1. / What is a Primary Key in Power BI?

- In Power BI, a **primary key** means a column (or set of columns) that uniquely identifies each row in a table.
- Unlike SQL databases, Power BI does not **enforce** primary key constraints.
- However, primary keys are essential when you build **relationships** between tables.

Example:

Customers table

 \circ CustomerID \rightarrow Primary Key (each customer has a unique ID).

Orders table

 \circ CustID \rightarrow Foreign Key (indicates which customer placed the order).

In Power BI's Model view:

Key Points:

- 1. Power BI doesn't require you to define a primary key like SQL.
- 2. But **choosing a unique identifier** for each table is critical.
- 3. Relationships in Power BI usually follow **one-to-many** (1 customer → many orders) or sometimes **many-to-many**.

2. In **Power BI**, there are two main types of table relationships:

1. One-to-Many (1:*):

- o The most common relationship.
- o Example: One Customer can have many Orders.
- Usually, the primary key is on the "one" side, and the foreign key is on the "many" side.

2. **Many-to-Many** (:):

- o Both tables can have multiple matching rows.
- Example: A Student can enroll in many Courses, and each Course can have many Students.

o In Power BI, this is handled using a **bridge table** or by enabling a direct many-to-many relationship.

There is also **One-to-One** (1:1), but it's less common and usually just means two tables share the same unique key (can often be combined into one table).

3. • Method 1: Using the Model View (most common)

- 1. Go to **Model View** (left-hand panel \rightarrow the third icon with relationships).
- 2. Drag a field (usually a **primary key**) from one table onto the corresponding field (usually a **foreign key**) in the other table.
- 3. The **Manage Relationships** dialog will open automatically.
- 4. Choose the **Cardinality** (One-to-Many, Many-to-Many, or One-to-One).
- 5. Choose **Cross filter direction** (Single or Both).
- 6. Click OK.

Method 2: Using the Manage Relationships dialog

- 1. On the **Home** tab \rightarrow click **Manage Relationships**.
- 2. Click New.
- 3. Select the two tables and the matching columns.
- 4. Configure Cardinality and Cross filter direction.
- 5. Click **OK** to create the relationship.

Example:

- Customers[CustomerID] (Primary Key) → Orders[CustomerID] (Foreign Key).
- Relationship type: One-to-Many (1:*).

4. A star schema is a type of database schema (data model) commonly used in Power BI, data warehousing, and analytics. It organizes data in a way that makes reporting and querying faster and easier to understand.

- Structure of a Star Schema
 - At the center, there is a Fact Table:
 - Stores quantitative data (numbers you want to measure, e.g., Sales Amount, Quantity, Revenue).
 - Contains foreign keys that link to dimension tables.
 - Around it, there are multiple Dimension Tables (the "points of the star"):
 - Contain descriptive attributes (e.g., Customer Name, Product Category, Order Date).
 - Provide context for the facts.
- Example (Sales Data Model in Power BI)
 - Fact Table: Sales
 - Columns: OrderID, CustomerID, ProductID, DateKey, SalesAmount, Quantity.
 - Dimension Tables:
 - o Customers → CustomerID, Name, Country
 - o Products → ProductID, Name, Category
 - o Dates → DateKey, Year, Month, Day
 - \circ Stores \rightarrow StoreID, City, Region
- **b** Visually, the fact table is in the middle, and dimensions branch out like a star.
- Benefits of Star Schema in Power BI
 - Simple and intuitive model for users.
 - Efficient performance (optimized for queries).
 - Works well with Power BI's DAX and relationships.
 - Reduces data redundancy compared to flat tables.

5. In a sales dataset, the Fact Table is typically the Sales (or Orders) table.

Why?

- It stores the measurable, numeric values you want to analyze.
- Examples of columns in the Fact Table (Sales):
 - o OrderID
 - \circ CustomerID (foreign key \rightarrow Customers table)
 - \circ ProductID (foreign key \rightarrow Products table)
 - DateKey (foreign key → Date table)
 - o Quantity
 - SalesAmount
 - Discount
 - o Revenue

Dimension tables in the same model:

- Customers (CustomerID, Name, Country)
- Products (ProductID, Category, Brand)
- **Dates** (DateKey, Year, Month, Day)
- Stores (StoreID, Location, Region)
- For So, in short: The Sales (or Orders) table is the Fact Table in a sales dataset.
- 6. In a Sales.csv table (fact table), ProductID is considered a foreign key because:

Reason:

- A foreign key is a column that links to a primary key in another table (usually a dimension table).
- In **Sales.csv**, ProductID does not describe the product itself (name, category, price).
- Instead, it just stores an **ID** that points to the **Products table**, where the full product information lives.

ii Example:

Products.csv (Dimension table)

ProductID ProductName Category Price

101 Laptop Electronics 800

Mouse Accessories 20

Sales.csv (Fact table)

SaleID CustomerID ProductID Quantity SalesAmount

1 201 101 2 1600

2 202 102 3 60

Here:

- ProductID in Sales.csv → Foreign Key
- ProductID in **Products.csv** → **Primary Key**

This allows **Sales** to link with **Products**, so you can answer questions like:

***** "How many Laptops did we sell last month?"

★ In Power BI, this becomes a One-to-Many relationship:

- **Products[ProductID]** (One side, primary key)
- Sales[ProductID] (Many side, foreign key)

- In Sales.csv, ProductID might be Whole Number (Int64).
- In **Products.csv**, ProductID might be loaded as **Text** (e.g., "101", "102").
- Power BI only allows relationships if **both columns are the same data type**.

✓ How to Fix It

Option 1: In Power Query

- 1. Go to **Power Query Editor** (Transform Data).
- 2. Open Sales table \rightarrow Right-click on ProductID column \rightarrow Change Type \rightarrow Whole Number.
- 3. Open **Products** table \rightarrow Do the same, set ProductID to **Whole Number**.
- 4. Click Close & Apply.

- 5. Recreate the relationship:
 - Products[ProductID] (Primary Key) → Sales[ProductID] (Foreign Key).

Option 2: In Model View

- You can't directly change types here, but you'll see a warning (data type mismatch).
- Always fix it in **Power Query**.
- 8. A **star schema** is the recommended modeling approach in **Power BI** (and most analytics tools) because it makes queries simpler and **improves performance**. Here's why:

Why Star Schema Improves Performance

1. Clear Separation of Fact and Dimension Tables

- Fact table → numeric, transactional data (Sales, Orders, Revenue).
- **Dimension tables** → descriptive attributes (Products, Customers, Dates).
- This separation reduces duplication and keeps the fact table lean.

Example: Instead of storing ProductName, Category, Price in every Sales row, you just store ProductID.

2. Smaller Fact Tables = Faster Queries

- Fact tables contain millions of rows, so keeping them only with numeric values and foreign keys makes them much smaller.
- Dimension tables are smaller and only store descriptive data once.
- Power BI's VertiPaq engine compresses these better, making aggregations very fast.

3. Efficient Relationships

- In a star schema, relationships are **simple One-to-Many** (Dimension → Fact).
- Power BI handles 1: joins* very efficiently.

• If you use a flat table (all data in one), or snowflake schema (too many joins), queries become slower.

4. Optimized for DAX Calculations

• Measures like SUM(Sales[Amount]), AVERAGE(Sales[Quantity]), or CALCULATE(..., Products[Category]) run faster, because Power BI only scans the **fact table** and uses dimension filters.

5. Better Compression (Storage Engine Benefit)

- Repeating text values (like "Laptop", "Mouse", "Laptop") in a flat table takes more memory.
- In a star schema, text lives once in the **dimension table**, and the fact table only stores small integer keys.
- This makes the in-memory model much smaller and faster.

Simple Example:

Flat table (slow):

| OrderID | CustomerName | ProductName | Category | Amount |

Star schema (fast):

- Sales (Fact) → OrderID, CustomerID, ProductID, Amount
- Customers (Dimension) → CustomerID, Name, Country
- **Products (Dimension)** → ProductID, Name, Category

☑ In short: Star schema improves performance because it reduces data duplication, keeps fact tables small, enables efficient compression, and makes relationships simpler (1:*).

11. A **circular relationship** (relationship loop) happens when your tables are connected in such a way that Power BI cannot decide how to filter them (it creates ambiguity).

Example of Circular Relationship

Suppose you have:

- Sales → Customers → Regions
- Sales \rightarrow Products \rightarrow Regions

Now Regions is linked twice, and Power BI forms a **loop**.

✓ How to Resolve Circular Relationships

1. Remove redundant relationships

- Keep only **one clear path** between fact and dimensions.
- If two paths exist, decide which one makes more business sense.
- For example, link Customers \rightarrow Regions, but not Products \rightarrow Regions.

2. Use a Bridge Table

- Create an intermediate table to break the loop.
- Example: instead of linking Regions directly to both Customers and Products, create a Region Bridge with only unique RegionID, and link Customers and Products to it.

3. Deactivate extra relationships

- In Model view, set some relationships to **inactive**.
- Then activate them only when needed using **DAX USERELATIONSHIP()**.
- Example:
- Sales by Ship Date =
- CALCULATE(
- SUM(Sales[Amount]),
- USERELATIONSHIP(Sales[ShipDate], Dates[Date])
-)

4. Normalize dimension tables (Star Schema)

- Always aim for a **star schema** (Fact in center, Dimensions around).
- Avoid snowflake-like chains of dimensions if possible.

5. Use Power Query merges instead of relationships

- If unavoidable, flatten some lookups in Power Query (merge Customers with Regions).
- This reduces relationship complexity and eliminates loops.

Best Practice

- In Power BI never leave a loop in place.
- Star Schema with One-to-Many relationships → fastest, cleanest, easiest to maintain.

14. • When NOT to use bidirectional filtering

- In a **clean star schema** (Fact in the middle, Dimensions around it) → you almost never need bidirectional filters.
- Dimensions should filter facts (One \rightarrow Many), not the other way around.
- If you turn on "Both" everywhere, you risk circular relationships and unexpected totals.

When it is appropriate

1. Many-to-Many Relationships

If you truly need to filter **both directions** (e.g., Customers ↔ Sales Territories where multiple customers can belong to multiple territories), sometimes **Both** is required.

But usually better to resolve with a **bridge table**.

2. Helper (Bridge) Tables

When you build a **bridge table** between two dimensions and need filters to flow both ways.

Example:

- Sales fact table
- Customers dimension
- Products dimension
- A custom bridge (e.g., MarketingCampaigns linked to both Customers and Products).

Here, bidirectional filtering allows campaigns to filter both Customers and Products.

3. Role-Playing Dimensions (sometimes)

If you have multiple Date relationships and want one slicer to control multiple date roles, you may use bidirectional filters.

(Although often better solved with USERELATIONSHIP or duplicate Date tables).

4. Specialized Analysis Needs

- You want to see **which customers bought which products** directly (Customer ↔ Product through Sales).
- Normally you'd query via the Sales fact, but with Both filtering you can sometimes shortcut.
- A Should still prefer explicit bridge tables.

Best Practice

- Use Single direction (One → Many) by default.
- Turn on **Both** only for *specific* analytical needs.
- If you enable Both, **document why**, and check for **ambiguity loops** in Model View.

15. This is a smart way to make your Power BI model more robust when referential integrity (foreign key ↔ primary key) is broken — e.g., a CustomerID in **Sales** does not exist anymore in **Customers**.

Since Power BI doesn't enforce referential integrity automatically (unlike SQL), you handle it with **DAX**.

Option 1: Flag invalid Sales rows

Create a calculated column in Sales:

```
CustomerValid =

IF (

ISBLANK ( RELATED ( Customers[CustomerID] ) ),

"Orphan", -- Customer does not exist
```

```
"Valid"
```

- RELATED() → tries to pull CustomerID from **Customers**.
- If it's blank \rightarrow that means the Customer was deleted (no match).

Now you can filter out "Orphan" rows in reports.

Option 2: Default replacement for missing Customers

Sometimes you want to show "Unknown Customer" instead of blanks. Create a new column in **Sales**:

```
CustomerNameSafe =
IF (
    ISBLANK ( RELATED ( Customers[CustomerName] ) ),
    "Unknown Customer",
    RELATED ( Customers[CustomerName] )
)
```

This way, reports won't break when a Customer is deleted.

✓ Option 3: Measure-level enforcement

If you only care about measures, you can wrap aggregations with TREATAS or COALESCE:

```
Total Sales (Valid Customers) =

CALCULATE (

SUM ( Sales[SalesAmount] ),

TREATAS ( VALUES ( Customers[CustomerID] ), Sales[CustomerID] )

)
```

This ensures only Sales linked to real Customers are included.

4 Best Practice

- **If possible**: fix it at the data source (SQL constraints or Power Query cleaning).
- In DAX: use ISBLANK(RELATED()) to catch broken foreign keys.

• Show "Unknown Customer" instead of dropping rows → more user-friendly.

b Do you want me to also show you how to **prevent these orphan rows in Power Query** (before loading the data), or should we stick to DAX handling only?