

```
In [2]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')
```

```
In [3]: df = pd.read_csv('application_record.csv')
df = pd.DataFrame(df)
```

```
In [6]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 438557 entries, 0 to 438556
Data columns (total 19 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  int64
18   STATUS                              36457 non-null   object
dtypes: float64(1), int64(9), object(9)
memory usage: 63.6+ MB
```

```
In [82]: df.shape
```

```
Out[82]: (438557, 19)
```

```
In [8]: df[df.duplicated()]
```

Out[8]:

```
In [9]: df.isnull().sum()
```

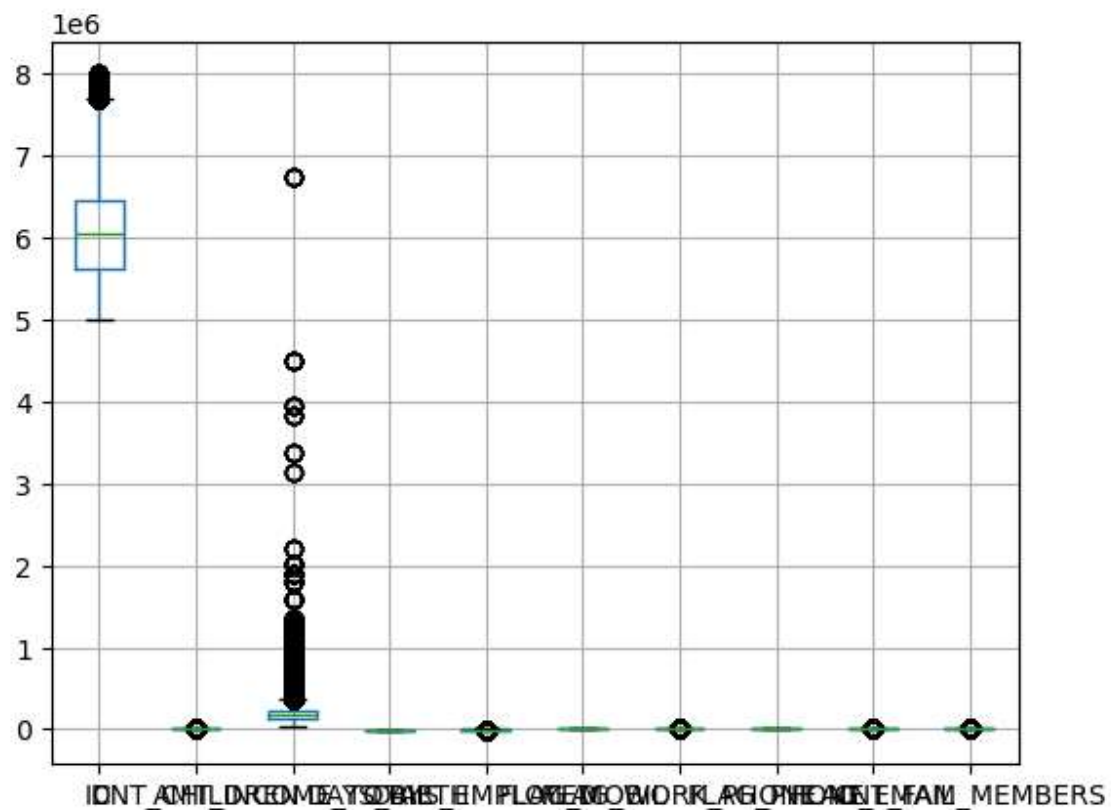
```
Out[9]: ID                                0
        CODE_GENDER                      0
        FLAG_OWN_CAR                     0
        FLAG_OWN_REALTY                  0
        CNT_CHILDREN                     0
        AMT_INCOME_TOTAL                 0
        NAME_INCOME_TYPE                 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
        STATUS                           402100
        dtype: int64
```

```
In [4]: df.dropna(subset = ['OCCUPATION_TYPE'],inplace = True , ignore_index= True)
```

```
In [84]: df.isnull().sum()
```

```
Out[84]: 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                  0
        CNT_FAM_MEMBERS                  0
        STATUS                           279220
        dtype: int64
```

```
In [22]: df.boxplot()  
plt.show()
```



```
In [5]: test = df[df['STATUS'].isna()]
```

```
In [6]: train = df[df['STATUS'].notna()]
```

```
In [28]: test.isnull().sum()
```

```
Out[28]: 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                   0
        CNT_FAM_MEMBERS                   0
        STATUS                           279220
        dtype: int64
```

```
In [27]: train.isnull().sum()
```

```
Out[27]: 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                   0
        CNT_FAM_MEMBERS                   0
        STATUS                           0
        dtype: int64
```

```
In [7]: from sklearn.preprocessing import LabelEncoder
```

```
In [8]: train.shape , test.shape
```

```
Out[8]: ((25134, 19), (279220, 19))
```

```
In [9]: X_train = train.iloc[:, train.columns != 'STATUS']  
y_train = train[['STATUS']]
```

```
In [10]: X_test = test.iloc[:, test.columns != 'STATUS']  
y_test = test[['STATUS']]
```

```
In [11]: for col in X_train.columns:  
    if X_train[col].dtypes == 'object':  
        X_train[col] = LabelEncoder().fit_transform(X_train[col])
```

```
In [12]: for col in X_test.columns:  
    if X_test[col].dtypes == 'object':  
        X_test[col] = LabelEncoder().fit_transform(X_test[col])
```

```
In [13]: from sklearn.tree import DecisionTreeClassifier  
from sklearn.model_selection import cross_val_score, GridSearchCV  
from sklearn.metrics import accuracy_score
```

```
In [14]: dtr = DecisionTreeClassifier(random_state=42)  
param = {'criterion': ['gini', 'entropy'],  
        'max_depth': range(1,16)}  
gscv = GridSearchCV(dtr, param, scoring='accuracy', cv=5)  
gscv.fit(X_train, y_train)
```

```
Out[14]: GridSearchCV(cv=5, estimator=DecisionTreeClassifier(random_state=42),  
                    param_grid={'criterion': ['gini', 'entropy'],  
                                'max_depth': range(1, 16)},  
                    scoring='accuracy')
```

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```
In [15]: dt = gscv.best_estimator_
```

```
In [16]: dt
```

```
Out[16]: DecisionTreeClassifier(max_depth=2, random_state=42)
```

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```
In [17]: gscv.feature_names_in_
```

```
Out[17]: array(['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_TYP  
E',  
              'DAYS_BIRTH', 'DAYS_EMPLOYED', 'FLAG_MOBIL', 'FLAG_WORK_PHONE',  
              'FLAG_PHONE', 'FLAG_EMAIL', 'OCCUPATION_TYPE', 'CNT_FAM_MEMBER  
S'],  
          dtype=object)
```

```
In [18]: dt.fit(X_train , y_train)  
score = cross_val_score(dt , X_train , y_train , scoring='accuracy' , c  
pred = dt.predict(X_test)
```

```
In [19]: score
```

```
Out[19]: 0.49013288449297787
```

```
In [20]: ypred = dt.predict(X_train)  
accuracy = accuracy_score(y_train , ypred)
```

```
In [21]: accuracy
```

```
Out[21]: 0.4901328877218111
```

```
In [22]: pred
```

```
Out[22]: array(['C', 'C', 'C', ..., 'C', 'C', 'C'], dtype=object)
```

```
In [23]: X_test.index
```

```
Out[23]: Index([      7,      35,      36,      37,      53,      57,     123,     124,  
              125,  
              149,  
              ...  
              304344, 304345, 304346, 304347, 304348, 304349, 304350, 304351,  
              304352,  
              304353],  
              dtype='int64', length=279220)
```

```
In [24]: ts = pd.DataFrame(pred , index = X_test.index , columns = ['STATUS'])
```

```
In [25]: ts.nunique()
```

```
Out[25]: STATUS      1  
dtype: int64
```

In [26]: `ts.value_counts()`

Out[26]: STATUS
C 279220
Name: count, dtype: int64

In [27]: `X_test`

Out[27]:

	ID	CODE_GENDER	FLAG_OWN_CAR	FLAG_OWN_REALTY	CNT_CHILDREN	AM
7	6153651	1	1	1	0	
35	6153733	1	1	1	0	
36	6153734	1	1	1	0	
37	6153735	1	1	1	0	
53	6153736	0	1	1	2	
...
304349	6837707	1	0	1	0	
304350	6839936	1	1	1	1	
304351	6840222	0	0	0	0	
304352	6841878	0	0	0	0	
304353	6842885	0	0	1	0	

279220 rows × 18 columns



```
In [28]: pd.merge(X_test , ts , how = 'left' , left_index= True , right_index= Tr
```

```
Out[28]:
```

	ID	CODE_GENDER	FLAG_OWN_CAR	FLAG_OWN_REALTY	CNT_CHILDREN	AM
	7	6153651	1	1	1	0
	35	6153733	1	1	1	0
	36	6153734	1	1	1	0
	37	6153735	1	1	1	0
	53	6153736	0	1	1	2

	304349	6837707	1	0	1	0
	304350	6839936	1	1	1	1
	304351	6840222	0	0	0	0
	304352	6841878	0	0	0	0
	304353	6842885	0	0	1	0

279220 rows × 19 columns



```
In [29]: df['STATUS'].fillna('C',inplace = True)
```

```
In [30]: df.isnull().sum()
```

```
Out[30]: 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    0
CNT_FAM_MEMBERS    0
STATUS            0
dtype: int64
```

```
In [31]: df.shape
```

```
Out[31]: (304354, 19)
```


In [34]: `df.dtypes`

```
Out[34]: ID                int64
CODE_GENDER              object
FLAG_OWN_CAR              object
FLAG_OWN_REALTY           object
CNT_CHILDREN              int64
AMT_INCOME_TOTAL          float64
NAME_INCOME_TYPE          object
NAME_EDUCATION_TYPE       object
NAME_FAMILY_STATUS        object
NAME_HOUSING_TYPE         object
DAYS_BIRTH                int64
DAYS_EMPLOYED             int64
FLAG_MOBIL                int64
FLAG_WORK_PHONE           int64
FLAG_PHONE                int64
FLAG_EMAIL                int64
OCCUPATION_TYPE           object
CNT_FAM_MEMBERS           int64
STATUS                    object
dtype: object
```

In [29]: `df.head()`

```
Out[29]:
```

	ID	CODE_GENDER	FLAG_OWN_CAR	FLAG_OWN_REALTY	CNT_CHILDREN	AMT_INCOME_TOTAL
0	5008806	M	Y	Y	0	150000
1	5008808	F	N	Y	0	160000
2	5008809	F	N	Y	0	170000
3	5008810	F	N	Y	0	180000
4	5008811	F	N	Y	0	190000

```
In [43]: df.isnull().sum()
```

```
Out[43]: 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                 0
          CNT_FAM_MEMBERS                 0
          STATUS                          0
          dtype: int64
```

```
In [32]: df1 = df.copy()
```

```
In [33]: df['TARGET'] = 'NAN'
```

```
In [34]: df['TARGET'] = np.where(df['STATUS'] == 'X' , 'approve' , df['TARGET'])
```

```
In [35]: df['TARGET'] = np.where(df['STATUS'] == 'C' , 'approve' , df['TARGET'])
```

```
In [36]: df['TARGET'] = np.where(df['STATUS'] == '0' , 'approve' , df['TARGET'])
```

```
In [37]: df['TARGET'] = np.where(df['STATUS'] == '1' , 'notapprove' , df['TARGET'])
```

```
In [38]: df['TARGET'] = np.where(df['STATUS'] == '2' , 'notapprove' , df['TARGET'])
```

```
In [39]: df['TARGET'] = np.where(df['STATUS'] == '3' , 'notapprove' , df['TARGET'])
```

```
In [40]: df['TARGET'] = np.where(df['STATUS'] == '4' , 'notapprove' , df['TARGET'])
```

```
In [41]: df['TARGET'] = np.where(df['STATUS'] == '5' , 'notapprove' , df['TARGET'])
```

```
In [42]: df['TARGET'] = np.where(df['STATUS'] == '6' , 'notapprove' , df['TARGET'])
```

In [43]: `df.head()`

Out[43]:

	ID	CODE_GENDER	FLAG_OWN_CAR	FLAG_OWN_REALTY	CNT_CHILDREN	AMT_INCOME_TOTAL
0	5008806	M	Y	Y	0	
1	5008808	F	N	Y	0	
2	5008809	F	N	Y	0	
3	5008810	F	N	Y	0	
4	5008811	F	N	Y	0	

In [44]: `df.dtypes`

Out[44]:

ID	int64
CODE_GENDER	object
FLAG_OWN_CAR	object
FLAG_OWN_REALTY	object
CNT_CHILDREN	int64
AMT_INCOME_TOTAL	float64
NAME_INCOME_TYPE	object
NAME_EDUCATION_TYPE	object
NAME_FAMILY_STATUS	object
NAME_HOUSING_TYPE	object
DAYS_BIRTH	int64
DAYS_EMPLOYED	int64
FLAG_MOBIL	int64
FLAG_WORK_PHONE	int64
FLAG_PHONE	int64
FLAG_EMAIL	int64
OCCUPATION_TYPE	object
CNT_FAM_MEMBERS	int64
STATUS	object
TARGET	object
dtype:	object

In [45]: `df['STATUS'] = df['STATUS'].astype('str')`

In [46]: `df.drop(['CODE_GENDER'],axis = 1 , inplace= True)`

In [48]: `X = df.iloc[: , df.columns != 'TARGET']`
`y = df[['TARGET']]`

```
In [49]: ➤ for col in X.columns:
            if X[col].dtypes == 'object':
                X[col] = LabelEncoder().fit_transform(X[col])
```

```
In [50]: ➤ y.value_counts()
```

```
Out[50]: TARGET
approve      304073
notapprove    281
Name: count, dtype: int64
```

```
In [51]: ➤ from imblearn.over_sampling import SMOTE
```

```
In [52]: ➤ X_re , y_re = SMOTE(random_state=42).fit_resample(X , y)
```

```
In [53]: ➤ y_re.value_counts()
```

```
Out[53]: TARGET
approve      304073
notapprove    304073
Name: count, dtype: int64
```

```
In [54]: ➤ from sklearn.model_selection import train_test_split , GridSearchCV , cv
```

```
In [55]: ➤ X_train , X_test , y_train , y_test =train_test_split(X_re , y_re , tes
```

```
In [56]: ➤ from sklearn.ensemble import AdaBoostClassifier , GradientBoostingClass
```

```
In [ ]: ➤ ac = AdaBoostClassifier()
param = {'n_estimators': [20 , 30 , 70] ,
         'learning_rate':[0.001, 0.5 , 0.1,1,10,100,0.8]}
gscv = GridSearchCV(ac , param , scoring='accuracy' , cv = 5 )
gscv.fit(X_train , y_train)
```

```
In [57]: ➤ ac = AdaBoostClassifier()
```

```
In [58]: ➤ ac.fit(X_train , y_train)
```

```
Out[58]: AdaBoostClassifier()

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representation or trust the notebook.
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```

```
In [59]: ➤ yp = ac.predict(X_test)
```

```
In [60]: accuracy = accuracy_score(y_test , yp)
```

```
In [61]: accuracy
```

```
Out[61]: 1.0
```

```
In [62]: ga = GradientBoostingClassifier()
```

```
In [63]: ga.fit(X_train , y_train)
```

```
Out[63]: GradientBoostingClassifier()
```

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```
In [64]: ypg = ga.predict(X_test)
```

```
In [65]: accuracyg = accuracy_score(y_test , ypg)
accuracyg
```

```
Out[65]: 1.0
```

```
In [ ]: ## by using voting classifier
```

```
In [66]: from sklearn.ensemble import VotingClassifier
```

```
In [67]: vc = VotingClassifier([('dt' , DecisionTreeClassifier() ,
                                ('ab' , AdaBoostClassifier() ,
                                ('gb' , GradientBoostingClassifier()))])
```

```
In [68]: vc.fit(X_train , y_train)
```

```
Out[68]: VotingClassifier(estimators=[('dt', DecisionTreeClassifier()),
                                      ('ab', AdaBoostClassifier()),
                                      ('gb', GradientBoostingClassifier())])
```

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```
In [69]: vp = vc.predict(X_test)
```

```
In [70]: accuracy = accuracy_score(y_test , vp)
```

```
In [71]: accuracy
```

```
Out[71]: 1.0
```

```
In [72]: vpp = pd.DataFrame(vp)
```

```
In [73]: vpp.value_counts()
```

```
Out[73]: approve      121978  
notapprove    121281  
Name: count, dtype: int64
```

```
In [74]: from sklearn.metrics import classification_report
```

```
In [75]: cm = classification_report(y_test , vp)
```

```
In [76]: print(cm)
```

	precision	recall	f1-score	support
approve	1.00	1.00	1.00	121978
notapprove	1.00	1.00	1.00	121281
accuracy			1.00	243259
macro avg	1.00	1.00	1.00	243259
weighted avg	1.00	1.00	1.00	243259

```
In [ ]:
```