

# Introduction to Data Wrangling in Python Using Pandas

Data wrangling is a critical step in the data analysis process, especially in psychological research. Before drawing insights from data, researchers must clean, structure, and transform raw data into a usable format. In this tutorial, we will use Python's Pandas library to perform common data wrangling tasks on a simulated psychology dataset.

We will cover:

1. Loading and inspecting data
2. Handling missing values
3. Transforming and categorizing data
4. Filtering and merging datasets

Let's get started!

```
In [1]: # Import necessary libraries
import pandas as pd
import numpy as np
```

```
In [2]: # Simulating a psychology survey dataset
np.random.seed(42)
data = {
    "Participant_ID": range(1, 11),
    "Age": [23, 45, 34, np.nan, 29, 40, 36, 28, 33, np.nan],
    "Gender": ["F", "M", "M", "F", "F", "M", "F", "M", "M", "F"],
    "Stress_Level": [3, 7, 5, 6, 2, 8, 4, 5, np.nan, 6],
    "Anxiety_Score": [np.nan, 12, 8, 15, 9, 14, 6, 7, 10, 11]
}
df = pd.DataFrame(data)
```

```
In [3]: # Display the first few rows
print("Original Dataset:")
print(df)
```

Original Dataset:

|   | Participant_ID | Age  | Gender | Stress_Level | Anxiety_Score |
|---|----------------|------|--------|--------------|---------------|
| 0 | 1              | 23.0 | F      | 3.0          | NaN           |
| 1 | 2              | 45.0 | M      | 7.0          | 12.0          |
| 2 | 3              | 34.0 | M      | 5.0          | 8.0           |
| 3 | 4              | NaN  | F      | 6.0          | 15.0          |
| 4 | 5              | 29.0 | F      | 2.0          | 9.0           |
| 5 | 6              | 40.0 | M      | 8.0          | 14.0          |
| 6 | 7              | 36.0 | F      | 4.0          | 6.0           |
| 7 | 8              | 28.0 | M      | 5.0          | 7.0           |
| 8 | 9              | 33.0 | M      | NaN          | 10.0          |
| 9 | 10             | NaN  | F      | 6.0          | 11.0          |

## Example 1: Handling Missing Values

```
In [4]: # Checking for missing values
def check_missing_values(df):
    print("\nMissing Values:")
    print(df.isnull().sum())

check_missing_values(df)
```

```
Missing Values:
Participant_ID    0
Age              2
Gender           0
Stress_Level     1
Anxiety_Score    1
dtype: int64
```

```
In [5]: # Filling missing values in Age with the median
# Filling missing values in Stress_Level and Anxiety_Score with the mean
df["Age"].fillna(df["Age"].median(), inplace=True)
df["Stress_Level"].fillna(df["Stress_Level"].mean(), inplace=True)
df["Anxiety_Score"].fillna(df["Anxiety_Score"].mean(), inplace=True)

print("\nDataset after handling missing values:")
print(df)
```

```
Dataset after handling missing values:
   Participant_ID  Age  Gender  Stress_Level  Anxiety_Score
0                1  23.0      F      3.000000      10.222222
1                2  45.0      M      7.000000      12.000000
2                3  34.0      M      5.000000       8.000000
3                4  33.5      F      6.000000     15.000000
4                5  29.0      F      2.000000       9.000000
5                6  40.0      M      8.000000     14.000000
6                7  36.0      F      4.000000       6.000000
7                8  28.0      M      5.000000       7.000000
8                9  33.0      M      5.111111     10.000000
9               10  33.5      F      6.000000     11.000000
```

## Example 2: Categorizing Psychological Data

```
In [6]: # Creating a categorical variable based on Anxiety Score
def categorize_anxiety(score):
    if score < 8:
        return "Low"
    elif 8 <= score < 12:
        return "Moderate"
    else:
        return "High"

df["Anxiety_Category"] = df["Anxiety_Score"].apply(categorize_anxiety)

print("\nDataset with Anxiety Categories:")
print(df)
```

Dataset with Anxiety Categories:

|   | Participant_ID | Age  | Gender | Stress_Level | Anxiety_Score | Anxiety_Category |
|---|----------------|------|--------|--------------|---------------|------------------|
| 0 | 1              | 23.0 | F      | 3.000000     | 10.222222     | Moderate         |
| 1 | 2              | 45.0 | M      | 7.000000     | 12.000000     | High             |
| 2 | 3              | 34.0 | M      | 5.000000     | 8.000000      | Moderate         |
| 3 | 4              | 33.5 | F      | 6.000000     | 15.000000     | High             |
| 4 | 5              | 29.0 | F      | 2.000000     | 9.000000      | Moderate         |
| 5 | 6              | 40.0 | M      | 8.000000     | 14.000000     | High             |
| 6 | 7              | 36.0 | F      | 4.000000     | 6.000000      | Low              |
| 7 | 8              | 28.0 | M      | 5.000000     | 7.000000      | Low              |
| 8 | 9              | 33.0 | M      | 5.111111     | 10.000000     | Moderate         |
| 9 | 10             | 33.5 | F      | 6.000000     | 11.000000     | Moderate         |

### Example 3: Filtering Data

```
In [7]: # Selecting participants with high stress levels (>=6)
high_stress_df = df[df["Stress_Level"] >= 6]

print("\nParticipants with High Stress Levels:")
print(high_stress_df)

### Example 4: Merging Data

# Creating a new dataset with additional participant information
demographics_data = {
    "Participant_ID": [1, 2, 3, 4, 5, 6, 7, 8, 9, 10],
    "Education_Level": ["Bachelor", "Master", "PhD", "Bachelor", "High School"]
}
demographics_df = pd.DataFrame(demographics_data)

# Merging datasets on Participant_ID
df_merged = pd.merge(df, demographics_df, on="Participant_ID")

print("\nMerged Dataset:")
print(df_merged)
```

Participants with High Stress Levels:

|   | Participant_ID | Age  | Gender | Stress_Level | Anxiety_Score | Anxiety_Category |
|---|----------------|------|--------|--------------|---------------|------------------|
| 1 | 2              | 45.0 | M      | 7.0          | 12.0          | High             |
| 3 | 4              | 33.5 | F      | 6.0          | 15.0          | High             |
| 5 | 6              | 40.0 | M      | 8.0          | 14.0          | High             |
| 9 | 10             | 33.5 | F      | 6.0          | 11.0          | Moderate         |

Merged Dataset:

|   | Participant_ID | Age  | Gender | Stress_Level | Anxiety_Score | Anxiety_Category |
|---|----------------|------|--------|--------------|---------------|------------------|
| \ |                |      |        |              |               |                  |
| 0 | 1              | 23.0 | F      | 3.000000     | 10.222222     | Moderate         |
| 1 | 2              | 45.0 | M      | 7.000000     | 12.000000     | High             |
| 2 | 3              | 34.0 | M      | 5.000000     | 8.000000      | Moderate         |
| 3 | 4              | 33.5 | F      | 6.000000     | 15.000000     | High             |
| 4 | 5              | 29.0 | F      | 2.000000     | 9.000000      | Moderate         |
| 5 | 6              | 40.0 | M      | 8.000000     | 14.000000     | High             |
| 6 | 7              | 36.0 | F      | 4.000000     | 6.000000      | Low              |
| 7 | 8              | 28.0 | M      | 5.000000     | 7.000000      | Low              |
| 8 | 9              | 33.0 | M      | 5.111111     | 10.000000     | Moderate         |
| 9 | 10             | 33.5 | F      | 6.000000     | 11.000000     | Moderate         |

|   | Education_Level |
|---|-----------------|
| 0 | Bachelor        |
| 1 | Master          |
| 2 | PhD             |
| 3 | Bachelor        |
| 4 | High School     |
| 5 | Master          |
| 6 | PhD             |
| 7 | Bachelor        |
| 8 | Master          |
| 9 | High School     |

## Exercise

Now it's your turn! Complete the following exercises to practice data wrangling on the psychology dataset.

1. **Filtering:** Create a new DataFrame that includes only participants with a "High" Anxiety Category. Display the resulting DataFrame.
2. **Data Transformation:** Add a new column called "Mental\_Wellbeing\_Score" by calculating the difference between Stress\_Level and Anxiety\_Score. Higher scores indicate better mental well-being. Display the updated dataset.

Write your code below each question and run the cells to test your solution!

### #1 Filtering

```
In [ ]: #Looking to see a suitable threshold for high anxiety
print(df.max(axis=0)['Anxiety_Score'])
```

```
print(df.min(axis=0)['Anxiety_Score'])
```

15.0

6.0

```
In [11]: highStress = df[df['Anxiety_Score'] >= 10]
highStress
```

```
Out[11]:
```

|   | Participant_ID | Age  | Gender | Stress_Level | Anxiety_Score | Anxiety_Category |
|---|----------------|------|--------|--------------|---------------|------------------|
| 0 | 1              | 23.0 | F      | 3.000000     | 10.222222     | Moderate         |
| 1 | 2              | 45.0 | M      | 7.000000     | 12.000000     | High             |
| 3 | 4              | 33.5 | F      | 6.000000     | 15.000000     | High             |
| 5 | 6              | 40.0 | M      | 8.000000     | 14.000000     | High             |
| 8 | 9              | 33.0 | M      | 5.111111     | 10.000000     | Moderate         |
| 9 | 10             | 33.5 | F      | 6.000000     | 11.000000     | Moderate         |

## #2 Data Transformation

Add a new column called "Mental\_Wellbeing\_Score" by calculating the difference between Stress\_Level and Anxiety\_Score. Higher scores indicate better mental well-being. Display the updated dataset.

```
In [12]: df['Mental_Wellbeing_Score'] = df['Stress_Level'] - df['Anxiety_Score']
df
```

```
Out[12]:
```

|   | Participant_ID | Age  | Gender | Stress_Level | Anxiety_Score | Anxiety_Category | Ment |
|---|----------------|------|--------|--------------|---------------|------------------|------|
| 0 | 1              | 23.0 | F      | 3.000000     | 10.222222     | Moderate         |      |
| 1 | 2              | 45.0 | M      | 7.000000     | 12.000000     | High             |      |
| 2 | 3              | 34.0 | M      | 5.000000     | 8.000000      | Moderate         |      |
| 3 | 4              | 33.5 | F      | 6.000000     | 15.000000     | High             |      |
| 4 | 5              | 29.0 | F      | 2.000000     | 9.000000      | Moderate         |      |
| 5 | 6              | 40.0 | M      | 8.000000     | 14.000000     | High             |      |
| 6 | 7              | 36.0 | F      | 4.000000     | 6.000000      | Low              |      |
| 7 | 8              | 28.0 | M      | 5.000000     | 7.000000      | Low              |      |
| 8 | 9              | 33.0 | M      | 5.111111     | 10.000000     | Moderate         |      |
| 9 | 10             | 33.5 | F      | 6.000000     | 11.000000     | Moderate         |      |