# EXPLORATORY DATA ANALYSIS (EDA) REPORT

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Exploratory Data Analysis (EDA) is an approach of analyzing data sets to summarize their main characteristics, often using statistical graphics and other data visualization methods.

In our case, I have found counts of different categorical data and also been able to make some visual data analysis using pandas and matplotlib.

#### Dataset "a.csv"

This dataset contains 7 variables(features) which are of 2 types, i.e., categorical data and numeric data.

- <u>Categorical Data</u> status, pay\_mode, product, marker, type
- Numeric Data log time, phone

#### Unique values in 'status' columns are:

```
['assigned' 'purchase' 'AA' 'AB' 'AC' 'AD' 'Switched off' 'Not interested' 'Already purchased' 'Follow-up later' 'User is Interested' 'Converted' 'New product potential' 'Line Busy' 'Has complaints' 'Invalid Number' 'Not reachable' 'Partially interested' 'Other' 'Assigned' 'Could not call' 'Not picking up' 'none']
```

Assigned	516435
purchase	261558
•	
Not picking up	61454
Not interested	47341
Partially interested	19780
Follow-up later	18244
User is Interested	17756
Not reachable	9774
Invalid Number	9327
Line Busy	8628
Already purchased	8123
Switched off	6316
New product potential	5938
Converted	5579
Other	1655
Could not call	632
assigned	252
Has complaints	22
AC	2
AB	2
AA	2
AD	1
none	1
Name - Charles - dr. mar - Santo	

Name: status, dtype: int64

# Unique values in 'type" columns are:

[1002 1001 1003 1005 1004 1006 2005 2209 2001 2102 2103 2104 2206 2207 2208 2010 2011 2012 2013 2105 2106]

```
1002
       516685
1001
        261752
2106
        61455
2005
        47341
2208
       19780
2206
        18242
2207
        17753
2102
         9774
2001
          9327
2103
          8627
2010
          7927
2104
          6315
2209
          5937
2011
          5578
2013
          1656
2105
           632
2012
            21
1005
             8
1003
             7
1004
             4
1006
             1
Name: type, dtype: int64
```

#### Unique values in 'pay\_mode' columns are:

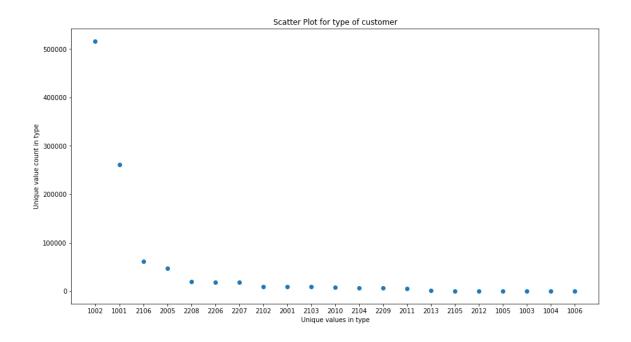
```
[nan 'cc-dc' 'cp-nb' 'PayPal' 'ptm' 'upi' 'paypal' 'wallet' 'UPI' 'WL'
'PTM' 'gbp' 'RY' 'paycancel_cc-dc_HDFC' 'rupay' 'RU' 'MN2942952864BEB'
'AVAC125921Dfc' 'MN294346739b3aa' 'INVC1557ZTXO' 'USD' 'usd'
'MN294496345f5db' 'gpay' 'MN294568094c4ab' 'rz' 'MN294589903D2FB'
'MN29467013811BB' 'MN294695472D3CD' 'MN2946970583D2d' 'AVFD1359951acb5'
'MN294783007CC1b' 'no_mode' 'jpay' 'MN294802793d1F3' 'MN294811647ac31'
'MN294854885AC1c' 'GPAY' 'MN294955924cd1A' 'paytm' 'MN295018279AA1e'
'CP-NB' 'CC-DC' 'paycancel_WL_PAYU' 'PhonePe' 'PHONEPE' 'CC' 'RZ' 'bank'
'paycancel_CC_HDFC' 'invoice' 'NB' 'paycancel_NB_INST'
'paycancel_CC_HDFC' 'paycancel_CC_INST' 'paycancel_NB_CITRUS' 'PTMMini'
'paycancel_CC_RZ' 'paycancel_CC_PAYU' 'paycancel_NB_RZ'
'paycancel_NB_PAYU' 'paycancel_UPI_RZ' 'PYPL']
```

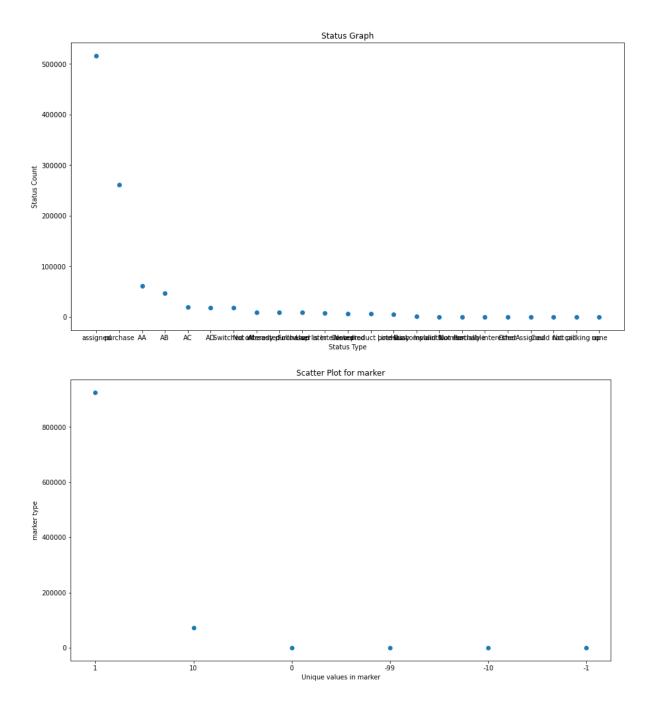
cc-dc	79030		
cp-nb	57755		
gpay	18825		
UPI	12459		
ptm	12332		
MN294955924cd1A	1		
MN2942952864BEB	1		
MN295018279AA1e	1		
CP-NB	1		
PYPL	1		
Name: pay mode,	Length: 62,	dtype:	int64

## Unique values in 'marker' columns are:

[ 6	1 -99	-1 -10	10]
1	925749		
10	72274		
0	488		
-99	273		
-10	36		
-1	2		
Name	· marker	dtvne:	int6

From the above dataset, some of the visualizations that can be made are:-





# Dataset "b.csv"

This dataset contains 5 variables, i.e., uuid, beacon\_type, beacon\_value, log\_data, status.

- <u>Categorical Data</u> beacon\_type, status
- Numeric Data uuid, log\_date, beacon\_value

The 'status' column contains just one value of 1 for all rows without any null values.

# The data summary can be described using describe() to get some of the understanding from data.

```
In [114]:
            df.describe()
Out[114]:
                            uuid beacon value
                                                    status
             count 3.900932e+07
                                  3.900933e+07 39009332.0
             mean 4.922558e+06
                                  5.610764e+00
                                                       1.0
               std
                   2.985628e+06
                                  1.102048e+01
                                                       0.0
                   0.000000e+00
                                 1.000000e+00
                                                       1.0
               min
                   2.232751e+06
                                 1.000000e+00
                                                       1.0
              50%
                   4.901919e+06
                                  2.000000e+00
                                                       1.0
              75% 7.526520e+06
                                  4.000000e+00
                                                       1.0
              max 1.005815e+07 9.990000e+02
                                                       1.0
```

#### The unique values present in the 'beacon\_type' column are:

```
array(['user stay', 'bottom banner', 'buy button FH', 'pay button cc-dc',
          pay_button_cp-nb', 'pay_button_paypal', 'buy_button_PP',
          'masked_content', 'buy_button_top', 'buy_button_autoPopupE',
          'buy_button_autopopup90p', 'pay_button_upi', 'pay_button_ptm',
          'pay_button_wallet', 'buy_button_InReportMaskToast',
          'buy_button_AdBannerRight', 'purchased in-depth,AVOL3218525210124',
          'pay_button_gbp', 'buy_button_CR', 'pay_button_undefined',
         'pay_button_PP', 'pay_fail', 'buy_button_undefined',
'pay_button_usd', 'pay_button_gpay', 'mask_area', 'top_banner',
'pay_button_rz', 'pay_button_PayPal', 'pay_button_RZ',
          'pay_button_PTM', 'pay_button_IM', 'pay_button_WL', 'pay_button_NB', 'pay_button_CC', 'pay_button_PYPL'
          'pay_button_RU', 'pay_button_RY', 'pay_button_UPI', 'pay_button_',
          'pay_button_ebs', 'pay_button_EBSG', 'pay_button_jpay',
          'pay_button_GPAY', 'buy_button_WL', 'pay_button_PAYU',
         'pay_button_rupay', 'pay_button_avupi', 'pay_button_juspay', 'pay_button_ebsg', 'pay_button_CTRS', 'pay_button_hdfc', 'pay_button_USD', 'pay_button_paytm', 'pay_button_EUR',
          'pay_button_PTMMini', 'social_proof_banner', 'pay_button_phonepe', 'pay_button_PhonePe', 'pay_button_HDFC', 'pay_button_EBS',
          'pay_button_payu', '6', nan, 'pay_button_NA', 'pay_button_invoice',
          'pay_button_bank'], dtype=object)
```

# Dataset "c.csv"

This dataset consists of 5 variables only in the form of numerical data.

Different columns of this dataset contain information regarding users about their email, phone numbers, etc.

#### Null values in the data are:

Visualizations that can be made on this dataset are between user id (x axia) and their profile\_submit\_count (y-axis).

```
In [34]: l=df['id'].to_list()
l1=df['profile_submit_count'].to_list()
plt.scatter(l, l1)
plt.show()
```

## **Dataset "ct.csv"**

This dataset contains 5 variables.

- Categorical data status
- Numerical data id, cid, timestamp, amount

#### The 'status' column contains status of the payment process as:

#### Counts of different categories in 'status' column are:

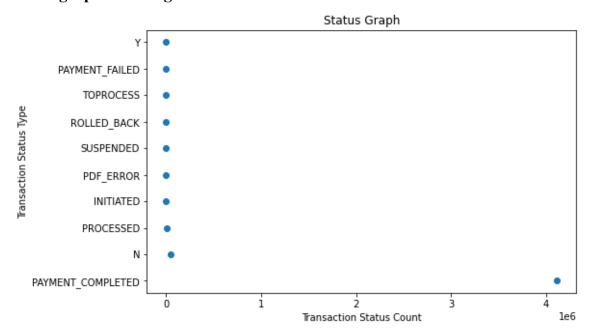
N	4115487
PROCESSED	48328
PAYMENT_COMPLETED	9226
SUSPENDED	526
INITIATED	279
ROLLED_BACK	85
PDF_ERROR	55
PAYMENT_FAILED	16
TOPROCESS	10
Υ	1
Name: status, dtype:	int64

#### Statistical Summary about the data is:

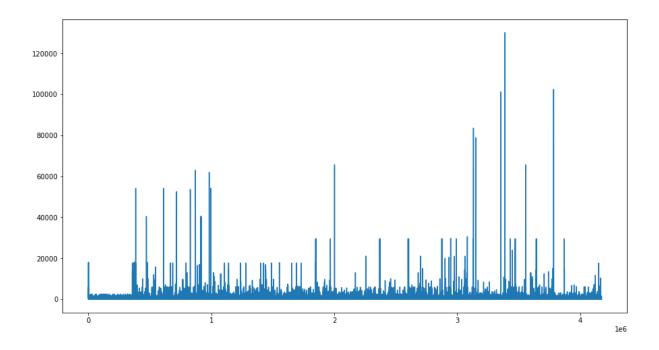
	id	cid	amount
count	4.174013e+06	4.174013e+06	4.174013e+06
mean	2.087048e+06	1.767756e+06	3.592679e+01
std	1.204940e+06	1.375565e+06	3.090433e+02
min	4.000000e+00	1.000000e+00	0.000000e+00
25%	1.043544e+06	5.536010e+05	0.000000e+00
50%	2.087049e+06	1.517702e+06	0.000000e+00
75%	3.130556e+06	2.748210e+06	0.000000e+00
max	4.174059e+06	4.867896e+06	1.301137e+05

#### Visualizations that can be made from this data are:

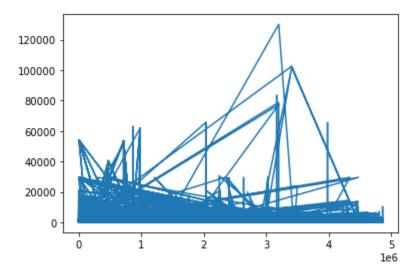
#### Scatter graph for categorical data in 'status' column.



Plot for id vs amount is:



#### Plot for cid vs amount is:



# Dataset "s.csv"

This dataset contains 10 variables.

• Categorical Data - language, status, report\_type, device

Categories and their counts in each of categorical variable are:

#### 'gender':

Male 5336491 Female 3722945 MALE 31319 M 45 FEMALE 20 F 17

Name: gender, dtype: int64

#### 'language':

TAM 2842756 ENG 2096781 HIN 1399821 KAN 666861 MAL 644473 TEL 635283 304390 MAR BEN 234800 ORI 218320 GUJ 44658 SIN 7048 Н 4 3 NIL HINDI 2 SAN 2 Е 1 1

Name: language, dtype: int64

#### <u>'status'</u>:

1 9095602

Name: status, dtype: int64

#### 'report\_type':

```
LS-MT 7908843
LS-MP
           680885
            243535
LS-CR
LS-FH
             40034
LS-SC
             34223
1029-ASK-1
                 1
1029-GEN-P
                 1
MPYS
                1
CMYM
1029-TEL-C
                1
Name: report_type, Length: 81, dtype: int64
```

#### 'device':

mobile 8165158 pc 914283 desktop 15513 MOBILE 275 PC 186

Name: device, dtype: int64

# **Dataset "tp.csv"**

Contains only categorical data with ctid as numerical variables.

Categories and their counts in each of categorical variable are:

#### 'variant':

basic 4	008071
premium	168432
premiumplus	2134
premiumpluscolor	370
premiumplusconsultancy	10
premiumplusleather	7
Name: variant, dtype: int64	

#### 'language':

eng	1253528
hin	995944
tam	643472
mal	309591
tel	293184
kan	225973
mar	171512
ben	130427
ori	87416
guj	42696
sin	18163
Eng	2679
Hin	799
Tam	459
Mal	336
Tel	291
Kan	205
Mar	177
Ben	138
0ri	44
Guj	44
Sin	34
ENG	33
en	18
0	14
nil	12
san	5
m	2
Hi	2
	2
	_

Name: language, dtype: int64

## <u>'status':</u>

50085
577
253
85
55

Name: status, dtype: int64