

DATA SCIENCE LAB EXPERIMENT 1

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AIM: Introduction to Data science and Data preparation using Pandas steps.

THEORY:

Pandas: Pandas is an open-source Python library used for data manipulation and analysis. It provides high-performance data structures and functions for efficiently handling structured data.

Key Pandas Functions for Data Cleaning

1. Handling Missing Data

- `df.isnull().sum()` → Check the number of missing values in each column.
- `df.dropna()` → Remove rows with missing values.
- `df.fillna(value, inplace=True)` → Fill missing values with a specific value (e.g., mean or median).

2. Removing Duplicates

- `df.duplicated()` → Identify duplicate rows.
- `df.drop_duplicates(inplace=True)` → Remove duplicate rows.

3. Handling Incorrect Data Formats

- `df['column'] = pd.to_datetime(df['column'])` → Convert a column to a datetime format.
- `df['column'] = df['column'].astype(int/float/str)` → Change data types.

Topic: [Bengaluru Housing Prices](#)

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1. Loading Data in Pandas:

```

aids1.py > ...
1  import pandas as pd
2
3
4  data = pd.read_excel('Bengaluru_House_Data.xlsx') # Load the dataset
5  print(data.head())

```

PROBLEMS OUTPUT DEBUG CONSOLE **TERMINAL** PORTS COMMENTS

File "parsers.pyx", line 891, in pandas._libs.parsers.TextReader._check_tokenize_status
File "parsers.pyx", line 2053, in pandas._libs.parsers.raise_parser_error
UnicodeDecodeError: 'utf-8' codec can't decode byte 0xb0 in position 10: invalid start byte

- PS C:\Users\lauki\OneDrive\Desktop\dataset> python aids1.py
- PS C:\Users\lauki\OneDrive\Desktop\dataset> python aids1.py

	area_type	availability	location	...	bath	balcony	price
0	Super built-up Area	2025-12-19 00:00:00	Electronic City Phase II	...	2.0	1.0	39.07
1	Plot Area	Ready To Move	Chikka Tirupathi	...	5.0	3.0	120.00
2	Built-up Area	Ready To Move	Uttarahalli	...	2.0	3.0	62.00
3	Super built-up Area	Ready To Move	Lingadheeranahalli	...	3.0	1.0	95.00
4	Super built-up Area	Ready To Move	Kothanur	...	2.0	1.0	51.00

2. Description of the Dataset:

```
print(data.head())
print(data.describe())
```

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[5 rows x 9 columns]

	bath	balcony	price
count	13247.000000	12711.000000	13320.000000
mean	2.692610	1.584376	112.565627
std	1.341458	0.817263	148.971674
min	1.000000	0.000000	8.000000
25%	2.000000	1.000000	50.000000
50%	2.000000	2.000000	72.000000
75%	3.000000	2.000000	120.000000
max	40.000000	3.000000	3600.000000

3. Drop columns that are not useful:

```
1 import pandas as pd
2
3
4 data = pd.read_excel('Bengaluru_House_Data.xlsx')
5 data = data.drop(columns=['bath'])
6 print(data.head())
7
```

PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL PORTS COMMENTS

powershell + v [] [] ... ^

```
5%      2.000000      1.000000      50.000000
0%      2.000000      2.000000      72.000000
5%      3.000000      2.000000      120.000000
max     40.000000      3.000000     3600.000000
```

C:\Users\lauki\OneDrive\Desktop\dataset> python aids1.py

	area_type	availability	location	...	total_sqft	balcony	price
Super built-up	Area	2025-12-19 00:00:00	Electronic City Phase II	...	1056	1.0	39.07
Plot	Area	Ready To Move	Chikka Tirupathi	...	2600	3.0	120.00
Built-up	Area	Ready To Move	Uttarahalli	...	1440	3.0	62.00
Super built-up	Area	Ready To Move	Lingadheeranahalli	...	1521	1.0	95.00
Super built-up	Area	Ready To Move	Kothanur	...	1200	1.0	51.00

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After dropping Number of bathrooms column:

```
PS C:\Users\lauki\OneDrive\Desktop\dataset> python aids1.py
```

		area_type	availability	location
0	Super built-up	Area	2025-12-19 00:00:00	Electronic City Phase II
1	Plot	Area	Ready To Move	Chikka Tirupathi
2	Built-up	Area	Ready To Move	Uttarahalli
3	Super built-up	Area	Ready To Move	Lingadheeranahalli
4	Super built-up	Area	Ready To Move	Kothanur

	size	society	total_sqft	balcony	price
0	2 BHK	Coomee	1056	1.0	39.07
1	4 Bedroom	Theanmp	2600	3.0	120.00
2	3 BHK	NaN	1440	3.0	62.00
3	3 BHK	Soiewre	1521	1.0	95.00
4	2 BHK	NaN	1200	1.0	51.00

4. Drop rows with maximum missing values:

Before Dropping:

	size	society	total_sqft	balcony	price
0	2 BHK	Coomee	1056	1.0	39.07
1	4 Bedroom	Theanmp	2600	3.0	120.00
2	3 BHK	NaN	1440	3.0	62.00
3	3 BHK	Soiewre	1521	1.0	95.00
4	2 BHK	NaN	1200	1.0	51.00

Sheet Size: (13320, 8)

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```
7
8 # Drop rows with too many missing values (e.g., more than 50% missing)
9 data = data.dropna(thresh=len(data.columns) / 2)
10
11 pd.set_option('display.max_columns', None)
12 print(data.head())
13
```

PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL PORTS COMMENTS powershell + v

3	Super built-up Area	Ready To Move	Lingadheeranahalli
4	Super built-up Area	Ready To Move	Kothanur

	size	society	total_sqft	balcony	price
0	2 BHK	Coomee	1056	1.0	39.07
1	4 Bedroom	Theanmp	2600	3.0	120.00
2	3 BHK	NaN	1440	3.0	62.00
3	3 BHK	Soiewre	1521	1.0	95.00
4	2 BHK	NaN	1200	1.0	51.00

After Dropping:

	size	society	total_sqft	balcony	price
0	2 BHK	Coomee	1056	1.0	39.07
1	4 Bedroom	Theanmp	2600	3.0	120.00
2	3 BHK	NaN	1440	3.0	62.00
3	3 BHK	Soiewre	1521	1.0	95.00
4	2 BHK	NaN	1200	1.0	51.00

Sheet Size: (13320, 8)

Since there are no rows with maximum missing values (more than 50% of the cells being empty), no rows were dropped.

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5. Take care of missing data:

Dropping rows if society name is missing

Before Dropping:

```
      size  society  total_sqft  balcony  price
0    2 BHK   Coomee      1056      1.0   39.07
1  4 Bedroom  Theanmp      2600      3.0  120.00
2    3 BHK     NaN      1440      3.0   62.00
3    3 BHK  Soiewre      1521      1.0   95.00
4    2 BHK     NaN      1200      1.0   51.00
Sheet Size: (13320, 8)
```

After Dropping:

```
12 # Drop rows where 'society' column has missing values
13 data = data.dropna(subset=['society'])
14
15 pd.set_option('display.max_columns', None)
16 print(data.head())
17 print("Sheet Size:", data.shape)
18
```

PROBLEMS	OUTPUT	DEBUG CONSOLE	TERMINAL	PORTS	COMMENTS
6	Super built-up Area	2025-05-18 00:00:00			Old Airport Road
	size	society	total_sqft	balcony	price
0	2 BHK	Coomee	1056	1.0	39.07
1	4 Bedroom	Theanmp	2600	3.0	120.00
3	3 BHK	Soiewre	1521	1.0	95.00
5	2 BHK	DuenaTa	1170	1.0	38.00
6	4 BHK	Jaades	2732	NaN	204.00
Sheet Size: (7818, 8)					

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6. Creating Dummy variables for the balcony column:

In data science, dummy values (or dummy variables) are used to represent categorical data in a numerical format so that machine learning models can process them effectively. Most machine learning models cannot handle categorical data directly. Converting categorical variables into dummy (binary) variables allows models to interpret them numerically.

```
# Convert 'balcony' column into dummy variables
if 'balcony' in data.columns:
    data = pd.get_dummies(data, columns=['balcony'])
```

```

    size  society  total_sqft  price  balcony_0.0  balcony_1.0  \
0    2 BHK   Coomee      1056   39.07          False          True
1  4 Bedroom  Theanmp      2600  120.00          False          False
3    3 BHK   Soiewre      1521   95.00          False          True
5    2 BHK   Duenata      1170   38.00          False          True
6    4 BHK    Jaades      2732  204.00          False          False

    balcony_2.0  balcony_3.0
0          False          False
1          False          True
3          False          False
5          False          False
6          False          False
Sheet Size: (7818, 11)
PS C:\Users\lauki\OneDrive\Desktop\dataset>
```

7. Finding Outliers:

The IQR method is used to find the outliers manually. The IQR (Interquartile Range) method is a statistical technique used to detect and handle outliers in a dataset. It is based on the spread of the middle 50% of the data.

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```
def find_outliers_iqr(data):
    Q1 = np.percentile(data, 25)
    Q3 = np.percentile(data, 75)
    IQR = Q3 - Q1
    lower_bound = Q1 - 1.5 * IQR
    upper_bound = Q3 + 1.5 * IQR
    return data[(data < lower_bound) | (data > upper_bound)]

# Apply to specific column
outliers = find_outliers_iqr(data['price']) # Replace 'column_name'
print(outliers)
```

```
PS C:\Users\lauki\OneDrive\Desktop\dataset> python aids1.py
6          204.0
7          600.0
11         295.0
18         290.0
22         380.0
...
13268      221.0
13269      201.0
13290      450.0
13315      231.0
13318      488.0
Name: price, Length: 670, dtype: float64
```

8. Standardization and Normalization of columns

Standardization is to ensure that all the features are transformed such that the mean is 0 and standard deviation is 1.

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```
# Identify and remove non-numeric columns
numeric_cols = data.select_dtypes(include=['number']).columns
data_numeric = data[numeric_cols] # Keep only numeric columns

# Standardizing only numeric columns
scaler = StandardScaler()
df_standardized = pd.DataFrame(scaler.fit_transform(data_numeric), columns=numeric_cols)
```

This code will help to print the standardized values:

$X(\text{standardized}) = (X - \mu) / \sigma$

```
PS C:\Users\lauki\OneDrive\Desktop\dataset> python aids1.py
      price
0 -0.535859
1  0.159350
2 -0.055406
3 -0.545051
4  0.880931
Sheet Size: (7818, 11)
```

Normalization is the process of scaling all the features to a range [0, 1]. It is also called min-max scaling.

```
# Initialize MinMaxScaler (default range [0,1])
scaler = MinMaxScaler()

# Apply Min-Max Normalization
data_numeric = pd.DataFrame(scaler.fit_transform(data_numeric), columns=numeric_cols)

# Print first few rows of the normalized data
print(data_numeric.head())
```

This code will help us to perform min max normalization and scale the features to range between [0, 1].

$X(\text{normalized}) = (X - X(\min)) / (X(\max) - X(\min))$

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```
PS C:\Users\lauki\OneDrive\Desktop\dataset> python aids1.py
      price
0  0.011542
1  0.041605
2  0.032318
3  0.011144
4  0.072808
Sheet Size: (7818, 11)
```

Conclusion: Thus we have successfully prepared the data from an unclean dataset using Pandas. It helps us in loading data from various file formats (e.g., CSV, Excel, SQL) into a structured DataFrame for easier manipulation, cleaning, and analysis. Removing irrelevant or redundant columns helps to reduce dimensionality and focus on important features, improving model performance. Remove irrelevant or redundant columns to reduce dimensionality and focus on important features, improving model performance. Remove irrelevant or redundant columns to reduce dimensionality and focus on important features, improving model performance. This experiment has helped us to understand these concepts efficiently.