

```
# Step 1: Import libraries
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.preprocessing import MinMaxScaler, StandardScaler
from google.colab import files
```

```
# Step 2: Upload dataset
print("Please choose your CSV file...")
uploaded = files.upload()
```

Please choose your CSV file...

<IPython.core.display.HTML object>

Saving StudentsPerformance.csv to StudentsPerformance.csv

```
# Get uploaded filename
filename = list(uploaded.keys())[0]
df = pd.read_csv(filename)

print("\nFile loaded successfully!")
print(f"Shape of dataset: {df.shape}")
print(df.head())
```

File loaded successfully!

Shape of dataset: (1005, 8)

	gender	race/ethnicity	parental level of education	lunch	
0	female	group B	bachelor's degree	standard	
1	female	group C	some college	standard	
2	female	group B	master's degree	standard	
3	male	group A	associate's degree	free/reduced	
4	male	group C	some college	standard	

	test preparation course	math score	reading score	writing score
0	none	72	72	74
1	completed	69	90	88
2	none	90	95	93
3	none	47	57	44
4	none	76	78	75

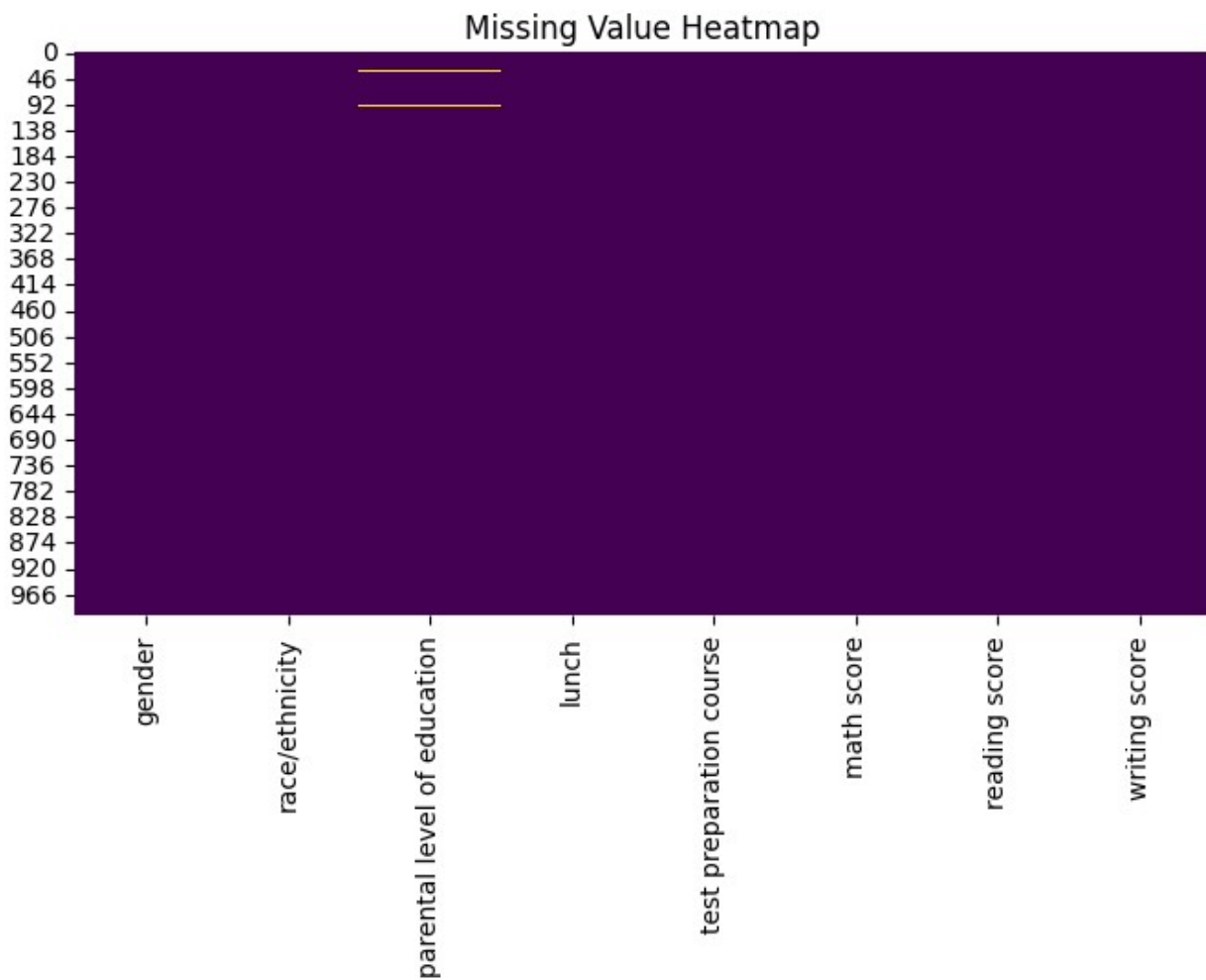
```
# -----
# 1. Find Missing Values
# -----
print("\nMissing Values in Each Column:")
print(df.isnull().sum())
```

```
# Heatmap for missing values
plt.figure(figsize=(8, 4))
```

```
sns.heatmap(df.isnull(), cbar=False, cmap="viridis")
plt.title("Missing Value Heatmap")
plt.show()
```

Missing Values in Each Column:

```
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
```



```
# -----
# 2. Imputation of Missing Values
# -----
```

```
# Numeric columns → fill with mean
num_cols = df.select_dtypes(include=['float64', 'int64']).columns
for col in num_cols:
    df[col] = df[col].fillna(df[col].mean())

cat_cols = df.select_dtypes(include=['object']).columns
for col in cat_cols:
    df[col] = df[col].fillna(df[col].mode()[0])

print("\nMissing Values After Imputation:")
print(df.isnull().sum())
```

```
Missing Values After Imputation:
gender                                0
race/ethnicity                        0
parental level of education           0
lunch                                 0
test preparation course               0
math score                           0
reading score                         0
writing score                         0
dtype: int64
```

```
# -----
# 3. Remove Duplicates
# -----
print(f"\nRows before removing duplicates: {len(df)}")
df.drop_duplicates(inplace=True)
print(f"Rows after removing duplicates: {len(df)}")
```

```
Rows before removing duplicates: 1005
Rows after removing duplicates: 1000
```

```
print("\nData Types After Conversion:")
print(df.dtypes)
```

```
Data Types After Conversion:
gender                                object
race/ethnicity                        object
parental level of education           object
lunch                                 object
test preparation course               object
math score                           int64
reading score                         int64
writing score                         int64
dtype: object
```

```
scaler_minmax = MinMaxScaler()
df_minmax = pd.DataFrame(scaler_minmax.fit_transform(df[num_cols]),
columns=num_cols)
```

```
# Z-score Standardization
```

```
scaler_zscore = StandardScaler()
df_zscore = pd.DataFrame(scaler_zscore.fit_transform(df[num_cols]),
columns=num_cols)
```

```
print("\nFirst 5 Rows After Min-Max Normalization:")
```

```
print(df_minmax.head())
```

```
print("\nFirst 5 Rows After Z-score Standardization:")
```

```
print(df_zscore.head())
```

```
First 5 Rows After Min-Max Normalization:
```

	math score	reading score	writing score
0	0.72	0.662651	0.711111
1	0.69	0.879518	0.866667
2	0.90	0.939759	0.922222
3	0.47	0.481928	0.377778
4	0.76	0.734940	0.722222

```
First 5 Rows After Z-score Standardization:
```

	math score	reading score	writing score
0	0.390024	0.193999	0.391492
1	0.192076	1.427476	1.313269
2	1.577711	1.770109	1.642475
3	-1.259543	-0.833899	-1.583744
4	0.653954	0.605158	0.457333

```
# -----
```

```
# 6. Visualization in Seaborn (Histograms Only)
```

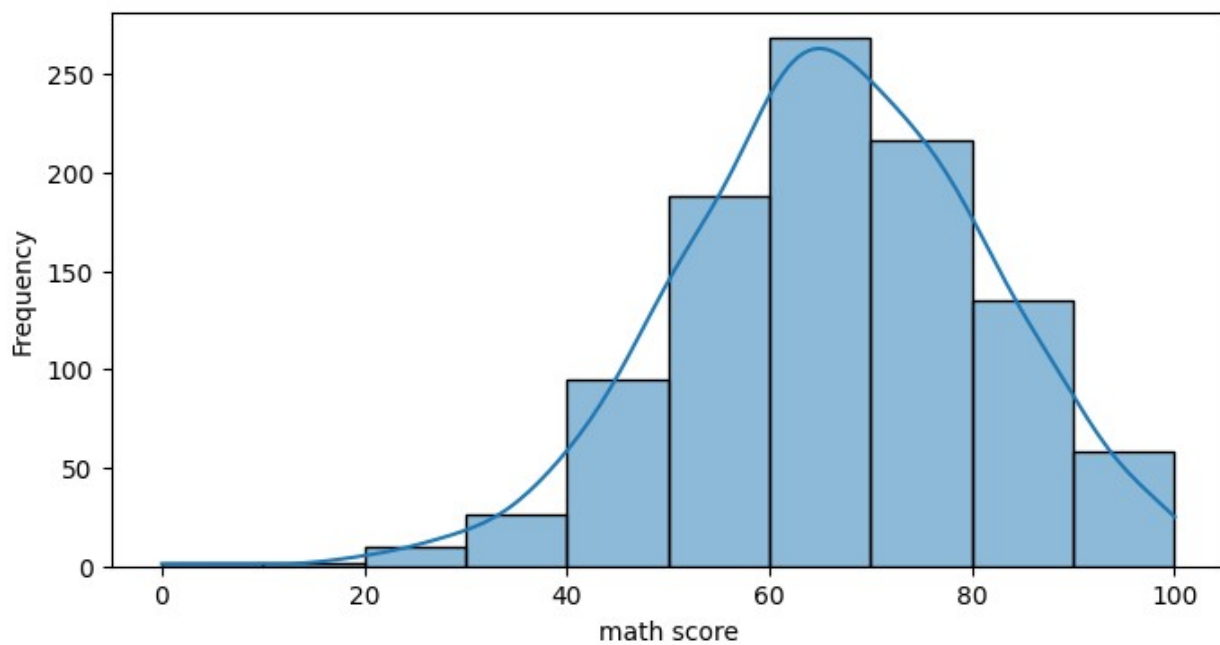
```
# -----
```

```
# Plot histograms for all numeric columns
```

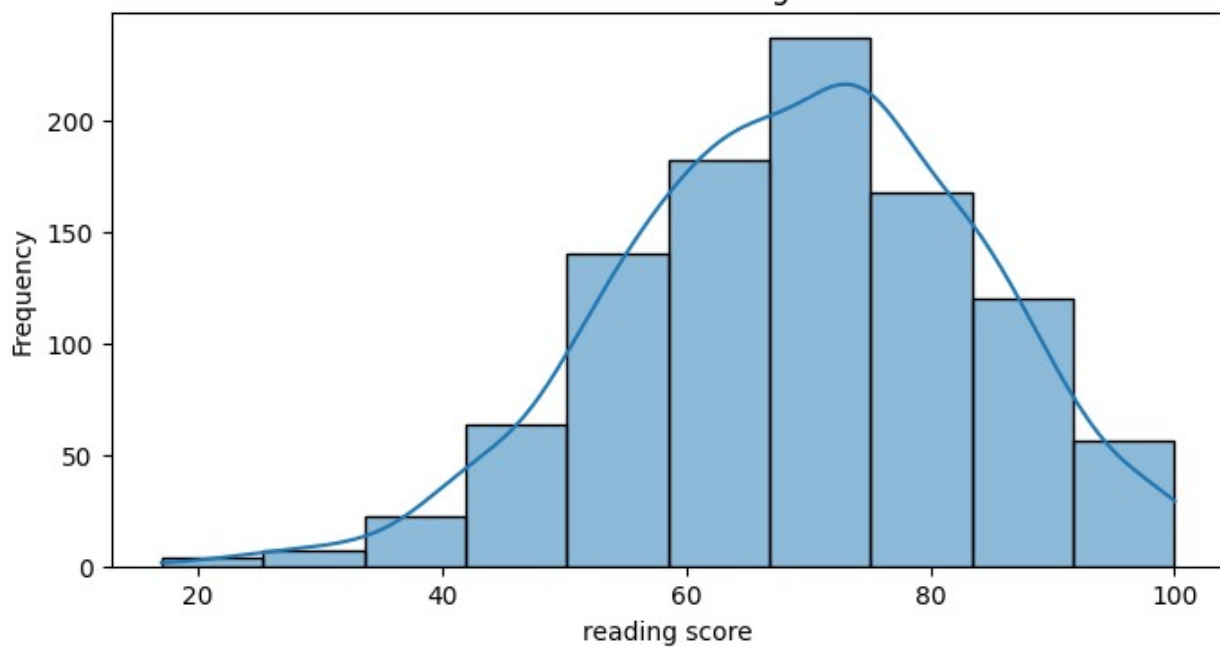
```
for col in num_cols:
    plt.figure(figsize=(8, 4))
    sns.histplot(df[col], kde=True, bins=10)
    plt.title(f"Distribution of {col}")
    plt.xlabel(col)
    plt.ylabel("Frequency")
    plt.show()
```

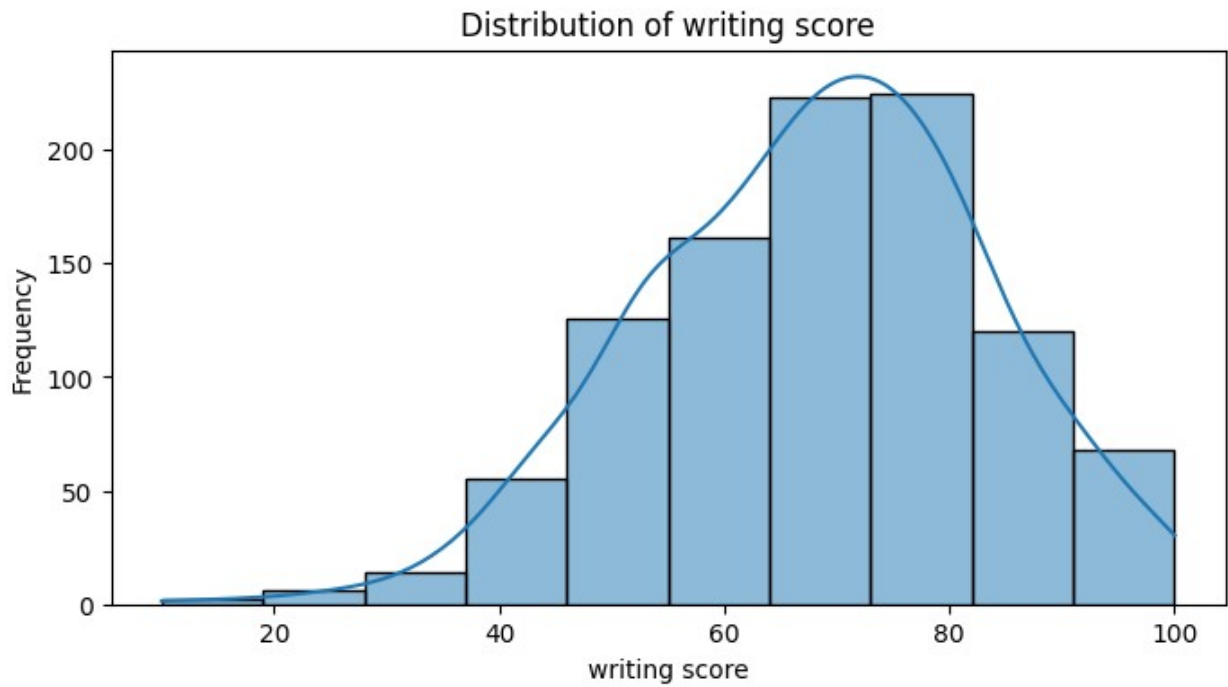
```
print("\n Histograms for numeric features displayed successfully!")
```

Distribution of math score



Distribution of reading score





□ Histograms for numeric features displayed successfully!