EXP NO:

EDA-DATA CLEANING

AIM

To clean data by handling missing values, duplicates, data types, and normalization.

PROBLEM STATEMENT

Clean a dataset by removing nulls, duplicates, and normalizing numeric fields.

ALGORITHM

- 1. Load dataset.
- 2. Detect missing values (isnull).
- 3. Fill or drop missing values.
- 4. Remove duplicates.
- 5. Convert data types.
- 6. Normalize numeric columns.

SAMPLE CODE

import pandas as pd

from sklearn.preprocessing import StandardScaler, MinMaxScaler import matplotlib.pyplot as plt

Step 1: Load dataset

df = pd.read_csv('StudentsPerformance.csv')

df.head()

	gender	race/ethnicity	level of education	lunch	preparation course	math score	reading score	writing score
0	female	group B	bachelor's degree	standard	none	72	72	74
1	female	group C	some college	standard	completed	69	90	88
2	female	group B	master's degree	standard	none	90	95	93
3	male	group A	associate's degree	free/reduced	none	47	57	44
4	male	group C	some college	standard	none	76	78	75

```
df.shape
(1005, 8)
# Step 2: Handle Missing Values
# Detect
missing info = df.isnull().sum()
print("Missing values:\n", missing info)
Missing values:
 gender
                                      0
race/ethnicity
                                     0
parental level of education
                                     7
lunch
                                     0
test preparation course
                                     0
math score
                                     0
reading score
                                     0
writing score
                                     0
dtype: int64
# Fill or Drop (based on context)
df.fillna({
  'parental level of education': df['parental level of education'].mode()[0],
  'lunch': df['lunch'].mode()[0]
}, inplace=True)
missing info = df.isnull().sum()
missing_info
                                     0
gender
race/ethnicity
                                     0
parental level of education
                                     0
lunch
                                     0
test preparation course
                                     0
math score
reading score
                                     0
writing score
                                     0
dtype: int64
duplicates = df[df.duplicated()]
duplicates
```

	gender	race/ethnicity	parental level of education	lunch	test preparation course	math score	reading score	writing score
1000	male	group D	some college	standard	none	76	64	66
1001	male	group C	associate's degree	standard	none	46	43	42
1002	female	group B	bachelor's degree	standard	none	67	86	83
1003	male	group E	some high school	standard	none	92	87	78
1004	male	group C	bachelor's degree	standard	completed	83	82	84

duplicates.shape

(5, 8)

Drop duplicates

df.drop_duplicates(inplace=True)

df.shape

Step 4: Convert Data Types (if needed)

For consistency, make sure string columns are lowercase

categorical_cols = ['gender', 'race/ethnicity', 'parental level of education', 'lunch', 'test preparation course']

for col in categorical_cols:

df[col] = df[col].astype(str).str.lower().str.strip()

categorical_cols

['gender',

'race/ethnicity',

'parental level of education',

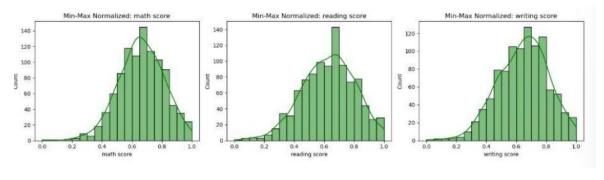
'lunch',

'test preparation course']

numeric_cols = ['math score', 'reading score', 'writing score']
numeric_cols

```
['math score', 'reading score', 'writing score']
plt.figure(figsize=(15, 4))
for i, col in enumerate(numeric_cols):
  plt.subplot(1, 3, i+1)
  sns.histplot(df[col], kde=True, bins=20)
  plt.title(f'Before Normalization: {col}')
plt.tight_layout()
plt.show()
          Before Normalization: math score
                                              Before Normalization: reading score
                                                                                   Before Normalization: writing score
                                                                            120
                                      120
 120
                                                                            100
minmax scaler = MinMaxScaler()
df_minmax = df.copy()
df_minmax[numeric_cols] = minmax_scaler.fit_transform(df[numeric_cols])
plt.figure(figsize=(15, 4))
for i, col in enumerate(numeric_cols):
  plt.subplot(1, 3, i+1)
  sns.histplot(df_minmax[col], kde=True, bins=20, color='green')
  plt.title(f'Min-Max Normalized: {col}')
plt.tight_layout()
```

plt.show()



```
# Standard Scaling (Z-score)

zscore_scaler = StandardScaler()

df_zscore = df.copy()

df_zscore[numeric_cols] = zscore_scaler.fit_transform(df[numeric_cols])

plt.figure(figsize=(15, 4))

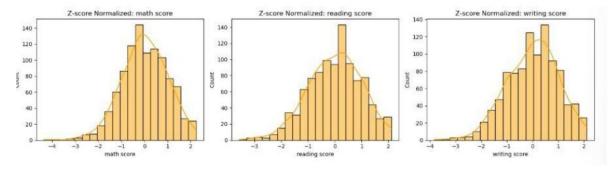
for i, col in enumerate(numeric_cols):

    plt.subplot(1, 3, i+1)

    sns.histplot(df_zscore[col], kde=True, bins=20, color='orange')

    plt.title(f'Z-score Normalized: {col}')

plt.tight_layout()
```



RESULT:

plt.show()

Thus, the program successfully created a Jupyter Notebook showcasing Python code handling missing values, removing duplicates and unnecessary data, Data type conversion and normalizing data.