

DataFrame Creation and Basic Operations

1. Create a DataFrame from a list of dictionaries with keys `ID`, `Name`, and `Age`. Display the DataFrame.
2. Convert a NumPy array of shape (3, 4) into a Pandas DataFrame with column names `A`, `B`, `C`, and `D`.
3. Create a DataFrame from a CSV file named `sales_data.csv` and display the last 5 rows.
4. Construct a DataFrame from a list of lists where each sublist represents a row. Assign column names `Product`, `Quantity`, and `Price`.
5. Read an Excel file named `financials.xlsx` into a DataFrame. Display the first 3 rows.

Data Selection and Filtering

1. Select rows where the value in the `Salary` column is less than 60000.
2. Filter a DataFrame to include only rows where the `Department` column is 'Sales' and `Age` is greater than 30.
3. Select specific columns `Name` and `Age` from a DataFrame and display them.
4. Use `.loc` to select rows where the `Score` is between 50 and 75 inclusive.
5. Use `.iloc` to select the first 5 rows and columns 2 to 4 of a DataFrame.

Data Manipulation

1. Add a new column `Experience` to a DataFrame with values calculated as the difference between the current year and the `Year_Joined` column.
2. Remove rows with duplicate values in the `Email` column.
3. Create a new column `Total` by multiplying the `Quantity` and `Price` columns.
4. Rename the columns of a DataFrame from `A1`, `A2`, `A3` to `B1`, `B2`, `B3`.
5. Replace all occurrences of `NaN` in a DataFrame with the median value of their respective columns.

Data Aggregation

1. Group a DataFrame by `Department` and calculate the average `Salary` for each department.
2. Compute the total sales for each product category using a DataFrame with columns `Product` and `Sales`.
3. Create a pivot table to show the average `Sales` for each `Region` and `Quarter`.
4. Calculate the sum of `Quantity` for each `Product` using a DataFrame.
5. Find the maximum value in the `Price` column and identify the corresponding row.

Time Series Data

1. Create a DataFrame with a `Date` column and set it as the index. Display the data for the year 2023.
2. Resample a time series DataFrame to show monthly data and calculate the mean of `Sales` for each month.
3. Create a DataFrame with `Date` and `Sales` columns. Add a new column that shows the rolling average of `Sales` with a window of 7 days.
4. Convert a `Date` column in a DataFrame from string format to datetime format.

5. Plot a time series of `Sales` data from a `DataFrame` using Pandas built-in plotting functions.

Advanced Data Operations

1. Merge two `DataFrames` on a common column `Employee_ID` using an inner join.
2. Concatenate two `DataFrames` along the columns and reset the index.
3. Perform a left join between two `DataFrames` on `Customer_ID` and display the result.
4. Use the `query` method to select rows where `Sales` is greater than the mean `Sales` of the `DataFrame`.
5. Apply a custom function to each row in a `DataFrame` that calculates a new value based on existing columns.

Data Cleaning

1. Identify and fill missing values in a `DataFrame` with the forward fill method.
2. Use the `dropna` method to remove rows with any missing values in a `DataFrame`.
3. Replace specific values in a column with new values (e.g., replace 'Unknown' with 'N/A').
4. Detect outliers in the `Score` column using the IQR method.
5. Standardize the values in a numerical column to have a mean of 0 and a standard deviation of 1.

Data Visualization

1. Plot a histogram of the `Age` column from a `DataFrame` using Pandas plotting functions.
2. Create a scatter plot of `Quantity` vs `Price` from a `DataFrame`.
3. Generate a bar plot showing the average `Sales` for each `Region`.
4. Plot a box plot of `Salary` grouped by `Department`.
5. Create a line plot showing the trend of `Revenue` over time.

Miscellaneous

1. Convert a `DataFrame` into a dictionary with `orient='records'`.
2. Use the `sample` method to randomly select 5 rows from a `DataFrame`.
3. Save a `DataFrame` to an Excel file named `output.xlsx`.
4. Convert a `DataFrame` to a NumPy array and display its shape.
5. Check the data types of all columns in a `DataFrame` and convert a specific column to `float`.
6. Use the `melt` function to unpivot a `DataFrame` from wide to long format.
7. Create a new `DataFrame` by selecting rows from an existing `DataFrame` based on a condition in another `DataFrame`.
8. Get the count of unique values in the `Category` column of a `DataFrame`.
9. Calculate the correlation matrix of numerical columns in a `DataFrame`.
10. Create a `DataFrame` with `Name`, `Age`, and `Score` columns and save it as a JSON file.