## 4. Change Data Types:

- Click the icon in the column header
- Or: select column → Transform tab → Any Column → Data Type
- Common types: Text, Number, Date, True/False

## **Data Cleaning Operations**

## 1. Replace Values:

- Select column → Transform tab → Replace Values
- Enter value to find and replacement value
- Example: Replace "N/A" with null

#### 2. Remove Errors:

- Select column → Home tab → Reduce Rows → Remove Errors
- Or: filter column to exclude errors

## 3. Remove Duplicates:

- Select column(s) → Home tab → Reduce Rows → Remove Duplicates
- Can be based on one or multiple columns

## 4. Fill Down/Up:

- Select column → Transform tab → Fill → Down/Up
- Fills empty cells with value from above/below

## **Structural Transformations**

#### 1. Pivot Column:

- Transform tab → Pivot Column
- Converts rows to columns
- Great for transforming survey data or cross-tabulations

## 2. Unpivot Columns:

- Select columns → Transform tab → Unpivot Columns
- Converts columns to rows
- Excellent for fixing "wide" data formats

## 3. **Group By**:

- Home tab → Group By
- Aggregates data by categories
- Similar to creating a pivot table in Excel

## 4. Add Custom Column:

- Add Column tab → Custom Column
- Create calculated fields using M formula language

# **Understanding Applied Steps**

Each transformation you perform creates a "step" in the Query Settings pane. This is crucial to understand because:

- Steps run in sequence from top to bottom
- You can select any step to see how your data looked at that point
- You can rename steps for clarity (right-click → Rename)
- You can delete steps you no longer need
- You can insert new steps between existing ones
- You can disable steps temporarily (right-click → Enable/Disable)

This step-by-step approach makes it easy to track changes and fix issues.

# **M Formula Language**

Power Query uses a language called "M" to define transformations. While you don't need to know M to use Power Query, understanding the basics helps:

- Each step generates M code automatically
- You can view the code in the formula bar
- You can edit the code directly for advanced transformations
- M is case-sensitive and uses "#" for comments

Simple M example for a custom column that concatenates first and last names:

```
= [FirstName] & " " & [LastName]
```

# Real-World Example

Imagine you receive a monthly sales report export with these issues:

- Column names have extra spaces
- Date field is in text format (MM/DD/YYYY)
- Product codes contain a mix of uppercase and lowercase
- Some rows have "NULL" as text instead of actual null values
- The data has multiple header rows

In Power Query, you can clean this by:

- 1. Remove extra header rows (Home → Reduce Rows → Remove Rows → Remove Top Rows)
- 2. Use first row as headers (Home → Transform → Use First Row as Headers)
- 3. Rename and trim column names (right-click → Rename)
- 4. Convert date column to Date type (click column data type icon → Date)
- 5. Standardize product codes (select column → Transform → Format → UPPERCASE)
- 6. Replace "NULL" text with actual null values (Transform → Replace Values)

Now every month when you get new data, simply refresh and all these cleaning steps happen automatically!

# **Step-by-Step: Basic Data Cleaning**

Let's clean a typical sales dataset:

# 1. Open Power Query Editor:

Home tab → Transform Data

#### 2. Fix Column Names:

- Right-click the column header → Rename
- Or: Transform tab → Transform → Format → Capitalize Each Word

# 3. Change Data Types:

- Set date columns to Date type
- Set numerical columns to appropriate number types (whole number, decimal, etc.)
- Set text columns to Text type

## 4. Remove Unnecessary Columns:

- Select columns you don't need → right-click → Remove
- Or: Home tab → Manage Columns → Remove Columns

## 5. Check and Apply Changes:

- Review the preview to make sure data looks correct
- Click "Close & Apply" in the Home tab to apply changes and return to Power BI

# **Practice Exercise: Power Query Basics**

**Objective**: Use Power Query Editor to clean and transform a dataset.

**Prerequisites**: Import a new file called "MessySalesData.xlsx" (your instructor will provide this file with intentional errors).

## Tasks:

## 1. Open Power Query Editor:

- After importing the file, click "Transform Data" in the Navigator
- Or if already imported, select Home → Transform Data

## 2. Clean Column Names:

- Rename columns to follow a consistent naming convention
- Remove any unnecessary columns

## 3. Fix Data Types:

- Set appropriate data types for each column
- Pay special attention to dates, numbers, and currency fields

#### 4. Clean Data Values:

- Find and replace any text versions of NULL or N/A with actual null values
- Fix any obvious errors in text fields (misspellings, inconsistent capitalization)

#### 5. **Document Your Process**:

- Rename each step in the Applied Steps pane with descriptive names
- Take notes on what issues you found and how you fixed them

**Bonus Challenge**: Create a custom column that combines information from two or more existing columns (like a full name from first and last names).

# **Removing Errors, Filtering, Splitting Columns**

# **Advanced Data Cleaning Techniques**

While basic transformations help with simple issues, real-world data often requires more advanced cleaning techniques. In this module, we'll focus on three powerful methods: handling errors, filtering data, and splitting columns.

# **Identifying and Handling Errors**

Errors in Power Query appear as red text with the word "Error" in cells. They occur when:

- Data can't be converted to the assigned type
- Calculations encounter problems (like division by zero)
- Functions receive invalid parameters

## **Ways to Handle Errors**

#### 1. Remove Errors:

- Select the column with errors
- Home tab → Reduce Rows → Remove Errors
- Best when errors represent invalid data you want to exclude

## 2. Replace Errors:

- Select the column with errors
- Transform tab → Replace Errors
- Enter a replacement value (like 0, null, or "Unknown")
- Useful when you need to keep the rows but fix the problematic values

#### 3. Convert to Null:

- Transform tab → Replace Errors → (leave replacement value empty)
- Good middle-ground approach

## 4. Error Inspection:

- Filter column to show only errors
- Try to understand the pattern causing errors
- Fix the root cause rather than just the symptoms

# **Filtering Data**

Filtering lets you focus on specific portions of your data or remove unwanted records.

## **Filter Types in Power Query**

#### 1. Basic Filters:

- Click the drop-down arrow in column header
- Select/deselect values to include/exclude
- Great for categorical data with few unique values

#### 2. Text Filters:

- Select "Text Filters" from dropdown
- Options include:
  - Begins With/Ends With
  - Contains/Does Not Contain
  - Equals/Does Not Equal
  - Custom Filter (combine multiple conditions)

## 3. Number Filters:

- Select "Number Filters" from dropdown
- Options include:
  - Equals/Does Not Equal
  - Greater Than/Less Than
  - Between

Top N/Bottom N

#### 4. Date Filters:

- Select "Date Filters" from dropdown
- Options include:
  - Before/After/Between
  - In the Past/In the Next
  - Date range options (This Month, Last Year, etc.)

## 5. Advanced Filter:

- Add Column tab → Custom Column
- Create a formula that returns true/false
- Then filter to show only "true" values

#### **Filter Locations**

- Column Header Filters: Apply to individual columns
- **Keep/Remove Rows**: Home tab → Reduce Rows
- Filter Rows: Select rows → right-click → Keep/Remove Rows

# **Splitting Columns**

Splitting columns is essential when dealing with data that combines multiple pieces of information in a single field.

## **Common Split Scenarios**

- Full names into first and last names
- Addresses into street, city, state, and zip
- Product codes that combine category and number
- **Date-times** that need separate date and time components

## **Splitting Methods**

## 1. Split by Delimiter:

- Select column → Transform tab → Split Column → By Delimiter
- Choose the character that separates values:
  - Common delimiters: comma, space, tab, semicolon
  - Custom delimiter option for other characters
- Choose split options:
  - Left-most delimiter (splits at first occurrence)

- Right-most delimiter (splits at last occurrence)
- Each occurrence (creates multiple columns)

## 2. Split by Number of Characters:

- Select column → Transform tab → Split Column → By Number of Characters
- Specify how many characters per column
- Useful for fixed-width data formats

## 3. Split by Positions:

- Select column → Transform tab → Split Column → By Positions
- Specify exact positions where splits should occur
- Good for complex or irregular patterns

## 4. Advanced Splits using Custom Columns:

- Add Column tab → Custom Column
- Use Text.Start, Text.Middle, Text.End functions
- Example: (Text.Start([FullName], 1)) for first initial

# **Handling Split Results**

After splitting columns:

- Rename the resulting columns with meaningful names
- Set appropriate data types
- Consider removing the original column if no longer needed

## **Real-World Example**

Imagine you have customer data with these issues:

#### 1. Combined Name Column:

- Original: "Smith, John A."
- Needed: Last Name = "Smith", First Name = "John", Middle Initial = "A"

## 2. Invalid Phone Numbers:

- Some entries have letters or are too short
- These appear as errors when converted to phone number format

## 3. Need Only US Customers:

Data includes international customers you want to exclude

Using Power Query, you could:

## 1. Split the Name:

- Split by delimiter (comma) to separate last name
- Split the remainder by the last space to separate middle initial
- Rename columns appropriately

## 2. Handle Phone Errors:

- Convert column to text type
- Replace errors with "Invalid"
- Or filter to remove rows with invalid phone numbers

#### 3. Filter for US Customers:

- Apply a filter on Country column to show only "USA" or "US"
- Or use a contains filter on the Postal Code to match US format

# **Step-by-Step: Advanced Data Cleaning**

Let's walk through cleaning a customer dataset:

#### 1. Handle Name Column:

- Select "FullName" column
- Transform tab → Split Column → By Delimiter
- Choose space as delimiter and split at the first occurrence
- Rename resulting columns to "FirstName" and "LastName"

## 2. Clean Phone Numbers:

- Select "Phone" column
- Transform tab → Replace Values
- Replace "-", "(", ")" and spaces with nothing to standardize format
- Filter to show only values with 10+ characters (valid phone length)

## 3. Extract Domain from Email:

- Select "Email" column
- Transform tab → Split Column → By Delimiter
- Choose "@" as delimiter
- Rename second column to "EmailDomain"
- You can now analyze which email providers are most common

## 4. Fix Date Format Issues:

- Select "JoinDate" column
- If errors appear when converting to Date type
- Transform tab → Replace Errors → null

• Or create formula to handle different date formats

# **Practice Exercise: Advanced Data Cleaning**

**Objective**: Apply advanced cleaning techniques to prepare data for analysis.

Prerequisites: Import a file called "CustomerData.xlsx" (your instructor will provide this).

#### Tasks:

#### 1. Handle Error Values:

- Identify columns with errors
- For each error column, decide whether to:
  - Remove rows with errors
  - Replace errors with null or another value
  - Fix the root cause of the errors

## 2. Apply Filters:

- Filter the data to include only:
  - Active customers (Status = "Active")
  - Customers who joined within the last 2 years
  - Customers with complete contact information

## 3. Split Complex Columns:

- Find the "Address" column which contains full address in one field
- Split it into components (Street, City, State, ZIP)
- Set appropriate data types for each new column

## 4. Create Clean Customer Dataset:

- Ensure all columns have appropriate names and data types
- Remove any unnecessary columns
- Document your cleaning process

**Bonus Challenge**: Create a custom column that categorizes customers based on their email domain (e.g., "Gmail User", "Business User", "Other").

# **Combining Data (Merge & Append)**

## The Power of Data Combination

In real-world scenarios, the data you need for analysis often lives in multiple files or tables. Power Query excels at combining data through two key operations:

1. **Merge**: Joins tables horizontally by matching related rows (like SQL JOIN)

2. **Append**: Combines tables vertically by stacking rows (like SQL UNION)

Understanding when and how to use these operations is essential for comprehensive data analysis.

# **Merging Tables**

Merging combines two tables based on a common field or key, bringing columns from the second table into the first.

## When to Use Merge

- Combine product details with sales data
- Add customer information to order history
- Link employee data to performance metrics
- Connect lookup tables with transaction data

## **Merge Types Explained**

Power Query offers several merge types that determine which rows appear in the result:

## 1. Left Outer (most common):

- Keeps all rows from first table
- Adds matching data from second table
- Creates null values where no match exists.
- Use when: You need all records from primary table regardless of matches

## 2. Right Outer:

- Keeps all rows from second table
- Adds matching data from first table
- Creates null values where no match exists
- Use when: You need all records from secondary table regardless of matches

## 3. Full Outer:

- Keeps all rows from both tables
- Creates null values where no match exists
- Use when: You need all records from both tables

#### 4. Inner:

- Keeps only rows that have matches in both tables
- Use when: You only want complete records with data from both tables

#### 5. Left Anti:

Keeps only rows from first table that don't match second table

• Use when: Finding exceptions or missing relationships

## 6. Right Anti:

- Keeps only rows from second table that don't match first table
- Use when: Finding exceptions or missing relationships

## **Step-by-Step Merge Process**

Let's merge Sales data with Product details:

#### 1. Start the Merge:

- In Power Query Editor, select the table to merge into (e.g., Sales)
- Home tab → Combine → Merge Queries
- Select "Merge Queries" for in-place merge or "Merge Queries as New" to create a new query

#### 2. Select Tables and Join Columns:

- Your first table is pre-selected
- Choose the second table from dropdown (e.g., Products)
- Click column(s) that match in both tables (e.g., ProductID)
- Select join type (e.g., Left Outer)
- Click OK

## 3. Work with Merged Data:

- A new column appears containing related rows from second table
- Click the expand button (double arrow icon) in column header
- Select which columns to bring in from second table
- Uncheck "Use original column name as prefix" for cleaner naming
- Click OK

## 4. Finalize the Merge:

- Rename columns if needed
- Set proper data types
- Apply any additional transformations

## **Common Merge Issues and Solutions**

## 1. Data Type Mismatches:

- Issue: Join columns have different data types (e.g., text vs. number)
- Solution: Make sure join columns have matching types before merging

## 2. Duplicate Keys:

• Issue: Multiple matching rows create duplicate or expanded results

• Solution: Remove duplicates from key columns or use aggregation before merging

## 3. Missing Matches:

- Issue: Expected matches are missing
- Solution: Check for spaces, case differences, or formatting issues in join columns

## 4. Performance:

- Issue: Large merges are slow
- Solution: Filter tables before merging to include only necessary data

# **Appending Tables**

Appending combines tables with similar structures by adding rows from one table to another.

## When to Use Append

- Combine monthly data files into a single table
- Merge historical and current data with the same structure
- Integrate data from different sources with compatible columns
- Combine regional data into a company-wide view

## **Append Process**

Let's append quarterly sales tables:

## 1. Start the Append:

- In Power Query Editor, select one of the tables to append
- Home tab → Combine → Append Queries
- Select "Append Queries" for in-place append or "Append Queries as New" to create a new query

## 2. Select Tables to Append:

- Choose append mode:
  - "Two tables" to append one table to another
  - "Three or more tables" to append multiple tables at once
- Select tables to append from the dropdown(s)
- Click OK

## 3. Handle Column Differences:

- If tables have different columns:
  - Matching columns will align automatically
  - Non-matching columns will get null values where needed
- Consider adding a source identifier (see next section)

## 4. Finalize the Append:

- Set proper data types
- Remove any duplicate rows if needed
- Apply additional transformations

## **Tracking Data Sources**

When appending data, it's often useful to track which source each row came from:

# 1. Before Appending:

- For each source table, add a custom column
- Add Column tab → Custom Column
- Name it "Source" or "DataSource"
- Enter a static value that identifies the table (e.g., "Q1 Sales", "2022 Data")

## 2. After Appending:

- All source identifiers will be preserved in the combined table
- You can filter or analyze by source as needed

# **Real-World Example**

Imagine you're a business analyst who receives:

- 1. Monthly sales files (12 separate Excel files with identical structure)
- 2. A product catalog with details like cost, category, and supplier
- 3. A store locations file with address and region information

To create a comprehensive sales analysis:

- 1. **Append** all 12 monthly sales files into a single sales table
  - Add a "Month" column to each before appending
- 2. Merge the combined sales with product catalog
  - Left outer join on ProductID
  - Bring in Category, Cost, Supplier
- 3. **Merge** again with store locations
  - Left outer join on StoreID
  - Bring in Region, City, StoreSize

Now you have a complete dataset for analysis with:

- Sales transactions for the entire year
- Product details for each sale

• Store information for regional analysis

# **Step-by-Step: Combining Multiple Data Sources**

Let's walk through combining regional sales data:

## 1. Append Regional Sales Tables:

- Home tab → Combine → Append Queries as New
- Select "Three or more tables"
- Add all regional sales tables (North, South, East, West)
- Name the new query "Combined Sales"

#### 2. Add Source Identification:

- Go back to each regional query
- Add Column tab → Custom Column
- Name: "Region"
- Formula: "North" (adjust for each table)
- Go back to Combined Sales and refresh

## 3. Merge with Product Information:

- With Combined Sales selected
- Home tab → Combine → Merge Queries
- Select Products table
- Join on ProductID columns
- Choose Left Outer join
- Expand to bring in needed product columns

# **Practice Exercise: Combining Data**

**Objective**: Combine multiple related datasets to create a comprehensive analysis table.

**Prerequisites**: Import the following files (your instructor will provide these):

- "SalesQ1.xlsx", "SalesQ2.xlsx", "SalesQ3.xlsx", "SalesQ4.xlsx"
- "ProductCatalog.xlsx"
- "CustomerList.xlsx"

#### Tasks:

## 1. Append Quarterly Sales Data:

- Create a unified sales table from all four quarterly files
- Add a quarter identifier to each table before appending

• Ensure data types are consistent across all tables

## 2. Merge with Product Catalog:

- Combine the sales data with product details
- Use appropriate join type (consider what should happen with unknown products)
- Bring in relevant product fields (description, category, cost)

## 3. Merge with Customer Information:

- Add customer details to your combined table
- Use appropriate join type
- Bring in relevant customer fields (name, segment, region)

## 4. Create Analysis-Ready Table:

- Ensure all columns have descriptive names
- Set appropriate data types
- Create any additional calculated columns needed for analysis
- Document the data combination process

**Bonus Challenge**: Calculate how many sales transactions had missing product information or customer details, and determine if there's a pattern to these gaps.

# **Relationships and Model View**

# **Understanding Data Relationships**

In Power BI, relationships connect tables together, enabling cross-filtering between visuals based on different tables. A well-structured data model with proper relationships is the foundation of an effective Power BI report.

# What is a Relationship?

A relationship in Power BI is a connection between two tables based on related columns. This tells Power BI how data in one table corresponds to data in another table.

## For example:

- A Products table might have a ProductID column
- A Sales table also has a ProductID column
- A relationship connects these tables through their ProductID columns
- This lets you analyze Sales data by Product attributes

# The Importance of Relationships

Without relationships:

- Tables exist in isolation
- You can't create visuals combining fields from different tables
- Filters from one table won't affect visuals based on another table

With proper relationships:

- Cross-filtering works automatically
- You can use fields from multiple tables in a single visual
- Your reports become truly interactive

# **Understanding Cardinality**

Cardinality describes how records in one table relate to records in another. Power BI supports several relationship types:

## 1. **One-to-Many** (most common):

- One record in the first table can relate to many in the second
- Example: One Product can have Many Sales
- The "one" side contains unique values (like a primary key)
- The "many" side can contain duplicate values (like a foreign key)

## 2. Many-to-One:

- Same as one-to-many but with table order reversed
- Direction matters for filter flow in Power BI

#### 3. One-to-One:

- One record in the first table relates to exactly one in the second
- Both columns contain unique values
- Example: EmployeeID in both Employee and EmployeeDetails tables

## 4. Many-to-Many:

- Multiple records in first table relate to multiple in second
- Requires special handling in Power BI
- Often implemented using an intermediate table

#### **Model View in Power Bl**

Model view provides a diagram of your data model, showing tables and their relationships.

## **Accessing Model View**

- Click the "Model view" icon in the left navigation bar (third icon)
- Or go to View tab → Model view

#### **Model View Interface Elements**

- **Tables**: Shown as boxes containing field lists
- **Relationships**: Shown as lines connecting tables
- Cardinality: Indicated by 1 or \* (asterisk) symbols on relationship lines
- **Filter direction**: Shown by arrows on relationship lines

# **Creating and Managing Relationships**

## **Automatic Relationship Detection**

When you load multiple tables, Power BI tries to detect relationships automatically based on:

- Matching column names
- Compatible data types
- Uniqueness of values

However, auto-detection isn't perfect and often needs adjustment.

## **Creating Relationships Manually**

## 1. In Model View:

- Click and drag from a column in one table to the related column in another table
- Configure relationship settings in the dialog box

## 2. Using Manage Relationships Dialog:

- Home tab → Relationships → Manage Relationships
- Click "New" to create a relationship
- Select tables and columns to connect
- Configure cardinality and other settings

# **Relationship Settings**

When creating or editing a relationship, you'll see several options:

## 1. Cardinality:

• Select the appropriate type (usually "Many to One")

## 2. Cross-filter direction:

- Single: Filters flow from the "one" side to the "many" side only (default)
- Both: Filters flow in both directions
- Both is useful in specific scenarios but can cause performance issues

#### 3. Make this relationship active:

- Only one active relationship can exist between two tables
- Inactive relationships can be used in specific DAX calculations

# **Common Relationship Scenarios**

## **Fact-to-Dimension Relationships**

The most common pattern in analytical models:

- Fact table: Contains business events or measurements (sales, orders)
- **Dimension tables**: Contain descriptive attributes (products, customers)
- **Relationship**: Many-to-one from fact to dimension tables
- Example: Sales (fact) related to Products, Customers, and Dates (dimensions)

## **Date Table Relationships**

Almost every model needs a date table:

- Date tables contain calendar attributes (year, month, quarter)
- Create relationships between transaction dates and the date table
- This enables time intelligence functions and date-based filtering

## **Role-Playing Dimensions**

When the same dimension serves multiple roles:

- Example: A Date table related to both OrderDate and ShipDate in an Orders table
- Solution: Create multiple relationships but only one active
- Use inactive relationships in specific measures with USERELATIONSHIP function

# **Troubleshooting Relationship Issues**

## **Common Problems and Solutions**

## 1. "The relationship couldn't be created":

- Check for data type mismatches between columns
- Verify the "one" side contains unique values
- Look for empty or null values in key columns

## 2. Ambiguous paths between tables:

- Problem: Multiple paths between tables cause filter confusion
- Solution: Make some relationships inactive or restructure your model

## 3. Circular dependencies:

Problem: Filter paths create a circular loop

• Solution: Break the circle by making one relationship inactive

## 4. Missing relationships:

- Visuals combining tables don't show the right data
- Create necessary relationships between tables

# **Real-World Example**

Imagine you're analyzing a retail business with these tables:

- 1. Sales: Contains transaction details (ProductID, CustomerID, StoreID, Date, Quantity, Revenue)
- 2. **Products**: Contains product information (ProductID, Name, Category, Cost)
- 3. Customers: Contains customer data (CustomerlD, Name, City, Segment)
- 4. Stores: Contains store information (StoreID, Location, Size)
- 5. Calendar: Contains date attributes (Date, Day, Month, Quarter, Year)

Your relationship model would have:

- Sales to Products: Many-to-One on ProductID
- Sales to Customers: Many-to-One on CustomerID
- Sales to Stores: Many-to-One on StoreID
- Sales to Calendar: Many-to-One on Date

#### With this model:

- You can analyze sales by product category
- You can filter all visuals by customer segment
- You can compare performance across store locations
- You can view trends over time periods

# Step-by-Step: Building a Relational Model

Let's create a basic sales analysis model:

#### 1. Examine Your Tables:

- Switch to Data view to understand each table's structure
- Identify primary keys (unique identifiers) in each table
- Identify foreign keys (references to other tables)

#### 2. Switch to Model View:

Click the Model view icon on the left navigation bar

## 3. Create Basic Relationships:

Click and drag from Sales[ProductID] to Products[ProductID]

- Click and drag from Sales[CustomerID] to Customers[CustomerID]
- Configure each as Many-to-One with single filter direction

## 4. Check Relationship Status:

- Verify relationships appear with correct cardinality symbols
- Test relationships by creating a visual with fields from different tables

## 5. Add Date Relationship:

- Create or import a date table
- Create relationship between Sales[Date] and Calendar[Date]

# **Practice Exercise: Building a Data Model**

**Objective**: Create an effective data model with proper relationships between tables.

**Prerequisites**: Import the following tables (your instructor will provide these):

- "Orders" (contains OrderID, CustomerID, ProductID, Date, Quantity, Amount)
- "Customers" (contains CustomerID, Name, City, Country)
- "Products" (contains ProductID, ProductName, Category, UnitPrice)
- "Calendar" (contains Date and date attributes like Year, Month, Quarter)

#### Tasks:

#### 1. Examine Table Structure:

- Review each table in Data view
- Identify primary keys and foreign keys
- Check data types of key columns

## 2. Create Relationships:

- Switch to Model view
- Create appropriate relationships between:
  - Orders and Customers
  - Orders and Products
  - Orders and Calendar
- Set proper cardinality for each relationship

#### 3. Test Your Model:

- Switch to Report view
- Create a visual showing Orders Amount by Product Category
- Create another visual showing Orders by Customer Country

Verify that clicking items in one visual filters the other

#### 4. Document Your Model:

- Take a screenshot of your model diagram
- List each relationship you created and its purpose
- Note any challenges you encountered

**Bonus Challenge**: Add a calculated column in the Orders table that calculates profit by subtracting product cost from order amount. Then create a visual showing profit by product category.

# Star Schema & Lookup Tables

# **Understanding Data Modeling Concepts**

A well-designed data model is the foundation of an effective Power BI solution. The Star Schema is one of the most popular and efficient data modeling approaches for analytics and reporting.

#### What is a Star Schema?

A Star Schema is a data modeling technique that organizes data into:

- One central **Fact Table** containing business measurements or events
- Multiple Dimension Tables containing descriptive attributes
- Relationships that connect dimensions to the fact table

The name "Star Schema" comes from the shape it creates in a diagram, with the fact table in the center and dimension tables arranged around it like points of a star.

#### **Benefits of Star Schema**

- Simplified Queries: Clear paths between tables
- Improved Performance: Optimized for analytical operations
- Intuitive Structure: Easy for users to understand
- **Enhanced Reporting Flexibility**: Analyze by any dimension
- Consistent Results: Single version of truth

# **Fact Tables Explained**

Fact tables record business events or measurements that you want to analyze.

#### **Characteristics of Fact Tables**

- **Contain measures** (numeric values to analyze)
- Typically have **many rows** (thousands to millions)
- Include **foreign keys** to dimension tables

- Are often date/time-stamped
- Are usually **narrow** (few columns but many rows)

## **Examples of Fact Tables**

- Sales transactions
- Website visits
- Manufacturing output
- Survey responses
- Financial transactions

#### **Common Measures in Fact Tables**

- Quantities (units sold, number of visits)
- Monetary values (revenue, cost, profit)
- Duration (call time, processing time)
- Rates and ratios (conversion rate, error rate)

# **Dimension Tables Explained**

Dimension tables provide context for the measurements in fact tables.

## **Characteristics of Dimension Tables**

- Contain descriptive attributes
- Typically have fewer rows than fact tables
- Include a **primary key** (unique identifier)
- Are usually **wider** (more columns but fewer rows)
- Provide hierarchical relationships within dimensions

# **Examples of Dimension Tables**

- Products (with attributes like category, brand, size)
- Customers (with attributes like name, segment, location)
- Dates (with attributes like year, quarter, month, day)
- Locations (with attributes like city, region, country)
- Employees (with attributes like department, job title, hire date)

# **Creating Lookup Tables**

Lookup tables are a type of dimension table that normalize your data model by removing redundant information.

## When to Create Lookup Tables

- When a column contains repeating values
- When descriptive attributes belong to a distinct entity
- When you need to standardize categorization
- When you want to improve model efficiency

## **Process for Creating Lookup Tables**

# 1. Identify Candidates:

- Columns with repeating values (like ProductCategory)
- Columns with multiple related attributes (like ProductCategory, ProductSubcategory)

## 2. Extract to a New Table:

- Use Power Query's "Remove Duplicates" to create distinct values
- Add a key column if one doesn't exist
- Add descriptive columns as needed

## 3. Create Relationships:

- Connect the lookup table to your fact or main table
- Ensure proper cardinality (usually one-to-many)

## **Date Dimensions**

A date dimension