# pivoting or reshaping

Pivoting and reshaping are fundamental techniques in data wrangling that allow you to transform the structure of your data to better suit analysis, visualization, or reporting needs. These operations are especially common when working with tabular data in tools like Excel, pandas (Python), or R's tidyverse.

## **Pivoting**

**Pivoting** involves rotating your data to switch between "wide" and "long" formats, typically to make data easier to analyze or visualize.

#### 1. Wide Format:

- **Structure**: Each subject or entity has a single row with multiple columns representing different variables or time points.
- **Use Case**: Useful for presenting summarized data or when each row represents a unique entity with multiple attributes.

## Example:

Employee	Q1_Sales	Q2_Sales	Q3_Sales	Q4_Sales
Alice	100	150	200	250
Bob	80	120	160	200

## 2. Long Format:

- Structure: Each row represents a single observation or measurement, often with key-value pairs.
- **Use Case**: Ideal for time series analysis, grouping, and visualization where repeated measures are needed.

### Example:

Employee	Quarter	Sales
Alice	Q1	100
Alice	Q2	150
Alice	Q3	200
Alice	Q4	250
Bob	Q1	80
Bob	Q2	120
Bob	Q3	160
Bob	Q4	200

### **Pivot Operations:**

- Pivoting Wider: Converting long data to a wide format by spreading a key-value pair across multiple columns.
  - In pandas: df.pivot(index='Employee', columns='Quarter', values='Sales')
- Pivoting Longer: Converting wide data to a long format by collapsing multiple columns into key-value pairs.
  - In pandas: df.melt(id\_vars=['Employee'], var\_name='Quarter', value\_name='Sales')

## Reshaping

**Reshaping** is a broader term that encompasses various transformations to change the structure or layout of your data without necessarily focusing on the wide or long formats. This can include:

- 1. Stacking and Unstacking:
  - Stack: Converts columns into rows, often reducing <u>dimensionality</u>.
  - Unstack: Converts rows into columns, often increasing dimensionality.
  - In pandas: df.stack(), df.unstack()
- 2. Transposing:

- Description: Flips the DataFrame over its diagonal, turning rows into columns and vice versa.
- **Use Case**: Useful when the orientation of data needs to be changed for analysis or presentation.
- In pandas: df.transpose() or df.T

#### 3. Merging and Joining:

- Description: Combining multiple datasets based on a common key or index.
- Use Case: Integrating data from different sources or tables.
- In pandas: pd.merge(), df.join()

### 4. Aggregating and Grouping:

- Description: Summarizing data by grouping it based on one or more keys and applying aggregate functions.
- Use Case: Creating summary statistics or performing calculations on grouped data.
- In pandas: df.groupby()

## **Key Differences**

## • Scope:

- Pivoting is specifically about rotating data between wide and long formats.
- Reshaping includes pivoting but also covers a wider range of transformations like stacking, unstacking, transposing, merging, and more.

## • Purpose:

- Pivoting is often used to prepare data for specific types of analysis or visualization.
- Reshaping addresses broader structural changes needed for data cleaning, integration, or preparation.

## **Practical Example in pandas**

Suppose you have sales data in a long format and want to pivot it to a wide format:

```
import pandas as pd

# Long format DataFrame
data = {
    'Employee': ['Alice', 'Alice', 'Bob', 'Bob'],
    'Quarter': ['Q1', 'Q2', 'Q1', 'Q2'],
    'Sales': [100, 150, 80, 120]
}
df_long = pd.DataFrame(data)

# Pivot to wide format
df_wide = df_long.pivot(index='Employee', columns='Quarter',
values='Sales').reset_index()

print(df_wide)
```

### **Output**:

```
Quarter Employee Q1 Q2
0 Alice 100 150
1 Bob 80 120
```

## Conclusion

Pivoting and reshaping are essential for organizing data in a way that aligns with your analytical goals. Pivoting is a specific type of reshaping focused on transforming between wide and long formats, while reshaping encompasses a broader set of data transformation techniques. Mastery of these operations allows for greater flexibility and efficiency in data analysis workflows.