

taiwan-credit-data-preproc

August 19, 2022

1 Taiwan Credit: Data Preprocessing for ML and Neural Network

```
[1]: from google.colab import drive
drive.mount('/content/gdrive/')
```

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

```
[2]: %cd /content/gdrive/MyDrive/Github/ml-blog
```

/content/gdrive/MyDrive/Github/ml-blog

```
[ ]: !pip install xlrd==1.2.0
!pip install matplotlib==3.5.3
```

```
[3]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import os
import xlrd

filename = 'default of credit card clients.xls'

DATA = os.path.relpath('/content/gdrive/MyDrive/Github/ml-blog/credit/data/' +
    filename)

df = pd.read_excel(DATA, 'Data', index_col=0, header=[1], na_values='NA')
df.head()
```

```
[3]:
```

	LIMIT_BAL	SEX	EDUCATION	MARRIAGE	AGE	PAY_0	PAY_2	PAY_3	PAY_4	\
ID										
1	20000	2	2	1	24	2	2	-1	-1	
2	120000	2	2	2	26	-1	2	0	0	
3	90000	2	2	2	34	0	0	0	0	
4	50000	2	2	1	37	0	0	0	0	
5	50000	1	2	1	57	-1	0	-1	0	

	PAY_5	...	BILL_AMT4	BILL_AMT5	BILL_AMT6	PAY_AMT1	PAY_AMT2	PAY_AMT3	\
ID		...							
1	-2	...	0	0	0	0	689	0	
2	0	...	3272	3455	3261	0	1000	1000	
3	0	...	14331	14948	15549	1518	1500	1000	
4	0	...	28314	28959	29547	2000	2019	1200	
5	0	...	20940	19146	19131	2000	36681	10000	

	PAY_AMT4	PAY_AMT5	PAY_AMT6	default	payment	next	month
ID							
1	0	0	0				1
2	1000	0	2000				1
3	1000	1000	5000				0
4	1100	1069	1000				0
5	9000	689	679				0

[5 rows x 24 columns]

```
[4]: X = df.iloc[:, :23]
      Y = df.iloc[:, 23]
      X.head(), Y.head()
```

	LIMIT_BAL	SEX	EDUCATION	MARRIAGE	AGE	PAY_0	PAY_2	PAY_3	PAY_4	\
ID										
1	20000	2	2	1	24	2	2	-1	-1	
2	120000	2	2	2	26	-1	2	0	0	
3	90000	2	2	2	34	0	0	0	0	
4	50000	2	2	1	37	0	0	0	0	
5	50000	1	2	1	57	-1	0	-1	0	

	PAY_5	...	BILL_AMT3	BILL_AMT4	BILL_AMT5	BILL_AMT6	PAY_AMT1	\
ID		...						
1	-2	...	689	0	0	0	0	
2	0	...	2682	3272	3455	3261	0	
3	0	...	13559	14331	14948	15549	1518	
4	0	...	49291	28314	28959	29547	2000	
5	0	...	35835	20940	19146	19131	2000	

	PAY_AMT2	PAY_AMT3	PAY_AMT4	PAY_AMT5	PAY_AMT6
ID					
1	689	0	0	0	0
2	1000	1000	1000	0	2000
3	1500	1000	1000	1000	5000
4	2019	1200	1100	1069	1000
5	36681	10000	9000	689	679

```
[5 rows x 23 columns], ID
1    1
2    1
3    0
4    0
5    0
Name: default payment next month, dtype: int64)
```

```
[10]: from sklearn.compose import ColumnTransformer
      from sklearn.preprocessing import StandardScaler, OneHotEncoder
```

```
[18]: ohe_cols = ['SEX', 'EDUCATION', 'MARRIAGE', 'PAY_0', 'PAY_2', 'PAY_3',
                'PAY_4', 'PAY_5', 'PAY_6']

      num_cols = ['LIMIT_BAL', 'AGE', 'BILL_AMT1', 'BILL_AMT2', 'BILL_AMT3',
                  'BILL_AMT4',
                  'BILL_AMT5', 'BILL_AMT6', 'PAY_AMT1', 'PAY_AMT2', 'PAY_AMT3',
                  'PAY_AMT4', 'PAY_AMT5', 'PAY_AMT6']
```

```
[24]: column_transform = ColumnTransformer(
      [ ('category', OneHotEncoder(handle_unknown='ignore'), ohe_cols),
        ('nums', StandardScaler(), num_cols)],
      remainder='drop')
```

```
[25]: column_transform.fit(X)
```

```
[25]: ColumnTransformer(transformers=[('category',
                                       OneHotEncoder(handle_unknown='ignore'),
                                       ['SEX', 'EDUCATION', 'MARRIAGE', 'PAY_0',
                                        'PAY_2', 'PAY_3', 'PAY_4', 'PAY_5',
                                        'PAY_6']),
                                       ('nums', StandardScaler(),
                                       ['LIMIT_BAL', 'AGE', 'BILL_AMT1', 'BILL_AMT2',
                                        'BILL_AMT3', 'BILL_AMT4', 'BILL_AMT5',
                                        'BILL_AMT6', 'PAY_AMT1', 'PAY_AMT2',
                                        'PAY_AMT3', 'PAY_AMT4', 'PAY_AMT5',
                                        'PAY_AMT6'])])
```

```
[26]: column_transform.get_feature_names_out()
```

```
[26]: array(['category__SEX_1', 'category__SEX_2', 'category__EDUCATION_0',
            'category__EDUCATION_1', 'category__EDUCATION_2',
            'category__EDUCATION_3', 'category__EDUCATION_4',
            'category__EDUCATION_5', 'category__EDUCATION_6',
            'category__MARRIAGE_0', 'category__MARRIAGE_1',
            'category__MARRIAGE_2', 'category__MARRIAGE_3',
            'category__PAY_0_-2', 'category__PAY_0_-1', 'category__PAY_0_0',
```

```

'category__PAY_0_1', 'category__PAY_0_2', 'category__PAY_0_3',
'category__PAY_0_4', 'category__PAY_0_5', 'category__PAY_0_6',
'category__PAY_0_7', 'category__PAY_0_8', 'category__PAY_2_-2',
'category__PAY_2_-1', 'category__PAY_2_0', 'category__PAY_2_1',
'category__PAY_2_2', 'category__PAY_2_3', 'category__PAY_2_4',
'category__PAY_2_5', 'category__PAY_2_6', 'category__PAY_2_7',
'category__PAY_2_8', 'category__PAY_3_-2', 'category__PAY_3_-1',
'category__PAY_3_0', 'category__PAY_3_1', 'category__PAY_3_2',
'category__PAY_3_3', 'category__PAY_3_4', 'category__PAY_3_5',
'category__PAY_3_6', 'category__PAY_3_7', 'category__PAY_3_8',
'category__PAY_4_-2', 'category__PAY_4_-1', 'category__PAY_4_0',
'category__PAY_4_1', 'category__PAY_4_2', 'category__PAY_4_3',
'category__PAY_4_4', 'category__PAY_4_5', 'category__PAY_4_6',
'category__PAY_4_7', 'category__PAY_4_8', 'category__PAY_5_-2',
'category__PAY_5_-1', 'category__PAY_5_0', 'category__PAY_5_2',
'category__PAY_5_3', 'category__PAY_5_4', 'category__PAY_5_5',
'category__PAY_5_6', 'category__PAY_5_7', 'category__PAY_5_8',
'category__PAY_6_-2', 'category__PAY_6_-1', 'category__PAY_6_0',
'category__PAY_6_2', 'category__PAY_6_3', 'category__PAY_6_4',
'category__PAY_6_5', 'category__PAY_6_6', 'category__PAY_6_7',
'category__PAY_6_8', 'nums__LIMIT_BAL', 'nums__AGE',
'nums__BILL_AMT1', 'nums__BILL_AMT2', 'nums__BILL_AMT3',
'nums__BILL_AMT4', 'nums__BILL_AMT5', 'nums__BILL_AMT6',
'nums__PAY_AMT1', 'nums__PAY_AMT2', 'nums__PAY_AMT3',
'nums__PAY_AMT4', 'nums__PAY_AMT5', 'nums__PAY_AMT6'], dtype=object)

```

```

[27]: X_prep = pd.DataFrame(column_transform.transform(X).toarray(),
    ↪ columns=column_transform.get_feature_names_out(), index=X.index)

```

```

[30]: X_prep.head()

```

```

[30]:
category__SEX_1  category__SEX_2  category__EDUCATION_0  \
ID
1                0.0                1.0                0.0
2                0.0                1.0                0.0
3                0.0                1.0                0.0
4                0.0                1.0                0.0
5                1.0                0.0                0.0

category__EDUCATION_1  category__EDUCATION_2  category__EDUCATION_3  \
ID
1                0.0                1.0                0.0
2                0.0                1.0                0.0
3                0.0                1.0                0.0
4                0.0                1.0                0.0
5                0.0                1.0                0.0

```

	category__EDUCATION_4	category__EDUCATION_5	category__EDUCATION_6	\
ID				
1	0.0	0.0	0.0	
2	0.0	0.0	0.0	
3	0.0	0.0	0.0	
4	0.0	0.0	0.0	
5	0.0	0.0	0.0	

	category__MARRIAGE_0	...	nums__BILL_AMT3	nums__BILL_AMT4	\
ID		...			
1	0.0	...	-0.667993	-0.672497	
2	0.0	...	-0.639254	-0.621636	
3	0.0	...	-0.482408	-0.449730	
4	0.0	...	0.032846	-0.232373	
5	0.0	...	-0.161189	-0.346997	

	nums__BILL_AMT5	nums__BILL_AMT6	nums__PAY_AMT1	nums__PAY_AMT2	\
ID					
1	-0.663059	-0.652724	-0.341942	-0.227086	
2	-0.606229	-0.597966	-0.341942	-0.213588	
3	-0.417188	-0.391630	-0.250292	-0.191887	
4	-0.186729	-0.156579	-0.221191	-0.169361	
5	-0.348137	-0.331482	-0.221191	1.335034	

	nums__PAY_AMT3	nums__PAY_AMT4	nums__PAY_AMT5	nums__PAY_AMT6
ID				
1	-0.296801	-0.308063	-0.314136	-0.293382
2	-0.240005	-0.244230	-0.314136	-0.180878
3	-0.240005	-0.244230	-0.248683	-0.012122
4	-0.228645	-0.237846	-0.244166	-0.237130
5	0.271165	0.266434	-0.269039	-0.255187

[5 rows x 91 columns]

[31]: X_prep.info()

```
<class 'pandas.core.frame.DataFrame'>
```

```
Int64Index: 30000 entries, 1 to 30000
```

```
Data columns (total 91 columns):
```

#	Column	Non-Null Count	Dtype
0	category__SEX_1	30000 non-null	float64
1	category__SEX_2	30000 non-null	float64
2	category__EDUCATION_0	30000 non-null	float64
3	category__EDUCATION_1	30000 non-null	float64
4	category__EDUCATION_2	30000 non-null	float64
5	category__EDUCATION_3	30000 non-null	float64
6	category__EDUCATION_4	30000 non-null	float64

7	category__EDUCATION_5	30000	non-null	float64
8	category__EDUCATION_6	30000	non-null	float64
9	category__MARRIAGE_0	30000	non-null	float64
10	category__MARRIAGE_1	30000	non-null	float64
11	category__MARRIAGE_2	30000	non-null	float64
12	category__MARRIAGE_3	30000	non-null	float64
13	category__PAY_0_-2	30000	non-null	float64
14	category__PAY_0_-1	30000	non-null	float64
15	category__PAY_0_0	30000	non-null	float64
16	category__PAY_0_1	30000	non-null	float64
17	category__PAY_0_2	30000	non-null	float64
18	category__PAY_0_3	30000	non-null	float64
19	category__PAY_0_4	30000	non-null	float64
20	category__PAY_0_5	30000	non-null	float64
21	category__PAY_0_6	30000	non-null	float64
22	category__PAY_0_7	30000	non-null	float64
23	category__PAY_0_8	30000	non-null	float64
24	category__PAY_2_-2	30000	non-null	float64
25	category__PAY_2_-1	30000	non-null	float64
26	category__PAY_2_0	30000	non-null	float64
27	category__PAY_2_1	30000	non-null	float64
28	category__PAY_2_2	30000	non-null	float64
29	category__PAY_2_3	30000	non-null	float64
30	category__PAY_2_4	30000	non-null	float64
31	category__PAY_2_5	30000	non-null	float64
32	category__PAY_2_6	30000	non-null	float64
33	category__PAY_2_7	30000	non-null	float64
34	category__PAY_2_8	30000	non-null	float64
35	category__PAY_3_-2	30000	non-null	float64
36	category__PAY_3_-1	30000	non-null	float64
37	category__PAY_3_0	30000	non-null	float64
38	category__PAY_3_1	30000	non-null	float64
39	category__PAY_3_2	30000	non-null	float64
40	category__PAY_3_3	30000	non-null	float64
41	category__PAY_3_4	30000	non-null	float64
42	category__PAY_3_5	30000	non-null	float64
43	category__PAY_3_6	30000	non-null	float64
44	category__PAY_3_7	30000	non-null	float64
45	category__PAY_3_8	30000	non-null	float64
46	category__PAY_4_-2	30000	non-null	float64
47	category__PAY_4_-1	30000	non-null	float64
48	category__PAY_4_0	30000	non-null	float64
49	category__PAY_4_1	30000	non-null	float64
50	category__PAY_4_2	30000	non-null	float64
51	category__PAY_4_3	30000	non-null	float64
52	category__PAY_4_4	30000	non-null	float64
53	category__PAY_4_5	30000	non-null	float64
54	category__PAY_4_6	30000	non-null	float64

```

55 category__PAY_4_7      30000 non-null float64
56 category__PAY_4_8      30000 non-null float64
57 category__PAY_5_-2     30000 non-null float64
58 category__PAY_5_-1     30000 non-null float64
59 category__PAY_5_0      30000 non-null float64
60 category__PAY_5_2      30000 non-null float64
61 category__PAY_5_3      30000 non-null float64
62 category__PAY_5_4      30000 non-null float64
63 category__PAY_5_5      30000 non-null float64
64 category__PAY_5_6      30000 non-null float64
65 category__PAY_5_7      30000 non-null float64
66 category__PAY_5_8      30000 non-null float64
67 category__PAY_6_-2     30000 non-null float64
68 category__PAY_6_-1     30000 non-null float64
69 category__PAY_6_0      30000 non-null float64
70 category__PAY_6_2      30000 non-null float64
71 category__PAY_6_3      30000 non-null float64
72 category__PAY_6_4      30000 non-null float64
73 category__PAY_6_5      30000 non-null float64
74 category__PAY_6_6      30000 non-null float64
75 category__PAY_6_7      30000 non-null float64
76 category__PAY_6_8      30000 non-null float64
77 nums__LIMIT_BAL       30000 non-null float64
78 nums__AGE              30000 non-null float64
79 nums__BILL_AMT1        30000 non-null float64
80 nums__BILL_AMT2        30000 non-null float64
81 nums__BILL_AMT3        30000 non-null float64
82 nums__BILL_AMT4        30000 non-null float64
83 nums__BILL_AMT5        30000 non-null float64
84 nums__BILL_AMT6        30000 non-null float64
85 nums__PAY_AMT1         30000 non-null float64
86 nums__PAY_AMT2         30000 non-null float64
87 nums__PAY_AMT3         30000 non-null float64
88 nums__PAY_AMT4         30000 non-null float64
89 nums__PAY_AMT5         30000 non-null float64
90 nums__PAY_AMT6         30000 non-null float64
dtypes: float64(91)
memory usage: 21.1 MB

```

```
[33]: X_prep.to_csv('./credit/data/taiwan-credit-col-transform-X.csv', header=True)
```

```
[ ]:
```