

CS-23334 FUNDAMENTALS OF DATA SCIENCE

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EXPERIMENT 3

3.A Handling Missing Data In a Dataset

Aim:

To demonstrate an experiment to handle missing data and inappropriate data using pandas

Algorithm:

Step 1: Identify Missing Data

Step 2: Quantify and Visualize Missingness

Step 3: Decide on a Strategy (Drop, Impute, Flag)

Step 4: Apply the Chosen Method and Validate

Given Dataset:

| CustomerID | Age_Group | Rating(1-5) | Hotel | FoodPreference | Bill | NoOfPax | EstimatedSalary | Age_Group |
|------------|-----------|-------------|-----------|----------------|------|---------|-----------------|-----------|
| 1 | 20-25 | 4 | Ibis | veg | 1300 | 2 | 40000 | 20-25 |
| 2 | 30-35 | 5 | LemonTree | Non-Veg | 2000 | 3 | 59000 | 30-35 |
| 3 | 25-30 | 6 | RedFox | Veg | 1322 | 2 | 30000 | 25-30 |
| 4 | 20-25 | -1 | LemonTree | Veg | 1234 | 2 | 120000 | 20-25 |
| 5 | 35+ | 3 | Ibis | Vegetarian | 989 | 2 | 45000 | 35+ |
| 6 | 35+ | 3 | Ibys | Non-Veg | 1909 | 2 | 122220 | 35+ |
| 7 | 35+ | 4 | RedFox | Vegetarian | 1000 | -1 | 21122 | 35+ |
| 8 | 20-25 | 7 | LemonTree | Veg | 2999 | -10 | 345673 | 20-25 |
| 9 | 25-30 | 2 | Ibis | Non-Veg | 3456 | 3 | -99999 | 25-30 |
| 9 | 25-30 | 2 | Ibis | Non-Veg | 3456 | 3 | -99999 | 25-30 |
| 10 | 30-35 | 5 | RedFox | non-Veg | 6755 | 4 | 87777 | 30-35 |

About Dataset:

No.of Columns =9 (called as series – Customer ID, Age Group, Rating(1-5),Hotel, Food Preference, Bill, No Of Pax, Estimated Salary)

CutomerID: Numerical Continuous data

Age: Categorical Data

Rating (1-5): Numerical Discrete Data

Hotel: Categorical Data

Food: Categorical Data

Bill: Numerical Continuous data

NoOfPax: Numerical Discrete

EstimatedSalary: Numerical Continuous data

Code with Output:

```
import numpy as np
import pandas as pd
df=pd.read_csv(r"D:\REC 2nd Year\Data Science\Data Sets\
Hotel_Dataset.csv")
df
```

| | CustomerID | Age_Group | Rating(1-5) | Hotel | FoodPreference | Bill |
|----|------------|-----------|-------------|-----------|----------------|-------|
| 0 | 1 | 20-25 | 4 | Ibis | veg | 1300 |
| 1 | 2 | 30-35 | 5 | LemonTree | Non-Veg | 2000 |
| 2 | 3 | 25-30 | 6 | RedFox | Veg | 1322 |
| 3 | 4 | 20-25 | -1 | LemonTree | Veg | 1234 |
| 4 | 5 | 35+ | 3 | Ibis | Vegetarian | 989 |
| 5 | 6 | 35+ | 3 | Ibys | Non-Veg | 1909 |
| 6 | 7 | 35+ | 4 | RedFox | Vegetarian | 1000 |
| 7 | 8 | 20-25 | 7 | LemonTree | Veg | 2999 |
| 8 | 9 | 25-30 | 2 | Ibis | Non-Veg | 3456 |
| 9 | 9 | 25-30 | 2 | Ibis | Non-Veg | 3456 |
| 10 | 10 | 30-35 | 5 | RedFox | non-Veg | -6755 |

```
df.duplicated()
```

```
0    False
1    False
2    False
3    False
4    False
```

```
5    False
6    False
7    False
8    False
9     True
10   False
dtype: bool
```

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 11 entries, 0 to 10
Data columns (total 9 columns):
#   Column                Non-Null Count  Dtype
---  -
0   CustomerID            11 non-null    int64
1   Age_Group             11 non-null    object
2   Rating(1-5)           11 non-null    int64
3   Hotel                 11 non-null    object
4   FoodPreference         11 non-null    object
5   Bill                  11 non-null    int64
6   NoOfPax               11 non-null    int64
7   EstimatedSalary       11 non-null    int64
8   Age_Group.1           11 non-null    object
dtypes: int64(5), object(4)
memory usage: 924.0+ bytes
```

```
df.drop_duplicates(inplace=True)
df
```

| | CustomerID | Age_Group | Rating(1-5) | Hotel | FoodPreference | Bill |
|----|------------|-----------|-------------|-----------|----------------|-------|
| 0 | 1 | 20-25 | 4 | Ibis | veg | 1300 |
| 1 | 2 | 30-35 | 5 | LemonTree | Non-Veg | 2000 |
| 2 | 3 | 25-30 | 6 | RedFox | Veg | 1322 |
| 3 | 4 | 20-25 | -1 | LemonTree | Veg | 1234 |
| 4 | 5 | 35+ | 3 | Ibis | Vegetarian | 989 |
| 5 | 6 | 35+ | 3 | Ibys | Non-Veg | 1909 |
| 6 | 7 | 35+ | 4 | RedFox | Vegetarian | 1000 |
| 7 | 8 | 20-25 | 7 | LemonTree | Veg | 2999 |
| 8 | 9 | 25-30 | 2 | Ibis | Non-Veg | 3456 |
| 10 | 10 | 30-35 | 5 | RedFox | non-Veg | -6755 |

```
len(df)

10

index=np.array(list(range(0,len(df))))
df.set_index(index,inplace=True)
index
```

```
array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
```

```
df
```

| | CustomerID | Age_Group | Rating(1-5) | Hotel | FoodPreference | Bill |
|-----------|------------|-----------|-------------|-----------|----------------|-------|
| NoOfPax \ | | | | | | |
| 0 | 1 | 20-25 | 4 | Ibis | veg | 1300 |
| 2 | | | | | | |
| 1 | 2 | 30-35 | 5 | LemonTree | Non-Veg | 2000 |
| 3 | | | | | | |
| 2 | 3 | 25-30 | 6 | RedFox | Veg | 1322 |
| 2 | | | | | | |
| 3 | 4 | 20-25 | -1 | LemonTree | Veg | 1234 |
| 2 | | | | | | |
| 4 | 5 | 35+ | 3 | Ibis | Vegetarian | 989 |
| 2 | | | | | | |
| 5 | 6 | 35+ | 3 | Ibys | Non-Veg | 1909 |
| 2 | | | | | | |
| 6 | 7 | 35+ | 4 | RedFox | Vegetarian | 1000 |
| -1 | | | | | | |
| 7 | 8 | 20-25 | 7 | LemonTree | Veg | 2999 |
| -10 | | | | | | |
| 8 | 9 | 25-30 | 2 | Ibis | Non-Veg | 3456 |
| 3 | | | | | | |
| 9 | 10 | 30-35 | 5 | RedFox | non-Veg | -6755 |
| 4 | | | | | | |

```
df.drop(['Age_Group.1'],axis=1,inplace=True)
df
```

| | CustomerID | Age_Group | Rating(1-5) | Hotel | FoodPreference | Bill |
|-----------|------------|-----------|-------------|-----------|----------------|-------|
| NoOfPax \ | | | | | | |
| 0 | 1 | 20-25 | 4 | Ibis | veg | 1300 |
| 2 | | | | | | |
| 1 | 2 | 30-35 | 5 | LemonTree | Non-Veg | 2000 |
| 3 | | | | | | |
| 2 | 3 | 25-30 | 6 | RedFox | Veg | 1322 |
| 2 | | | | | | |
| 3 | 4 | 20-25 | -1 | LemonTree | Veg | 1234 |
| 2 | | | | | | |
| 4 | 5 | 35+ | 3 | Ibis | Vegetarian | 989 |
| 2 | | | | | | |
| 5 | 6 | 35+ | 3 | Ibys | Non-Veg | 1909 |
| 2 | | | | | | |
| 6 | 7 | 35+ | 4 | RedFox | Vegetarian | 1000 |
| -1 | | | | | | |
| 7 | 8 | 20-25 | 7 | LemonTree | Veg | 2999 |
| -10 | | | | | | |
| 8 | 9 | 25-30 | 2 | Ibis | Non-Veg | 3456 |
| 3 | | | | | | |
| 9 | 10 | 30-35 | 5 | RedFox | non-Veg | -6755 |
| 4 | | | | | | |

```
df.CustomerID.loc[df.CustomerID<0]=np.nan
df.Bill.loc[df.Bill<0]=np.nan
```

```
df.EstimatedSalary.loc[df.EstimatedSalary<0]=np.nan
df
```

| | CustomerID | Age_Group | Rating(1-5) | Hotel | FoodPreference | Bill |
|---|------------|-----------|-------------|-----------|----------------|--------|
| 0 | 1.0 | 20-25 | 4 | Ibis | veg | 1300.0 |
| 1 | 2.0 | 30-35 | 5 | LemonTree | Non-Veg | 2000.0 |
| 2 | 3.0 | 25-30 | 6 | RedFox | Veg | 1322.0 |
| 3 | 4.0 | 20-25 | -1 | LemonTree | Veg | 1234.0 |
| 4 | 5.0 | 35+ | 3 | Ibis | Vegetarian | 989.0 |
| 5 | 6.0 | 35+ | 3 | Ibys | Non-Veg | 1909.0 |
| 6 | 7.0 | 35+ | 4 | RedFox | Vegetarian | 1000.0 |
| 7 | 8.0 | 20-25 | 7 | LemonTree | Veg | 2999.0 |
| 8 | 9.0 | 25-30 | 2 | Ibis | Non-Veg | 3456.0 |
| 9 | 10.0 | 30-35 | 5 | RedFox | non-Veg | NaN |

```
df['NoOfPax'].loc[(df['NoOfPax']<1) | (df['NoOfPax']>20)]=np.nan
df
```

| | CustomerID | Age_Group | Rating(1-5) | Hotel | FoodPreference | Bill |
|---|------------|-----------|-------------|-----------|----------------|--------|
| 0 | 1.0 | 20-25 | 4 | Ibis | veg | 1300.0 |
| 1 | 2.0 | 30-35 | 5 | LemonTree | Non-Veg | 2000.0 |
| 2 | 3.0 | 25-30 | 6 | RedFox | Veg | 1322.0 |
| 3 | 4.0 | 20-25 | -1 | LemonTree | Veg | 1234.0 |
| 4 | 5.0 | 35+ | 3 | Ibis | Vegetarian | 989.0 |
| 5 | 6.0 | 35+ | 3 | Ibys | Non-Veg | 1909.0 |
| 6 | 7.0 | 35+ | 4 | RedFox | Vegetarian | 1000.0 |
| 7 | 8.0 | 20-25 | 7 | LemonTree | Veg | 2999.0 |
| 8 | 9.0 | 25-30 | 2 | Ibis | Non-Veg | 3456.0 |
| 9 | 10.0 | 30-35 | 5 | RedFox | non-Veg | NaN |

```

df.Age_Group.unique()
array(['20-25', '30-35', '25-30', '35+'], dtype=object)

df.Hotel.unique()
array(['Ibis', 'LemonTree', 'RedFox', 'Ibys'], dtype=object)

df.FoodPreference.unique
<bound method Series.unique of 0          veg
1      Non-Veg
2          Veg
3          Veg
4  Vegetarian
5      Non-Veg
6  Vegetarian
7          Veg
8      Non-Veg
9      non-Veg
Name: FoodPreference, dtype: object>

df.FoodPreference.replace(['Vegetarian', 'veg'], 'Veg', inplace=True)
df.FoodPreference.replace(['non-Veg'], 'Non-Veg', inplace=True)

```

Result:

Thus the process of missing data values handling is carried out using pandas library in Python.

3.B Data Preprocessig In Data Science

Aim:

To understand the data preprocessing in Data Science and understand the importance of data preprocessing in data science.

Algorithm:

Step 1: Data Cleaning

Step 2: Data Transformation

Step 3: Feature Engineering

Step 4: Data Scaling and Encoding

Code With Output:

```
import numpy as np
import pandas as pd
df=pd.read_csv(r"D:\REC 2nd Year\Data Science\Data Sets\
Pre_Process_Data.csv")
df
```

| | 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 |

```
df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10 entries, 0 to 9
Data columns (total 4 columns):
#   Column      Non-Null Count  Dtype
---  -
0   Country     10 non-null    object
1   Age         9 non-null     float64
2   Salary      9 non-null     float64
3   Purchased   10 non-null    object
dtypes: float64(2), object(2)
memory usage: 452.0+ bytes
```

```
df['Country'].mode()

0    France
Name: Country, dtype: object

df.Country.mode()[0]

'France'
```

```
df.Country.fillna(df.Country.mode()[0],inplace=True)
df.Age.fillna(df.Age.median(),inplace=True)
df.Salary.fillna(round(df.Salary.mean()),inplace=True)
df
```

```
df.Salary.fillna(round(df.Salary.mean()),inplace=True)
```

| | 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 | 63778.0 | Yes |
| 5 | France | 35.0 | 58000.0 | Yes |
| 6 | Spain | 38.0 | 52000.0 | No |
| 7 | France | 48.0 | 79000.0 | Yes |
| 8 | Germany | 50.0 | 83000.0 | No |
| 9 | France | 37.0 | 67000.0 | Yes |


```
pd.get_dummies(df.Country)
```

| | France | Germany | Spain |
|---|--------|---------|-------|
| 0 | True | False | False |
| 1 | False | False | True |
| 2 | False | True | False |
| 3 | False | False | True |
| 4 | False | True | False |
| 5 | True | False | False |
| 6 | False | False | True |
| 7 | True | False | False |
| 8 | False | True | False |
| 9 | True | False | False |

```
updated_dataset=pd.concat([pd.get_dummies(df.Country),df.iloc[:,  
[1,2,3]]],axis=1)  
updated_dataset
```

| | France | Germany | Spain | Age | Salary | Purchased |
|---|--------|---------|-------|------|---------|-----------|
| 0 | True | False | False | 44.0 | 72000.0 | No |
| 1 | False | False | True | 27.0 | 48000.0 | Yes |
| 2 | False | True | False | 30.0 | 54000.0 | No |
| 3 | False | False | True | 38.0 | 61000.0 | No |
| 4 | False | True | False | 40.0 | 63778.0 | Yes |
| 5 | True | False | False | 35.0 | 58000.0 | Yes |
| 6 | False | False | True | 38.0 | 52000.0 | No |
| 7 | True | False | False | 48.0 | 79000.0 | Yes |
| 8 | False | True | False | 50.0 | 83000.0 | No |
| 9 | True | False | False | 37.0 | 67000.0 | Yes |

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 10 entries, 0 to 9  
Data columns (total 4 columns):  
#   Column      Non-Null Count  Dtype  
---  ---  
0   Country     10 non-null     object  
1   Age         10 non-null     float64  
2   Salary      10 non-null     float64  
3   Purchased   10 non-null     object  
dtypes: float64(2), object(2)  
memory usage: 452.0+ bytes
```

```
updated_dataset.Purchased.replace(['No','Yes'],[0,1],inplace=True)  
updated_dataset
```

| | France | Germany | Spain | Age | Salary | Purchased |
|---|--------|---------|-------|------|---------|-----------|
| 0 | True | False | False | 44.0 | 72000.0 | 0 |
| 1 | False | False | True | 27.0 | 48000.0 | 1 |
| 2 | False | True | False | 30.0 | 54000.0 | 0 |
| 3 | False | False | True | 38.0 | 61000.0 | 0 |
| 4 | False | True | False | 40.0 | 63778.0 | 1 |
| 5 | True | False | False | 35.0 | 58000.0 | 1 |
| 6 | False | False | True | 38.0 | 52000.0 | 0 |
| 7 | True | False | False | 48.0 | 79000.0 | 1 |
| 8 | False | True | False | 50.0 | 83000.0 | 0 |
| 9 | True | False | False | 37.0 | 67000.0 | 1 |

3.C. Create And Save a CSV file Using Pandas

Aim:

To create a CSV File and save the file using Pandas Library in python

Algorithm:

1. Import Pandas
2. Prepare Data
3. Create a Dataframe
4. Save to CSV File

Description:

Follow Steps to create and save a file a CSV File using Pandas Library

Code With Output:

```
import pandas as pd
import random
from datetime import datetime, timedelta
```

```

# Step 1: Define columns
columns = [
    "Member ID", "Name", "Age", "Gender",
    "Plan_Type", "Join_Date", "Expiry_Date", "Trainer_Name"
]

# Step 2: Sample data lists
names = [
    "Aarav", "Ananya", "Vihaan", "Ishita", "Advait",
    "Meera", "Rohan", "Priya", "Kabir", "Simran",
    "Arjun", "Neha", "Yash", "Sanya", "Kunal",
    "Ritika", "Omkar", "Ira", "Aditya", "Tanya",
    "Siddharth", "Pooja", "Manav", "Shruti", "Aditi"
]
genders = ["Male", "Female"]
plans = ["Monthly", "Quarterly", "Yearly"]
trainers = ["Raj", "Priya", "Amit", "Sonal", "Vikram"]

# Step 3: Generate dataset
data = []
start_date = datetime(2025, 1, 1)

for i in range(25):
    member_id = f"M{100+i}"
    name = names[i]
    age = random.randint(18, 50)
    gender = random.choice(genders)
    plan = random.choice(plans)

    join_date = start_date + timedelta(days=random.randint(0, 60))

    if plan == "Monthly":
        expiry_date = join_date + timedelta(days=30)
    elif plan == "Quarterly":
        expiry_date = join_date + timedelta(days=90)
    else:
        expiry_date = join_date + timedelta(days=365)

    trainer = random.choice(trainers)

    data.append([
        member_id, name, age, gender, plan,
        join_date.strftime("%Y-%m-%d"),
        expiry_date.strftime("%Y-%m-%d"),
        trainer
    ])

```

```

# Step 4: Create DataFrame
df = pd.DataFrame(data, columns=columns)

```

```
# Step 5: Save to CSV
df.to_csv("gym_members.csv", index=False)

print(" gym members.csv file created with 25 entries.")
print(df.head())
```

gym_members.csv file created with 25 entries.

| | Member ID | Name | Age | Gender | Plan Type | Join Date | Expiry Date | \ |
|---|-----------|--------|-----|--------|-----------|------------|-------------|---|
| 0 | M100 | Aarav | 41 | Female | Yearly | 2025-02-02 | 2026-02-02 | |
| 1 | M101 | Ananya | 41 | Female | Yearly | 2025-02-22 | 2026-02-22 | |
| 2 | M102 | Vihaan | 49 | Male | Quarterly | 2025-02-22 | 2025-05-23 | |
| 3 | M103 | Ishita | 22 | Female | Yearly | 2025-01-12 | 2026-01-12 | |
| 4 | M104 | Advait | 35 | Male | Yearly | 2025-02-23 | 2026-02-23 | |

| | Trainer Name |
|---|--------------|
| 0 | Sonal |
| 1 | Sonal |
| 2 | Raj |
| 3 | Priya |
| 4 | Raj |

Result:

Thus a dataset is created in CSV format and saved using Pandas Library