

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
- **Shipping Lag** (purchase → delivery)
- **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	<code>`raw`</code>	Landing zone & schema-evolved ingest	
Staging	<code>`stg`</code>	Light cleaning, renaming, type casting (<code>`stg_*`</code>)	
Core	<code>`core`</code>	Star schema dims/facts (incremental: <code>`d_*`</code> , <code>`f_*`</code>)	
Mart	<code>`mart`</code>	Business aggregations & KPI views (<code>`rep_*`</code>)	

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.

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_shipping_lag**
Average & median shipping lag by date/shop.
- **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.
- Optimize performance: clustering keys on fact tables, right-size warehouses.