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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
- **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** | `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.

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• 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; we 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 $1990 01 01 \rightarrow 70$ 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.