

Analyse des défauts de paiement des prêts : Impact des caractéristiques des emprunteurs

Import Librairies

```
In [1]: import pandas as pd
from pandasql import sqldf
```

Data Preparation

```
In [2]: data = pd.read_csv("BA-ProjectData.csv", sep=',', header=0)
```

```
In [3]: data
```

```
Out[3]:
```

	CaseNo	Default	CreditBalPerc	DebtPerc	LateUpto60Days	Late60to90Days	LateOver
0	2.0	0	0.957151	0.121876	0	0	
1	4.0	0	0.233810	0.036050	0	0	
2	5.0	0	0.907239	0.024926	1	0	
3	8.0	0	0.754464	0.209940	0	0	
4	10.0	0	0.189169	0.606291	0	0	
...
75750	149974.0	0	1.026395	0.494819	0	0	
75751	149980.0	1	0.224711	0.057235	0	0	
75752	149982.0	0	0.810012	0.121752	0	0	
75753	149983.0	0	0.021046	0.250272	0	0	
75754	149986.0	0	0.954409	0.324962	0	0	

75755 rows x 15 columns

```
In [4]: data.sample(5)
```

```
Out[4]:
```

	CaseNo	Default	CreditBalPerc	DebtPerc	LateUpto60Days	Late60to90Days	LateOver
44185	82887.0	0	0.247048	1.255310	0	0	
72481	142869.0	0	0.001857	1.373528	0	0	
2989	5594.0	0	0.000000	0.225320	0	0	
67517	132154.0	0	0.022485	0.284559	0	0	
21031	39231.0	0	0.028999	0.431009	0	0	

```
In [5]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 75755 entries, 0 to 75754
Data columns (total 15 columns):
#   Column                Non-Null Count  Dtype
---  -
0   CaseNo                 75755 non-null  float64
1   Default                75755 non-null  int64
2   CreditBalPerc          75755 non-null  float64
3   DebtPerc               75755 non-null  float64
4   LateUpto60Days         75755 non-null  int64
5   Late60to90Days        75755 non-null  int64
6   LateOver90Days         75755 non-null  int64
7   Income                 75755 non-null  float64
8   NoOfOpenLoans          75755 non-null  int64
9   NoOfHomeLoans          75755 non-null  int64
10  Dependents             75755 non-null  int64
11  Age                   75755 non-null  int64
12  Job                   75755 non-null  object
13  Status                 75755 non-null  object
14  Education              75755 non-null  object
dtypes: float64(4), int64(8), object(3)
memory usage: 8.7+ MB
```

```
In [6]: data.describe().round(2)
```

```
Out[6]:
```

	CaseNo	Default	CreditBalPerc	DebtPerc	LateUpto60Days	Late60to90Days	LateOver90Days
count	75755.00	75755.00	75755.00	75755.00	75755.00	75755.00	75755.00
mean	72415.58	0.10	0.36	0.41	0.16	0.05	0.05
std	42980.21	0.29	0.36	0.41	0.55	0.30	0.30
min	2.00	0.00	0.00	0.00	0.00	0.00	0.00
25%	35348.50	0.00	0.05	0.17	0.00	0.00	0.00
50%	70921.00	0.00	0.21	0.31	0.00	0.00	0.00
75%	108985.00	0.00	0.61	0.49	0.00	0.00	0.00
max	149986.00	1.00	4.00	2.98	7.00	6.00	6.00

```
In [7]: data.duplicated().sum()
```

```
Out[7]: 0
```

```
In [8]: data.nunique()
```

```
Out[8]: CaseNo           75755
Default             2
CreditBalPerc       65456
DebtPerc            70823
LateUpto60Days       8
Late60to90Days       7
LateOver90Days       8
Income              11002
NoOfOpenLoans        18
NoOfHomeLoans         6
Dependents           6
Age                  50
Job                  10
Status                3
Education             4
dtype: int64
```

```
In [9]: data[['Default', 'LateUpto60Days', 'Late60to90Days', 'LateOver90Days', 'NoOfHomeLoans',
```

```
Out [9]: Default                                [0, 1]
LateUpto60Days                                [0, 1, 7, 2, 3, 5, 6, 4]
Late60to90Days                                [0, 1, 2, 5, 3, 4, 6]
LateOver90Days                                [0, 1, 2, 4, 3, 7, 5, 6]
NoOfHomeLoans                                [0, 1, 4, 2, 3, 5]
Dependents                                    [1, 0, 2, 3, 4, 5]
Status                                         [married, single, divorced]
Education                                     [tertiary, secondary, unknown, primary]
dtype: object
```

```
In [10]: #Ajout ratio entre le solde de crédit et le revenu
data['CreditBalance_Income_Ratio']= data['CreditBalPerc']/data['Income']
```

```
In [ ]:
```

```
In [11]: #Ajout colonne Category Revenu (faible,moyen,elevé)
R1 = data['Income'].quantile(0.25)
R3 = data['Income'].quantile(0.75)

def cat_income(income):
    if (income<= R1):
        return 'Weak income'
    elif (income<=R3) and (income > R1):
        return 'middle income'
    else:
        return 'high income'

data['Cat_income']= data['Income'].apply(cat_income)
```

DATA ANALYSIS

```
In [18]: #TOP Income Job
data.sort_values(by='Income', ascending=False).head(5)[['Age', 'Job', 'Income']]
```

```
Out [18]:
```

	Age	Job	Income
23221	59	admin.	287662.0
57888	44	technician	261666.0
12723	57	self-employed	261666.0
49599	50	blue-collar	261666.0
51010	52	blue-collar	251608.0

```
In [19]: #Worst Income Job
data.sort_values(by='Income', ascending=True).head(5)[['Age', 'Job', 'Income']]
```

```
Out [19]:
```

	Age	Job	Income
54821	36	self-employed	500.0
66066	44	technician	500.0
6151	34	unemployed	500.0
29461	51	unemployed	500.0
66214	54	unemployed	500.0

```
In [26]: #percentage of clients defaulted from the loan
percent_default=round((len(data[data['Default']==1])/len(data)*100),2)
print(percent_default)
```

```
In [27]: #percentage of employed clients are loan defaulters?  
data['Job'].unique()
```

```
Out[27]: array(['technician', 'entrepreneur', 'blue-collar', 'management',  
              'admin.', 'services', 'retired', 'self-employed', 'unemployed',  
              'student'], dtype=object)
```

```
In [33]: #percentage of employed clients are loan defaulters?  
employed_client= data[(data["Job"]!='student') & (data["Job"]!='unemployed')]  
percent_employ=round((len(employed_client[employed_client['Default']==1])/len(data)*100)  
print(percent_employ)
```

3.42

```
In [34]: #percentage of unemployed clients are loan defaulters?  
print(round(percent_default-percent_employ,2))
```

6.12

```
In [43]: #percentage of married clients are loan defaulters?  
married_client=data[data['Status']=='married']  
percent_married_default=round(len(married_client[married_client['Default']==1])/len(data)*100)  
print(percent_married_default)
```

5.71

```
In [48]: #top 3 jobs according to employee counts  
data["Job"].value_counts().sort_values(ascending=False).head(3)
```

```
Out[48]: blue-collar      19660  
         management      13078  
         technician      11864  
         Name: Job, dtype: int64
```

```
In [70]: #3 job types have the most loan defaulters  
data_default=data[data['Default']==1]  
top_job_default=data_default['Job'].value_counts().sort_values(ascending=False).head(3)  
top_job_default
```

```
Out[70]: unemployed      4605  
         self-employed    883  
         services         390  
         Name: Job, dtype: int64
```

```
In [71]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 75755 entries, 0 to 75754  
Data columns (total 17 columns):
```

#	Column	Non-Null Count	Dtype
0	CaseNo	75755 non-null	float64
1	Default	75755 non-null	int64
2	CreditBalPerc	75755 non-null	float64
3	DebtPerc	75755 non-null	float64
4	LateUpto60Days	75755 non-null	int64
5	Late60to90Days	75755 non-null	int64
6	LateOver90Days	75755 non-null	int64
7	Income	75755 non-null	float64
8	NoOfOpenLoans	75755 non-null	int64
9	NoOfHomeLoans	75755 non-null	int64
10	Dependents	75755 non-null	int64
11	Age	75755 non-null	int64
12	Job	75755 non-null	object
13	Status	75755 non-null	object
14	Education	75755 non-null	object
15	CreditBalance_Income_Ratio	75755 non-null	float64
16	Cat_income	75755 non-null	object

```
dtypes: float64(5), int64(8), object(4)
```

```
memory usage: 9.8+ MB
```

Sauvegarder les modifications

```
In [72]: data.to_csv('data_modified.csv', index=False)
```

Dashboard : Loan Default Analysis

Number of case

75,76K

Filtre by Age

21

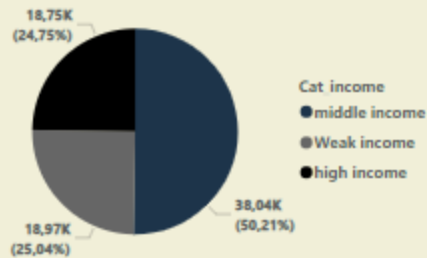
70

Filtre by Default

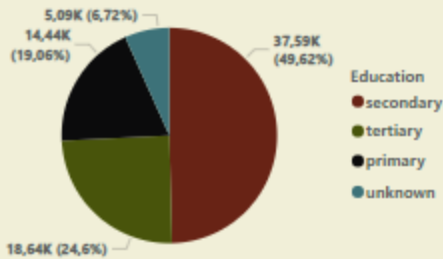
0

1

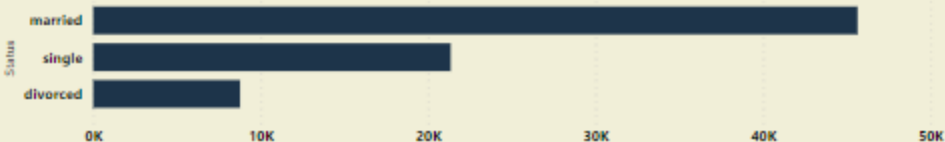
Income Category



Education



Count of Status



Count of Job

