

## P01\_1\_EDA\_Categ

```
In [1]: import pandas as pd
import numpy as np
import regex as re
import warnings
warnings.filterwarnings('ignore')
from nltk.corpus import stopwords
from nltk.stem.porter import PorterStemmer
from nltk.stem import WordNetLemmatizer
import pickle

import matplotlib.pyplot as plt
import re

In [2]: import seaborn as sns
from sklearn.model_selection import train_test_split

In [3]: %config InlineBackend.figure_format = 'retina'

In [4]: from sklearn.impute import SimpleImputer
from sklearn.linear_model import LogisticRegression
from sklearn.pipeline import make_pipeline

from sklearn.preprocessing import FunctionTransformer

from sklearn.compose import ColumnTransformer

In [5]: import warnings
warnings.filterwarnings('ignore')

In [6]: import pandas as pd
from sklearn.compose import make_column_transformer
from sklearn.preprocessing import OneHotEncoder
from sklearn.linear_model import LogisticRegression
from sklearn.pipeline import make_pipeline
from sklearn.model_selection import cross_val_score
```

## Read data set

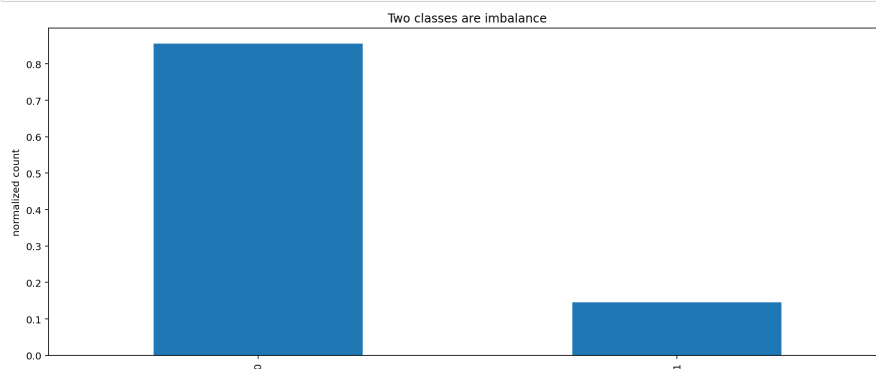
```
In [7]: file_path = "../DataSet/"
file_name = "exercise_40_train.csv"
df = pd.read_csv(file_path+file_name)

In [8]: df.shape

Out[8]: (40000, 101)

In [9]: df['y'].value_counts(normalize=True).plot(kind='bar', figsize=(15,6))
plt.ylabel('normalized count')
plt.title('Two classes are imbalance')
plt.savefig("../Figures/plot_01_1_imbalancedata.png")

plt.show()
```



Check out the object columns, number of uniques and number of NaNs

```
In [10]: df_categ = df.select_dtypes(include=['object'])
print(df_categ.columns)
columns_obj_dtype = df_categ.columns

# print('columns_obj_dtype = ', columns_obj_dtype)

print(100*'=')

number_of_Obj_cols_with_nan = 0
columns_obj_dtype_NaN = []

for col in columns_obj_dtype:
    if df_categ[col].isnull().sum() != 0:
        number_of_Obj_cols_with_nan = number_of_Obj_cols_with_nan + 1

        columns_obj_dtype_NaN.append(col)

    print('NaNs % in ', col, '=', (df_categ[col].isnull().sum())*100/(df_categ.shape[0]), '%',
          '\nNumber of uniqs =', df_categ[col].nunique(), '\n')
print('number_of_Obj_cols_with_nan ', number_of_Obj_cols_with_nan)

Index(['x3', 'x7', 'x19', 'x24', 'x31', 'x33', 'x39', 'x60', 'x65', 'x77',
      'x93', 'x99'],
      dtype='object')
=====
NaNs % in  x24 = 9.64 %
Number of uniqs = 2

NaNs % in  x33 = 17.9275 %
Number of uniqs = 51

NaNs % in  x77 = 23.1425 %
Number of uniqs = 7

NaNs % in  x99 = 32.09 %
Number of uniqs = 1

number_of_Obj_cols_with_nan 4

In [11]: df_categ['target'] = df[['y']]

In [12]: # target =
# df_categ = df_categ.join(target)
df_categ.head()

Out[12]:
```

		x3	x7		x19	x24	x31		x33	x39		x60	x65		x77	x93	x99	target
0	Wed	0.0062%	\$-908.650758424405	female	no	Colorado	5-10 miles	August	farmers	mercedes	no	yes	0					
1	Friday	0.0064%	\$-1864.9622875143	male	no	Tennessee	5-10 miles	April	allstate	mercedes	no	yes	1					
2	Thursday	-8e-04%	\$-543.187402955527	male	no	Texas	5-10 miles	September	geico	subaru	no	yes	1					
3	Tuesday	-0.0057%	\$-182.626380634258	male	no	Minnesota	5-10 miles	September	geico	nissan	no	yes	0					
4	Sunday	0.0109%	\$967.007090837503	male	yes	New York	5-10 miles	January	geico	toyota	yes	yes	0					

```
In [13]: df_categ = df_categ.drop(['x7', 'x19'],axis=1)
df_categ.head()

Out[13]:
```

		x3	x24	x31		x33	x39		x60	x65		x77	x93	x99	target
0	Wed	female	no	Colorado	5-10 miles	August	farmers	mercedes	no	yes	0				
1	Friday	male	no	Tennessee	5-10 miles	April	allstate	mercedes	no	yes	1				
2	Thursday	male	no	Texas	5-10 miles	September	geico	subaru	no	yes	1				
3	Tuesday	male	no	Minnesota	5-10 miles	September	geico	nissan	no	yes	0				
4	Sunday	male	yes	New York	5-10 miles	January	geico	toyota	yes	yes	0				

check out the unique values

```
In [14]: df_categ['x3'].unique()

Out[14]: array(['Wed', 'Friday', 'Thursday', 'Tuesday', 'Sunday', 'Saturday',
                'Sat', 'Wednesday', 'Sun', 'Tue', 'Thur', 'Monday', 'Fri', 'Mon'],
               dtype=object)
```

```
In [15]: for col in df_categ.columns:
        print(df_categ[col].unique(),'\n')

['Wed' 'Friday' 'Thursday' 'Tuesday' 'Sunday' 'Saturday' 'Sat' 'Wednesday'
 'Sun' 'Tue' 'Thur' 'Monday' 'Fri' 'Mon']

['female' 'male' nan]

['no' 'yes']

['Colorado' 'Tennessee' 'Texas' 'Minnesota' 'New York' 'Florida'
 'Nebraska' 'California' nan 'North Dakota' 'Arizona' 'Alabama' 'Ohio'
 'Pennsylvania' 'Iowa' 'Indiana' 'Vermont' 'Arkansas' 'Massachusetts'
 'Illinois' 'Georgia' 'West Virginia' 'Connecticut' 'Virginia'
 'North Carolina' 'Montana' 'New Mexico' 'New Hampshire' 'Michigan' 'DC'
 'Washington' 'Louisiana' 'Kentucky' 'Utah' 'Missouri' 'Oregon' 'Oklahoma'
 'Nevada' 'Wisconsin' 'New Jersey' 'Maryland' 'Maine' 'Alaska' 'Idaho'
 'Wyoming' 'Rhode Island' 'South Dakota' 'Mississippi' 'Kansas' 'Delaware'
 'Hawaii' 'South Carolina']

['5-10 miles']

['August' 'April' 'September' 'January' 'December' 'March' 'July'
 'November' 'June' 'February' 'October' 'May']

['farmers' 'allstate' 'geico' 'progressive' 'esurance']

['mercedes' 'subaru' 'nissan' 'toyota' nan 'chevrolet' 'buick' 'ford']

['no' 'yes']

['yes' nan]

[0 1]

In [16]: # https://stackoverflow.com/questions/60237488/python-replace-only-exact-word-in-string
def fix_x3(df_categ):
    for i, string in enumerate(df_categ['x3'].values):
        df_categ['x3'].values[i] = re.sub(r'\bTue\b', 'Tuesday', string)

    for i, string in enumerate(df_categ['x3'].values):
        df_categ['x3'].values[i] = re.sub(r'\bSun\b', 'Sunday', string)

    for i, string in enumerate(df_categ['x3'].values):
        df_categ['x3'].values[i] = re.sub(r'\bSat\b', 'Saturday', string)

    for i, string in enumerate(df_categ['x3'].values):
        df_categ['x3'].values[i] = re.sub(r'\bFri\b', 'Friday', string)

    for i, string in enumerate(df_categ['x3'].values):
        df_categ['x3'].values[i] = re.sub(r'\bWed\b', 'Wednesday', string)

    for i, string in enumerate(df_categ['x3'].values):
        df_categ['x3'].values[i] = re.sub(r'\bMon\b', 'Monday', string)

    for i, string in enumerate(df_categ['x3'].values):
        df_categ['x3'].values[i] = re.sub(r'\bThur\b', 'Thursday', string)

In [17]: df_categ['x3'] = df_categ['x3'].apply(lambda x: re.sub(r'\bTue\b', 'Tuesday', x))
df_categ['x3'] = df_categ['x3'].apply(lambda x: re.sub(r'\bSun\b', 'Sunday', x))
df_categ['x3'] = df_categ['x3'].apply(lambda x: re.sub(r'\bSat\b', 'Saturday', x))
df_categ['x3'] = df_categ['x3'].apply(lambda x: re.sub(r'\bFri\b', 'Friday', x))
df_categ['x3'] = df_categ['x3'].apply(lambda x: re.sub(r'\bWed\b', 'Wednesday', x))
df_categ['x3'] = df_categ['x3'].apply(lambda x: re.sub(r'\bMon\b', 'Monday', x))
df_categ['x3'] = df_categ['x3'].apply(lambda x: re.sub(r'\bThur\b', 'Thursday', x))

In [18]: df_categ['x3'].unique()

Out[18]: array(['Wednesday', 'Friday', 'Thursday', 'Tuesday', 'Sunday', 'Saturday',
               'Monday'], dtype=object)

In [19]: targets = df_categ['target']

In [20]: df_categ.head(2)

Out[20]:
   x3    x24  x31    x33    x39  x60  x65  x77  x93  x99  target
0  Wednesday  female  no  Colorado  5-10 miles  August  farmers  mercedes  no  yes  0
1   Friday   male  no  Tennessee  5-10 miles   April  allstate  mercedes  no  yes  1

In [21]: for col in df_categ.columns:
        if df_categ[col].isnull().sum() != 0:
            print('column is ',col, 'number of nulls is ', df_categ[col].isnull().sum())
            df_categ[col].fillna(df_categ[col].mode()[0], inplace=True)
            print('column was ',col, 'number of nulls is ', df_categ[col].isnull().sum(),'\n')

column is  x24 number of nulls is  3856
column was  x24 number of nulls is  0

column is  x33 number of nulls is  7171
column was  x33 number of nulls is  0

column is  x77 number of nulls is  9257
column was  x77 number of nulls is  0

column is  x99 number of nulls is  12836
column was  x99 number of nulls is  0

group data and check out the differences

In [22]: target_grp = df_categ.groupby(['target'])
```

X3

```
In [23]: target_grp['x3'].value_counts(normalize=True)
```

```
Out[23]: target  x3
0      Tuesday    0.177442
      Wednesday  0.176039
      Monday     0.159605
      Friday     0.136679
      Saturday   0.132000
      Thursday   0.111121
      Sunday     0.107115
1      Wednesday  0.164225
      Friday     0.158711
      Monday     0.152680
      Saturday   0.149750
      Tuesday    0.136998
      Sunday     0.128382
      Thursday   0.109254
Name: x3, dtype: float64
```

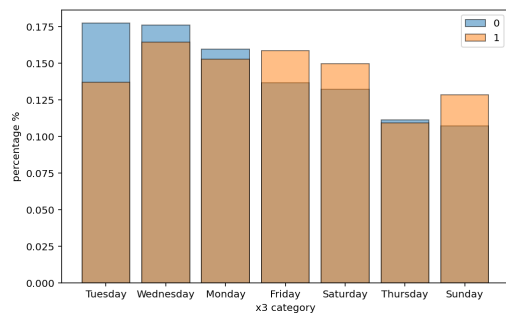
```
In [24]: target_grp_x3_0 = target_grp['x3'].value_counts(normalize=True).loc[0]
target_grp_x3_1 = target_grp['x3'].value_counts(normalize=True).loc[1]

target_grp_x3_0_pd = pd.DataFrame(target_grp['x3'].value_counts(normalize=True).loc[0])
target_grp_x3_1_pd = pd.DataFrame(target_grp['x3'].value_counts(normalize=True).loc[1])

plt.figure(figsize=(8,5))
plt.bar(target_grp_x3_0_pd.index,
        target_grp_x3_0_pd['x3'],
        alpha=0.5, ec='k', label='0')

plt.bar(target_grp_x3_1_pd.index,
        target_grp_x3_1_pd['x3'],
        alpha=0.5, ec='k', label='1')

plt.xlabel('x3 category'); plt.ylabel('percentage %')
plt.legend()
plt.savefig("../Figures/plot_01_1_EDA_1.png")
plt.show()
```



clearly is seen that the days including Tuesday, Wednesday, and Monday have more than 0 and in 3 days including Friday, Saturday, and Sunday the number of 1 are higher. make a bar function as follow and call it on column.

```
In [25]: def plot_bars(df, col_name, fig_number, figsize):

    target_grp_0 = target_grp[col_name].value_counts(normalize=True).loc[0]
    target_grp_1 = target_grp[col_name].value_counts(normalize=True).loc[1]

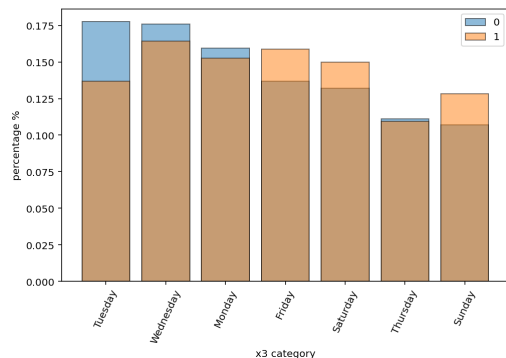
    target_grp_0_pd = pd.DataFrame(target_grp[col_name].value_counts(normalize=True).loc[0])
    target_grp_1_pd = pd.DataFrame(target_grp[col_name].value_counts(normalize=True).loc[1])

    plt.figure(figsize=figsize)
    plt.bar(target_grp_0_pd.index,
            target_grp_0_pd[col_name],
            alpha=0.5, ec='k', label='0')

    plt.bar(target_grp_1_pd.index,
            target_grp_1_pd[col_name],
            alpha=0.5, ec='k', label='1')

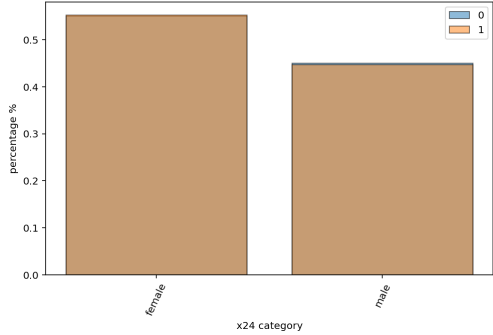
    plt.xticks(rotation=65)
    plt.xlabel(col_name+ ' category'); plt.ylabel('percentage %')
    plt.legend()
    plt.savefig(f"../Figures/plot_01_1_EDA_{fig_number}.png")
    plt.show()
```

```
In [26]: plot_bars(df_categ, col_name='x3', fig_number=1, figsize=(8,5))
```



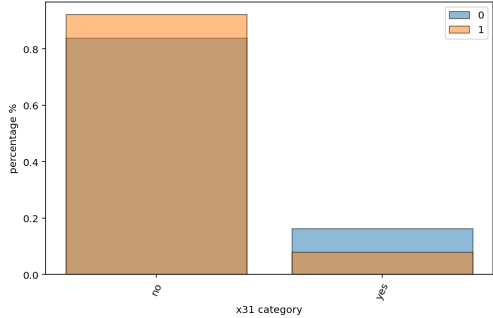
the first three days the chance of having 0 is higher. compare to the rest of the week that the chance of 1 is higher.

```
In [27]: plot_bars(df_categ, col_name='x24', fig_number=2, figsize=(8,5))
```



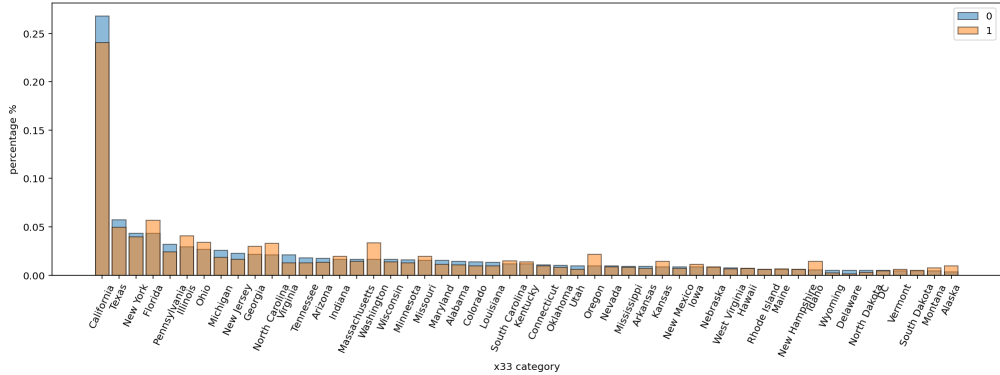
no difference between female and male is observed.

```
In [28]: plot_bars(df_categ, col_name='x31', fig_number=3, figsize=(8,5))
```



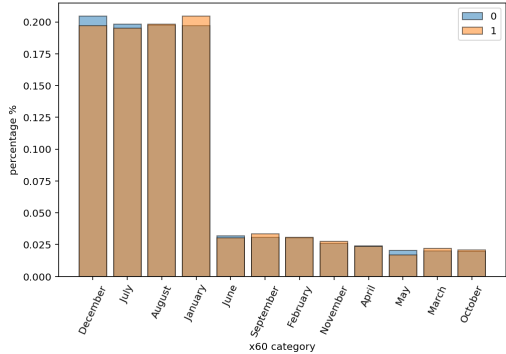
if the x31 value is no, it is higher chance for getting 1 as target.

```
In [29]: plot_bars(df_categ, col_name='x33', fig_number=4, figsize=(17,5))
```



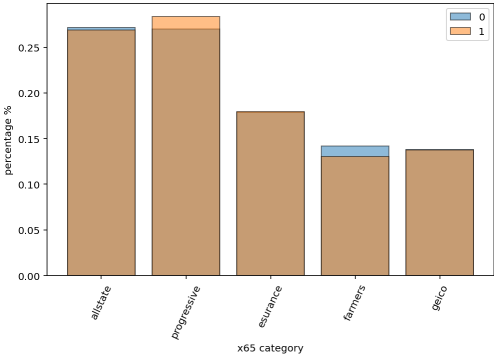
california and texas have more 0 number compare to 1. so if x33 is texas, there is higher chance to get 0 as target value.

```
In [30]: plot_bars(df_categ, col_name='x60', fig_number=5, figsize=(8,5))
```



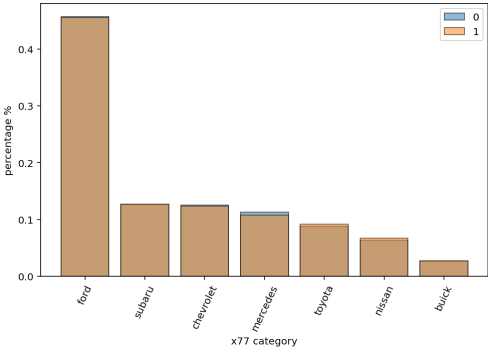
the chance for getting 0 as target is higher in december.

```
In [31]: plot_bars(df_categ, col_name='x65', fig_number=6, figsize=(8,5))
```



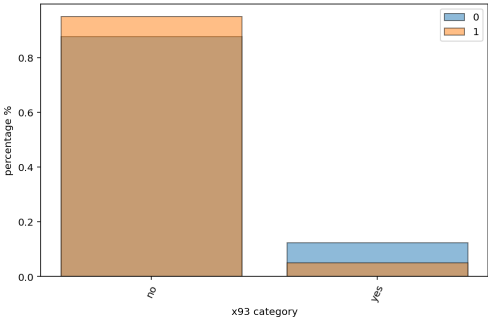
when the x65 category is farmers, the chance of getting 0 is higher. for the rest, I don't see any difference.

```
In [32]: plot_bars(df_categ, col_name='x77', fig_number=6, figsize=(8,5))
```



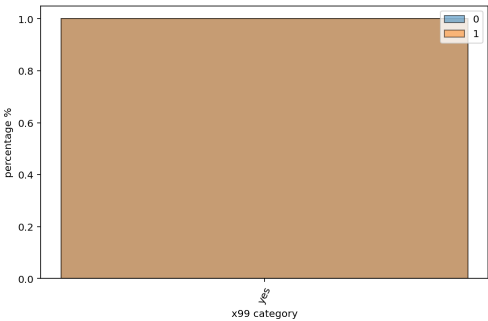
Mercedes has a slightly higher chance to get 0 as target value.

```
In [33]: plot_bars(df_categ, col_name='x93', fig_number=6, figsize=(8,5))
```



clearly there is difference between yes and no in column x93.

```
In [34]: plot_bars(df_categ, col_name='x99', fig_number=6, figsize=(8,5))
```



no difference is observed. this is a feature with 0 variance and should be dropped for final modeling.

i stop the EDA here and proceed into next steps

In [ ]: