

Build a machine learning model to predict if an applicant is 'good' or 'bad' client, different from other tasks, the definition of 'good' or 'bad' is not given. You should use some technique, such as vintage analysis to construct you label. Also, unbalance data problem is a big problem in this task

```
from google.colab import drive
drive.mount('/content/drive')
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

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

```
df = pd.read_csv('/content/drive/MyDrive/Colab Notebooks/credit_record.csv')
```

df



	ID	MONTHS_BALANCE	STATUS
0	5001711	0	X
1	5001711	-1	0
2	5001711	-2	0
3	5001711	-3	0
4	5001712	0	C
...
1048570	5150487	-25	C
1048571	5150487	-26	C
1048572	5150487	-27	C
1048573	5150487	-28	C
1048574	5150487	-29	C

1048575 rows × 3 columns

df.describe()

	ID	MONTHS_BALANCE
count	1.048575e+06	1.048575e+06
mean	5.068286e+06	-1.913700e+01
std	4.615058e+04	1.402350e+01
min	5.001711e+06	-6.000000e+01
25%	5.023644e+06	-2.900000e+01
50%	5.062104e+06	-1.700000e+01
75%	5.113856e+06	-7.000000e+00
max	5.150487e+06	0.000000e+00

df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1048575 entries, 0 to 1048574
Data columns (total 3 columns):
#   Column          Non-Null Count  Dtype
---  -
0    ID              1048575 non-null  int64
1   MONTHS_BALANCE  1048575 non-null  int64
2    STATUS         1048575 non-null  object
dtypes: int64(2), object(1)
memory usage: 24.0+ MB
```

```
df.isnull().count()
```

```
ID          1048575
MONTHS_BALANCE 1048575
STATUS        1048575
dtype: int64
```

```
df = df.dropna()
```

```
df.isnull().sum()
```

```
ID          0
MONTHS_BALANCE 0
STATUS        0
dtype: int64
```

```
data = pd.read_csv('/content/drive/MyDrive/Colab Notebooks/application_record.csv')
```

```
data
```

	ID	CODE_GENDER	FLAG_OWN_CAR	FLAG_OWN_REALTY	CNT_CHILDREN	AMT_INCOME_TC
0	5008804	M	Y	Y	0	4275
1	5008805	M	Y	Y	0	4275
2	5008806	M	Y	Y	0	1125
3	5008808	F	N	Y	0	2700
4	5008809	F	N	Y	0	2700
...	
438552	6840104	M	N	Y	0	1350
438553	6840222	F	N	N	0	1035
438554	6841878	F	N	N	0	540
438555	6842765	F	N	Y	0	720
438556	6842885	F	N	Y	0	1215

438557 rows × 18 columns

```
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 438557 entries, 0 to 438556
Data columns (total 18 columns):
#   Column                Non-Null Count  Dtype
---  -
0   ID                    438557 non-null  int64
1   CODE_GENDER          438557 non-null  object
2   FLAG_OWN_CAR          438557 non-null  object
3   FLAG_OWN_REALTY      438557 non-null  object
4   CNT_CHILDREN          438557 non-null  int64
5   AMT_INCOME_TOTAL     438557 non-null  float64
6   NAME_INCOME_TYPE      438557 non-null  object
7   NAME_EDUCATION_TYPE   438557 non-null  object
8   NAME_FAMILY_STATUS    438557 non-null  object
9   NAME_HOUSING_TYPE     438557 non-null  object
10  DAYS_BIRTH            438557 non-null  int64
11  DAYS_EMPLOYED         438557 non-null  int64
12  FLAG_MOBIL            438557 non-null  int64
13  FLAG_WORK_PHONE       438557 non-null  int64
14  FLAG_PHONE            438557 non-null  int64
15  FLAG_EMAIL            438557 non-null  int64
```

```
16 OCCUPATION_TYPE      304354 non-null object
17 CNT_FAM_MEMBERS      438557 non-null float64
dtypes: float64(2), int64(8), object(8)
memory usage: 60.2+ MB
```

```
data.isnull().sum()
```

```
ID                0
CODE_GENDER       0
FLAG_OWN_CAR      0
FLAG_OWN_REALTY   0
CNT_CHILDREN      0
AMT_INCOME_TOTAL  0
NAME_INCOME_TYPE  0
NAME_EDUCATION_TYPE 0
NAME_FAMILY_STATUS 0
NAME_HOUSING_TYPE 0
DAYS_BIRTH        0
DAYS_EMPLOYED     0
FLAG_MOBIL        0
FLAG_WORK_PHONE   0
FLAG_PHONE        0
FLAG_EMAIL        0
OCCUPATION_TYPE   134203
CNT_FAM_MEMBERS   0
dtype: int64
```

```
data.dropna()
```

	ID	CODE_GENDER	FLAG_OWN_CAR	FLAG_OWN_REALTY	CNT_CHILDREN	AMT_INCOME_T
2	5008806	M	Y	Y	0	1125
3	5008808	F	N	Y	0	2700
4	5008809	F	N	Y	0	2700
5	5008810	F	N	Y	0	2700
6	5008811	F	N	Y	0	2700
...	
438541	6837707	M	N	Y	0	2025
438548	6839936	M	Y	Y	1	1350
438553	6840222	F	N	N	0	1035
438554	6841878	F	N	N	0	540
438556	6842885	F	N	Y	0	1215

304354 rows × 18 columns

```
data.isnull().sum()
```

```
ID                0
CODE_GENDER       0
FLAG_OWN_CAR      0
FLAG_OWN_REALTY   0
CNT_CHILDREN      0
AMT_INCOME_TOTAL  0
NAME_INCOME_TYPE  0
NAME_EDUCATION_TYPE 0
NAME_FAMILY_STATUS 0
NAME_HOUSING_TYPE 0
DAYS_BIRTH        0
DAYS_EMPLOYED     0
FLAG_MOBIL        0
```

```
FLAG_WORK_PHONE      0
FLAG_PHONE            0
FLAG_EMAIL            0
OCCUPATION_TYPE      134203
CNT_FAM_MEMBERS       0
dtype: int64
```

```
data.isna().sum()
```

```
ID                0
CODE_GENDER       0
FLAG_OWN_CAR      0
FLAG_OWN_REALTY   0
CNT_CHILDREN      0
AMT_INCOME_TOTAL  0
NAME_INCOME_TYPE  0
NAME_EDUCATION_TYPE 0
NAME_FAMILY_STATUS 0
NAME_HOUSING_TYPE 0
DAYS_BIRTH        0
DAYS_EMPLOYED     0
FLAG_MOBIL        0
FLAG_WORK_PHONE   0
FLAG_PHONE        0
FLAG_EMAIL        0
CNT_FAM_MEMBERS   0
dtype: int64
```

```
data.drop(['OCCUPATION_TYPE'], axis=1, inplace=True)
```

```
data.isnull().sum()
```

```
ID                0
CODE_GENDER       0
FLAG_OWN_CAR      0
FLAG_OWN_REALTY   0
CNT_CHILDREN      0
AMT_INCOME_TOTAL  0
NAME_INCOME_TYPE  0
NAME_EDUCATION_TYPE 0
NAME_FAMILY_STATUS 0
NAME_HOUSING_TYPE 0
DAYS_BIRTH        0
DAYS_EMPLOYED     0
FLAG_MOBIL        0
FLAG_WORK_PHONE   0
FLAG_PHONE        0
FLAG_EMAIL        0
CNT_FAM_MEMBERS   0
dtype: int64
```

```
join = pd.merge(df,data)
```

```
join
```

```

    ID  MONTHS_BALANCE  STATUS  CODE_GENDER  FLAG_OWN_CAR  FLAG_OWN_REALTY  CNT
0    5008804           0      C             M             Y             Y
1    5008804          -1      C             M             Y             Y
2    5008804          -2      C             M             Y             Y
...
join1 = pd.merge(df,data, on = 'ID', how='inner')
join1
```

	ID	MONTHS_BALANCE	STATUS	CODE_GENDER	FLAG_OWN_CAR	FLAG_OWN_REALTY	CNT
0	5008804	0	C	M	Y	Y	
1	5008804	-1	C	M	Y	Y	
2	5008804	-2	C	M	Y	Y	
3	5008804	-3	C	M	Y	Y	
4	5008804	-4	C	M	Y	Y	
...	
777710	5150487	-25	C	M	Y	N	
777711	5150487	-26	C	M	Y	N	
777712	5150487	-27	C	M	Y	N	
777713	5150487	-28	C	M	Y	N	
777714	5150487	-29	C	M	Y	N	

777715 rows × 19 columns

```

join1.drop(['MONTHS_BALANCE','STATUS'], axis=1, inplace=True)

join1.info()
```

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 777715 entries, 0 to 777714
Data columns (total 17 columns):
#   Column                Non-Null Count  Dtype
---  -
0   ID                    777715 non-null  int64
1   CODE_GENDER           777715 non-null  object
2   FLAG_OWN_CAR           777715 non-null  object
3   FLAG_OWN_REALTY        777715 non-null  object
4   CNT_CHILDREN           777715 non-null  int64
5   AMT_INCOME_TOTAL       777715 non-null  float64
6   NAME_INCOME_TYPE       777715 non-null  object
7   NAME_EDUCATION_TYPE     777715 non-null  object
8   NAME_FAMILY_STATUS     777715 non-null  object
9   NAME_HOUSING_TYPE       777715 non-null  object
10  DAYS_BIRTH             777715 non-null  int64
11  DAYS_EMPLOYED           777715 non-null  int64
12  FLAG_MOBIL             777715 non-null  int64
13  FLAG_WORK_PHONE         777715 non-null  int64
14  FLAG_PHONE              777715 non-null  int64
15  FLAG_EMAIL              777715 non-null  int64
16  CNT_FAM_MEMBERS         777715 non-null  float64
dtypes: float64(2), int64(8), object(7)
memory usage: 106.8+ MB
```

```

##AMT_INCOME_TOTAL, NAME_EDUCATION_TYPE ,NAME_INCOME_TYPE ,NAME_FAMILY_STATUS, NAME_HOUSING_TYPE
##student are considered as bad
###commercial associate and state_servent
```

```
join1['NAME_INCOME_TYPE'].unique()

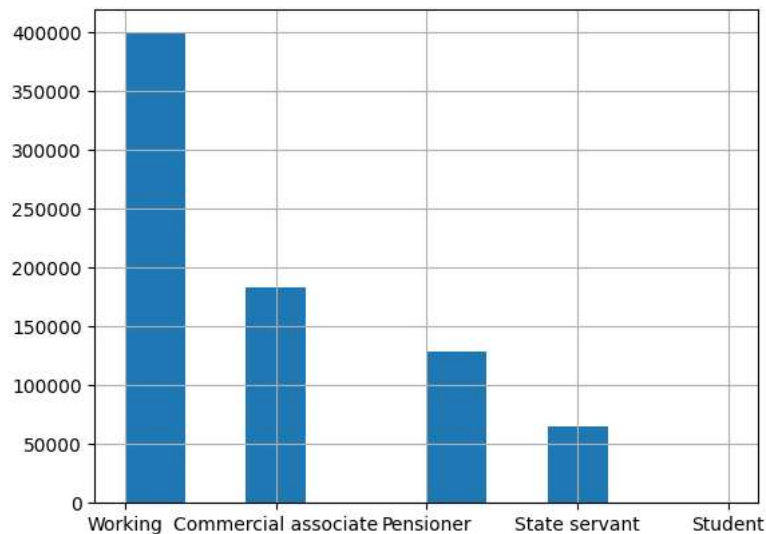
array(['Working', 'Commercial associate', 'Pensioner', 'State servant',
       'Student'], dtype=object)

join1.isnull().sum()

ID                0
CODE_GENDER       0
FLAG_OWN_CAR      0
FLAG_OWN_REALTY   0
CNT_CHILDREN      0
AMT_INCOME_TOTAL  0
NAME_INCOME_TYPE  0
NAME_EDUCATION_TYPE 0
NAME_FAMILY_STATUS 0
NAME_HOUSING_TYPE 0
DAYS_BIRTH        0
DAYS_EMPLOYED     0
FLAG_MOBIL        0
FLAG_WORK_PHONE   0
FLAG_PHONE        0
FLAG_EMAIL        0
CNT_FAM_MEMBERS   0
dtype: int64
```

```
join1['NAME_INCOME_TYPE'].hist()
```

<Axes: >



```
from sklearn.preprocessing import LabelEncoder
features = ['AMT_INCOME_TOTAL', 'NAME_EDUCATION_TYPE', 'NAME_INCOME_TYPE', 'NAME_FAMILY_STATUS', 'NAME_HOUSING_TYPE']

le = LabelEncoder()

for col in features:
    join1[col] = le.fit_transform(join1[col])

join1.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 777715 entries, 0 to 777714
Data columns (total 17 columns):
#   Column                Non-Null Count  Dtype
---  -
0   ID                    777715 non-null  int64
1   CODE_GENDER          777715 non-null  object
2   FLAG_OWN_CAR          777715 non-null  object
3   FLAG_OWN_REALTY      777715 non-null  object
4   CNT_CHILDREN         777715 non-null  int64
5   AMT_INCOME_TOTAL     777715 non-null  int64
6   NAME_INCOME_TYPE     777715 non-null  int64
7   NAME_EDUCATION_TYPE  777715 non-null  int64
8   NAME_FAMILY_STATUS   777715 non-null  int64
9   NAME_HOUSING_TYPE    777715 non-null  int64
10  DAYS_BIRTH           777715 non-null  int64
11  DAYS_EMPLOYED        777715 non-null  int64
```

```

12 FLAG_MOBIL          777715 non-null int64
13 FLAG_WORK_PHONE     777715 non-null int64
14 FLAG_PHONE          777715 non-null int64
15 FLAG_EMAIL          777715 non-null int64
16 CNT_FAM_MEMBERS     777715 non-null float64
dtypes: float64(1), int64(13), object(3)
memory usage: 106.8+ MB

```

```
join1['Applicant_type'] = ""
```

```
join1.head()
```

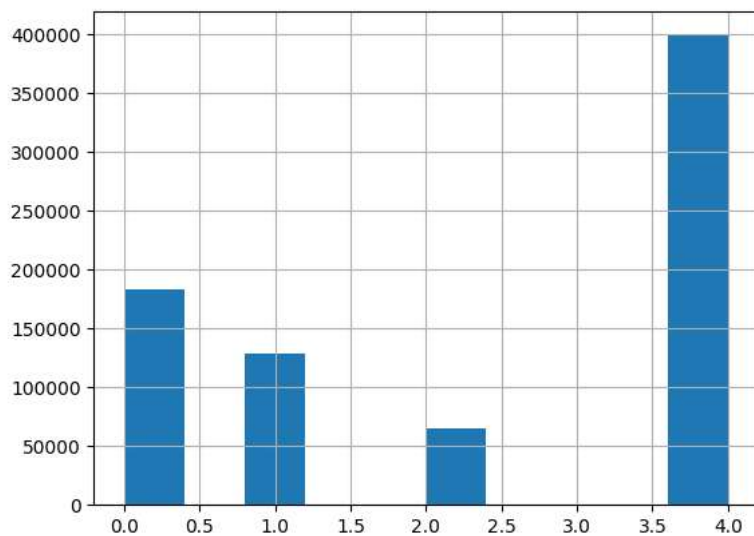
	ID	CODE_GENDER	FLAG_OWN_CAR	FLAG_OWN_REALTY	CNT_CHILDREN	AMT_INCOME_TOTAL
0	5008804	M	Y	Y	0	226
1	5008804	M	Y	Y	0	226
2	5008804	M	Y	Y	0	226
3	5008804	M	Y	Y	0	226
4	5008804	M	Y	Y	0	226

```
join1['NAME_INCOME_TYPE'].unique()
```

```
array([4, 0, 1, 2, 3])
```

```
join1['NAME_INCOME_TYPE'].hist()
```

<Axes: >



✓ class 0 is for commercial associate

class 1 for pensioner

class 2 for state_servant

class 3 for students

class 4 for working

```
join1['Applicant_type'] = join1['NAME_INCOME_TYPE'].apply(lambda x: 'Bad' if x == 3 else 'Good')
```

```
join1.head()
```

	ID	CODE_GENDER	FLAG_OWN_CAR	FLAG_OWN_REALTY	CNT_CHILDREN	AMT_INCOME_TOTAL
0	5008804	M	Y	Y	0	226
1	5008804	M	Y	Y	0	226
2	5008804	M	Y	Y	0	226
3	5008804	M	Y	Y	0	226
4	5008804	M	Y	Y	0	226

```
join1['Applicant_type'] = le.fit_transform(join1['Applicant_type'])
```

```
join1.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 777715 entries, 0 to 777714
Data columns (total 18 columns):
#   Column                Non-Null Count  Dtype
---  -
0   ID                    777715 non-null  int64
1   CODE_GENDER          777715 non-null  object
2   FLAG_OWN_CAR         777715 non-null  object
3   FLAG_OWN_REALTY      777715 non-null  object
4   CNT_CHILDREN         777715 non-null  int64
5   AMT_INCOME_TOTAL    777715 non-null  int64
6   NAME_INCOME_TYPE     777715 non-null  int64
7   NAME_EDUCATION_TYPE  777715 non-null  int64
8   NAME_FAMILY_STATUS   777715 non-null  int64
9   NAME_HOUSING_TYPE    777715 non-null  int64
10  DAYS_BIRTH           777715 non-null  int64
11  DAYS_EMPLOYED        777715 non-null  int64
12  FLAG_MOBIL           777715 non-null  int64
13  FLAG_WORK_PHONE      777715 non-null  int64
14  FLAG_PHONE           777715 non-null  int64
15  FLAG_EMAIL           777715 non-null  int64
16  CNT_FAM_MEMBERS      777715 non-null  float64
17  Applicant_type       777715 non-null  int64
dtypes: float64(1), int64(14), object(3)
memory usage: 112.7+ MB
```

```
temp = join1.drop(columns = ['DAYS_BIRTH', 'FLAG_EMAIL', 'FLAG_WORK_PHONE', 'FLAG_MOBIL', 'CNT_CHILDREN'])
```

```
temp.duplicated().sum()
```

```
741258
```

```
Y = ['Applicant_type']
```

```
X = join1.drop(['Applicant_type'], axis=1)
```

```
X['CODE_GENDER'] = le.fit_transform(X['CODE_GENDER'])
```

```
X['FLAG_OWN_CAR'] = le.fit_transform(X['FLAG_OWN_CAR'])
```

```
X['FLAG_OWN_REALTY'] = le.fit_transform(X['FLAG_OWN_REALTY'])
```

```
X.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 777715 entries, 0 to 777714
Data columns (total 18 columns):
#   Column                Non-Null Count  Dtype
---  -
0   ID                    777715 non-null  int64
1   CODE_GENDER          777715 non-null  int64
2   FLAG_OWN_CAR         777715 non-null  int64
3   FLAG_OWN_REALTY      777715 non-null  int64
4   CNT_CHILDREN         777715 non-null  int64
5   AMT_INCOME_TOTAL    777715 non-null  int64
6   NAME_INCOME_TYPE     777715 non-null  int64
7   NAME_EDUCATION_TYPE  777715 non-null  int64
8   NAME_FAMILY_STATUS   777715 non-null  int64
```



```

9  NAME_HOUSING_TYPE    777715 non-null int64
10 DAYS_BIRTH          777715 non-null int64
11 DAYS_EMPLOYED       777715 non-null int64
12 FLAG_MOBIL          777715 non-null int64
13 FLAG_WORK_PHONE     777715 non-null int64
14 FLAG_PHONE          777715 non-null int64
15 FLAG_EMAIL          777715 non-null int64
16 CNT_FAM_MEMBERS     777715 non-null float64
17 FLAG_OWN_CAR        777715 non-null int64
dtypes: float64(1), int64(17)
memory usage: 112.7 MB

```

```
X.value_counts()
```

```

ID      CODE_GENDER  FLAG_OWN_CAR  FLAG_OWN_REALTY  CNT_CHILDREN  AMT_INCOME_TOTAL  NAME_INCOME_TYPE  NAME_EDUCATION_TYPE
NAME_FAMILY_STATUS  NAME_HOUSING_TYPE  DAYS_BIRTH  DAYS_EMPLOYED  FLAG_MOBIL  FLAG_WORK_PHONE  FLAG_PHONE  FLAG_EMAIL  CNT_FAM_MEMBERS
FLAG_OWN_CAR
5148819  0          1          1          0          104          4          1          0
1          -19841      -4428          1          1          1          0          2.0          1          61
5115964  1          1          1          2          104          4          4          1          1
1          -14677      -3938          1          1          0          0          4.0          1          61
5061741  0          0          1          0          104          1          4          3
1          -23929      365243          1          0          1          0          1.0          0          61
5078799  0          0          1          0          193          4          1          0
1          -19808      -390          1          1          0          0          2.0          0          61
5061685  1          0          0          0          192          4          4          1
1          -11822      -4246          1          0          0          0          2.0          0          61

..
5139553  0          1          1          2          120          0          1          1
1          -13584      -6337          1          0          1          0          4.0          1          1
5069020  0          0          1          0          158          4          4          1
1          -20295      -3700          1          0          0          0          2.0          0          1
5097025  1          1          0          0          157          0          1          1
1          -13643      -2956          1          0          0          0          2.0          1          1
5023604  0          0          1          0          104          4          4          1
1          -20323      -1727          1          0          0          0          2.0          0          1
5092141  1          0          0          0          89          4          1          3
1          -11162      -1327          1          1          1          0          1.0          0          1
Length: 36457, dtype: int64

```