

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

```
In [2]: data=pd.read_csv(r"C:\Users\DELL\Downloads\ML Feature Encoding Resource16931320640.csv")
```

In [3]: data

Out[3]:

	age	workclass	fnlwgt	education	educational-num	marital-status	occupation	relationship	race	gender	capital-gain	capital-loss	hours-per-week	
0	25	Private	226802	11th	7	Never-married	Machine-op-inspct	Own-child	Black	Male	0	0	40	1
1	38	Private	89814	HS-grad	9	Married-civ-spouse	Farming-fishing	Husband	White	Male	0	0	50	1
2	28	Local-gov	336951	Assoc-acdm	12	Married-civ-spouse	Protective-serv	Husband	White	Male	0	0	40	1
3	44	Private	160323	Some-college	10	Married-civ-spouse	Machine-op-inspct	Husband	Black	Male	7688	0	40	1
4	18	?	103497	Some-college	10	Never-married	?	Own-child	White	Female	0	0	30	1
...
48837	27	Private	257302	Assoc-acdm	12	Married-civ-spouse	Tech-support	Wife	White	Female	0	0	38	1
48838	40	Private	154374	HS-grad	9	Married-civ-spouse	Machine-op-inspct	Husband	White	Male	0	0	40	1
48839	58	Private	151910	HS-grad	9	Widowed	Adm-clerical	Unmarried	White	Female	0	0	40	1
48840	22	Private	201490	HS-grad	9	Never-married	Adm-clerical	Own-child	White	Male	0	0	20	1
48841	52	Self-emp-inc	287927	HS-grad	9	Married-civ-spouse	Exec-managerial	Wife	White	Female	15024	0	40	1

48842 rows × 15 columns



```
In [4]: data["income"].unique()
```

```
Out[4]: array(['<=50K', '>50K'], dtype=object)
```

```
In [14]: data["income"].replace(['<=50K', '>50K'],[0,1],inplace=True)
```

In [15]: data

Out[15]:

	age	workclass	fnlwgt	education	educational-num	marital-status	occupation	relationship	race	gender	capital-gain	capital-loss	hours-per-week	
0	25	Private	226802	11th	7	Never-married	Machine-op-inspct	Own-child	Black	Male	0	0	40	1
1	38	Private	89814	HS-grad	9	Married-civ-spouse	Farming-fishing	Husband	White	Male	0	0	50	1
2	28	Local-gov	336951	Assoc-acdm	12	Married-civ-spouse	Protective-serv	Husband	White	Male	0	0	40	1
3	44	Private	160323	Some-college	10	Married-civ-spouse	Machine-op-inspct	Husband	Black	Male	7688	0	40	1
4	18	?	103497	Some-college	10	Never-married	?	Own-child	White	Female	0	0	30	1
...
48837	27	Private	257302	Assoc-acdm	12	Married-civ-spouse	Tech-support	Wife	White	Female	0	0	38	1
48838	40	Private	154374	HS-grad	9	Married-civ-spouse	Machine-op-inspct	Husband	White	Male	0	0	40	1
48839	58	Private	151910	HS-grad	9	Widowed	Adm-clerical	Unmarried	White	Female	0	0	40	1
48840	22	Private	201490	HS-grad	9	Never-married	Adm-clerical	Own-child	White	Male	0	0	20	1
48841	52	Self-emp-inc	287927	HS-grad	9	Married-civ-spouse	Exec-managerial	Wife	White	Female	15024	0	40	1

48842 rows × 15 columns



```
In [16]: data["income"].replace([0,1],['<=50K', '>50K'],inplace=True)
```

```
In [17]: data
```

```
Out[17]:
```

	age	workclass	fnlwgt	education	educational-num	marital-status	occupation	relationship	race	gender	capital-gain	capital-loss	hours-per-week	c
0	25	Private	226802	11th	7	Never-married	Machine-op-inspct	Own-child	Black	Male	0	0	40	1
1	38	Private	89814	HS-grad	9	Married-civ-spouse	Farming-fishing	Husband	White	Male	0	0	50	1
2	28	Local-gov	336951	Assoc-acdm	12	Married-civ-spouse	Protective-serv	Husband	White	Male	0	0	40	1
3	44	Private	160323	Some-college	10	Married-civ-spouse	Machine-op-inspct	Husband	Black	Male	7688	0	40	1
4	18	?	103497	Some-college	10	Never-married	?	Own-child	White	Female	0	0	30	1
...
48837	27	Private	257302	Assoc-acdm	12	Married-civ-spouse	Tech-support	Wife	White	Female	0	0	38	1
48838	40	Private	154374	HS-grad	9	Married-civ-spouse	Machine-op-inspct	Husband	White	Male	0	0	40	1
48839	58	Private	151910	HS-grad	9	Widowed	Adm-clerical	Unmarried	White	Female	0	0	40	1
48840	22	Private	201490	HS-grad	9	Never-married	Adm-clerical	Own-child	White	Male	0	0	20	1
48841	52	Self-emp-inc	287927	HS-grad	9	Married-civ-spouse	Exec-managerial	Wife	White	Female	15024	0	40	1

48842 rows × 15 columns



```
In [18]: data.head(2)
```

```
Out[18]:
```

	age	workclass	fnlwgt	education	educational-num	marital-status	occupation	relationship	race	gender	capital-gain	capital-loss	hours-per-week	native-country
0	25	Private	226802	11th	7	Never-married	Machine-op-inspct	Own-child	Black	Male	0	0	40	United States
1	38	Private	89814	HS-grad	9	Married-civ-spouse	Farming-fishing	Husband	White	Male	0	0	50	United States

```
In [11]: data["workclass"].unique()
```

```
Out[11]: array(['Private', 'Local-gov', '?', 'Self-emp-not-inc', 'Federal-gov',
                'State-gov', 'Self-emp-inc', 'Without-pay', 'Never-worked'],
              dtype=object)
```

```
In [13]: from sklearn.preprocessing import LabelEncoder
```

```
In [20]: encoder=LabelEncoder()
```

```
In [21]: a=encoder.fit_transform(data["workclass"])
```

```
In [22]: encoder.inverse_transform(a)
```

```
Out[22]: array(['Private', 'Private', 'Local-gov', ..., 'Private', 'Private',
                'Self-emp-inc'], dtype=object)
```

```
In [19]: data["gender"].unique()
```

```
Out[19]: array(['Male', 'Female'], dtype=object)
```

```
In [23]: z=pd.get_dummies(data["gender"])
```

```
In [24]: z
```

Out[24]:

	Female	Male
0	False	True
1	False	True
2	False	True
3	False	True
4	True	False
...
48837	True	False
48838	False	True
48839	True	False
48840	False	True
48841	True	False

48842 rows × 2 columns

In [25]: `z=pd.from_dummies(data["gender"])`

```
-----
TypeError                                Traceback (most recent call last)
Cell In[25], line 1
----> 1 z=pd.from_dummies(data["gender"])

File ~\anaconda3\Lib\site-packages\pandas\core\reshape\encoding.py:450, in from_dummies(data, sep, default_
category)
    447 from pandas.core.reshape.concat import concat
    449 if not isinstance(data, DataFrame):
--> 450     raise TypeError(
    451         "Expected 'data' to be a 'DataFrame'; "
    452         f"Received 'data' of type: {type(data).__name__}"
    453     )
    455 if data.isna().any().any():
    456     raise ValueError(
    457         "Dummy DataFrame contains NA value in column: "
    458         f"{data.isna().any().idxmax()}"
    459     )
```

TypeError: Expected 'data' to be a 'DataFrame'; Received 'data' of type: Series

In [26]: `z.idxmax(axis=1)`

Out[26]:

0	Male
1	Male
2	Male
3	Male
4	Female
	...
48837	Female
48838	Male
48839	Female
48840	Male
48841	Female

Length: 48842, dtype: object


```
In [27]: from sklearn.preprocessing import OneHotEncoder
```

```
In [28]: encoder1=OneHotEncoder()
```

```
In [31]: df=pd.DataFrame(encoder1.fit_transform(data[["gender"]]))
```

```
In [32]: df
```

```
Out[32]:
```

	0
0	(0, 1)\t1.0
1	(0, 1)\t1.0
2	(0, 1)\t1.0
3	(0, 1)\t1.0
4	(0, 0)\t1.0
...	...
48837	(0, 0)\t1.0
48838	(0, 1)\t1.0
48839	(0, 0)\t1.0
48840	(0, 1)\t1.0
48841	(0, 0)\t1.0

48842 rows × 1 columns

```
In [35]: b=encoder1.fit_transform(data[["gender"]]).toarray()
```

```
In [36]: encoder1.inverse_transform(b)
```

```
Out[36]: array([[ 'Male'],  
                [ 'Male'],  
                [ 'Male'],  
                ...,  
                [ 'Female'],  
                [ 'Male'],  
                [ 'Female']], dtype=object)
```

```
In [37]: b
```

```
Out[37]: array([[0., 1.],  
                [0., 1.],  
                [0., 1.],  
                ...,  
                [1., 0.],  
                [0., 1.],  
                [1., 0.]])
```

```
In [ ]: pd.DataFrame
```