12_ML - Label Encoder

12.1 Label encoding on Nominal data

```
In [3]: import pandas as pd
         from sklearn.preprocessing import LabelEncoder
 In [5]: df = pd.DataFrame({'name':['Rashid', 'Lion', 'Computer', 'Gym', 'Plant']})
Out[5]:
                name
         0
               Rashid
                 Lion
         2 Computer
                 Gym
                 Plant
 In [7]: le = LabelEncoder()
         df['en_name'] = le.fit_transform(df['name'])
 In [8]:
 Out[8]:
                name en_name
         0
               Rashid
                             4
                 Lion
                             2
         2 Computer
                             0
         3
                 Gym
         4
                 Plant
                             3
         Now work on real time data
 In [9]: dataset = pd.read_csv('loan.csv')
In [10]: dataset.head(3)
```

```
Out[10]:
             Loan_ID Gender Married Dependents Education Self_Employed ApplicantIncome (
         0 LP001002
                        Male
                                                   Graduate
                                                                                    5849
                                  No
                                                                     No
         1 LP001003
                        Male
                                               1
                                                   Graduate
                                                                                    4583
                                 Yes
                                                                      No
         2 LP001005
                        Male
                                                   Graduate
                                                                                    3000
                                 Yes
                                                                     Yes
In [14]: # To check number of data
         dataset['Property_Area'].unique()
Out[14]: array(['Urban', 'Rural', 'Semiurban'], dtype=object)
In [12]: la = LabelEncoder()
         la.fit(dataset['Property_Area'])
Out[12]:
         ▼ LabelEncoder
         LabelEncoder()
In [13]: la.transform(dataset['Property_Area'])
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                1, 0, 0, 2, 1, 2, 1, 2, 0, 1, 0, 1, 0, 2, 1, 0, 0, 1, 1, 0, 1, 0,
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                2, 1, 1, 2, 1, 2, 0, 2, 1, 2, 1, 0, 0, 0, 2, 1, 1, 1, 1, 1, 1, 0,
                2, 0, 0, 1, 0, 2, 2, 0, 2, 0, 1, 2, 1, 0, 0, 0, 0, 2, 2, 1])
In [15]: # to replace the property data with encoding data
         dataset['Property_Area'] = la.transform(dataset['Property_Area'])
```

18]:	<pre>dataset.head(3)</pre>												
]:		Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	C				
	0	LP001002	Male	No	0	Graduate	No	5849					
	1	LP001003	Male	Yes	1	Graduate	No	4583					
	2	LP001005	Male	Yes	0	Graduate	Yes	3000					

12.1 Label encoding on Ordinal data

12.1.1 Ordincal encoding through Cyclic Line

```
In [21]: dfo = pd.DataFrame({'size': ['s', 'm', 'l', 'xl', 'xxl', 's', 's', 'xl', 'm',
         dfo.head(3)
Out[21]:
            size
         0
               S
              m
         2
In [22]: ord_data = [['s', 'm', 'l', 'xl', 'xxl']]
In [24]: from sklearn.preprocessing import OrdinalEncoder
In [25]: # oe = OrdinalEncoder(): This will encode the data alphabatically
         oe = OrdinalEncoder(categories=ord_data)
         oe.fit(dfo[['size']])
Out[25]:
                                 OrdinalEncoder
         OrdinalEncoder(categories=[['s', 'm', 'l', 'xl', 'xxl']])
In [27]: oe.transform(dfo[['size']])
Out[27]: array([[0.],
                 [1.],
                [2.],
                 [3.],
                 [4.],
                 [0.],
                 [0.],
                [0.],
                [3.],
                 [1.],
                 [2.]])
```

```
In [29]: dfo['size_en'] = oe.transform(dfo[['size']])
          dfo
Out[29]:
              size size_en
           0
                        0.0
                 S
           1
                        1.0
                m
           2
                 2.0
                χl
                        3.0
           3
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               xxl
                        4.0
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                m
                        1.0
          10
                 2.0
```

12.1.2 Ordincal encoding through Map function

```
In [30]: # In map function, you can manually assign numbers to each data type, for example
         # You can assign any number
         ord_data1 = {'s':0, 'm':1, 'l':2, 'xl':3, 'xxl':4}
In [32]: dfo['size'].map(ord_data1)
Out[32]: 0
                0
         1
                1
          2
                2
          3
                3
          4
                4
          5
                0
         6
                0
         7
                0
                3
          9
                1
          10
                2
         Name: size, dtype: int64
In [33]: dfo['size_en_map'] = dfo['size'].map(ord_data1)
```

Out[33]:		size	size_en	size_en_map
	0	S	0.0	0
	1	m	1.0	1
	2	I	2.0	2
	3	xl	3.0	3
	4	xxl	4.0	4
	5	S	0.0	0
	6	S	0.0	0
	7	S	0.0	0
	8	xl	3.0	3
	9	m	1.0	1
	10	1	2.0	2

12.2 Perform Ordinal Encoding on big data

```
In [35]: dataset = pd.read_csv('loan.csv')
In [36]: dataset.head(3)
Out[36]:
             Loan_ID Gender Married Dependents Education Self_Employed ApplicantIncome (
         0 LP001002
                        Male
                                   No
                                                    Graduate
                                                                        No
                                                                                       5849
          1 LP001003
                        Male
                                                    Graduate
                                  Yes
                                                                        No
                                                                                       4583
         2 LP001005
                        Male
                                  Yes
                                                    Graduate
                                                                       Yes
                                                                                       3000
In [37]: dataset['Property_Area'].unique()
Out[37]: array(['Urban', 'Rural', 'Semiurban'], dtype=object)
In [40]: # if there is nan (missing data)m, you can fill it
         # if the data is categorical then you should do mode filling
         dataset['Property_Area'].fillna(dataset['Property_Area'].mode()[0], inplace=True)
In [42]: en_data_loan = [['Urban', 'Rural', 'Semiurban']]
In [45]: oen = OrdinalEncoder(categories=en_data_loan)
         oen.fit_transform(dataset[['Property_Area']])
```

```
Out[45]: array([[0.],
                  [1.],
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```
dataset['Property_Area'] = oen.fit_transform(dataset[['Property_Area']])
In [49]: dataset.head(3)
Out[49]:
             Loan_ID Gender Married Dependents Education Self_Employed ApplicantIncome C
         0 LP001002
                        Male
                                                0
                                                    Graduate
                                                                                       5849
                                   No
                                                                       No
         1 LP001003
                                  Yes
                        Male
                                                    Graduate
                                                                                       4583
                                                                        No
         2 LP001005
                                                                                       3000
                        Male
                                                0
                                                    Graduate
                                  Yes
                                                                       Yes
```