

Loading 1st dataset which is CRX data

In [1]:

```

1 # importing essential libraries to do the following task.
2 import numpy as np
3 import pandas as pd

```

In [2]:

```

1 # crx_data = pd.read_csv("crx.data",names=["col"+str(i) for i in range(0,16)])
2 crx_data = pd.read_csv("crx.data",names=["col"+str(i+1) for i in range(0,16)])

```

In [3]:

```

1 # visualizing the first 5 rows to know whether the data is loaded in correct manner or not
2 crx_data.head(5)

```

Out[3]:

	col1	col2	col3	col4	col5	col6	col7	col8	col9	col10	col11	col12	col13	col14	col15
0	b	30.83	0.000	u	g	w	v	1.25	t	t	1	f	g	00202	
1	a	58.67	4.460	u	g	q	h	3.04	t	t	6	f	g	00043	56
2	a	24.50	0.500	u	g	q	h	1.50	t	f	0	f	g	00280	82
3	b	27.83	1.540	u	g	w	v	3.75	t	t	5	t	g	00100	
4	b	20.17	5.625	u	g	w	v	1.71	t	f	0	f	s	00120	

Displaying the last 10 rows of crx_data

In [4]:

```

1 # tail is used to denote the values from last
2 crx_data.tail(10)

```

Out[4]:

	col1	col2	col3	col4	col5	col6	col7	col8	col9	col10	col11	col12	col13	col14	col15
680	b	19.50	0.290	u	g	k	v	0.290	f	f	0	f	g	00280	
681	b	27.83	1.000	y	p	d	h	3.000	f	f	0	f	g	00176	
682	b	17.08	3.290	u	g	i	v	0.335	f	f	0	t	g	00140	
683	b	36.42	0.750	y	p	d	v	0.585	f	f	0	f	g	00240	
684	b	40.58	3.290	u	g	m	v	3.500	f	f	0	t	s	00400	
685	b	21.08	10.085	y	p	e	h	1.250	f	f	0	f	g	00260	
686	a	22.67	0.750	u	g	c	v	2.000	f	t	2	t	g	00200	
687	a	25.25	13.500	y	p	ff	ff	2.000	f	t	1	t	g	00200	
688	b	17.92	0.205	u	g	aa	v	0.040	f	f	0	f	g	00280	
689	b	35.00	3.375	u	g	c	h	8.290	f	f	0	t	g	00000	

Replace the '?' with Not-a-Number

In [7]: 1 crx_data.replace('?',np.nan)

Out[7]:

	col1	col2	col3	col4	col5	col6	col7	col8	col9	col10	col11	col12	col13	col14
0	b	30.83	0.000	u	g	w	v	1.25	t	t	1	f	g	00202
1	a	58.67	4.460	u	g	q	h	3.04	t	t	6	f	g	00043
2	a	24.50	0.500	u	g	q	h	1.50	t	f	0	f	g	00280
3	b	27.83	1.540	u	g	w	v	3.75	t	t	5	t	g	00100
4	b	20.17	5.625	u	g	w	v	1.71	t	f	0	f	s	00120
...
685	b	21.08	10.085	y	p	e	h	1.25	f	f	0	f	g	00260
686	a	22.67	0.750	u	g	c	v	2.00	f	t	2	t	g	00200
687	a	25.25	13.500	y	p	ff	ff	2.00	f	t	1	t	g	00200
688	b	17.92	0.205	u	g	aa	v	0.04	f	f	0	f	g	00280
689	b	35.00	3.375	u	g	c	h	8.29	f	f	0	t	g	00000

690 rows × 16 columns

In [8]: 1 crx_data.isnull.sum()

AttributeError Traceback (most recent call last)
 <ipython-input-8-de0b73093ffc> in <module>
 ----> 1 crx_data.isnull.sum()

AttributeError: 'function' object has no attribute 'sum'

In [9]: 1 crx_data.isnull().sum()

Out[9]:

col1	0
col2	0
col3	0
col4	0
col5	0
col6	0
col7	0
col8	0
col9	0
col10	0
col11	0
col12	0
col13	0
col14	0
col15	0
col16	0
	dtype: int64

In [10]: 1 crx_data=crx_data.replace('?',np.nan)

In [11]: 1 crx_data.isnull().sum()

Out[11]: col1 12
 col2 12
 col3 0
 col4 6
 col5 6
 col6 9
 col7 9
 col8 0
 col9 0
 col10 0
 col11 0
 col12 0
 col13 0
 col14 13
 col15 0
 col16 0
 dtype: int64

Comment on the datatype of variables

In [12]: 1 # the info method of pandas dataframe gives detailed information about the columns and their datatypes
 2 crx_data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 690 entries, 0 to 689
Data columns (total 16 columns):
 #   Column Non-Null Count Dtype  
 --- 
 0   col1    678 non-null   object 
 1   col2    678 non-null   object 
 2   col3    690 non-null   float64 
 3   col4    684 non-null   object 
 4   col5    684 non-null   object 
 5   col6    681 non-null   object 
 6   col7    681 non-null   object 
 7   col8    690 non-null   float64 
 8   col9    690 non-null   object 
 9   col10   690 non-null   object 
 10  col11   690 non-null   int64  
 11  col12   690 non-null   object 
 12  col13   690 non-null   object 
 13  col14   677 non-null   object 
 14  col15   690 non-null   int64  
 15  col16   690 non-null   object 
dtypes: float64(2), int64(2), object(12)
memory usage: 86.4+ KB
```

In [18]:

```
1 for i in crx_data:
2     print(type(i))
```

```
<class 'str'>
```

In [17]:

```
1 # as the above code displayed the names of column it prints column's name type which we can use to identify the type of data
2 for i in crx_data:
3     print(crx_data[i].dtype)
```

```
object
object
float64
object
object
object
object
float64
object
object
int64
object
object
object
int64
object
```

In [19]:

```
1 float_,int_,object_ = [],[],[]
2 for i in crx_data:
3     if crx_data[i].dtype == "object":
4         object_.append(i)
5     elif crx_data[i].dtype == "float64":
6         float_.append(i)
7     elif crx_data[i].dtype == "int64":
8         int_.append(i)
9     else:
10        print(i,"something went wrong")
```

```
In [27]: 1 print("there are", len(float_), "columns having datatype as float and they are :- ",end=" ")
2 for i in float_:
3     print(i,end=" , ")
4 print("\nthere are", len(int_), "columns having datatype as int and they are :- ",end=" ")
5 for i in int_:
6     print(i,end=" , ")
7 print("\nthere are", len(object_), "columns having datatype as string and they are :- ",end=" ")
8 for i in object_:
9     print(i,end=" , ")
```

there are 2 columns having datatype as float and they are :- col3 , col8 ,
 there are 2 columns having datatype as int and they are :- col11 , col15 ,
 there are 12 columns having datatype as string and they are :- col1 , col2 , col4 , col5 , col6 ,
 col7 , col9 , col10 , col12 , col13 , col14 , col16 ,

The col16 has + and -, replace them 'P' and 'N' respectively

```
In [28]: 1 crx_data['col16'].head()
```

```
Out[28]: 0   +
1   +
2   +
3   +
4   +
Name: col16, dtype: object
```

```
In [30]: 1 # checking though if any another value exists or not
2 crx_data["col16"].unique()
```

```
Out[30]: array(['+', '-'], dtype=object)
```

```
In [31]: 1 # using replace method to replace + with P and - with N
2 crx_data["col16"] = crx_data["col16"].replace("+","P").replace("-","N")
```

```
In [32]: 1 crx_data["col16"]
```

```
Out[32]: 0    P
1    P
2    P
3    P
4    P
..
685   N
686   N
687   N
688   N
689   N
Name: col16, Length: 690, dtype: object
```

In [33]: 1 crx_data["col16"].unique()

Out[33]: array(['P', 'N'], dtype=object)

Find and display the number of variables of type 'Object'

In [34]: 1 print("the object types columns are :-")
2 for i in object_:
3 print(i)

the object types are :-

col1
col2
col4
col5
col6
col7
col9
col10
col12
col13
col14
col16

loading 2nd Dataset which is loan.csv

In [35]: 1 loan_data = pd.read_csv("loan.csv")

In [36]: 1 loan_data.head(5)

Out[36]:

	customer_id	disbursed_amount	interest	market	employment	time_employed	householder
0	0	23201.5	15.4840	C	Teacher	<=5 years	RENT
1	1	7425.0	11.2032	B	Accountant	<=5 years	OWNER 1
2	2	11150.0	8.5100	A	Statistician	<=5 years	RENT
3	3	7600.0	5.8656	A	Other	<=5 years	RENT 1
4	4	31960.0	18.7392	E	Bus driver	>5 years	RENT

Display the mean of any two variables with continuous values

```
In [38]: 1 # there are 3 continuous variables that are disbursed_amount,interest and income
          2 # hence printing mean of disbursed_amount and interest
          3 loan_data['disbursed_amount'].mean()
```

Out[38]: 14132.2755

```
In [39]: 1 loan_data['interest'].mean()
```

Out[39]: 12.678819440000039

Print the number of discrete variables

```
In [40]: 1 loan_data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999
Data columns (total 14 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   customer_id      10000 non-null   int64  
 1   disbursed_amount 10000 non-null   float64 
 2   interest         10000 non-null   float64 
 3   market           10000 non-null   object  
 4   employment       9389 non-null   object  
 5   time_employed    9471 non-null   object  
 6   householder      10000 non-null   object  
 7   income           10000 non-null   float64 
 8   date_issued      10000 non-null   object  
 9   target           10000 non-null   int64  
 10  loan_purpose     10000 non-null   object  
 11  number_open_accounts 10000 non-null   float64 
 12  date_last_payment 10000 non-null   object  
 13  number_credit_lines_12 238 non-null   float64 
dtypes: float64(5), int64(2), object(7)
memory usage: 1.1+ MB
```

```
In [ ]: 1 # name_of_columns = []
          2 # for i in loan_data:
          3 #   if loan_data[i].dtype == "object":
          4 #     name_of_columns.append(i)
          5 # print("the columns that are discrete in nature are :- ",end=" ")
          6 # for i in name_of_columns:
          7 #   if (i == "date_last_payment" or i == 'date_issued'): # removing these columns as they
          8 #     pass
          9 #   else:
          10 #     print(i,end=" , ")
```

```
In [54]: 1 categorical_columns = [i for i in loan_data.columns if loan_data[i].dtype == "O"]
```

```
In [56]: 1 print("the columns that are discrete in nature are :-")
          2 for i in categorical_columns:
          3     print(i)
```

the columns that are discrete in nature are :-
market
employment
time_employed
householder
date_issued
loan_purpose
date_last_payment

Display the unique values of two variables with discrete values

```
In [52]: 1 loan_data["market"].unique()
```

Out[52]: array(['C', 'B', 'A', 'E', 'D'], dtype=object)

```
In [53]: 1 loan_data["employment"].unique()
```

Out[53]: array(['Teacher', 'Accountant', 'Statistician', 'Other', 'Bus driver',
 'Secretary', 'Software developer', 'Nurse', 'Taxi driver', nan,
 'Civil Servant', 'Dentist'], dtype=object)

Display the Month with most of loans issued date

In [57]: 1 month = loan_data['date_last_payment'].dt.month

```
-----
-----
AttributeError                                Traceback (most recent call last)
<ipython-input-57-8720de89590b> in <module>
      1 month = loan_data['date_last_payment'].dt.month

~\anaconda3\lib\site-packages\pandas\core\generic.py in __getattr__(self, name)
  5456     or name in self._accessors
  5457     ):
-> 5458         return object.__getattribute__(self, name)
  5459     else:
  5460         if self._info_axis._can_hold_identifiers_and_holds_name(name):

~\anaconda3\lib\site-packages\pandas\core\accessor.py in __get__(self, obj, cls)
  178     # we're accessing the attribute of the class, i.e., Dataset.geo
  179     return self._accessor
--> 180     accessor_obj = self._accessor(obj)
  181     # Replace the property with the accessor object. Inspired by:
  182     # https://www.pydanny.com/cached-property.html (https://www.pydanny.co
m/cached-property.html)

~\anaconda3\lib\site-packages\pandas\core\indexes\accessors.py in __new__(cls, data)
  492     return PeriodProperties(data, orig)
  493
--> 494     raise AttributeError("Can only use .dt accessor with datetimelike value
s")
```

AttributeError: Can only use .dt accessor with datetimelike values

In [79]: 1 loan_data['date_issued']=pd.to_datetime(loan_data['date_issued'])

In [80]: 1 month = loan_data['date_issued'].dt.month

In [81]: 1 month.value_counts()

Out[81]: 10 1277
7 1066
11 1017
12 882
8 852
4 816
5 749
9 734
1 700
6 700
3 623
2 584

Name: date_issued, dtype: int64

In [83]: 1 print(month.value_counts() > 1000)

```
10  True
 7  True
11  True
12  False
 8  False
 4  False
 5  False
 9  False
 1  False
 6  False
 3  False
 2  False
Name: date_issued, dtype: bool
```

In [85]: 1 for i in month.value_counts():
 2 if i > 1000:
 3 print(i)

```
1277
1066
1017
```

In [86]: 1 for i,x in zip(month.value_counts().keys(),month.value_counts()):
 2 if x > 1000:
 3 print("Month number :- ",i," Month Counts :- ",x)

```
Month number :- 10      Month Counts :- 1277
Month number :- 7       Month Counts :- 1066
Month number :- 11      Month Counts :- 1017
```

Display the count of 'Teacher' who are 'Owners'

In [97]: 1 new_df = loan_data[["employment",'householder']]

In [98]: 1 new_df = new_df.loc[new_df['employment'] == 'Teacher']

In [99]: 1 new_df = new_df.loc[new_df['householder'] == 'OWNER']

In [100]: 1 new_df.head()

Out[100]:

	employment	householder
71	Teacher	OWNER
85	Teacher	OWNER
171	Teacher	OWNER
672	Teacher	OWNER
1024	Teacher	OWNER

In [102]: 1 print("there are ",new_df.shape[0],"teacher who are owners")

there are 69 teacher who are owners

Display the 'Employment' of customers who mostly 'Rent'

In [107]: 1 new_df_1 = loan_data[["employment",'householder']]
2 new_df_1 = new_df_1.loc[new_df_1['householder'] == 'RENT']
3 print("there are ",new_df_1.shape[0],"employee who are on rent")

there are 4055 employee who are on rent

In [108]: 1 new_df_1

Out[108]:

	employment	householder
0	Teacher	RENT
2	Statistician	RENT
3	Other	RENT
4	Bus driver	RENT
6	Secretary	RENT
...
9991	Software developer	RENT
9992	Statistician	RENT
9994	Accountant	RENT
9996	Civil Servant	RENT
9998	Bus driver	RENT

4055 rows × 2 columns