applying the necessary libraries

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

import warnings
warnings.filterwarnings("ignore")

importing the dataset

df = pd.read_csv("credit_train.csv")

df.head()

	Loan ID	Customer ID	Loan Status	Current Loan Amount	Term	Credit Score	Annual Income	Years in current job
0	14dd8831- 6af5-400b- 83ec- 68e61888a048	981165ec- 3274-42f5- a3b4- d104041a9ca9	Fully Paid	445412.0	Short Term	709.0	1167493.0	8 years
1	4771cc26- 131a-45db- b5aa- 537ea4ba5342	2de017a3- 2e01-49cb- a581- 08169e83be29	Fully Paid	262328.0	Short Term	NaN	NaN	10+ years
2	4eed4e6a- aa2f-4c91- 8651- ce984ee8fb26	5efb2b2b-bf11- 4dfd-a572- 3761a2694725	Fully Paid	99999999.0	Short Term	741.0	2231892.0	8 years
3	77598f7b- 32e7-4e3b- a6e5- 06ba0d98fe8a	e777faab- 98ae-45af- 9a86- 7ce5b33b1011	Fully Paid	347666.0	Long Term	721.0	806949.0	3 years
4	d4062e70- befa-4995- 8643- a0de73938182	81536ad9- 5ccf-4eb8- befb- 47a4d608658e	Fully Paid	176220.0	Short Term	NaN	NaN	5 years

checking the information of the dataset

df.info()

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

RangeIndex: 100514 entries, 0 to 100513 Data columns (total 19 columns):

#	Column	Non-Null Count	Dtype
0	Loan ID	100000 non-null	object
1	Customer ID	100000 non-null	object
2	Loan Status	100000 non-null	object
3	Current Loan Amount	100000 non-null	float64
4	Term	100000 non-null	object
5	Credit Score	80846 non-null	float64
6	Annual Income	80846 non-null	float64
7	Years in current job	95778 non-null	object
8	Home Ownership	100000 non-null	object
9	Purpose	100000 non-null	object
10	Monthly Debt	100000 non-null	float64
11	Years of Credit History	100000 non-null	float64
12	Months since last delinquent	46859 non-null	float64
13	Number of Open Accounts	100000 non-null	float64
14	Number of Credit Problems	100000 non-null	float64
15	Current Credit Balance	100000 non-null	float64
16	Maximum Open Credit	99998 non-null	float64
17	Bankruptcies	99796 non-null	float64
18	Tax Liens	99990 non-null	float64

dtypes: float64(12), object(7)

memory usage: 14.6+ MB

descriptive data of the dataset

df.describe()

	Current Loan Amount	Credit Score	Annual Income	Monthly Debt	Years of Credit History	sir del
count	1.000000e+05	80846.000000	8.084600e+04	100000.000000	100000.000000	4685
mean	1.176045e+07	1076.456089	1.378277e+06	18472.412336	18.199141	3,
std	3.178394e+07	1475.403791	1.081360e+06	12174.992609	7.015324	2
min	1.080200e+04	585.000000	7.662700e+04	0.000000	3.600000	(
25%	1.796520e+05	705.000000	8.488440e+05	10214.162500	13.500000	10
50%	3.122460e+05	724.000000	1.174162e+06	16220.300000	16.900000	3:
75%	5.249420e+05	741.000000	1.650663e+06	24012.057500	21.700000	5
max	1.000000e+08	7510.000000	1.655574e+08	435843.280000	70.500000	170

droping the unnecessary columns

df.drop(["Loan ID","Customer ID"],axis=1, inplace = True)

checking any null values in the dataset

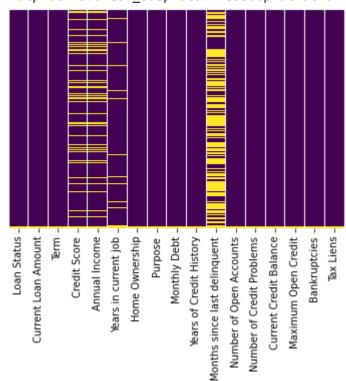
df.isna().sum()

Loan Status	514
Current Loan Amount	514
Term	514
Credit Score	19668
Annual Income	19668
Years in current job	4736
Home Ownership	514
Purpose	514
Monthly Debt	514
Years of Credit History	514
Months since last delinquent	53655
Number of Open Accounts	514
Number of Credit Problems	514
Current Credit Balance	514
Maximum Open Credit	516
Bankruptcies	718
Tax Liens	524
dtype: int64	

heatmap showing the null values

sns.heatmap(df.isnull(), yticklabels=False, cbar=False, cmap="viridis")





droping all the columns with null values

df.drop(["Months since last delinquent"],axis = 1, inplace = True)

```
df.drop(["Credit Score"],axis = 1, inplace = True)
df.drop(["Annual Income"],axis = 1, inplace = True)
```

df.dropna(inplace = True)

df.head()

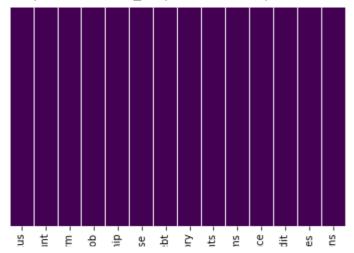
	Loan Status	Current Loan Amount	Term	Years in current job	Home Ownership	Purpose	Monthly Debt	Years of Credit History	Α
0	Fully Paid	445412.0	Short Term	8 years	Home Mortgage	Home Improvements	5214.74	17.2	
1	Fully Paid	262328.0	Short Term	10+ years	Home Mortgage	Debt Consolidation	33295.98	21.1	
2	Fully Paid	99999999.0	Short Term	8 years	Own Home	Debt Consolidation	29200.53	14.9	
3	Fully Paid	347666.0	Long Term	3 years	Own Home	Debt Consolidation	8741.90	12.0	
4	Fully Paid	176220.0	Short Term	5 years	Rent	Debt Consolidation	20639.70	6.1	

df.isna().sum()

Loan Status	0
Current Loan Amount	0
Term	0
Years in current job	0
Home Ownership	0
Purpose	0
Monthly Debt	0
Years of Credit History	0
Number of Open Accounts	0
Number of Credit Problems	0
Current Credit Balance	0
Maximum Open Credit	0
Bankruptcies	0
Tax Liens	0
dtype: int64	

sns.heatmap(df.isnull(), yticklabels = False, cbar = False, cmap = "viridis")

<matplotlib.axes._subplots.AxesSubplot at 0x7f7e2f1bb550>



df.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 95572 entries, 0 to 99998
Data columns (total 14 columns):

#	Column	Non-Null Count	Dtype				
0	Loan Status	95572 non-null	object				
1	Current Loan Amount	95572 non-null	float64				
2	Term	95572 non-null	object				
3	Years in current job	95572 non-null	object				
4	Home Ownership	95572 non-null	object				
5	Purpose	95572 non-null	object				
6	Monthly Debt	95572 non-null	float64				
7	Years of Credit History	95572 non-null	float64				
8	Number of Open Accounts	95572 non-null	float64				
9	Number of Credit Problems	95572 non-null	float64				
10	Current Credit Balance	95572 non-null	float64				
11	Maximum Open Credit	95572 non-null	float64				
12	Bankruptcies	95572 non-null	float64				
13	Tax Liens	95572 non-null	float64				
dtyp	<pre>dtypes: float64(9), object(5)</pre>						

checking the value counts of the column Loan Status

```
df["Loan Status"].value_counts()
```

memory usage: 10.9+ MB

Fully Paid 74257 Charged Off 21315

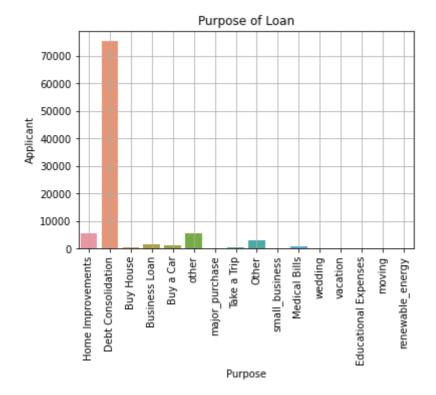
Name: Loan Status, dtype: int64

countplot of the Loan Status column

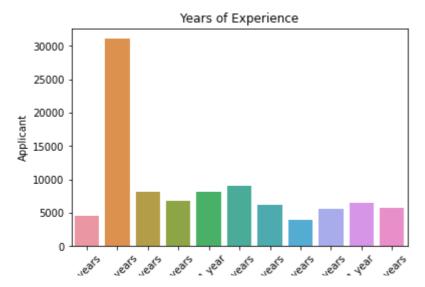
```
sns.countplot(data = df, x= "Loan Status")
plt.grid(True)
```



```
# Purpose of the Loan that was taken
sns.countplot(data = df, x= "Purpose")
plt.title("Purpose of Loan")
plt.xlabel("Purpose")
plt.ylabel("Applicant")
plt.xticks(rotation=90)
plt.grid(True)
```



```
sns.countplot(data = df, x= "Years in current job")
plt.title("Years of Experience")
plt.xlabel("Years")
plt.ylabel("Applicant")
plt.xticks(rotation=45)
plt.show()
```



checking the pairplot for the Dataset
sns.pairplot(df, hue="Loan Status")

dividing the dataset into categorical and numerical

```
df_cat = df.select_dtypes(object)
df_num = df.select_dtypes(["int64", "float64"])
```

using label encoder

from sklearn.preprocessing import LabelEncoder

label encoding the dataset

```
for cols in df_cat:
    le = LabelEncoder()
    df_cat[cols] = le.fit_transform(df_cat[cols])
```

concatinating the dataset

```
df = pd.concat([df_cat,df_num], axis = 1)
df.head()
```

```
dividing the dataset into x and y
```

```
x = df.iloc[:,1:].values
y = df.iloc[:,:1].values
from imblearn.over_sampling import RandomOverSampler
ros = RandomOverSampler()
x, y = ros.fit_resample(x,y)
x.shape
y.shape
from sklearn.model_selection import train_test_split
xtrain,xtest,ytrain,ytest = train_test_split(x,y, test_size = 0.3, random_state = 1)
using standard scaler to standardise the dataset
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
xtrain = sc.fit transform(xtrain)
xtest = sc.transform(xtest)
df.head()
```

```
import tensorflow as tf
```

training the model

```
ann = tf.keras.Sequential()
ann.add(tf.keras.layers.Dense(units=10, activation="relu"))
ann.add(tf.keras.layers.Dense(units=1, activation="sigmoid"))
ann.compile(optimizer="adam", loss="binary_crossentropy", metrics=["accuracy"])
ann.fit(xtrain, ytrain, batch_size=30, epochs=250)
ypred = ann.predict(xtest)
ypred = (ypred>0.5)
```

getting the classification report

from sklearn.metrics import classification_report
print(classification_report(ytest,ypred))

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