6_Encoding

January 14, 2025

[ ]:
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'''
    Encoding Categorical Data -->

    Categorical encoding refers to the process of converting categorical data
    (data that can take on a limited number of values,typically representing␣
 ↪categories or labels)
    into a numerical format that machine learning algorithms can understand and␣
 ↪work with

    Since many machine learning models can only handle numerical data,␣
 ↪categorical features need to be
    transformed into numbers. There are several techniques to achieve this,␣
 ↪depending on the type of categorical
    data and the specific problem. Here are some common methods for categorical␣
 ↪encoding
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[ ]:
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#   Methods -->

#   Label Encoding
#   One-Hot Encoding
#   Binary Encoding
#   Frequency Encoding
#   Mean Encoding
#   Ordinal Encoding
```

[1]:
```
import numpy as np
import pandas as pd
```

[ ]:
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'''
    Label Encoding -->

    Label encoding is another technique for converting categorical data into␣
 ↪numerical form,
    but unlike one-hot encoding, it assigns an integer to each category. Each␣
 ↪unique category
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    is given a distinct integer value, which allows algorithms to process the␣
 ↪data in numerical form.

    Example -->

    If you have the categories (Red, Blue, Green), label encoding would assign␣
 ↪-->

    Red = 0
    Blue = 1
    Green = 2
'''
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[3]:
```
#   Import Label Encoder From sklearn -->

from sklearn.preprocessing import LabelEncoder
```

[4]:
```
dataset = pd.read_csv('Data/Data.csv')
dataset
```

[4]:

	Country	Age	Salary	Purchased
0	France	44.0	72000.0	No
1	Spain	27.0	48000.0	Yes
2	Germany	30.0	54000.0	No
3	Spain	38.0	61000.0	No
4	Germany	40.0	NaN	Yes
5	France	35.0	58000.0	Yes
6	Spain	NaN	52000.0	No
7	France	48.0	79000.0	Yes
8	Germany	50.0	83000.0	No
9	France	37.0	67000.0	Yes

[5]:
```
encoder = LabelEncoder()
dataset['Country'] = encoder.fit_transform(dataset['Country'])
dataset
```

[5]:

	Country	Age	Salary	Purchased
0	0	44.0	72000.0	No
1	2	27.0	48000.0	Yes
2	1	30.0	54000.0	No
3	2	38.0	61000.0	No
4	1	40.0	NaN	Yes
5	0	35.0	58000.0	Yes
6	2	NaN	52000.0	No
7	0	48.0	79000.0	Yes
8	1	50.0	83000.0	No
9	0	37.0	67000.0	Yes

[ ]:
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'''
    One Hot Encoding -->

    One-hot encoding is a technique used to represent categorical data as␣
 ↪binary vectors.
    It transforms categorical variables, which might not have a numerical␣
 ↪relationship,
    into a format that can be provided to machine learning algorithms.

    Here's how it works ->

    Each category in the data is represented by a vector of 0s and 1s.
    The length of the vector equals the number of unique categories.
    For a given category, the corresponding position in the vector is marked as␣
 ↪1, and all other positions are marked as 0.

    Example -->

    Suppose you have a dataset with three categories:
    Red, Blue, and Green. One-hot encoding would represent these categories as:

    Red: [1, 0, 0]
    Blue: [0, 1, 0]
    Green: [0, 0, 1]
'''
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[16]:
```
data = pd.read_csv('Data/Data.csv')
data
```

[16]:

	Country	Age	Salary	Purchased
0	France	44.0	72000.0	No
1	Spain	27.0	48000.0	Yes
2	Germany	30.0	54000.0	No
3	Spain	38.0	61000.0	No
4	Germany	40.0	NaN	Yes
5	France	35.0	58000.0	Yes
6	Spain	NaN	52000.0	No
7	France	48.0	79000.0	Yes
8	Germany	50.0	83000.0	No
9	France	37.0	67000.0	Yes

[17]:
```
x_data = data.iloc[:, :-1].values
y_data = data.iloc[:, -1].values
```

[12]:
```
from sklearn.compose import ColumnTransformer
from sklearn.preprocessing import OneHotEncoder
```

[18]:
```
clt = ColumnTransformer(transformers = [('encoder', OneHotEncoder(), [0])],␣
 ↪remainder = 'passthrough')
x_data = np.array(clt.fit_transform(x_data))
x_data
```

[18]:
```
array([[1.0, 0.0, 0.0, 44.0, 72000.0],
       [0.0, 0.0, 1.0, 27.0, 48000.0],
       [0.0, 1.0, 0.0, 30.0, 54000.0],
       [0.0, 0.0, 1.0, 38.0, 61000.0],
       [0.0, 1.0, 0.0, 40.0, nan],
       [1.0, 0.0, 0.0, 35.0, 58000.0],
       [0.0, 0.0, 1.0, nan, 52000.0],
       [1.0, 0.0, 0.0, 48.0, 79000.0],
       [0.0, 1.0, 0.0, 50.0, 83000.0],
       [1.0, 0.0, 0.0, 37.0, 67000.0]], dtype=object)
```

[19]:
```
y_data = encoder.fit_transform(y_data)
y_data
```

[19]:
```
array([0, 1, 0, 0, 1, 1, 0, 1, 0, 1])
```