

11_ML - One Hot Encoding and Dummy Variables

- To convert categorical data into numerical data, as ML use mathematical formulas in its model so all string data should be converted into numerical data
- It is normally used when number of data is less

```
In [1]: import pandas as pd
```

```
In [2]: dataset = pd.read_csv('loan.csv')
```

```
In [3]: dataset.head(3)
```

```
Out[3]:
```

| | Loan_ID | Gender | Married | Dependents | Education | Self_Employed | ApplicantIncome | CoapplicantIncome | LoanAmount | Loan_Amount_Term | Credit_History | Property_Area | Loan_Status |
|---|----------|--------|---------|------------|-----------|---------------|-----------------|-------------------|------------|------------------|----------------|---------------|-------------|
| 0 | LP001002 | Male | No | 0 | Graduate | No | 5849 | 0 | 120000 | 36 | 1 | Urban | A |
| 1 | LP001003 | Male | Yes | 1 | Graduate | No | 4583 | 0 | 90000 | 36 | 1 | Urban | A |
| 2 | LP001005 | Male | Yes | 0 | Graduate | Yes | 3000 | 0 | 120000 | 36 | 1 | Urban | A |

11.1 Find Missing Values and Handle it

```
In [4]: dataset.isnull().sum()
```

```
Out[4]: Loan_ID      0
Gender      13
Married      3
Dependents  15
Education    0
Self_Employed 32
ApplicantIncome  0
CoapplicantIncome  0
LoanAmount    22
Loan_Amount_Term 14
Credit_History 50
Property_Area  0
Loan_Status    0
dtype: int64
```

```
In [8]: dataset['Gender'].mode()[0]
```

```
Out[8]: 'Male'
```

```
In [11]: # Gender column contains missing values so we will fill it using mode method
dataset['Gender'].fillna(dataset['Gender'].mode()[0], inplace=True)
```

```
In [12]: dataset.isnull().sum()
```

```
Out[12]: Loan_ID          0
Gender          0
Married         3
Dependents      15
Education       0
Self_Employed   32
ApplicantIncome 0
CoapplicantIncome 0
LoanAmount      22
Loan_Amount_Term 14
Credit_History  50
Property_Area   0
Loan_Status     0
dtype: int64
```

```
In [13]: # Married column contains missing values so we will fill it using mode method
dataset['Married'].fillna(dataset['Married'].mode()[0],inplace=True)
```

```
In [14]: dataset.isnull().sum()
```

```
Out[14]: Loan_ID          0
Gender          0
Married         0
Dependents      15
Education       0
Self_Employed   32
ApplicantIncome 0
CoapplicantIncome 0
LoanAmount      22
Loan_Amount_Term 14
Credit_History  50
Property_Area   0
Loan_Status     0
dtype: int64
```

11.2 Use One Hot Code to handle missing values

First separate Gender and Married data to perform encoding

```
In [16]: en_data = dataset[['Gender', 'Married']]
en_data
```

Out[16]:

| | Gender | Married |
|-----|--------|---------|
| 0 | Male | No |
| 1 | Male | Yes |
| 2 | Male | Yes |
| 3 | Male | Yes |
| 4 | Male | No |
| ... | ... | ... |
| 609 | Female | No |
| 610 | Male | Yes |
| 611 | Male | Yes |
| 612 | Male | Yes |
| 613 | Female | No |

614 rows × 2 columns

```
In [17]: pd.get_dummies(en_data)
```

Out[17]:

| | Gender_Female | Gender_Male | Married_No | Married_Yes |
|-----|---------------|-------------|------------|-------------|
| 0 | 0 | 1 | 1 | 0 |
| 1 | 0 | 1 | 0 | 1 |
| 2 | 0 | 1 | 0 | 1 |
| 3 | 0 | 1 | 0 | 1 |
| 4 | 0 | 1 | 1 | 0 |
| ... | ... | ... | ... | ... |
| 609 | 1 | 0 | 1 | 0 |
| 610 | 0 | 1 | 0 | 1 |
| 611 | 0 | 1 | 0 | 1 |
| 612 | 0 | 1 | 0 | 1 |
| 613 | 1 | 0 | 1 | 0 |

614 rows × 4 columns

```
In [18]: pd.get_dummies(en_data).info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 614 entries, 0 to 613
Data columns (total 4 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Gender_Female    614 non-null   uint8
1   Gender_Male      614 non-null   uint8
2   Married_No       614 non-null   uint8
3   Married_Yes      614 non-null   uint8
dtypes: uint8(4)
memory usage: 2.5 KB
```

```
In [19]: from sklearn.preprocessing import OneHotEncoder
```

```
In [20]: # fit_transform() converts categorical data into numerical data
ohe = OneHotEncoder()
ohe.fit_transform(en_data)
```

```
Out[20]: <614x4 sparse matrix of type '<class 'numpy.float64'>'
         with 1228 stored elements in Compressed Sparse Row format>

sparse matrix contains data in 0 and 1 form
```

```
In [25]: ohe = OneHotEncoder()
ar = ohe.fit_transform(en_data).toarray()
ar
```

```
Out[25]: array([[0., 1., 1., 0.],
                [0., 1., 0., 1.],
                [0., 1., 0., 1.],
                ...,
                [0., 1., 0., 1.],
                [0., 1., 0., 1.],
                [1., 0., 1., 0.]])
```

```
In [27]: # Convert the array data into dataframe

pd.DataFrame(ar, columns=['Gender_Female', 'Gender_Male', 'Married_No', 'Married_Yes'])
```

Out[27]:

| | Gender_Female | Gender_Male | Married_No | Married_Yes |
|------------|---------------|-------------|------------|-------------|
| 0 | 0.0 | 1.0 | 1.0 | 0.0 |
| 1 | 0.0 | 1.0 | 0.0 | 1.0 |
| 2 | 0.0 | 1.0 | 0.0 | 1.0 |
| 3 | 0.0 | 1.0 | 0.0 | 1.0 |
| 4 | 0.0 | 1.0 | 1.0 | 0.0 |
| ... | ... | ... | ... | ... |
| 609 | 1.0 | 0.0 | 1.0 | 0.0 |
| 610 | 0.0 | 1.0 | 0.0 | 1.0 |
| 611 | 0.0 | 1.0 | 0.0 | 1.0 |
| 612 | 0.0 | 1.0 | 0.0 | 1.0 |
| 613 | 1.0 | 0.0 | 1.0 | 0.0 |

614 rows × 4 columns

You can see out of 2 column 4 column are produced, so to avoid this, we will use 'drop first' to delete first column after encoding i.e. Gender_Female and Married_No, so use it as follows:

```
In [28]: ohe = OneHotEncoder(drop='first')
ar = ohe.fit_transform(en_data).toarray()
ar
```

```
Out[28]: array([[1., 0.],
                [1., 1.],
                [1., 1.],
                ...,
                [1., 1.],
                [1., 1.],
                [0., 0.]])
```

```
In [29]: # Convert the array data into dataframe

pd.DataFrame(ar, columns=['Gender_Male', 'Married_Yes'])
```

Out[29]:

| | Gender_Male | Married_Yes |
|-----|-------------|-------------|
| 0 | 1.0 | 0.0 |
| 1 | 1.0 | 1.0 |
| 2 | 1.0 | 1.0 |
| 3 | 1.0 | 1.0 |
| 4 | 1.0 | 0.0 |
| ... | ... | ... |
| 609 | 0.0 | 0.0 |
| 610 | 1.0 | 1.0 |
| 611 | 1.0 | 1.0 |
| 612 | 1.0 | 1.0 |
| 613 | 0.0 | 0.0 |

614 rows × 2 columns

In []:

In []: