Q3] A line graph for “Level 2” for the different “Coming from” over the months of the year 2021 & 2022.

Source code:

```
import pandas as pd
import matplotlib.pyplot as plt
df=pd.read_csv('test_DataScience.csv')
#df = df.head()
data = pd.DataFrame(df, columns=["Year", "Level 2", "Coming_from"])
data=data.loc["2020":"2021"]
# plot the dataframe
data.plot(x="Year", y=["Level 2", "Coming_from"], kind="line", figsize=(10, 7))
# print bar graph
plt.show()
```

OUTPUT:

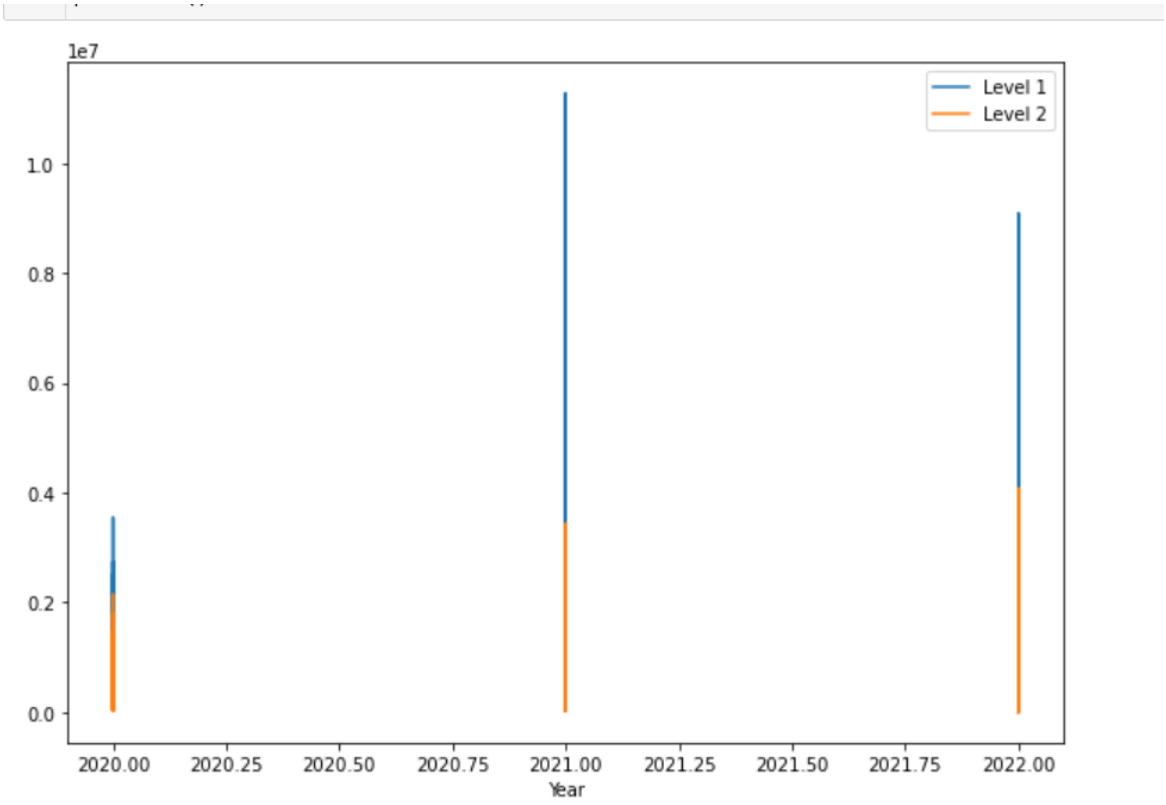


Q4] A line graph for “Level 1” and “Level 2” over the months of the year 2020, 2021 & 2022.

Source code:

```
import pandas as pd
import matplotlib.pyplot as plt
df=pd.read_csv('test_DataScience.csv')
#df = df.head()
data = pd.DataFrame(df, columns=[“Year”, “Level 1”, “Level 2”])
data=data.loc["2020":"2022"]
# plot the dataframe
data.plot(x=”Year”, y=[“Level 1”, “Level 2”], kind=”line”, figsize=(10, 7))
# print bar graph
plt.show()
```

OUTPUT:

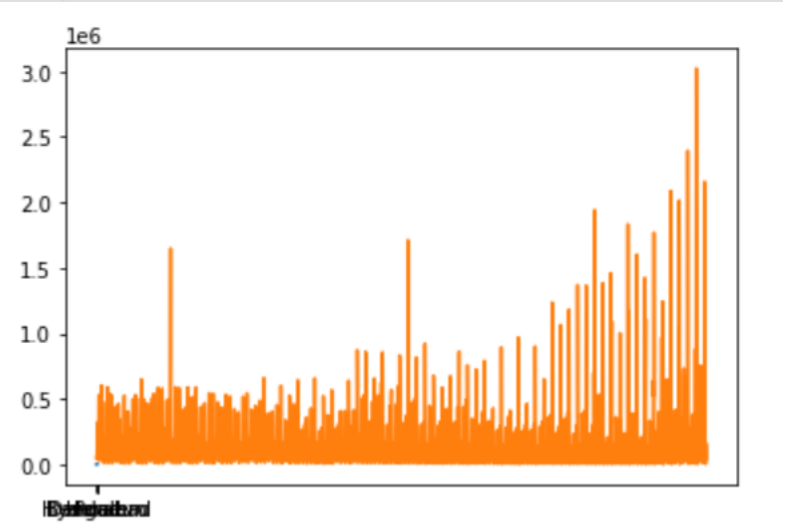


Q5] A line graph for “Level 3” foyearslace_in_India” over the months of the year 2020 and 2021.

Source code:

```
import pandas as pd
import matplotlib.pyplot as plt
df=pd.read_csv('test_DataScience.csv')
df=df.loc["2020":"2021"]
plt.plot( df["Place_in_India"],df["Year"], df["Level 3"])
plt.show()
```

OUTPUT:



Q6] Please add any insights you could derive from all the graphs above.

Answer: As we can see, the market was high during pandemic that is in 2021. Most of the users are from Pune and Bengaluru. And sales are high in the month of May-2021.

Part 6: About the Previous projects

● Please describe any interesting project you did in the Data Science domain in more than 250 words. Attach Github links if possible.

Answer: I have built a database in Healthcare, Banking, Ashram database using MySQL server. I'll share my drive link which includes complete database.

In June-2022, our team worked on database to build fitness freaks website. We collected data and segregated it based on different weight type. And recommended diet accordingly. We worked on bionic sensors which take input as muscle strain and through API, this data is saved on cloud and reports are generated accordingly. The report includes weekly diet, exercises according to BMI. Sport biomechanics represents an important research field aimed at analysing sport movements in order to quantitatively evaluate athlete performance, offer useful tools and guidelines for coaches to apply during athlete training and prevent or minimize the risk of injury. Recent technological innovations allow the performance of movement analysis during sporting activities thanks to the compact wearable sensors that do not influence the technical movements of athletes. The aim of this project is to present the design and development of a wearable multi-sensor system that is affordable for all types of users and can be used for a long time for the application of exercise monitoring. Wearable sensors are widely used in healthcare, due to their hardware capacity, small footprint and lower cost compared to equivalent medical instruments capable of monitoring the same vital signs.Our device includes combination of sensors that is MC sensors, Strain sensors and cloud.

I have also designed graphs, gantt chart, stacked column charts, pert chart, pie chart,etc in Power BI Desktop Visualisation tool.

There is one more project on which me and my team is working, basically we are collecting database. Processing and filtering will be done in the month of January, registered for SIH-2023.

Part 7: Time management

I managed time to solve this assignment after college hours (9:30 to 5:30-college). The same way I'll adjust and prepare new schedule for this internship. I will work from 6:00pm. I will complete my academics work and projects in college hours. And after college can work efficiently till night, at least for 6hours everyday. I will make sure that my work doesn't get affected due to academics. As there have been many incidents where I have been working on multiple thinks. I will adjust and make separate schedule for daily tasks. However we learn same concepts in college session, so this internship will more be like implementing what I have learnt throughout these 3 years of engineering. Its fun working on things which you love.