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Welcome to your new dbt project!

# Webshop Data Warehouse POC

#### 1. Overview

This project is a **proof-of-concept data warehouse** built on Snowflake, orchestrated with dbt. It demonstrates an end-to-end pipeline from raw CSV landing through business-ready marts, supporting the CEO's key metrics:

- Order Intake & Revenue by shop, location, time
- **Drop-off Rate** from intake to shipped revenue
- New vs. Returning Customers over time

# 2. Architecture & Ingestion

- 2.1 PoC Ingestion (Raw Schema)
  - Schema: raw
  - Method: Snowflake Dynamic Tables + ENABLE\_SCHEMA\_EVOLUTION
  - Process:
    - 1. Stage CSV files in an internal stage (@raw/...).
    - 2. Create dynamic tables on top of the stage; Snowflake infers schema and picks up new columns automatically.
    - 3. Changed or added files automatically insert or evolve columns on reload.

#### Why?

Quick to set up, no external orchestration required, perfect for a 3-hour PoC.

### 2.2 Production-Grade Ingestion (Future)

- Schema: raw
- Method: External S3 stage + Snowpipe + schema evolution + PIPES
- Process:
  - 1. Drop CSVs into an S3 bucket.
  - 2. Snowpipe auto-ingests new files into raw tables.
  - 3. Schema evolution handles new columns.
  - 4. Provides event-driven, near real-time loading.

## 3. dbt Layered Schema Design

```
raw → stg → core → mart

| Layer | Schema | Purpose | | ------|
```

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```
| **Raw** | `raw` | Landing zone & schema-evolved ingest |
| **Staging** | `stg` | Light cleaning, renaming, type casting (`stg_*`)
|
| **Core** | `core` | Star schema dims/facts (incremental: `d_*`, `f_*`) |
| **Mart** | `mart` | Business aggregations & KPI views (`rep_*`) |
```

### 4. Naming Conventions

```
Staging models → schema stg, tables named stg_<source>

Core Dimensions → schema core, tables named d_<business_object>

Core Facts → schema core, tables named f_<fact_name>

Marts → schema mart, views named rep_<metric_name>

dbt models folders mirror schemas: models/stg/, models/core/, models/mart/
```

### 5. Key Findings & Assumptions

- 5.1 Products (stg\_products → core.d\_products)
  - **product\_number** appears to be the *base product* code.
  - When is\_variant = TRUE, each sku\_id/product\_id is a variant (e.g. color, size).
  - When is\_variant = FALSE but product\_number still has multiple rows, these are likely historical versions of the base product (tracked by updated\_at).
  - **Assumption:** Keep only the **latest** row per **sku\_id** and link every variant back to its base via product number.
  - **Surrogate key:** product\_sk generated via dbt\_utils.surrogate\_key.
- 5.2 Customers (stg\_customers → core.d\_customers)
  - Multiple customer\_ids may represent the same human (guest vs. registered).
  - No reliable email/name fields to dedupe; treat each customer\_id as distinct.
  - Future: Implement identity resolution via email/name fuzzy matching or an external MDM system.
  - Incremental logic: Keep only the latest row per customer\_id based on updated\_at.
  - Surrogate key: customer\_sk.
- 5.3 Orders (stg\_orders → core.f\_orders)
  - order date = event for Order Intake
  - delivery\_date = event for Revenue Recognition
  - Shipping Lag = DATEDIFF('day', order\_date, delivery\_date)
  - Various sales\_event statuses (e.g. shipped, return, failed\_payment, etc.).
    - **Drop-off Rate** uses only webshop\_order → shipped for intake vs. revenue.
    - Other statuses (returns, failed, etc.) are tracked separately or assumed out-of-scope for basic drop-off.

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### 5.4 Order Positions (stg\_order\_positions → core.f\_order\_lines)

- price & quantity used to compute position\_amount = quantity \* price.
- **Precision:** Cast to **DECIMAL(10, 4)** and **ROUND(price, 4)** for all monetary/unit fields; final line amounts stored as **DECIMAL(12, 2)**.
- Unit of Measure: Assumed price is per-unit in EUR.
- Ambiguity: If price were total per line, use price / quantity; I assume per-unit for the PoC.

#### 5.5 Shops (stg\_shops → core.d\_shops)

- Static lookup of shop metadata: shop\_id, shop, platform, locale, etc.
- Materialization: Full-refresh table; very small, so rebuild on every run.

### 5.6 Date Dimension (core.d\_date)

- Covers **1990-01-01** → 70 years out.
- All timestamps loaded as NTZ; business dates assume UTC.
- Surrogate key: date\_sk = YYYYMMDD.
- Supports both order\_date and ship\_date keys.

### 6. Marts & Example Metrics

- rep\_order\_intake\_vs\_revenue
   Intake, revenue, and drop-off rate by date/shop/location.
- rep\_customer\_retention
   Absolute & relative new vs. returning customers over time.

# 7. Next Steps & Production Considerations

- Automate ingestion with **Snowpipe** & external stages + pipes for true CDC.
- Improve customer dedupe via **MDM** or fuzzy-matching on email/name/address.
- Implement **SCD Type 2** for dimensions that require history.
- CI/CD & Testing: Add dbt tests (not\_null, unique, relationships) and integrate with GitHub Actions.
- Enforce **RBAC** and data governance in Snowflake.