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

In [4]: dataset=pd.read\_csv('loan.csv')

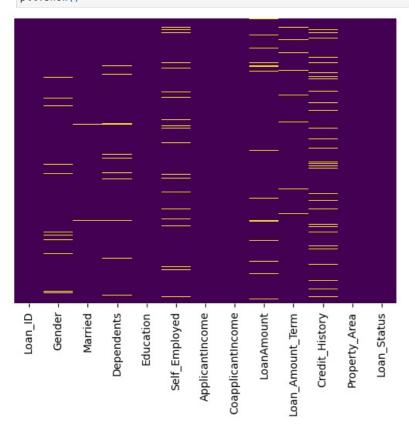
In [5]: dataset

Married Dependents Education Self\_Employed ApplicantIncome CoapplicantIncome LoanAmount Loan Out[5]: Loan\_ID Gender **0** LP001002 Male No Graduate No 5849 0.0 NaN **1** LP001003 4583 1508.0 128.0 Male Yes Graduate No 2 LP001005 Male 0 Graduate 3000 0.0 66.0 Yes Yes 3 LP001006 Male Yes 0 No 2583 2358.0 120.0 Graduate 4 LP001008 0 No 6000 0.0 141.0 Male No Graduate LP002978 Female No 0 Graduate No 2900 0.0 71.0 610 LP002979 Yes Graduate No 4106 0.0 40.0 **611** LP002983 8072 240.0 253.0 Male Yes Graduate No 612 LP002984 Male Yes Graduate No 7583 0.0 187.0 613 LP002990 Female Graduate 4583 0.0 133.0 No Yes

614 rows × 13 columns

•

In [9]: sns.heatmap(dataset.isnull(),yticklabels=False,cbar=False,cmap='viridis')
plt.show()



In [11]: dataset.describe()

	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History
count	614.000000	614.000000	592.000000	600.00000	564.000000
mean	5403.459283	1621.245798	146.412162	342.00000	0.842199
std	6109.041673	2926.248369	85.587325	65.12041	0.364878
min	150.000000	0.000000	9.000000	12.00000	0.000000
25%	2877.500000	0.000000	100.000000	360.00000	1.000000
50%	3812.500000	1188.500000	128.000000	360.00000	1.000000
75%	5795.000000	2297.250000	168.000000	360.00000	1.000000
max	81000.000000	41667.000000	700.000000	480.00000	1.000000

```
In [13]: dataset.isnull().sum()
```

```
Out[13]: Loan ID
                               13
          Gender
          Married
          Dependents
                               15
          Education
                                0
          Self Employed
                               32
          ApplicantIncome
                                0
          CoapplicantIncome
                                0
          LoanAmount
                               22
          Loan Amount Term
                               14
          Credit History
                               50
          Property_Area
                                0
          Loan Status
                                0
          dtype: int64
```

## In [15]: pd.crosstab(dataset['Credit History'], dataset['Loan Status'], margins=True)

# Out[15]: Loan\_Status N Y All Credit\_History 0.0 82 7 89 1.0 97 378 475 All 179 385 564

### In [17]: dataset['Gender'].fillna(dataset['Gender'].mode()[0],inplace=True)

C:\Users\Administrator\AppData\Local\Temp\ipykernel\_12460\2782724874.py:1: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method.

The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting values always behaves as a copy.

For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].method(value) instead, to perform the operation inplace on the original object.

dataset['Gender'].fillna(dataset['Gender'].mode()[0],inplace=True)

### In [19]: dataset['Married'].fillna(dataset['Married'].mode()[0],inplace=True)

C:\Users\Administrator\AppData\Local\Temp\ipykernel\_12460\568568310.py:1: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method. The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on w hich we are setting values always behaves as a copy.

For example, when doing 'df[col].method(value, inplace=True)', try using  $'df.method(\{col: value\}, inplace=True)'$  or df[col] = df[col].method(value) instead, to perform the operation inplace on the original object.

dataset['Married'].fillna(dataset['Married'].mode()[0],inplace=True)

# In [21]: dataset['Self\_Employed'].fillna(dataset['Self\_Employed'].mode()[0],inplace=True)

C:\Users\Administrator\AppData\Local\Temp\ipykernel\_12460\39551059.py:1: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method.

The behavior will change in pands 3.0. This inplace method will never work because the intermediate object on a

The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on w hich we are setting values always behaves as a copy.

For example, when doing 'df[col].method(value, inplace=True)', try using  $'df.method(\{col: value\}, inplace=True)'$  or df[col] = df[col].method(value) instead, to perform the operation inplace on the original object.

dataset['Self Employed'].fillna(dataset['Self Employed'].mode()[0],inplace=True)

```
In [23]: dataset['Dependents'].fillna(dataset['Dependents'].mode()[0],inplace=True)
        C:\Users\Administrator\AppData\Local\Temp\ipykernel 12460\2291896846.py:1: FutureWarning: A value is trying to b
        e set on a copy of a DataFrame or Series through chained assignment using an inplace method.
        The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on w
        hich we are setting values always behaves as a copy.
        For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)'
        or df[col] = df[col].method(value) instead, to perform the operation inplace on the original object.
         dataset['Dependents'].fillna(dataset['Dependents'].mode()[0],inplace=True)
In [25]: dataset['Loan Amount Term'].fillna(dataset['Loan Amount Term'].mode()[0],inplace=True)
        C:\Users\Administrator\AppData\Local\Temp\ipykernel 12460\3879141361.py:1: FutureWarning: A value is trying to b
        e set on a copy of a DataFrame or Series through chained assignment using an inplace method.
        The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on w
        hich we are setting values always behaves as a copy.
        For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)'
        or df[col] = df[col].method(value) instead, to perform the operation inplace on the original object.
        dataset['Loan Amount Term'].fillna(dataset['Loan Amount Term'].mode()[0],inplace=True)
In [27]: dataset['Credit History'].fillna(dataset['Credit History'].mode()[0],inplace=True)
        C:\Users\Administrator\AppData\Local\Temp\ipykernel 12460\979352118.py:1: FutureWarning: A value is trying to be
        set on a copy of a DataFrame or Series through chained assignment using an inplace method.
        The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on w
        hich we are setting values always behaves as a copy.
        For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)'
        or df[col] = df[col].method(value) instead, to perform the operation inplace on the original object.
         dataset['Credit History'].fillna(dataset['Credit History'].mode()[0],inplace=True)
In [29]: dataset.LoanAmount=dataset.LoanAmount.fillna(dataset.LoanAmount.mean())
In [31]: dataset.isnull().sum()
Out[31]: Loan ID
         Gender
                              Θ
         Married
                              0
         Dependents
                              0
         Education
                              0
         Self_Employed
                              0
         ApplicantIncome
                              0
         CoapplicantIncome
                              0
         LoanAmount
                              0
         Loan Amount Term
                              0
         Credit History
                              0
         Property_Area
                              0
                              0
         Loan Status
```

In [33]: dataset

dtype: int64

Out[33]:		Loan_ID	Gender	Married	Dependents	Education	Self_Emp	loyed	Applicantlnco	me Coapplican	tIncome	Loan	Amount	Loan
	0	LP001002	Male	No	0	Graduate		No	58	349	0.0	146	.412162	
	1	LP001003	Male	Yes	1	Graduate		No	45	583	1508.0	128	.000000	
	2	LP001005	Male	Yes	0	Graduate		Yes	30	000	0.0	66	.000000	
	3	LP001006	Male	Yes	0	Not Graduate		No	25	583	2358.0	120	.000000	
	4	LP001008	Male	No	0	Graduate		No	60	000	0.0	141	.000000	
	609	LP002978	Female	No	0	Graduate		No	29	900	0.0	71	.000000	
	610	LP002979	Male	Yes	3+	Graduate		No	41	06	0.0	40	.000000	
	611	LP002983	Male	Yes	1	Graduate		No	80	)72	240.0	253	.000000	
	612	LP002984	Male	Yes	2	Graduate		No	75	583	0.0	187	.000000	
	613	LP002990	Female	No	0	Graduate		Yes	45	583	0.0	133	.000000	
	614 rd	ows × 13 co	lumns											
	4													<b> </b>
<pre>In [35]: gender=pd.get_dummies(dataset['Gender'],drop_first=True)     married=pd.get_dummies(dataset['Married'],drop_first=True)     education=pd.get_dummies(dataset['Education'],drop_first=True)     self_employed=pd.get_dummies(dataset['Self_Employed'],drop_first=True)     dependents=pd.get_dummies(dataset['Dependents'],drop_first=True)     property_area=pd.get_dummies(dataset['Property_Area'],drop_first=True)     dataset.drop(['Gender','Married','Education','Self_Employed','Dependents','Property_Area']</pre>							Property_Area	a'],axis	=1,in	place=	True)			
In [37]:	data	set=pd.co	ncat([da	ataset,g	ender,marrie	d,educat	ion,self_	employ	ved, dependent	s,property_ar	rea],axi	s=1)		
In [39]:	data	set												
Out[39]:		Loan_ID	Applicar	ntIncome	CoapplicantInd	come Lo	anAmount	Loan_	Amount_Term	Credit_History	Loan_St	tatus	Male	Yes
	0	LP001002		5849		0.0 1	46.412162		360.0	1.0		Υ	True	False
	1	LP001003		4583	1:	508.0 1	28.000000		360.0	1.0		N	True	True
	2	LP001005		3000		0.0	66.000000		360.0	1.0		Υ	True	True
	3	LP001006		2583	23	358.0 1	20.000000		360.0	1.0		Υ	True	True
	4	LP001008		6000		0.0 1	41.000000		360.0	1.0		Υ	True	False
	609	LP002978		2900		0.0	71.000000		360.0	1.0		Υ	False	False
	610	LP002979		4106		0.0	40.000000		180.0	1.0		Υ	True	True
	611	LP002983		8072	:	240.0 2	53.000000		360.0	1.0		Υ	True	True
	612	LP002984		7583		0.0 1	87.000000		360.0	1.0		Υ	True	True
	613	LP002990		4583		0.0 1	33.000000		360.0	0.0		N	False	False
	614 rd	ows × 16 co	lumns											
	4													>
In [41]:	data	set['Yes'	]=datase	et['Yes'	e'].astype(int].astype(int ].astype(int set['Not Grad	)	astype(in	t)						

In [43]: dataset

3]:	Loan_ID	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History	Loan_Status	Male	Yes (
	<b>0</b> LP001002	5849	0.0	146.412162	360.0	1.0	Y	1	0
	<b>1</b> LP001003	4583	1508.0	128.000000	360.0	1.0	N	1	1
	<b>2</b> LP001005	3000	0.0	66.000000	360.0	1.0	Υ	1	1
	<b>3</b> LP001006	2583	2358.0	120.000000	360.0	1.0	Υ	1	1
	<b>4</b> LP001008	6000	0.0	141.000000	360.0	1.0	Υ	1	0
-									
60	<b>09</b> LP002978	2900	0.0	71.000000	360.0	1.0	Υ	0	0
61	<b>10</b> LP002979	4106	0.0	40.000000	180.0	1.0	Υ	1	1
61	<b>11</b> LP002983	8072	240.0	253.000000	360.0	1.0	Υ	1	1
61	<b>12</b> LP002984	7583	0.0	187.000000	360.0	1.0	Υ	1	1
61	13 LP002990	4583	0.0	133.000000	360.0	0.0	N	0	0
614	4 rows × 16 co	lumns							
4									)
			<pre>['Semiurban'].asty ban'].astype(int)</pre>						
da	ataset								
		ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History	Loan_Status	Male	Yes (
		ApplicantIncome 5849	CoapplicantIncome	LoanAmount 146.412162	Loan_Amount_Term	Credit_History	Loan_Status	Male 1	Yes 0
	Loan_ID								
	Loan_ID  0 LP001002	5849	0.0	146.412162	360.0	1.0	Y	1	0
	Loan_ID  0 LP001002  1 LP001003	5849 4583	0.0 1508.0	146.412162 128.000000	360.0 360.0	1.0	Y	1	0
	Loan_ID  1 LP001002 1 LP001003 2 LP001005	5849 4583 3000	0.0 1508.0 0.0	146.412162 128.000000 66.000000	360.0 360.0 360.0	1.0 1.0 1.0	Y N Y	1 1 1	0 1 1
	Loan_ID  Description LP001002 LP001003 LP001005 LP001006	5849 4583 3000 2583	0.0 1508.0 0.0 2358.0	146.412162 128.000000 66.000000 120.000000	360.0 360.0 360.0 360.0	1.0 1.0 1.0	Y N Y	1 1 1	0 1 1
	Loan_ID  0 LP001002  1 LP001003  2 LP001005  3 LP001006  4 LP001008	5849 4583 3000 2583 6000	0.0 1508.0 0.0 2358.0 0.0	146.412162 128.000000 66.000000 120.000000	360.0 360.0 360.0 360.0	1.0 1.0 1.0 1.0	Y N Y Y	1 1 1 1	0 1 1 1 0
. 60	Loan_ID  D LP001002 LP001003 LP001005 LP001006 LP001008	5849 4583 3000 2583 6000	0.0 1508.0 0.0 2358.0 0.0	146.412162 128.000000 66.000000 120.000000 141.000000	360.0 360.0 360.0 360.0	1.0 1.0 1.0 1.0 1.0	Y N Y Y	1 1 1 1 1	0 1 1 1 1 0
600	Loan_ID  1 LP001002 1 LP001003 2 LP001005 3 LP001006 4 LP001008 09 LP002978	5849 4583 3000 2583 6000 	0.0 1508.0 0.0 2358.0 0.0 	146.412162 128.000000 66.000000 120.000000 141.000000  71.000000	360.0 360.0 360.0 360.0  360.0	1.0 1.0 1.0 1.0 1.0 	Y N Y Y Y Y	1 1 1 1 1 	0 1 1 1 0 0
 600 611	Loan_ID  1 LP001002 1 LP001005 2 LP001006 4 LP001008 09 LP002978	5849 4583 3000 2583 6000  2900 4106	0.0 1508.0 0.0 2358.0 0.0  0.0	146.412162 128.000000 66.000000 120.000000 141.000000  71.000000 40.000000	360.0 360.0 360.0 360.0  360.0 180.0	1.0 1.0 1.0 1.0 1.0 1.0	Y N Y Y Y Y Y Y	1 1 1 1 1  0	0 1 1 1 0 0 1
600 61 61	Loan_ID  1 LP001002 1 LP001003 2 LP001006 4 LP001008 10 LP002978 11 LP002983	5849 4583 3000 2583 6000  2900 4106 8072	0.0 1508.0 0.0 2358.0 0.0  0.0 0.0 240.0	146.412162 128.000000 66.000000 120.000000 141.000000  71.000000 40.0000000 253.0000000	360.0 360.0 360.0 360.0  360.0 180.0 360.0	1.0 1.0 1.0 1.0 1.0  1.0 1.0	Y N Y Y Y Y  Y	1 1 1 1  0 1	0 1 1 1 0 0 1 1
60 61 61 61	Loan_ID  1 LP001002 1 LP001005 2 LP001006 4 LP001008 10 LP002978 11 LP002984 12 LP002984	5849 4583 3000 2583 6000  2900 4106 8072 7583 4583	0.0 1508.0 0.0 2358.0 0.0  0.0 240.0	146.412162 128.000000 66.000000 120.000000 141.000000 40.000000 253.000000 187.000000	360.0 360.0 360.0 360.0 360.0 180.0 360.0	1.0 1.0 1.0 1.0 1.0 1.0 1.0 1.0	Y N Y Y Y Y Y Y Y Y	1 1 1 1  0 1 1	0 1 1 1 0  0 1 1
600 611 611	Loan_ID  1 LP001002 1 LP001003 2 LP001006 4 LP001008 09 LP002978 10 LP002979 11 LP002983 12 LP002984 13 LP002990	5849 4583 3000 2583 6000  2900 4106 8072 7583 4583	0.0 1508.0 0.0 2358.0 0.0  0.0 240.0	146.412162 128.000000 66.000000 120.000000 141.000000 40.000000 253.000000 187.000000	360.0 360.0 360.0 360.0 360.0 180.0 360.0	1.0 1.0 1.0 1.0 1.0 1.0 1.0 1.0	Y N Y Y Y Y Y Y Y Y	1 1 1 1  0 1 1	0 1 1 1 0  0 1 1
60 61 61 61 614	Loan_ID  0 LP001002  1 LP001003  2 LP001006  4 LP001008   09 LP002979  11 LP002984  13 LP002984  14 rows × 16 co	5849 4583 3000 2583 6000 2900 4106 8072 7583 4583	0.0 1508.0 0.0 2358.0 0.0  0.0 240.0	146.412162 128.000000 66.000000 120.000000 141.000000 40.000000 253.000000 187.000000 133.000000	360.0 360.0 360.0 360.0  360.0 180.0 360.0 360.0	1.0 1.0 1.0 1.0 1.0 1.0 1.0 1.0	Y N Y Y Y Y Y Y Y Y	1 1 1 1  0 1 1	0 1 1 1 0  0 1 1

In [53]: dataset

Out[53]:		Loan_ID	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History	Male	Yes	Not Graduate	Yes
		LP001002	5849	0.0	146.412162	360.0	1.0	1	0	0	0
		LP001003	4583	1508.0	128.000000	360.0	1.0	1	1	0	0
		LP001005	3000	0.0	66.000000	360.0	1.0	1	1	0	1
		LP001006	2583	2358.0	120.000000	360.0	1.0	1	1	1	0
	4	LP001008	6000	0.0	141.000000	360.0	1.0	1	0	0	0
	609	LP002978	2900	0.0	71.000000	360.0	1.0	0	0	0	0
	610	LP002979	4106	0.0	40.000000	180.0	1.0	1	1	0	0
	611	LP002983	8072	240.0	253.000000	360.0	1.0	1	1	0	0
	612	LP002984	7583	0.0	187.000000	360.0	1.0	1	1	0	0
	613	LP002990	4583	0.0	133.000000	360.0	0.0	0	0	0	1
	614 r	ows × 16 co	lumns								
	4										Þ
[n [55]:	data	set['Y']=	dataset['Y'].as	type(int)							
[n [57]:	data	iset									
Out[57]:		Loan_ID	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History	Male	Yes	Not Graduate	Yes
	0	LP001002	5849	0.0	146.412162	360.0	1.0	1	0	0	0
	1	LP001003	4583	1508.0	128.000000	360.0	1.0	1	1	0	0
	2	LP001005	3000	0.0	66.000000	360.0	1.0	1	1	0	1
	3	LP001006	2583	2358.0	120.000000	360.0	1.0	1	1	1	0
	4	LP001008	6000	0.0	141.000000	360.0	1.0	1	0	0	0
	609	LP002978	2900	0.0	71.000000	360.0	1.0	0	0	0	0
	610	LP002979	4106	0.0	40.000000	180.0	1.0	1	1	0	0
	611	LP002983	8072	240.0	253.000000	360.0	1.0	1	1	0	0
	612	LP002984	7583	0.0	187.000000	360.0	1.0	1	1	0	0
	613	LP002990	4583	0.0	133.000000	360.0	0.0	0	0	0	1
	614 r	ows × 16 co	lumns								
	4										Þ
In [59]:	data	set.drop(	'Loan_ID',axis=	1,inplace= <b>True</b> )							
In [61]:	data	iset									
Out[61]:		Applicantli	ncome Coapplica	ntIncome LoanAmou	ınt Loan_Amo	unt_Term Credit_His	tory Male Yes	Grad	Not uate	Yes 1	2 3+

0 0.0 146.412162 360.0 5849 1.0 1 0 0 0 0 4583 1508.0 128.000000 360.0 1.0 0 2 1 3000 0.0 66.000000 360.0 1.0 1 0 0 0 0 3 2583 2358.0 120.000000 360.0 1.0 1 0 0 0 0 4 6000 0.0 141.000000 360.0 1.0 0 0 0 0 0 ... ... ... ... 2900 0.0 71.000000 360.0 0 0 0 609 1.0 0 0 0 0 610 4106 0.0 40.000000 180.0 1.0 1 0 0 0 611 8072 240.0 253.000000 360.0 1.0 1 0 0 0 0.0 0 0 0 7583 187.000000 360.0 1.0 0 612 1 0 0 0 613 4583 0.0 133.000000 360.0 0.0 0 0 0

614 rows × 15 columns

```
In [67]: from sklearn.linear_model import LogisticRegression
         from sklearn.model selection import train test split
         \textbf{from} \ \ \text{sklearn.metrics} \ \ \textbf{import} \ \ \text{accuracy\_score}, \\ \text{classification\_report}, \\ \text{roc\_auc\_score}
         from imblearn.over_sampling import SMOTE
In [69]: X=dataset.drop('Y',axis=1)
         y=dataset['Y']
In [71]: print(X.shape)
         print(X.isnull().sum())
        (614, 14)
        ApplicantIncome
        CoapplicantIncome
                              0
        LoanAmount
        Loan Amount Term
                              0
        Credit History
        Male
                              0
        Yes
        Not Graduate
                              0
        Yes
                              0
        1
                              0
        2
        3+
                              0
        Semiurban
        Urban
        dtype: int64
In [73]: print(y.shape)
         print(y.isnull().sum())
        (614,)
In [75]: X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.3,random state=101)
In [77]: smote=SMOTE(random state=42)
In [79]: X train balanced, y train balanced=smote.fit resample(X train, y train)
In [81]: model=LogisticRegression(class weight='balanced', max iter=1000)
In [83]: model.fit(X train balanced,y train balanced)
        C:\ProgramData\anaconda3\Lib\site-packages\sklearn\linear model\ logistic.py:469: ConvergenceWarning: lbfgs fail
        ed to converge (status=1):
        STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
        Increase the number of iterations (max iter) or scale the data as shown in:
            https://scikit-learn.org/stable/modules/preprocessing.html
        Please also refer to the documentation for alternative solver options:
            https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression
          n iter i = check optimize result(
Out[83]: 🔻
                               LogisticRegression
         LogisticRegression(class weight='balanced', max iter=1000)
In [85]: y pred=model.predict(X test)
In [87]: accuracy=accuracy score(y test,y pred)
In [89]: report=classification_report(y_test,y_pred)
In [91]: print("ACCURACY:",accuracy)
         print("CLASSIFICATION:\n", report)
         print("ROC-AUC Score",roc_auc_score(y_test,model.predict_proba(X_test)[:,1]))
         print("LOAN STATUS:",y_pred)
```

ACCURACY: 0.772972972973

CLASSIFICATION:

	precision	recall	f1-score	support
Θ	0.73	0.55	0.62	64
1	0.79	0.89	0.84	121
accuracy			0.77	185
macro avg weighted avg	0.76 0.77	0.72 0.77	0.73 0.76	185 185

ROC-AUC Score 0.750129132231405

 $0\ 1\ 1\ 1\ 1\ 1\ 0\ 1\ 0\ 0\ 0\ 1\ 1\ 1\ 1\ 0\ 0\ 1\ 1\ 1\ 1\ 0\ 0\ 1\ 0\ 1\ 1\ 1\ 0\ 0\ 1\ 1\ 1\ 0\ 0\ 1\ 1\ 1$ 

In [ ]:

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