

Understanding
Declining Sales of
Walmart in Brazil

| ORDERS | | | |
|-------------------------------|-------------|--|--|
| Field Name | Datatype | Description | |
| order_id | varchar(50) | Unique ID for each order | |
| customer_id | varchar(50) | Foreign key referencing the customer | |
| order_status | varchar(20) | Current status of the order | |
| order_purchase_timestamp | datetime | Date and time when the order was placed | |
| order_approved_at | datetime | Timestamp when the order was approved | |
| order_delivered_carrier_date | datetime | Date the order was handed to the carrier | |
| order_delivered_customer_date | datetime | Actual delivery date to the customer | |
| order_estimated_delivery_date | datetime | Expected delivery date | |

| ORDER_ITEMS | | | |
|---------------------|-------------|--|--|
| Field Name | Datatype | Description | |
| order_id | varchar(50) | Foreign key referencing the order | |
| order_item_id | int | Item ID number within the order | |
| product_id | varchar(50) | Foreign key referencing the product | |
| seller_id | varchar(50) | Foreign key referencing the seller | |
| shipping_limit_date | datetime | Deadline for the product to be shipped | |
| price | float | Selling price of the item | |
| freight value | float | Shipping cost paid by the customer | |

| PRODUCTS | | | |
|----------------------------|--------------|--|--|
| Field Name | Datatype | Description | |
| product_id | varchar(50) | Unique ID assigned to each product | |
| product_category | varchar(100) | Category or type of the product | |
| product_name_length | float | Length of the product name (in characters) | |
| product_description_length | float | Length of the product description | |
| product_photos_qty | float | Number of photos associated with the product | |
| product_weight_g | float | Weight of the product in grams | |
| product_length_cm | float | Length of the product in centimeters | |
| product_height_cm | float | Height of the product in centimeters | |
| product_width_cm | float | Width of the product in centimeters | |

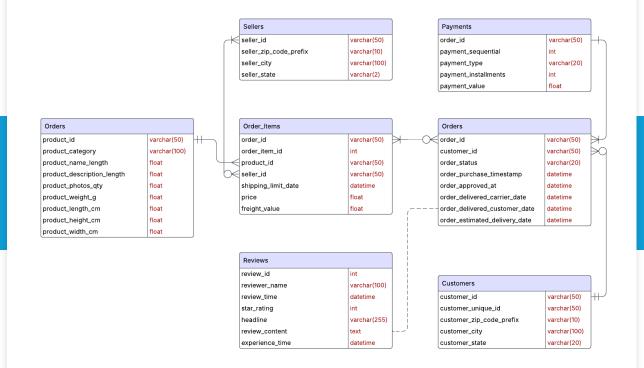
| REVIEWS | | | |
|-----------------|-------------|--|--|
| Field Name | Datatype | Description | |
| review_id | varchar(50) | Unique ID assigned to the review | |
| review_NAME | varchar(50) | Foreign key referencing the related order | |
| review_time | int | Rating given by the customer (typically 1 to 5) | |
| star_rating | text | Optional title of the review (may be null) | |
| headline | text | Customer's written feedback or review message | |
| review_content | datetime | Date the review was submitted | |
| experience time | datetime | Timestamp when the review was responded to (if applicable) | |

| CUSTOMERS | | |
|--------------------------|--------------|--|
| Field Name | Datatype | Description |
| customer_id | varchar(50) | Unique customer identifier (primary key) |
| customer_unique_id | varchar(50) | Anonymized persistent customer ID |
| customer_zip_code_prefix | varchar(10) | ZIP code prefix of the customer |
| customer_city | varchar(100) | City of the customer |
| customer_state | varchar(20) | State of the customer |

| PAYMENTS | | |
|----------------------|-------------|--|
| Field Name | Datatype | Description |
| order_id | varchar(50) | Foreign key referencing the order |
| payment_sequential | int | Payment sequence number (for split payments) |
| payment_type | varchar(20) | Type of payment (e.g., credit card, boleto) |
| payment_installments | int | Number of installments selected |
| payment_value | float | Total payment amount made |

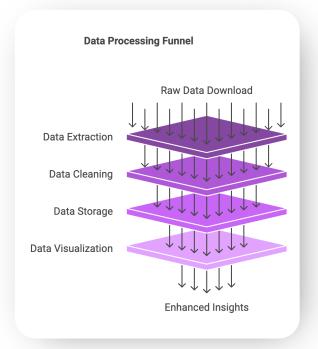
| SELLERS | | |
|------------------------|--------------|--|
| Field Name | Datatype | Description |
| seller_id | varchar(50) | Unique ID assigned to each seller |
| seller_zip_code_prefix | varchar(10) | ZIP code prefix of the seller's location |
| seller_city | varchar(100) | City where the seller is located |
| seller_state | varchar(20) | State where the seller is located |

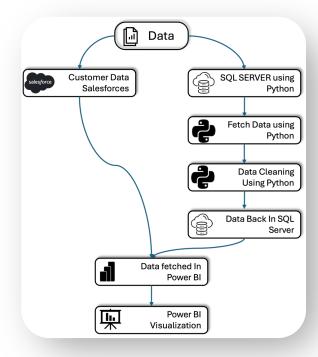
Data Dictionary



Entity Relationship Diagram

High Level Project Flow





Challenges Faced

```
1 select
        isnull(reviewer_name, 'No Name') as reviewer_name,
 2
 3
        coalesce(cast(review_time as date),experience_time) as review_time,
 4
 5
            when star_rating='Rated 1 out of 5 stars' then 1
 6
            when star_rating='Rated 2 out of 5 stars' then 2
            when star_rating='Rated 3 out of 5 stars' then 3
 7
 8
            when star_rating='Rated 4 out of 5 stars' then 4
 9
            when star_rating='Rated 5 out of 5 stars' then 5
10
            else 3 end as rating,
11
        headline,
12
        review content
    into refined_review
    from walmart_trustpilot_reviews
```

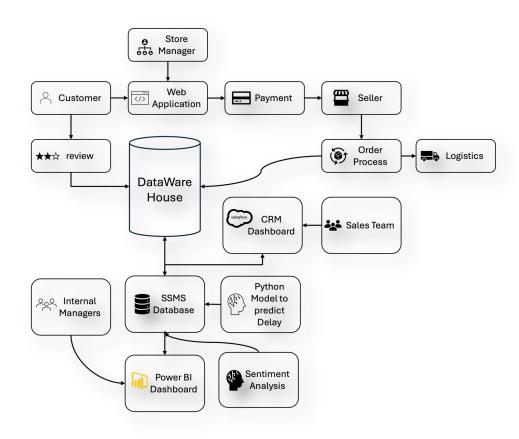
Review was In text Format

```
1 def convert_review_time(df):
     2
                                                     """Convert review_time and experience_time columns."""
     3
                                                    def clean_review_time(value):
                                                                           if isinstance(value, str) and re.match(r'^\d{4}-\d{2}-\d{2}:\d{2}:\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2}\cdot\d{2
     5
                                                                                                     return datetime.strptime(value[:-1], "%Y-%m-%dT%H:%M:%S.%f").strftime("%Y-%m-%d %H:%M:%S")
     6
                                                                            return None
    7
                                                    def clean_experience_time(value):
     8
    9
                                                                                                     return datetime.strptime(value, "%B %d, %Y").strftime("%Y-%m-%d")
10
                                                                           except (ValueError, TypeError):
11
                                                                                                     return None
12
13
                                                    df['review_time'] = df['review_time'].apply(clean_review_time)
14
                                                    df['experience_time'] = df['experience_time'].apply(clean_experience_time)
  15
                                                    return df
```

DateTime Format was aligning with SSMS

Walmart SWOT Analysis: Strategic Focus Areas **∀**⊃ Weaknesses Strengths -- High Rate of Late Deliveries Large Customer Base -**Poor Customer Reviews** Brand Recognition - -Siloed Data Systems Extensive Supply Chain Network .-' Walmart **SWOT** Opportunities Analysis Threats 🖰 - Leverage Business Intelligence Intense Competition from Amazon -Implement CRM Strategies Risk of Further Sales Decline .- ' - Utilize Predictive Analytics





Conclusion

4 Integrated Solutions

 Power BI, Salesforce CRM, and continuous improvement processes addresses sales decline and operational inefficiencies Strategic Impact

3

- Predictive inventory analytics reduce stockouts.
- Al-driven campaigns personalize customer engagement.
- Logistics improvements boost delivery reliability.

2 Marketing Wins

- Target younger audiences (25–34) via digital campaigns.
- Launch "Fresh Finds with Walmart" to promote groceries.
- Introduce "Walmart Rewards Brazil" to drive loyalty and repeat sales.

Expected Results

- +10% sales growth projected within one fiscal year.
- Improved customer satisfaction, online engagement, and foot traffic.
- Reclaimed market share and strengthened brand positioning



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