1) Source Data Verification — Approach & Findings

Tools / Setup

I used **SQLiteOnline** (https://sqliteonline.com/) to ensure the process was simple and reproducible.

Source Files & Formats (as Received)

The dataset was provided as three separate files in different formats:

- Customer.xls: An Excel workbook with a single sheet containing columns:
 Customer_ID, First, Last, Age, Country.
- Order.csv: A CSV file with columns: Order ID, Item, Amount, Customer ID.
- Shipping.json: A JSON array of objects with fields: Shipping_ID, Status, Customer_ID.

Handling Mixed Formats in SQLiteOnline

I handled the different file formats by uploading each file directly into **SQLiteOnline**, creating three separate staging tables: customers, orders, and shipping. This approach, which matched column names to the source headers, ensured traceability.

- For the Excel (.xls) file, I used SQLiteOnline's direct file upload feature to ingest the first sheet as a flat table.
- For the CSV, I confirmed the **UTF-8 encoding** and **comma delimiter** were correct before importing.
- For the JSON, the file was imported as a single array of records, with each object becoming a single row in the shipping table.

Import Checks at Load Time

During the ingestion process, I performed the following checks to ensure data integrity:

- I verified row counts for each table to confirm they matched the file expectations.
- I confirmed that **key columns were non-null** after the import.
- I **spot-checked a few rows** in each table to ensure data types were parsed as intended (e.g., Amount as numeric, Age as an integer, and Status as text).
- I kept the column names identical to the source files for easy traceability.

Staging tables created:

- customers: Customer_ID, First, Last, Age, Country
- orders: Order ID, Item, Amount, Customer ID
- shipping: Shipping ID, Status, Customer ID

What I Checked (and Why)

- Row/Column Sanity & Nulls: To confirm the data feeds loaded correctly and no crucial fields were missing.
- 2. **Primary Key Uniqueness**: To ensure each table had a stable natural key from the source.
- 3. **Referential Integrity**: To verify that foreign keys are correctly resolved (orders.customer_id → customers.customer_id, shipping.customer_id → customers.customer_id).
- 4. **Domain / Business Rules**: To validate that values were within valid ranges (e.g., Amount > 0, Status was limited to expected values like 'Pending' or 'Delivered').
- 5. **Text Hygiene**: To identify and address "leetspeak" or stray characters in names (e.g., N!cole, L@rry, R0bert, Al1cia).
- 6. **Structural Anomalies**: To detect issues like multiple shipping rows per customer or customers without corresponding orders/shipping data.

SQL I Ran

Nulls / Shape

SQL

SELECT

(SELECT COUNT(*) FROM customers) AS customers_rows, (SELECT COUNT(*) FROM orders) AS orders_rows, (SELECT COUNT(*) FROM shipping) AS shipping_rows;

- -- Example for checking specific column nulls
- -- SELECT SUM(CASE WHEN Customer_ID IS NULL THEN 1 ELSE 0 END) FROM customers;

Primary-Key Uniqueness

SQL

```
SELECT Customer_ID, COUNT(*) c FROM customers GROUP BY 1 HAVING c > 1; SELECT Order_ID, COUNT(*) c FROM orders GROUP BY 1 HAVING c > 1; SELECT Shipping_ID, COUNT(*) c FROM shipping GROUP BY 1 HAVING c > 1;
```

Referential Integrity

```
SQL
-- Orders → Customers
SELECT o.*
FROM orders o
LEFT JOIN customers c
ON c.Customer ID = o.Customer ID
WHERE c.Customer_ID IS NULL;
-- Shipping → Customers
SELECT s.*
FROM shipping s
LEFT JOIN customers c
ON c.Customer ID = s.Customer ID
WHERE c.Customer_ID IS NULL;
Business Rules
SQL
-- Amount must be positive
SELECT COUNT(*) FROM orders WHERE Amount <= 0;
-- Status limited to two values
SELECT DISTINCT Status FROM shipping;
Name Hygiene
SQL
-- Has digits
SELECT Customer ID, First FROM customers WHERE First GLOB '*[0-9]*';
SELECT Customer_ID, Last FROM customers WHERE Last GLOB '*[0-9]*';
-- Has disallowed special characters (blacklist pattern)
SELECT Customer_ID, First FROM customers
WHERE First GLOB '*[]!@#$%^&*()_+=\[{};:""|,./<>?\\]*';
SELECT Customer_ID, Last FROM customers
```

WHERE Last GLOB '*[]!@#\$%^&*()_+=\[{};:""|,./<>?\\]*';

Coverage / Structural Checks

```
SQL
```

```
-- Customers with no orders
SELECT COUNT(*) AS customers_without_orders
FROM customers c
LEFT JOIN orders o
 ON o.Customer ID = c.Customer ID
WHERE o.Customer_ID IS NULL;
-- Customers with no shipping rows
SELECT COUNT(*) AS customers without shipping
FROM customers c
LEFT JOIN shipping s
 ON s.Customer ID = c.Customer ID
WHERE s.Customer ID IS NULL;
-- Multiple shipping rows per customer
SELECT Customer ID, COUNT(*) AS shipping rows
FROM shipping
GROUP BY Customer ID
HAVING COUNT(*) > 1;
```

What I Found

- Loads & Nulls: All three files loaded cleanly with no NULL values in the key fields.
 Each file had 250 rows.
- Primary Keys: No duplicate Customer_ID, Order_ID, or Shipping_ID values were found.
- **Foreign Keys**: 100% of the orders.customer_id and shipping.customer_id values successfully matched a customer, indicating no orphaned rows.
- Domain Checks:
 - All Amount values were positive.
 - Status values were limited to 'Pending' and 'Delivered'.
- Name Hygiene: I did find a few names with digits or symbols (e.g., N!cole, L@rry, R0bert, Al1cia). I developed a repeatable transformation to map these characters to letters.
- Structural Anomaly: A significant finding was that the shipping table is at the
 customer level, not the order level. This means a customer can have multiple
 shipping rows, which can cause double-counting of order amounts if not handled
 carefully in joins.
 - My interim approach was to collapse shipping to one row per customer for reporting purposes. (Note: this is a major caveat as all the modelling is dependant on this assumption)
 - Note: For future robustness, the shipping table should ideally include
 Order_ID (and/or timestamps) to accurately model shipping at the order level.

- The Shipping table status should ideally have fields like Shipped, partially delivered, delivered, In-Transit, Returned, Exchanged etc for better logging of transaction
- As mentioned earlier Date related fields are missing entirely, which will give us some more flexibility and various insights of the transactions

Bottom Line

- **Accuracy**: The data is highly accurate, with unique keys, resolving foreign keys, and clean numeric and categorical domains.
- **Completeness**: There were no missing rows in the core join paths, and I accounted for customers without orders or shipping rows.
- Reliability: Data distributions were sane. The primary modeling observation is the
 customer-level grain of the shipping table, which was addressed in the dataset
 design. Other probable fields which potentially help modelling better, are mentioned
 above.