**Aim:**- To understand and apply data transformation and data discretization techniques on a given dataset using Python.

**Software Used: IDLE** 

Theory:

## **Dataset (dataset.csv):**

Age,Income,Education\_Level

23,50000,Bachelor

45,65000,Master

25,48000,Bachelor

34,52000,PhD

65,70000,PhD

42,62000,Master

21,45000,Bachelor

35,51000,PhD

32,59000,Master

40,60000,Bachelor

23,52000,Master

52,68000,PhD

30,58000,Master

45,62000,Bachelor

54,71000,PhD

24,49000,Bachelor

28,53000,Master

36,54000,PhD

## Code:

```
import pandas as pd
from sklearn.preprocessing import MinMaxScaler, StandardScaler
import matplotlib.pyplot as plt
import seaborn as sns
# Load dataset
df = pd.read_csv('dataset.csv')
# Display the original data
print("Original Data:")
print(df)
# Step 3: Data Transformation
# Normalization
scaler = MinMaxScaler()
df[['Age_Normalized', 'Income_Normalized']] = scaler.fit_transform(df[['Age', 'Income']])
# Standardization
scaler = StandardScaler()
df[['Age_Standardized', 'Income_Standardized']] = scaler.fit_transform(df[['Age', 'Income']])
print("\nData after Normalization and Standardization:")
print(df)
# Step 4: Data Discretization
df['Age_Bins'] = pd.cut(df['Age'], bins=[20, 30, 40, 50, 60, 70], labels=['20-30', '30-40', '40-50', '50-60',
'60-70'])
df['Income_Bins'] = pd.cut(df['Income'], bins=4, labels=['Low', 'Medium', 'High', 'Very High'])
print("\nData after Discretization:")
print(df)
# Step 5: Summary and Visualization
print("\nSummary of Transformed Data:")
print(df.describe())
# Visualization
plt.figure(figsize=(12, 6))
```

```
# Age distribution
plt.subplot(1, 2, 1)
sns.histplot(df['Age'], bins=5, kde=True)
plt.title('Age Distribution')
# Income distribution
plt.subplot(1, 2, 2)
sns.histplot(df['Income'], bins=4, kde=True)
plt.title('Income Distribution')
plt.tight_layout()
plt.show()
plt.figure(figsize=(12, 6))
# Age Bins
plt.subplot(1, 2, 1)
sns.countplot(x='Age_Bins', data=df)
plt.title('Age Bins')
# Income Bins
plt.subplot(1, 2, 2)
sns.countplot(x='Income_Bins', data=df)
plt.title('Income Bins')
plt.tight_layout()
plt.show()
```

## **Output-**

```
Summary of Transformed Data:
Age count 20.000000
Age Income ...
count 20.000000 20.000000 ...
mean 38.200000 58100.000000 ...
                                           ... Age_Standardized Income_Standardized
                                                        -2.331468e-16
                                                    -2.331468e-16
1.025978e+00
-1.347373e+00
-8.577754e-01
-2.115063e-01
6.306020e-01
2.099396e+00
       13.097207 8025.616881 ...
21.000000 45000.000000 ...
std
         21.000000 45000.000000
27.250000 51750.000000 ...
58500.000000 ...
min
                                                                                   -1.674677e+00
                                                                                -8.117709e-01
5.113518e-02
        35.500000 58500.000000
         46.250000
                        65000.000000
                                                                                    8.820818e-01
        46.250000 65000.000000 ...
65.000000 71000.000000 ...
                                                                                   1.649109e+00
[8 rows x 6 columns]
```

	Age	Income	Ag	e_Standardized	Income Standardized								
)	23	50000		-1.190702	-1.035487	Data	aft	er Disc	etization:				
L	45	65000		0.532682	0.882082		Age	Income	Education_Level		Income Standardized	Age_Bins	Income Bin
2	25	48000		-1.034031	-1.291163	0	23	50000	Bachelor		-1.035487	20-30	Lo
3	34	52000		-0.329010	-0.779811	1	45	65000	Master		0.882082	40-50	Very Hig
1	65	70000		2.099396	1.521272	2	25	48000			-1.291163	20-30	Lo
5	42	62000		0.297675	0.498568	3	34	52000			-0.779811	30-40	Mediu
5	21	45000		-1.347373	-1.674677	4	65	70000	PhD		1.521272	60-70	Very Hig
7	35	51000		-0.250674	-0.907649	5	42	62000	Master		0.498568	40-50	Hig
	32	59000		-0.485681	0.115054	6	21	45000	Bachelor		-1.674677	20-30	Lo
à	40	60000		0.141004	0.242892	8	35 32	51000			-0.907649	30-40 30-40	Lov
0	23	52000		-1.190702	-0.779811	9	40	59000 60000	Master Bachelor		0.115054 0.242892	30-40	High
11	52	68000		1.081032	1.265596	10	23	52000	Master	• • •	-0.779811	20-30	High Mediu
12	30	58000	• • •	-0.642352	-0.012784	11	52	68000	PhD		1.265596	50-60	Very High
L2 L3	45	62000	• • •	0.532682	0.498568	12	30	58000	Master		-0.012784	20-30	very migi Mediur
						13	45	62000			0.498568	40-50	High
L4	54	71000	• • •	1.237703	1.649109	14	54	71000			1.649109	50-60	Very High
L5	24	49000	• • •	-1.112366	-1.163325	15	24	49000	Bachelor		-1.163325	20-30	Lo
L6	28	53000		-0.799024	-0.651974	16	28	53000	Master		-0.651974	20-30	Mediu
L 7	36	54000		-0.172338	-0.524136	17	36	54000			-0.524136	30-40	Mediu
L8	50	65000		0.924361	0.882082	18	50	65000	Bachelor		0.882082	40-50	Very High
L9	60	68000		1.707717	1.265596	19	60	68000			1.265596	50-60	Very High

[20 rows x 7 columns]

[20 rows x 9 columns]

Ori	ginal	Data:	
	Age	Income	Education_Level
0	23	50000	Bachelor
1	45	65000	Master
	25	48000	Bachelor
	34	52000	PhD
	65	70000	PhD
5	42	62000	Master
6	21	45000	Bachelor
7	35	51000	PhD
8	32	59000	Master
9	40	60000	Bachelor
10	23	52000	Master
11	52	68000	PhD
12	30	58000	Master
13	45	62000	Bachelor
14	54	71000	PhD
15	24	49000	Bachelor
16	28	53000	Master
17	36	54000	PhD
18	50	65000	Bachelor
19	60	68000	Master
	0 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18	Age 0 23 1 25 2 25 3 34 4 65 5 42 66 21 7 7 35 8 32 9 9 40 10 23 11 52 112 30 13 45 14 54 16 28 17 36 50	0 23 50000 2 25 48000 2 25 48000 3 34 52000 5 42 62000 7 35 51000 9 40 60000 11 52 68000 12 30 58000 12 30 58000 13 45 62000 14 54 71000 15 24 49000 16 28 53000 17 36 54000 18 50 65000

