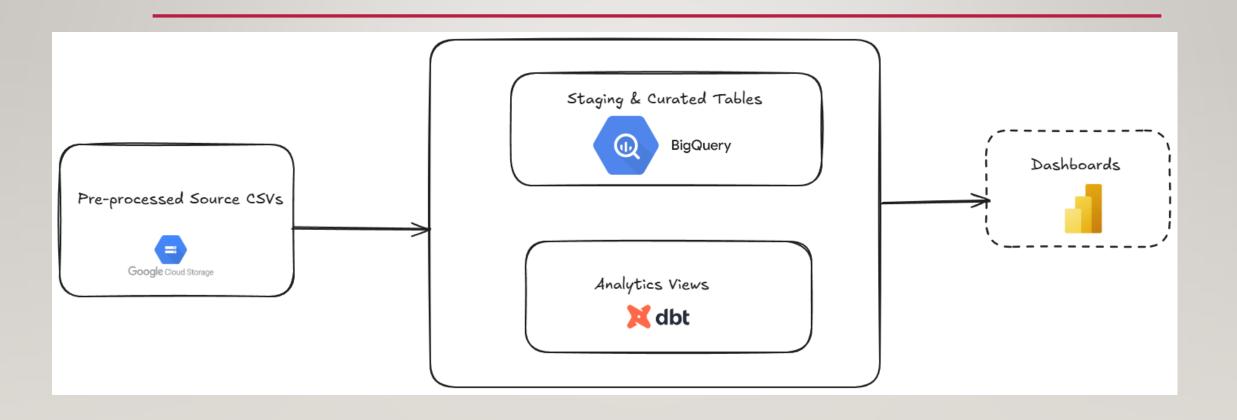
AMAZON REVIEWS DWH- JUST EAT

PRIYANKA KUPPUSWAMY

HIGH LEVEL ARCHITECTURE/TOOLS



EDA ON REVIEWS DATA

- reviewerID: Unique ID for each reviewer.
- asin: Amazon Standard Identification Number (product ID).
- reviewerName: Name of the reviewer (469 missing values).
- helpful: A list representing the helpfulness votes (e.g., [0,0] for no votes).
- reviewText: Full text of the review (24 missing values).
- overall: Rating given by the user (float, typically from 1 to 5).
- summary: Short review summary.
- unixReviewTime: Unix timestamp of the review.
- reviewTime: Formatted date of the review.

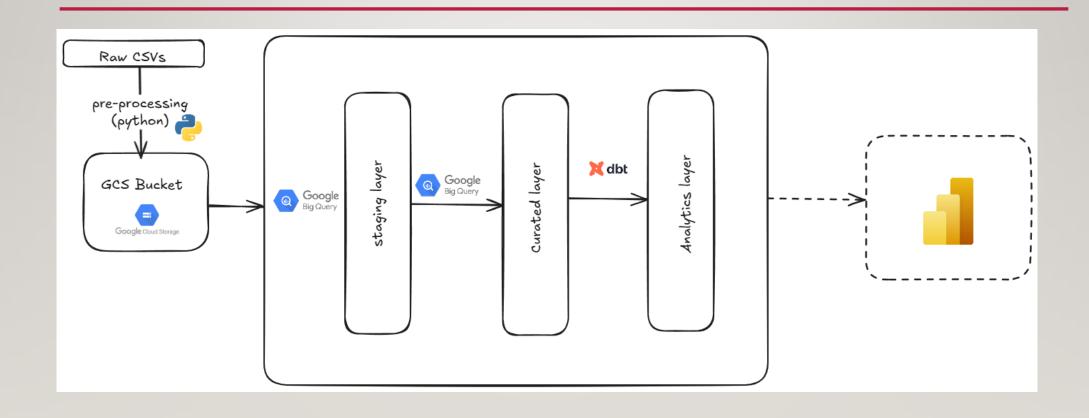
EDA ON METADATA

- metadataid: Unique identifier for each product metadata entry.
- asin: Amazon Standard Identification Number (product ID).
- salesrank: Dictionary-like structure storing the product's sales rank in categories.
- imurl: URL of the product image.
- categories: Category of the product (e.g., Clothing, Shoes & Jewelry).
- title: Product title (23 missing values).
- **description**: Product description.
- **price**: Product price.
- related: JSON-like structure listing related products (124 missing)
- brand: Product brand.

DATA QUALITY ISSUES - SOURCE

- Malformed Rows: Some rows did not have all columns, due to missing values or improper formatting(eg special characters: #).
- Unescaped Quotes: If a field contains unescaped quotes (") or line breaks, it might cause misalignment. So, I escape the quotes (\")
- **Delimiter Issues:** If a field contains a comma but is not enclosed in quotes so it is splitting into multiple columns incorrectly. So (,) inside fields was replaced with (;)
- Preprocessing python scripts help to fix these issues to successfully load in BigQuery
- Note: Most formatting problems are in the Description column of metadata table.

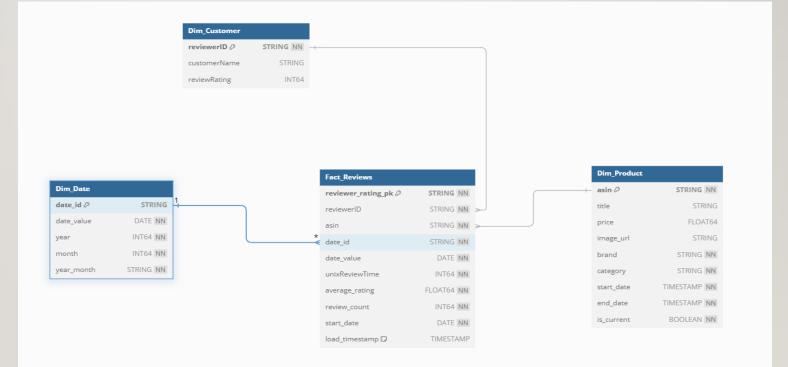
ETL SOLUTION DESIGN



ER DIAGRAM

Dim_customer is part of the ETL Solution Design but not implemented in Data

Warehouse.



STAGING TABLES

- Optimized CSV Integration: configure parsing parameters (skip_leading_rows, field_delimiter, quote, encoding) to ensure accurate and efficient data ingestion.
- Missing Data Management: Use COALESCE and NULLIF functions to handle null or empty values, maintaining data consistency
- Data Type Casting: Apply CAST functions to enforce correct data types, enhancing data integrity.
- **Date Standardization:** Convert string dates into DATE format with PARSE_DATE, ensuring uniformity across datasets.

DIM TABLES – SCD TYPE 2

- **Historical Tracking:** Utilized a MERGE statement to handle updates and insertions (upsert), ensuring changes in product attributes are captured over time.
- Record Updates: Set end_date to the current timestamp and is_current to FALSE for existing records when changes are detected.
- New Record Insertions: Added new records with start_date as the current timestamp and end_date set to a distant future timestamp (9999-12-31 23:59:59 UTC), marking them as current (is_current = TRUE).

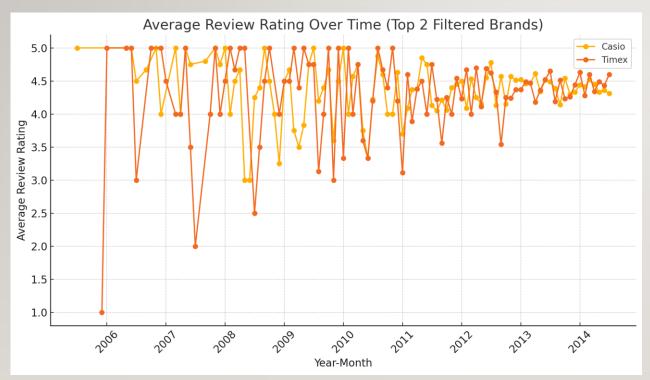
Dim_Product	
asin 🖉	STRING NN
title	STRING
price	FLOAT64
image_url	STRING
brand	STRING NN
category	STRING NN
start_date	TIMESTAMP NN
end_date	TIMESTAMP NN
is_current	BOOLEAN NN

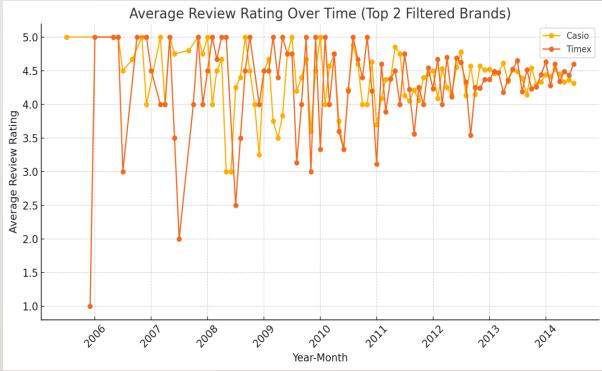
FACT TABLE

- Composite Primary Key: Generated a unique reviewer_rating_pk using a hash of key fields to accurately identify and merge records.
- Review Metrics Calculation: Computed average_rating and review_count for each product on a given date
- **Date Dimension Integration:** Joined with the Dim_Date table to enrich review data with standardized date attributes, supporting consistent time-based analysis.

I	Fact_Reviews	
ſ	reviewer_rating_pk 🛭	STRING NN
•	reviewerID	STRING NN
4	asin	STRING NN
4	date_id	STRING NN
	date_value	DATE NN
l	unixReviewTime	INT64 NN
l	average_rating	FLOAT64 NN
l	review_count	INT64 NN
	reviewRating	FLOAT64
	start_date	DATE NN
L	load_timestamp 🖸	TIMESTAMP

ANALYTICS VIEWS





DATA QUALITY CHECKS - IMPLEMENTED

- Total Record Count Check
- Null Value Check
- Duplicate Check
- MINUS Check (Data Mismatch Between External & Staged)
- Check for Default Value Assignments
- Referential Integrity
- Data Range Check

MERGE – CODE QUALITY

- Better readability and code maintainability
- MERGE ensures INSERT, UPDATE, and DELETE operations into a single action

OPTIMIZATION TECHNIQUES

Considerations for High-Volume, High-Frequency Data Loads:

- Partitioning and Clustering: Implemented partitioning by date_value and clustering by asin to enhance query performance in BigQuery.
- **Staging Tables:** Load new data into staging tables and then use selective queries to update or append only the necessary data.

WHAT COULD BE BETTER & FUTURE STEPS

- Create a Dim_Customer table to hold customer information & reviewRating
- Implement staging & curated layers in DBT (Currently only analytics views are built in BigQuery using DBT)
- Scheduling jobs using cron scripts

DEMO

THANK YOU!