

```

import pandas as pd

# Creating the dataset
data = {
    "CustomerID": ["C001", "C002", "C003", "C004", "C004", "C005", "C006", "C007", "C008", "C008"],
    "Name": ["John", "Alice", "BOB", "BOBY", "Eve", "eve", "Steve", "Ramu", "mary", "Bob"],
    "Age": [25, 34, 17, 29, 29, 120, -5, None, 220, 30],
    "JoinDate": ["12/1/2024", "11/15/2023", "6/1/2022", "6/1/2022", "12/5/2024", "invalid_date", None, "1/1/2024", "3/5/2023", "3/5/2023"],
    "MonthlyCharges": [29.85, 56.95, 4000, 75.5, 75.5, 45.99, 60, 49.99, -30, 55],
    "Churn": ["No", "Yes", "No", "No", "No", "Yes", "No", None, "Yes", "No"]
}

# Converting the dictionary into a pandas DataFrame
df = pd.DataFrame(data)

# Display the dataset
print(df)

# Checking for missing values using isnull()
print("Missing values using isnull():")
print(df.isnull())

# Checking for missing values using isna()
print("\nMissing values using isna():")
print(df.isna())

# Checking for non-missing values using notna()
print("\nNon-missing values using notna():")
print(df.notna())

# Checking for non-missing values using notnull()
print("\nNon-missing values using notnull():")
print(df.notnull())

# Checking if any column has missing values using any()
print("\nColumns with missing values using any():")
print(df.isnull().any())

# Checking if any column has missing values using sum()
print("\nCount of missing values in each column using sum():")
print(df.isnull().sum())

import missingno as msno
import pandas as pd

# Assuming 'df' is your DataFrame

# Visualizing missing values using msno.matrix()
msno.matrix(df)

```

| | CustomerID | Name | Age | JoinDate | MonthlyCharges | Churn |
|---|------------|-------|-------|--------------|----------------|-------|
| 0 | C001 | John | 25.0 | 12/1/2024 | 29.85 | No |
| 1 | C002 | Alice | 34.0 | 11/15/2023 | 56.95 | Yes |
| 2 | C003 | BOB | 17.0 | 6/1/2022 | 4000.00 | No |
| 3 | C004 | BOBY | 29.0 | 6/1/2022 | 75.50 | No |
| 4 | C004 | Eve | 29.0 | 12/5/2024 | 75.50 | No |
| 5 | C005 | eve | 120.0 | invalid_date | 45.99 | Yes |
| 6 | C006 | Steve | -5.0 | None | 60.00 | No |
| 7 | C007 | Ramu | NaN | 1/1/2024 | 49.99 | None |
| 8 | C008 | mary | 220.0 | 3/5/2023 | -30.00 | Yes |
| 9 | C008 | Bob | 30.0 | 3/5/2023 | 55.00 | No |

Missing values using isnull():

| | CustomerID | Name | Age | JoinDate | MonthlyCharges | Churn |
|---|------------|-------|-------|----------|----------------|-------|
| 0 | False | False | False | False | False | False |
| 1 | False | False | False | False | False | False |
| 2 | False | False | False | False | False | False |
| 3 | False | False | False | False | False | False |
| 4 | False | False | False | False | False | False |
| 5 | False | False | False | False | False | False |
| 6 | False | False | False | True | False | False |
| 7 | False | False | True | False | False | True |
| 8 | False | False | False | False | False | False |
| 9 | False | False | False | False | False | False |

Missing values using isna():

| | CustomerID | Name | Age | JoinDate | MonthlyCharges | Churn |
|---|------------|-------|-------|----------|----------------|-------|
| 0 | False | False | False | False | False | False |
| 1 | False | False | False | False | False | False |
| 2 | False | False | False | False | False | False |
| 3 | False | False | False | False | False | False |
| 4 | False | False | False | False | False | False |
| 5 | False | False | False | False | False | False |
| 6 | False | False | False | True | False | False |
| 7 | False | False | True | False | False | True |
| 8 | False | False | False | False | False | False |
| 9 | False | False | False | False | False | False |

Non-missing values using notna():

| | CustomerID | Name | Age | JoinDate | MonthlyCharges | Churn |
|---|------------|------|-------|----------|----------------|-------|
| 0 | True | True | True | True | True | True |
| 1 | True | True | True | True | True | True |
| 2 | True | True | True | True | True | True |
| 3 | True | True | True | True | True | True |
| 4 | True | True | True | True | True | True |
| 5 | True | True | True | True | True | True |
| 6 | True | True | True | False | True | True |
| 7 | True | True | False | True | True | False |
| 8 | True | True | True | True | True | True |
| 9 | True | True | True | True | True | True |

Non-missing values using notnull():

| | CustomerID | Name | Age | JoinDate | MonthlyCharges | Churn |
|---|------------|------|-------|----------|----------------|-------|
| 0 | True | True | True | True | True | True |
| 1 | True | True | True | True | True | True |
| 2 | True | True | True | True | True | True |
| 3 | True | True | True | True | True | True |
| 4 | True | True | True | True | True | True |
| 5 | True | True | True | True | True | True |
| 6 | True | True | True | False | True | True |
| 7 | True | True | False | True | True | False |
| 8 | True | True | True | True | True | True |
| 9 | True | True | True | True | True | True |

Columns with missing values using any():

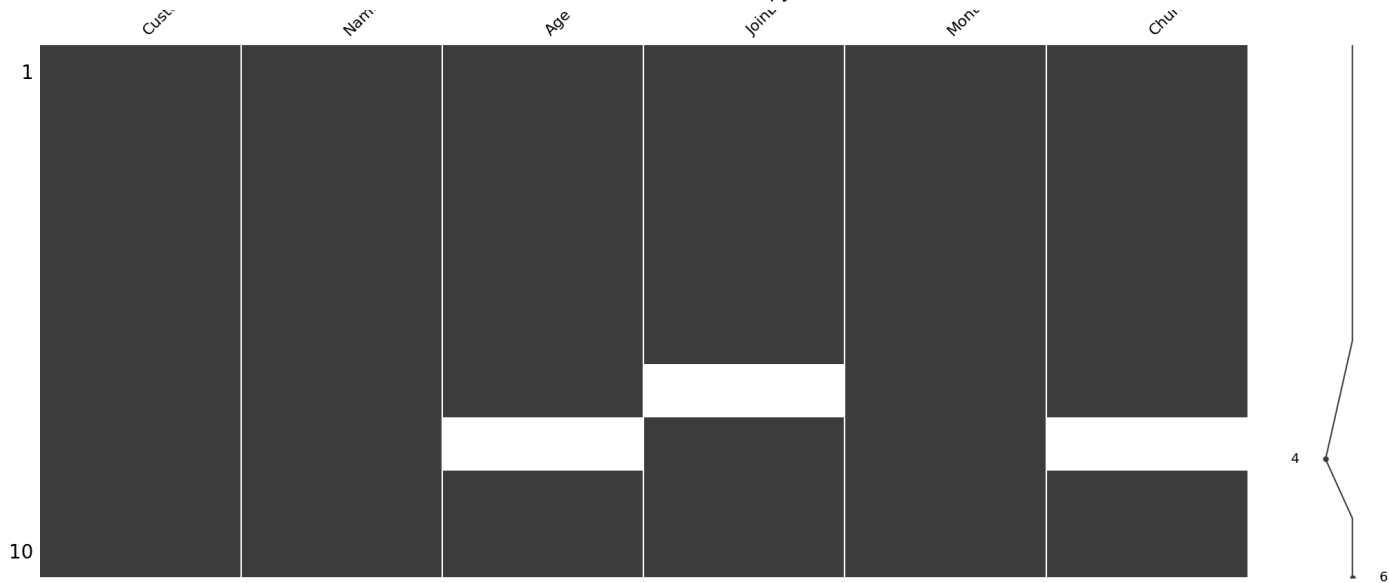
| | |
|----------------|-------|
| CustomerID | False |
| Name | False |
| Age | True |
| JoinDate | True |
| MonthlyCharges | False |
| Churn | True |
| dtype: | bool |

Count of missing values in each column using sum():

| | |
|----------------|---|
| CustomerID | 0 |
| Name | 0 |
| Age | 1 |
| JoinDate | 1 |
| MonthlyCharges | 0 |
| Churn | 1 |

dtype: int64

<Axes: >



```
import pandas as pd

# Assuming 'df' is your original DataFrame

# Fill missing values for 'Age' and 'MonthlyCharges' using mean
df['Age'] = df['Age'].fillna(df['Age'].mean())
df['MonthlyCharges'] = df['MonthlyCharges'].fillna(df['MonthlyCharges'].mean())

# Fill missing values for 'Churn' using mode (since it's categorical)
df['Churn'] = df['Churn'].fillna(df['Churn'].mode()[0])

# Fill missing values for 'JoinDate' using mode (since it's categorical)
df['JoinDate'] = df['JoinDate'].fillna(df['JoinDate'].mode()[0])

# Fill missing values for 'Name' using mode (since it's categorical)
df['Name'] = df['Name'].fillna(df['Name'].mode()[0])

# Display the new dataset with missing values filled
print(df)

df['Age'] = df['Age'].replace(-5, df['Age'].mean()) # Replace -5 with the mean of Age
df['MonthlyCharges'] = df['MonthlyCharges'].replace(-30, df['MonthlyCharges'].mean())
df['JoinDate'] = df['JoinDate'].replace('invalid_date', df['JoinDate'].mode()[0]) # Replace 'invalid_date' with the mode of JoinDate
# Display the new dataset with missing values filled
print(df)
```

| | CustomerID | Name | Age | JoinDate | MonthlyCharges | Churn |
|---|------------|-------|------------|--------------|----------------|-------|
| 0 | C001 | John | 25.000000 | 12/1/2024 | 29.85 | No |
| 1 | C002 | Alice | 34.000000 | 11/15/2023 | 56.95 | Yes |
| 2 | C003 | BOB | 17.000000 | 6/1/2022 | 4000.00 | No |
| 3 | C004 | BOBY | 29.000000 | 6/1/2022 | 75.50 | No |
| 4 | C004 | Eve | 29.000000 | 12/5/2024 | 75.50 | No |
| 5 | C005 | eve | 120.000000 | invalid_date | 45.99 | Yes |

| | | | | | | |
|---|------|-------|------------|----------|--------|-----|
| 6 | C006 | Steve | -5.000000 | 3/5/2023 | 60.00 | No |
| 7 | C007 | Ramu | 55.444444 | 1/1/2024 | 49.99 | No |
| 8 | C008 | mary | 220.000000 | 3/5/2023 | -30.00 | Yes |
| 9 | C008 | Bob | 30.000000 | 3/5/2023 | 55.00 | No |

| | CustomerID | Name | Age | JoinDate | MonthlyCharges | Churn |
|---|------------|-------|------------|------------|----------------|-------|
| 0 | C001 | John | 25.000000 | 12/1/2024 | 29.850 | No |
| 1 | C002 | Alice | 34.000000 | 11/15/2023 | 56.950 | Yes |
| 2 | C003 | BOB | 17.000000 | 6/1/2022 | 4000.000 | No |
| 3 | C004 | BOBY | 29.000000 | 6/1/2022 | 75.500 | No |
| 4 | C004 | Eve | 29.000000 | 12/5/2024 | 75.500 | No |
| 5 | C005 | eve | 120.000000 | 3/5/2023 | 45.990 | Yes |
| 6 | C006 | Steve | 55.444444 | 3/5/2023 | 60.000 | No |
| 7 | C007 | Ramu | 55.444444 | 1/1/2024 | 49.990 | No |
| 8 | C008 | mary | 220.000000 | 3/5/2023 | 441.878 | Yes |
| 9 | C008 | Bob | 30.000000 | 3/5/2023 | 55.000 | No |

```
import pandas as pd
```

```
# Assuming 'df' is your DataFrame
```

```
# Displaying the unique names in the 'Name' column
unique_names = df['Name'].unique()
```

```
# Printing the unique names
print(unique_names)
```

```
→ ['John' 'Alice' 'BOB' 'BOBY' 'Eve' 'eve' 'Steve' 'Ramu' 'mary' 'Bob']
```

```
import pandas as pd
```

```
# Assuming 'df' is your DataFrame
```

```
# Identify the duplicated rows based on 'CustomerID'
duplicated_rows = df[df.duplicated(subset='CustomerID', keep=False)]
```

```
# Display duplicated rows
print("Duplicated rows based on 'CustomerID':")
print(duplicated_rows)
```

```
# Remove duplicated rows based on 'CustomerID'
df_cleaned = df.drop_duplicates(subset='CustomerID', keep='first')
```

```
# Display the cleaned dataset
print("\nDataset after removing duplicated rows:")
print(df_cleaned)
```

```
→ Duplicated rows based on 'CustomerID':
```

| | CustomerID | Name | Age | JoinDate | MonthlyCharges | Churn |
|---|------------|------|-------|-----------|----------------|-------|
| 3 | C004 | BOBY | 29.0 | 6/1/2022 | 75.500 | No |
| 4 | C004 | Eve | 29.0 | 12/5/2024 | 75.500 | No |
| 8 | C008 | mary | 220.0 | 3/5/2023 | 441.878 | Yes |
| 9 | C008 | Bob | 30.0 | 3/5/2023 | 55.000 | No |

Dataset after removing duplicated rows:

| | CustomerID | Name | Age | JoinDate | MonthlyCharges | Churn |
|---|------------|-------|------------|------------|----------------|-------|
| 0 | C001 | John | 25.000000 | 12/1/2024 | 29.850 | No |
| 1 | C002 | Alice | 34.000000 | 11/15/2023 | 56.950 | Yes |
| 2 | C003 | BOB | 17.000000 | 6/1/2022 | 4000.000 | No |
| 3 | C004 | BOBY | 29.000000 | 6/1/2022 | 75.500 | No |
| 5 | C005 | eve | 120.000000 | 3/5/2023 | 45.990 | Yes |
| 6 | C006 | Steve | 55.444444 | 3/5/2023 | 60.000 | No |
| 7 | C007 | Ramu | 55.444444 | 1/1/2024 | 49.990 | No |
| 8 | C008 | mary | 220.000000 | 3/5/2023 | 441.878 | Yes |