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# LAB 02: ELT Ingestion & Transformations

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**Duration:** ~40 min

**Day:** 1

**After module:** M02: ELT Data Ingestion

**Difficulty:** Beginner-Intermediate

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## Scenario

*“The RetailHub data team received the first data export from the store system – CSV files with customer and product data, plus a JSON file with order history. Your task: load these files into the Lakehouse, apply basic transformations, and save them as Delta tables in the Bronze layer.”*

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## Objectives

After completing this lab you will be able to:

- Read CSV, JSON, and Parquet files using `spark.read`
- Compare `inferSchema` vs explicit schema definition
- Apply DataFrame transformations: `select`, `withColumn`, `filter`, `cast`
- Create temporary views and run SQL queries
- Write DataFrames as managed Delta tables using `saveAsTable()`

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## Prerequisites

- Completed LAB 01 (or just run `%run ../setup/00_setup`)
  - Dataset files available in Volume or `DATASET_PATH`
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## Part 1: Data Ingestion (~15 min)

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### Task 1: Read Customers CSV

Read the customers CSV file with explicit schema (don't use `inferSchema`).

**Requirements:** - Define a `StructType` schema with appropriate types - Use `.option("header", True)` - Verify column types with `printSchema()`

**Exam Tip:** `inferSchema` triggers an extra pass over the data. In production, always define schema explicitly for CSV files.

<screen = Notebook cell showing `printSchema()` output with correctly typed columns (string, integer, etc.)>

### Task 2: Read Orders JSON

Read the orders batch JSON file. JSON infers schema automatically, but verify nested structures.

**Requirements:** - Use `spark.read.format("json")` - Check if any columns contain nested types (struct, array)

### Task 3: Read Products CSV

Read the products file with `inferSchema` and then compare the schema with a manually defined one.

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## Part 2: Transformations (~15 min)

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### Task 4: Clean Customer Data

Apply the following transformations to `df_customers` : 1. `select` only: `customer_id, first_name, last_name, email, city, country` 2. `withColumn` – create `full_name` by concatenating `first_name + " " + last_name` 3. `filter` – keep only customers from a specific country 4. `cast` – ensure `customer_id` is `IntegerType`

## Task 5: Create Temp View + SQL Query

1. Register `df_customers` as a temporary view `v_customers`
2. Write a SQL query to count customers per country
3. Display the result sorted by count descending

**Exam Tip:** `createOrReplaceTempView()` is session-scoped.

`createOrReplaceGlobalTempView()` is accessible from any notebook on the same cluster via `global_temp.view_name`.

## Part 3: Save as Delta Tables (~10 min)

### Task 6: Write to Bronze Layer

Save all three DataFrames as Delta tables in the Bronze schema: - `bronze.customers` - `bronze.orders` - `bronze.products`

**Requirements:** - Use `.write.mode("overwrite").saveAsTable()` - Use fully qualified names: `{CATALOG}.{BRONZE_SCHEMA}.table_name`

### Task 7: Verify Tables

Query the tables using SQL to verify they were saved correctly.

<screen = SQL query result showing `SELECT count(*) FROM bronze.customers` with row count>

## Summary

In this lab you: - Loaded CSV and JSON files into Spark DataFrames - Applied transformations: `select`, `withColumn`, `filter`, `cast` - Created temporary views and ran SQL queries - Saved cleaned data as Delta tables in the Bronze layer

**What's next:** In LAB 03 you will use MERGE to handle incremental data updates and learn Time Travel for disaster recovery.