

1. Consider a dataset such as finance, education, marketing, healthcare etc. Load, inspect, and store data in various formats like CSV, Excel, JSON, and SQL using Pandas. Clean and prepare data by handling missing values, duplicates, normalization, and encoding categorical data.

### Assignment Steps

#### 1. Dataset Selection:

Choose a dataset relevant to a domain of interest (e.g., finance, education, marketing, or healthcare). Public datasets are available on platforms like [Kaggle](#) or [UCI Machine Learning Repository](#).

#### 2. Tasks:

##### a. Load and Inspect the Data:

- Load the dataset into a Pandas DataFrame from a CSV file.
- Display basic statistics using `.info()` and `.describe()`.
- Print the first five rows using `.head()`.

##### b. Store the Data in Various Formats:

- Save the dataset in **Excel**, **JSON**, and **SQL** formats using Pandas.
- Verify each saved format by reloading it back into a DataFrame.

##### c. Clean and Prepare the Data:

- Handle missing values by filling or dropping them using `.fillna()` or `.dropna()`.
- Remove duplicate rows using `.drop_duplicates()`.
- Normalize numerical columns by scaling them between 0 and 1.
- Encode categorical columns using one-hot encoding or label encoding.

#### 3. Code Template:

```
import pandas as pd
from sklearn.preprocessing import MinMaxScaler

# Step 1: Load Dataset
file_path = "Heart.csv" # Replace with your dataset path
df = pd.read_csv(file_path)

# Step 2: Inspect Data
print("Data Info:")
print(df.info())
print("\nData Statistics:")
print(df.describe())
print("\nFirst 5 Rows:")
print(df.head())

# Step 3: Save Data in Various Formats
df.to_excel("dataset.xlsx", index=False)
df.to_json("dataset.json", orient="records", indent=2)
from sqlalchemy import create_engine
engine = create_engine("sqlite:///memory:") # In-memory SQL database
df.to_sql("dataset", con=engine, index=False, if_exists="replace")

# Step 4: Reload Data to Verify
```

```

df_excel = pd.read_excel("dataset.xlsx")
df_json = pd.read_json("dataset.json")
df_sql = pd.read_sql("dataset", con=engine)

# Step 5: Clean and Prepare Data
# Handle Missing Values
df.fillna(method="ffill", inplace=True) # Forward fill missing values

# Remove Duplicates
df.drop_duplicates(inplace=True)

# Normalize Numerical Data
scaler = MinMaxScaler()
numerical_cols = df.select_dtypes(include=["number"]).columns
df[numerical_cols] = scaler.fit_transform(df[numerical_cols])

# Encode Categorical Data
df = pd.get_dummies(df, drop_first=True) # One-hot encoding

# Final Dataset
print("\nCleaned Data:")
print(df.head())

# Save the cleaned dataset
df.to_csv("cleaned_dataset.csv", index=False)

```

#### 4. Deliverables:

- Python code file (.py or .ipynb) implementing the above steps.
- Saved data files in **CSV**, **Excel**, **JSON**, and **SQL** formats.
- A brief report (in .docx or .pdf) summarizing the observations and challenges faced during the assignment.

#### Evaluation Criteria:

1. Proper loading and inspection of the dataset.
2. Successful storage and reloading of the dataset in all specified formats.
3. Effective handling of missing values and duplicates.
4. Proper normalization and encoding of data.
5. Clarity and efficiency of the Python code.
6. Quality of the report summarizing observations and challenges.

## OUTPUT SAMPLE :

### Data Info:

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 303 entries, 0 to 302

Data columns (total 15 columns):

# Column Non-Null Count Dtype

```
---
0 Unnamed: 0 303 non-null int64
1 Age 303 non-null int64
2 Sex 303 non-null int64
3 ChestPain 303 non-null object
4 RestBP 303 non-null int64
5 Chol 303 non-null int64
6 Fbs 303 non-null int64
7 RestECG 303 non-null int64
8 MaxHR 303 non-null int64
9 ExAng 303 non-null int64
10 Oldpeak 303 non-null float64
11 Slope 303 non-null int64
12 Ca 299 non-null float64
13 Thal 301 non-null object
14 AHD 303 non-null object
```

dtypes: float64(2), int64(10), object(3)

memory usage: 35.6+ KB

None

### Data Statistics:

	Unnamed: 0	Age	Sex	RestBP	Chol	Fbs \
count	303.000000	303.000000	303.000000	303.000000	303.000000	303.000000
mean	152.000000	54.438944	0.679868	131.689769	246.693069	0.148515
std	87.612784	9.038662	0.467299	17.599748	51.776918	0.356198
min	1.000000	29.000000	0.000000	94.000000	126.000000	0.000000
25%	76.500000	48.000000	0.000000	120.000000	211.000000	0.000000
50%	152.000000	56.000000	1.000000	130.000000	241.000000	0.000000
75%	227.500000	61.000000	1.000000	140.000000	275.000000	0.000000
max	303.000000	77.000000	1.000000	200.000000	564.000000	1.000000

	RestECG	MaxHR	ExAng	Oldpeak	Slope	Ca
count	303.000000	303.000000	303.000000	303.000000	303.000000	299.000000
mean	0.990099	149.607261	0.326733	1.039604	1.600660	0.672241
std	0.994971	22.875003	0.469794	1.161075	0.616226	0.937438
min	0.000000	71.000000	0.000000	0.000000	1.000000	0.000000
25%	0.000000	133.500000	0.000000	0.000000	1.000000	0.000000
50%	1.000000	153.000000	0.000000	0.800000	2.000000	0.000000
75%	2.000000	166.000000	1.000000	1.600000	2.000000	1.000000
max	2.000000	202.000000	1.000000	6.200000	3.000000	3.000000

### First 5 Rows:

	Unnamed: 0	Age	Sex	ChestPain	RestBP	Chol	Fbs	RestECG	MaxHR \
0	1	63	1	typical	145	233	1	2	150
1	2	67	1	asymptomatic	160	286	0	2	108

2	3	67	1	asymptomatic	120	229	0	2	129
3	4	37	1	nonanginal	130	250	0	0	187
4	5	41	0	nontypical	130	204	0	2	172

	ExAng	Oldpeak	Slope	Ca	Thal	AHD
0	0	2.3	3	0.0	fixed	No
1	1	1.5	2	3.0	normal	Yes
2	1	2.6	2	2.0	reversable	Yes
3	0	3.5	3	0.0	normal	No
4	0	1.4	1	0.0	normal	No

Cleaned Data:

	Unnamed: 0	Age	Sex	RestBP	Chol	Fbs	RestECG	MaxHR \
0	0.000000	0.708333	1.0	0.481132	0.244292	1.0	1.0	0.603053
1	0.003311	0.791667	1.0	0.622642	0.365297	0.0	1.0	0.282443
2	0.006623	0.791667	1.0	0.245283	0.235160	0.0	1.0	0.442748
3	0.009934	0.166667	1.0	0.339623	0.283105	0.0	0.0	0.885496
4	0.013245	0.250000	0.0	0.339623	0.178082	0.0	1.0	0.770992

	ExAng	Oldpeak	Slope	Ca	ChestPain_nonanginal \
0	0.0	0.370968	1.0	0.000000	False
1	1.0	0.241935	0.5	1.000000	False
2	1.0	0.419355	0.5	0.666667	False
3	0.0	0.564516	1.0	0.000000	True
4	0.0	0.225806	0.0	0.000000	False

	ChestPain_nontypical	ChestPain_typical	Thal_normal	Thal_reversable \
0	False	True	False	False
1	False	False	True	False
2	False	False	False	True
3	False	False	True	False
4	True	False	True	False

	AHD_Yes
0	False
1	True
2	True
3	False
4	False

[ ]: