

Data Cleaning, Transformation and Modelling

Question 1 : What is Data Loading in Power BI and why is it considered the first step of analysis?

Data Loading in Power BI, often initiated via the "Get Data" feature, is the foundational process of connecting the Power BI Desktop to external data sources and importing that information into the application. It acts as the bridge between raw data storage—whether in Excel files, SQL databases, cloud services, or web pages—and the analytical engine of Power BI.

It is considered the **first and most critical step** of analysis for several reasons:

- **Establishing the Foundation:** We cannot analyze what is not there. Data loading brings the raw material. Without this step, the subsequent stages of cleaning, modeling, and visualization simply cannot happen.
- **Defining Scope and Source:** This stage determines *what* data is available for analysis. It involves selecting specific tables or datasets, which sets the boundaries of your entire project.
- **Triggering ETL (Extract, Transform, Load):** Loading data immediately leads into the **Power Query Editor**.

This is where the raw data is cleaned and shaped. Since transformation logic relies on the structure of the incoming data, the loading phase dictates how accurate and efficient our data preparation will be.

Question 2 : Explain the difference between “Load” and “Transform Data” in Power BI.

When you connect to a new data source in Power BI, we are presented with a Navigator window containing two primary action buttons. The choice you make here dictates the immediate next step in your workflow.

1. Load

Selecting **Load** tells Power BI to take the data exactly as it currently appears in the source and import it directly into the **Power BI Data Model**.

- The window closes, and Power BI begins processing rows immediately. The data becomes available in the "Data" and "Report" views.
- We have to use this *only* if your data is already perfectly clean, formatted, and requires no changes (e.g., a highly curated SQL view or a verified Excel table).

- If the data has errors, unnecessary columns, or wrong data types, you will clutter your model and potentially slow down performance.

2. Transform Data (Recommended)

Selecting **Transform Data** (formerly "Edit") redirects the data to the **Power Query Editor** before it enters the Data Model.

- The Power BI interface opens a separate window (Power Query). Here, we can clean the data—filtering rows, renaming columns, changing data types, or replacing values—without affecting the original source.
- This is the **best practice** for most scenarios. It allows us to shape the data to fit your analytical needs, ensuring that only clean, relevant data is loaded into your report.
- It reduces the file size and improves report performance by filtering out unnecessary data *before* import.

Feature	Load	Transform Data
Destination	Direct to Data Model & Report View	Power Query Editor
Action	Imports data "as is"	Allows cleaning & shaping first
Best For	pre-processed data	Raw, messy, or large datasets
Memory Usage	High (imports everything)	Optimized (imports only what you keep)

Question 3 : What is a Fact Table and a Dimension Table?
Give examples from the dataset.

In Power BI and data modeling (specifically the Star Schema), tables are categorized into two types based on the data they hold:

1. Fact Table

A **Fact Table** contains the quantitative data or "metrics" of your business. These are the numbers you want to analyze (sum, average, count). It represents **events** or transactions that happened.

- Long and narrow (many rows, fewer columns), contains numbers and IDs (Foreign Keys) linking to dimension tables.

Example from the Dataset: If we were to model our given dataset, the **Fact Table** would be the central table containing the daily statistics.

- **Name:** Fact_Daily_Cases
- **Columns included:**
 - **Keys (to link data):** Date (or DateID), Location_ID (a generated ID linking to Country/State).
 - **Metrics (The "Facts"):**

- Confirmed_Cases
- Recovered_Cases
- Deaths
- Active_Cases
- Tests_Conducted
- Hospitalized

These columns contain numbers that you will aggregate (e.g., "Total Deaths in Brazil" or "Average Hospitalized in UK").

2. Dimension Table

A **Dimension Table** contains descriptive attributes or "context" related to the facts. These are the text fields you use to slice, dice, and filter your data. Short and wide (fewer rows, more descriptive columns), contains unique values and a Primary Key.

Example from your Dataset: From your file, we would extract the descriptive columns to create **Dimension Tables**.

- **Proposed Name:** Dim_Location
 - **Columns included:** Country, State.
 - **Usage:** Used to filter the report (e.g., a Slicer for "Country").
- **Proposed Name:** Dim_Vaccination_Status

- **Columns included:** Vaccination_Status.

Question 4 : Why is Star Schema preferred over Snowflake Schema in Power BI?

In Power BI, the **Star Schema** is universally preferred over the Snowflake Schema primarily due to **performance** and **usability**.

Power BI runs on the **VertiPaq engine**, which is an in-memory, columnar database. It is highly optimized for scanning single tables and compressing data. While Snowflake schemas save storage in traditional databases by normalizing data (splitting tables to avoid repetition), Power BI's compression engine handles data redundancy so efficiently that this storage benefit is negligible.

Instead, the Snowflake schema introduces a major penalty: **Joins**. Every relationship in Power BI has a computational cost. A Star Schema ensures that any dimension is just *one relationship* away from the fact table. A Snowflake schema often requires traversing chains of tables (e.g., Sales > Product > SubCategory > Category) to filter data. This "relationship hopping" significantly slows down query performance and report rendering.

Furthermore, Star Schemas are **simpler for users**. A business user wants to find "Product Category" in the "Product" table, not hunt for it in a separate linked table. This structure also simplifies **DAX formulas**, avoiding the need for complex, performance-killing bi-directional filters often required in Snowflake designs .

Question 5 : Identify and remove duplicate records based on Date, Country, and State.

1. Open Power Query

- We have to use the **Home** tab in Power BI Desktop.
- Click **Transform data**.

2. Select the Columns

- Hold down the **Ctrl** key on our keyboard.
- Click the headers of the three columns you want to check: Date, Country, and State.

3. Removing Duplicates

- With those three columns still highlighted, **Right-click** on one of the headers clicking the **Ctrl** key .
- Select **Remove Duplicates** from the menu.

4. Check our Work

- Look at the "Applied Steps" panel on the right. We will be able to see a new step called "**Removed Duplicates**".
- Click **Close & Apply** (top left) to save the changes.

Note : The date format in the raw data was incorrect. Directly changing the datatype in Power BI resulted in an "Error" for the entire column, even when attempting the change locale.

The screenshot shows the Power Query Editor interface with a table of data from the 'coronavirus_2020_dataset.csv' file. The table has columns: County, State, Date, Confirmed_Cases, Recovered_Cases, Deaths, Active_Cases, Tests_Conducted, and Hospitalized. The 'Date' column is highlighted with a red arrow pointing to it. In the 'APPLIED STEPS' pane on the right, there is a step named 'Removed Duplicates' with a red arrow pointing to it. The 'Properties' pane shows the query name is 'coronavirus_2020_dataset (1)'.

County	Date	Confirmed_Cases	Recovered_Cases	Deaths	Active_Cases	Tests_Conducted	Hospitalized
1 Brazil	01-01-2020	41240	56133	158	41424	123788	10288
2 UK	02-01-2020	88807	61288	1977	44002	98239	17709
3 Italy	03-01-2020	33434	38243	956	53744	27938	17543
4 UK	04-01-2020	69724	57384	2071	4795	281466	4650
5 UK	05-01-2020	75697	31653	403	10634	451930	4987
6 USA	06-01-2020	28732	26657	4611	4798	351040	15200
7 Italy	07-01-2020	11134	39952	2779	22420	63292	4697
8 Italy	08-01-2020	23954	21280	1387	51507	30055	16724
9 Italy	09-01-2020	57108	78411	3453	55039	306035	3828
10 UK	10-01-2020	87044	1622	955	29300	82912	18352
11 Brazil	11-01-2020	87505	48756	1712	24345	138660	17087
12 Italy	12-01-2020	56668	48089	4703	10749	353680	12605
13 Spain	13-01-2020	23061	16725	1273	38595	269448	5482
14 UK	14-01-2020	79252	61396	3811	25121	8345	18184
15 USA	15-01-2020	9749	51226	2625	37487	161372	5463
16 Brazil	16-01-2020	16907	29402	2382	28411	418475	13860
17 Spain	17-01-2020	18777	4263	1441	33361	92006	4912
18 Spain	18-01-2020	93229	46828	2979	21223	482769	4648
19 USA	19-01-2020	51745	47924	1797	7394	409568	855
20 Brazil	20-01-2020	91865	74182	4390	14056	130350	2866
21 UK	21-01-2020	81818	59052	4009	28380	165437	11834
22 India	22-01-2020	23624	11664	4850	26035	282590	16294
23 Brazil	23-01-2020	23321	35737	1394	30620	265480	16566
24 USA	24-01-2020	10488	29373	4281	19877	325176	3112
25 Spain	25-01-2020	22574	36397	1269	59945	207487	4492
26 UK	26-01-2020	5240	38709	4732	55316	98852	15114
27 Brazil	27-01-2020	43221	61120	3714	43404	493092	1517
28 India	28-01-2020	10478	19250	529	51824	139953	16018
29 India	29-01-2020	55645	21172	3623	53352	203779	17740
30 Italy	30-01-2020	19262	11555	2958	30911	155553	11451
31 Italy	31-01-2020	52796	78617	2914	32784	281618	8210

Question 6 : Identify and replace null values in Vaccination_Status

Step 1: Open Power Query Editor

1. We have to use the **Home** tab in Power BI Desktop.
2. Click **Transform data**. This opens the Power Query Editor window.

Step 2: Identify the Nulls

1. Locate the **Vaccination_Status** column.
2. We will see some cells that are literally empty.
3. To verify, click the small arrow (filter icon) on the column header. We will see (blank) in the list of values.

Step 3: Replace Values

1. **Right-click** the header of the **Vaccination_Status** column.
2. Select **Replace Values** from the menu.
3. In the dialog box that appears:
 - **Value To Find:** Type **null** (leave this empty as your cells are just blank).
 - **Replace With:** Type the value you want, such as **Not Mentioned**.

4. Click OK.

Step 4: Verify and Save

1. The *blank* values in the column should now instantly change to "Not Mentioned".
2. Click **Close & Apply** in the top-left corner to save these changes to your report.

The screenshot shows the Power Query Editor interface with the following details:

- File**: Untitled - Power Query Editor
- Queries**: [coronavirus_2020_dataset]
- Transform** ribbon tab selected.
- Applied Steps** pane on the right shows the step: **Removed Duplicates**.
- Properties** pane on the right shows the query name: **coronavirus_2020_dataset()**.
- Data** pane displays a table with columns: Country, State, Date, Confirmed_Cases, Recovered_Cases, Deaths, Active_Cases, Tests_Conducted, Hospitalized.
- A red arrow points to the **Removed Duplicates** step in the Applied Steps pane.

Question 7 : Create a new column to calculate Recovery Rate.

Method : Using DAX

Go to Data View: Click the Table icon on the left sidebar.

New Column:

- Right-click on your table name in the **Data** panel (right side).
- Select **New column**.

Enter Formula:

Recovery Rate =

```
DIVIDE('coronavirus_2020_dataset  
(1)[Recovered_Cases], 'coronavirus_2020_  
dataset (1)[Confirmed_Cases], 0)
```

Question 8 : Create a summarized table showing total confirmed cases by Country.

[Click here](#) to check answers (5 to 8)