

One-Hot-Encoding

Encoding -convert catagirical data into numarical data

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.preprocessing import OneHotEncoder

df=pd.read_csv(r"C:\Users\Bhoomika.G\OneDrive\Documents\
Salary_EDA.csv")
df.head()
```

	Age	Gender	Education Level	Job Title	Years of Experience
0	32.0	Male	Bachelor's	Software Engineer	5.0
1	28.0	Female	Master's	Data Analyst	3.0
2	45.0	Male	PhD	Senior Manager	15.0
3	36.0	Female	Bachelor's	Sales Associate	7.0
4	36.0	Female	Bachelor's	Sales Associate	7.0

	Salary
0	90000.0
1	65000.0
2	150000.0
3	60000.0
4	60000.0

Filter catogiracal feature

```
catogarical_cols=['Education Level'] # fit_transform-it convert the
non-numirical data to numirical data
```

Define and apply the encoder

```
encoder=OneHotEncoder(drop=None,sparse_output=False) # it not drop
hole data in that it show the unique

encoded_data=encoder.fit_transform(df[catogarical_cols]) # modify the
original row in matrix (convert numerical type)
```

#the encoded data is in the form of array dot now we need to convert the encoded data into dataframe with catagery as follows the name Encode dataframes

```
encoded_df=pd.DataFrame(encoded_data,columns=encoder.get_feature_names_out(catogorical_cols)) # get_feature_names_out it take unique values the help to encoder
```

```
encoded_df.head()
```

	Education Level_Bachelor's	Education Level_Master's	Education Level_PhD
0	1.0	0.0	0.0
1	0.0	1.0	0.0
2	0.0	0.0	1.0
3	1.0	0.0	0.0
4	1.0	0.0	0.0

	Education Level_nan
0	0.0
1	0.0
2	0.0
3	0.0
4	0.0

```
encoded_df.drop(columns=['Education Level_nan'], inplace=True) # it used to drop the columns
```

```
encoded_df.head()
```

	Education Level_Bachelor's	Education Level_Master's	Education Level_PhD
0	1.0	0.0	0.0
1	0.0	1.0	0.0
2	0.0	0.0	1.0
3	1.0	0.0	0.0
4	1.0	0.0	0.0

```
Fdf=pd.concat([df,encoded_df], axis=1) #axis=1 column in pandas,axis=0 row in pandas revers in numpy  
Fdf.head()
```

	Age	Gender	Education Level	Job Title	Years of Experience \
0	32.0	Male	Bachelor's	Software Engineer	5.0
1	28.0	Female	Master's	Data Analyst	3.0
2	45.0	Male	PhD	Senior Manager	15.0
3	36.0	Female	Bachelor's	Sales Associate	7.0
4	36.0	Female	Bachelor's	Sales Associate	7.0

	Salary	Education Level_Bachelor's	Education Level_Master's \
0	90000.0	1.0	0.0
1	65000.0	0.0	1.0
2	150000.0	0.0	0.0
3	60000.0	1.0	0.0
4	60000.0	1.0	0.0

	Education Level_PhD
0	0.0
1	0.0
2	1.0
3	0.0
4	0.0

Label encoding

```
from sklearn.preprocessing import LabelEncoder

df1=pd.read_csv(r"C:\Users\Bhoomika.G\OneDrive\Documents\Salary_EDA.csv")
df1.head()
```

	Age	Gender	Education Level	Job Title	Years of Experience \
0	32.0	Male	Bachelor's	Software Engineer	5.0
1	28.0	Female	Master's	Data Analyst	3.0
2	45.0	Male	PhD	Senior Manager	15.0
3	36.0	Female	Bachelor's	Sales Associate	7.0
4	36.0	Female	Bachelor's	Sales Associate	7.0

	Salary
0	90000.0
1	65000.0
2	150000.0
3	60000.0
4	60000.0

```
le1=LabelEncoder()
df1['Gender_encoder']=le1.fit_transform(df1['Gender']) #fit_transform
is used to convert the 0/1 value assened insted of male and female
df1.head()
```

	Age	Gender	Education Level	Job Title	Years of Experience \
0	32.0	Male	Bachelor's	Software Engineer	5.0
1	28.0	Female	Master's	Data Analyst	3.0
2	45.0	Male	PhD	Senior Manager	15.0
3	36.0	Female	Bachelor's	Sales Associate	7.0
4	36.0	Female	Bachelor's	Sales Associate	7.0

	Salary	Gender_encoder
0	90000.0	1
1	65000.0	0
2	150000.0	1
3	60000.0	0
4	60000.0	0

```
le2=LabelEncoder()
df1['Education level encoded']=le2.fit_transform(df1['Education
Level'])
df1.head()
```

	Age	Gender	Education Level	Job Title	Years of Experience \
0	32.0	Male	Bachelor's	Software Engineer	5.0
1	28.0	Female	Master's	Data Analyst	3.0
2	45.0	Male	PhD	Senior Manager	15.0
3	36.0	Female	Bachelor's	Sales Associate	7.0
4	36.0	Female	Bachelor's	Sales Associate	7.0

	Salary	Gender_encoder	Education level encoded
0	90000.0	1	0
1	65000.0	0	1
2	150000.0	1	2
3	60000.0	0	0
4	60000.0	0	0

Scaling

it will convert the bigger value to smaller value by dividing bigger value (up to 0 to 1)

```
from sklearn.preprocessing import MinMaxScaler
```

```
df2=pd.read_csv(r"C:\Users\Bhoomika.G\OneDrive\Documents\Salary_EDA.csv")
df2.head()
```

	Age	Gender	Education Level	Job Title	Years of Experience \
0	32.0	Male	Bachelor's	Software Engineer	5.0
1	28.0	Female	Master's	Data Analyst	3.0
2	45.0	Male	PhD	Senior Manager	15.0
3	36.0	Female	Bachelor's	Sales Associate	7.0
4	36.0	Female	Bachelor's	Sales Associate	7.0

	Salary
0	90000.0
1	65000.0
2	150000.0
3	60000.0
4	60000.0

```
sk1= MinMaxScaler()
df2['Salary_skl']=sk1.fit_transform(df2[['Salary']]) #fit_transform
is used to convert the 0/1 value assened insted of male and female
df2.head()
```

	Age	Gender	Education Level	Job Title	Years of Experience \
0	32.0	Male	Bachelor's	Software Engineer	5.0
1	28.0	Female	Master's	Data Analyst	3.0

2	45.0	Male	PhD	Senior Manager
15.0				
3	36.0	Female	Bachelor's	Sales Associate
7.0				
4	36.0	Female	Bachelor's	Sales Associate
7.0				

	Salary	Salary_skla
0	90000.0	0.359103
1	65000.0	0.258963
2	150000.0	0.599439
3	60000.0	0.238935
4	60000.0	0.238935

```
from sklearn.preprocessing import StandardScaler
```

```
df3=pd.read_csv(r"C:\Users\Bhoomika.G\OneDrive\Documents\Salary_EDA.csv")
df3.head()
```

	Age	Gender	Education Level	Job Title	Years of Experience \
0	32.0	Male	Bachelor's	Software Engineer	5.0
1	28.0	Female	Master's	Data Analyst	3.0
2	45.0	Male	PhD	Senior Manager	15.0
3	36.0	Female	Bachelor's	Sales Associate	7.0
4	36.0	Female	Bachelor's	Sales Associate	7.0

	Salary
0	90000.0
1	65000.0
2	150000.0
3	60000.0
4	60000.0

```
ss=StandardScaler()
df3['Salary_Standar']=ss.fit_transform(df3[['Salary']]) #fit_transform
is used to convert the 0/1 value assened insted of male and female
df3.head()
```

	Age	Gender	Education Level	Job Title	Years of Experience \
0	32.0	Male	Bachelor's	Software Engineer	5.0
1	28.0	Female	Master's	Data Analyst	3.0

2	45.0	Male	PhD	Senior Manager
15.0				
3	36.0	Female	Bachelor's	Sales Associate
7.0				
4	36.0	Female	Bachelor's	Sales Associate
7.0				

	Salary	Salary_Stander
0	90000.0	-0.211488
1	65000.0	-0.733148
2	150000.0	1.040496
3	60000.0	-0.837480
4	60000.0	-0.837480