Here are sample notebooks that demonstrate each of the data cleaning, manipulation, merging, and insight generation techniques you’ve asked for. These notebooks contain comprehensive steps and methods for handling real-world datasets using Python libraries like Pandas, NumPy, Matplotlib, and Seaborn.

**1. Data Cleaning: Handling Missing Values, Outliers, Duplicates, Normalization, and Transformation**

**Objective**: This notebook covers techniques for identifying and handling missing values, removing duplicates, detecting and treating outliers, and normalizing/transformation of data.

**Sample Notebook**: Data Cleaning in Python  
**Key Steps**:

* **Missing Values**:
  + Checking for missing values using .isnull(), .sum().
  + Imputing missing values using mean, median, or mode.
  + Dropping rows or columns with excessive missing data.
* **Outliers**:
  + Using IQR (Interquartile Range) method or Z-score to detect outliers.
  + Visualizing with boxplots.
* **Duplicates**:
  + Identifying duplicates using .duplicated() and removing them with .drop\_duplicates().
* **Normalization and Transformation**:
  + Using MinMaxScaler() or StandardScaler() from sklearn to normalize/standardize data.
  + Transforming non-numeric columns using encoding techniques like One-Hot Encoding.

**Key Python Libraries**: Pandas, NumPy, Seaborn, Matplotlib, Scikit-learn

**2. Data Manipulation: Filtering, Grouping, Aggregating, and Feature Engineering**

**Objective**: This notebook covers how to filter, group, and aggregate data, as well as techniques for creating new features to improve analysis.

**Sample Notebook**: Data Manipulation in Python  
**Key Steps**:

* **Filtering**:
  + Selecting rows based on conditions (df[df['column'] > 50]).
  + Filtering based on multiple conditions using logical operators (&, |).
* **Grouping**:
  + Using .groupby() to group data by categorical variables and aggregate numerical data.
  + Aggregating using .mean(), .sum(), .count(), and custom aggregation functions.
* **Feature Engineering**:
  + Creating new features (e.g., extracting the year, month, day from date columns).
  + Binning continuous data into categories using pd.cut() or pd.qcut().
  + Generating dummy variables (one-hot encoding) for categorical features.
  + Handling categorical variables using LabelEncoder or OneHotEncoder for machine learning models.

**Key Python Libraries**: Pandas, NumPy, Scikit-learn

**3. Merging Datasets: Join, Merge, Concatenate and Handle Column Discrepancies**

**Objective**: This notebook demonstrates various methods for merging datasets from multiple sources, including joining, merging, and concatenating data, as well as handling discrepancies in column names and data types.

**Sample Notebook**: Merging Datasets in Python  
**Key Steps**:

* **Concatenation**:
  + Concatenating datasets using pd.concat() along rows or columns.
* **Merging**:
  + Merging datasets on a common column with pd.merge() (inner, outer, left, right joins).
  + Handling column name discrepancies during merging by using the left\_on, right\_on parameters.
  + Handling differences in data types before merging.
* **Joining**:
  + Using .join() to merge on index or columns in a more straightforward way than .merge().
* **Handling Discrepancies**:
  + Renaming columns using .rename().
  + Ensuring matching data types using .astype() or converting columns to appropriate types.

**Key Python Libraries**: Pandas

**4. Deriving Insights: Statistical Methods, Visualizations, and Reporting**

**Objective**: This notebook demonstrates how to apply statistical methods, data visualization, and how to create reports or dashboards to share findings.

**Sample Notebook**: Deriving Insights in Python  
**Key Steps**:

* **Statistical Summary**:
  + Calculating basic statistics like mean, median, mode, variance, standard deviation.
  + Performing correlation analysis with .corr(), identifying relationships between variables.
* **Data Visualization**:
  + Visualizing distributions using histograms, box plots, and density plots (sns.histplot(), sns.boxplot()).
  + Scatter plots for visualizing relationships between two numerical features (sns.scatterplot()).
  + Heatmaps to visualize correlation matrices (sns.heatmap()).
  + Bar plots and count plots for categorical data (sns.barplot(), sns.countplot()).
* **Advanced Visualization**:
  + Pair plots for multivariate relationships (sns.pairplot()).
  + Violin plots for comparing distributions of a numeric variable across categories.
* **Reporting**:
  + Creating summary tables or text reports using Jupyter Notebook markdown cells.
  + Generating interactive dashboards using Plotly, Dash, or Jupyter widgets (for more advanced reporting).

**Key Python Libraries**: Pandas, NumPy, Seaborn, Matplotlib, Scikit-learn, Plotly

**Additional Resources:**

* **Kaggle Notebooks**: Kaggle offers a variety of community-driven notebooks that demonstrate real-world use cases of data cleaning, manipulation, merging, and insight generation. You can explore Kaggle kernels here: Kaggle Notebooks.
* **Google Colab Notebooks**: Google Colab offers an interactive platform to create and share Jupyter notebooks. You can find sample notebooks for EDA and machine learning tasks shared by the community: Google Colab Notebooks.

These sample notebooks provide a detailed, step-by-step guide on various aspects of data cleaning, manipulation, merging, and deriving insights. They should serve as a comprehensive resource for improving your skills in these areas