Experiment No. 3

Title: Data Preprocessing Reading the Dataset Handling Missing Data Conversion to the Tensor Format

Aim: Data Preprocessing Reading the Dataset Handling Missing Data Conversion to the Tensor Format

Theory:

Data preprocessing:

Data preprocessing is a crucial step in the data analysis and machine learning pipeline. It involves cleaning, transforming, and preparing the data to be used in machine learning models. Let's go through the steps of data preprocessing, including reading the dataset, handling missing data, and converting the data into the tensor format.

Step 1) Reading the Dataset:

To read a dataset into Python, you can use various libraries such as NumPy, Pandas, or TensorFlow (tf.data) depending on the format of your data. Pandas is a popular choice for handling tabular data.

```
In [2]: import pandas as pd

df = pd.read_csv('DataSet.csv') #Read a CSV file into a Pandas DataFrame
print(df)
```

```
        Name
        Age
        Gender
        Salary

        0
        Alice
        25.0
        Female
        50000

        1
        Bob
        30.0
        Male
        60000

        2
        Charlie
        NaN
        Male
        45000

        3
        David
        28.0
        Male
        70000

        4
        Eva
        35.0
        Female
        55000
```

Step 2) Handling Missing Data:

Missing data is a common issue in datasets. You need to handle missing data appropriately to avoid biased or incorrect results in your analysis or model training

```
In [3]: print(df.isnull().sum()) #Check for missing values in the DataFrame
        #Option 1: Drop rows with missing values
        #df = df.dropna()
        #Option 2: Fill missing values with a specific value (e.g. mean, median, or
        mean_age = df['Age'].mean()
        df['Age'].fillna(mean_age, inplace=True)
        #df = pd.DataFrame(df)
        df['Gender'] = df['Gender'].map({'Male':1, 'Female':0})
        print(df)
        Name
                  0
        Age
                  1
        Gender
                  0
        Salary
                  0
        dtype: int64
              Name
                     Age Gender Salary
        0
             Alice 25.0
                               0
                                   50000
        1
               Bob 30.0
                                   60000
                               1
          Charlie 29.5
        2
                               1
                                   45000
        3
             David 28.0
                               1
                                   70000
        4
               Eva 35.0
                                   55000
```

Step 3) Conversion to the Tensor Format:

Machine learning models, especially those built with TensorFlow, often require data in tensor format. A tensor is a multi-dimensional array

```
In [6]: import numpy as np
import tensorflow as tf

# Extract features and Labels from the DataFrame
X = df[{'Age', 'Gender'}].values
y = df['Salary'].values

#Convert to Tensorflow tensors
X_tensor = tf.constant(X, dtype=tf.float32)
y_tensor = tf.constant(y, dtype=tf.int32)

#Alternatively, you can use Nupy arrays directly as Tensorflow tensors
X_tensor = tf.convert_to_tensor(X, dtype=tf.float32)
y_tensor = tf.convert_to_tensor(y, dtype=tf.int32)

print("\n X tensor \n", X_tensor)
print("\n X tensor \n", X_tensor)
print("\n Y tensor \n", y_tensor)
```

```
X tensor
tf.Tensor(
[[ 0. 25. ]
[ 1. 30. ]
[ 1. 29.5]
[ 1. 28. ]
[ 0. 35. ]], shape=(5, 2), dtype=float32)

Y tensor
tf.Tensor([50000 60000 45000 70000 55000], shape=(5,), dtype=int32)

C:\Users\Sanika\AppData\Local\Temp\ipykernel_13576\2201011612.py:6: Future
Warning: Passing a set as an indexer is deprecated and will raise in a fut ure version. Use a list instead.
```

Conclusion:

X = df[{'Age', 'Gender'}].values

Successful installation of essential library, Data Preprocessing Reading the Dataset Handling Missing Data Conversion to the Tensor Format.