Global Superstore Sales Analysis



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Introduction

In today's competitive business environment, data-driven decision-making is crucial in optimizing sales performance and improving efficiency. This report presents an in-depth analysis of Global Superstore Sales Data, covering various aspects of data preprocessing, visualization in Power BI, statistical hypothesis testing, and business insights.

This project aimed to explore sales data, identify key trends, and address challenges affecting business growth. By leveraging data preprocessing techniques, we ensured data quality and consistency. Interactive dashboards were then developed to visualize sales patterns and enhance business intelligence. Furthermore, hypothesis testing was conducted to validate assumptions and uncover meaningful statistical relationships. Finally, actionable recommendations were proposed to help improve sales, optimize inventory, and enhance customer satisfaction.

This study provides a comprehensive approach to analyzing large-scale retail data and demonstrates the importance of data analytics in making strategic business decisions.

Dataset Overview and Structure

The Global Superstore Sales Dataset contains transactional data related to sales, customers, shipping, and product details. It provides valuable insights into various aspects of business operations, such as sales trends, customer segments, shipping preferences, and product performance.

Key Features of the Dataset

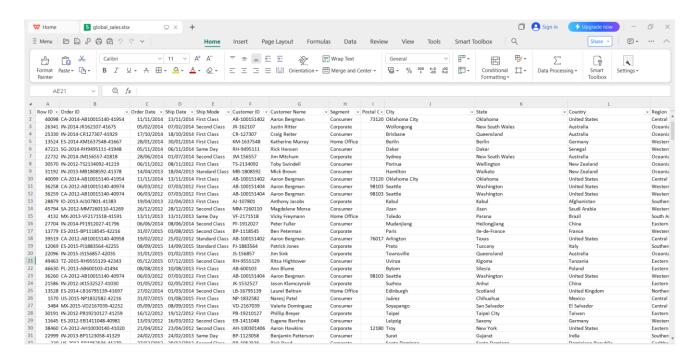
The dataset consists of multiple columns, each representing specific information about an order. Below are some of the key attributes:

- Order ID: A unique identifier for each transaction.
- Order Date & Ship Date: The date an order was placed and the corresponding shipping date.
- Ship Mode: The type of shipping service used (e.g., First Class, Standard Class).
- Customer Details: Includes Customer ID, Name, and Segment (e.g., Consumer, Corporate).
- Location Information: Covers Postal Code, City, State, Country, and Region where the order was placed.
- Market: The geographical market segment (e.g., USCA for the United States and Canada).
- **Product Details:** Contains Product ID, Category, Sub-Category, and Product Name.
- Sales & Quantity: The total sales amount and the number of units sold.
- **Discount & Profit:** The discount applied and profit earned from the order.
- Shipping Cost: The cost incurred for shipping the order.
- Order Priority: The priority level assigned to the order (e.g., High, Medium, Low).

Initial Observations

- The dataset contains a diverse range of products across multiple categories such as Technology, Furniture, and Office Supplies.
- Customers belong to different segments, including Consumer, Corporate, and Home Office.
- Sales data is recorded across various geographical locations, providing opportunities to analyze regional trends.
- Different shipping modes affect delivery time and costs, impacting customer satisfaction.
- The presence of discounts and varying profit margins suggests the need for pricing strategy analysis.

Understanding this dataset forms the foundation for further analysis, including data preprocessing, dashboard creation, and hypothesis testing, to extract meaningful business insights.



Data Preprocessing in Power BI

Introduction to Data Pre-processing

Data Pre-processing is a crucial step in data analysis that involves cleaning, transforming, and preparing data for analysis and visualization. In Power BI, data preprocessing is performed using Power Query Editor, which allows users to handle missing values, remove duplicates, format data, and more.

Why is Preprocessing Important?

- Ensures data consistency and accuracy.
- Removes errors and missing values that can impact analysis.
- Standardizes data for better insights and visualization.
- Improves data quality for decision-making.

Checking and Handling Missing Values

Different strategies are used for different columns:

Column	Missing Value Strategy
Postal code	Fill with N/A
Sales, profit, Quantity, Shipping cost	Replace with 0
Discount	If missing set to 0%

Removing Duplicates

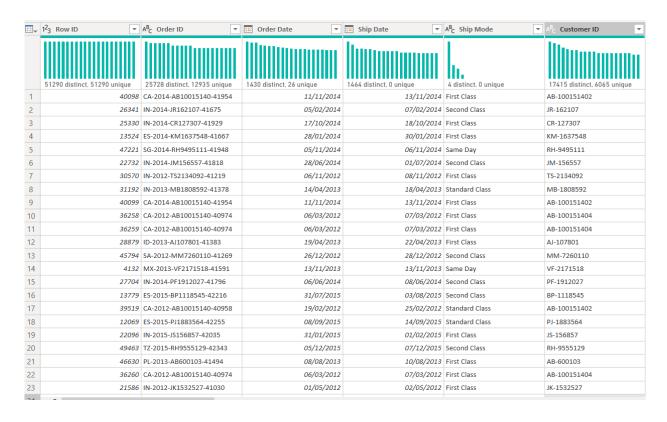
To ensure data accuracy, duplicate entries were identified and removed based on Order ID and other relevant attributes. This step prevented redundancy and maintained the integrity of sales records.

Correcting Data Types

Incorrect data types can lead to computational errors and inconsistencies in visualization. For example:

- Text → Customer Name, Order ID
- Date → Order Date, Ship Date
- Number → Sales, Profit, Quantity
- Currency → Sales, Profit, Shipping Cost

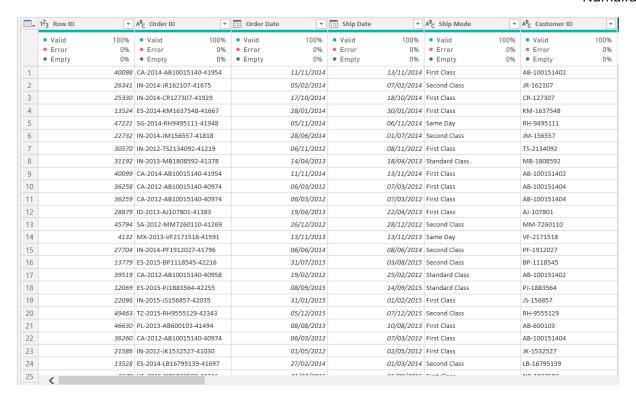
Error Free data Images with some statistics



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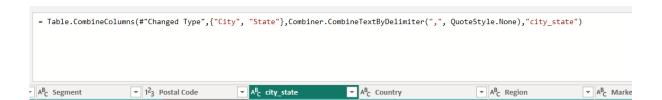
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Table.TransformColumnTypes(#"Promoted Headers", {{"Row ID", Int64.Type}, {"Order ID", type text}, {"Order Date", type date}, {"Ship Date", type date}, {"Ship Mode", type text}, {"Customer ID", type text}, {"Customer ID", type text}, {"State", type text}, {"Country", type text}, {"Region", type text}, {"Sub-Category", type text}, {"Product Name", type text}, {"Sales", type number}, {"Quantity", Int64.Type}, {"Order Priority", type text}}, {"Sales", type number}, {"Order Priority", type text}}) ▼ A^B_C State ▼ A^B_C Region ▼ A^B_C Market ▼ A^B_r Product ID AB_C City ValidError ValidErrorEmpty ValidErrorEmpty ValidErrorEmpty ValidErrorEmpty ValidErrorEmpty ValidError 0% • Empty Empty 0% 0% 0% Oklahoma City Oklahoma TEC-PH-5816 Technology United States Central US USCA FUR-CH-5379 Wollongong New South Wales Australia Oceania Asia Pacific Furniture Brisbane Queensland Australia Oceania Asia Pacific TEC-PH-5356 Technology Berlin Berlin Germany Western Europe Europe TEC-PH-5267 Technology Dakar Dakar Senegal Western Africa Africa TEC-CO-6011 Technology Sydney New South Wales Australia Asia Pacifi TEC-PH-5842 Ship Date ▼ A^BC Ship Mode ▼ A^B_C Segment ▼ 1²₃ Postal Code ☐ ✓ ☐ Order Date ▼ A^B_C Customer Nan ▼ A^B_C Customer ID 11/11/2014 13/11/2014 First Cl AB-100151402 Aaron Bergman 05/02/2014 07/02/2014 Second Class JR-162107 Justin Ritter Corporate 17/10/2014 18/10/2014 First Class CR-127307 Craig Reiter 28/01/2014 30/01/2014 First Class KM-1637548 Katherine Murray Home Office 06/11/2014 Same Day 05/11/2014 RH-9495111 Rick Hansen 28/06/2014 01/07/2014 Second Class JM-156557 Jim Mitchum Corporate 06/11/2012 08/11/2012 First Class TS-2134092 Toby Swindell Consumer 18/04/2013 Standard Class 14/04/2013 MB-1808592 Mick Brown 11/11/2014 13/11/2014 First Class AB-100151402 Aaron Bergman Consume 10 < Value distribution Column statistics Standard Class Count Second Class First Class Empty Distinct Same Day Unique 0 Empty string 0 Min First Class Max Standar... ▼ A^B_C Order Priority 40.77 High 923.63 Critical 915.49 Medium 5175.171 919.971 910.16 Medium 2892.51 0.1 -96.54 2832.96 2832.96 2862.675 1822.08 311.52 763.275 564.84 903.04 Critical 897.35 Critical 894.77 Critical 5244.84 878.38 High 25.27 High 54.7136 341.96 < Column statistics Value distribution 51290 Error Empty Distinct Unique NaN Zero Min Max 0.85 0.14290.. Standard deviation 0.21227.. 0.5 0.5 0.17 0.47 ▼ A^B_C Country ▼ A^B_C State ▼ A^B_C Market ▼ A^B_C Product ID TEC-PH-5816 Oklahoma **United States** Central US USCA Technology New South Wales Australia Oceania Asia Pacific FUR-CH-5379 Furniture Queensland Australia Oceania Asia Pacific TEC-PH-5356 Technology 4 Berlin Germany Europe TEC-PH-5267 Technology Western Europe Dakar Africa TEC-CO-6011 Technology Western Africa Senegal Australia Technology Wellington New Zealand Oceania Asia Pacific FUR-CH-5378 Furniture Waikato Oceania Asia Pacific FUR-TA-3764 Furniture New Zealand Oklahoma United States Central US USCA FUR-BO-5957 Furniture > 10 < Value distribution Count 0 Error Empty 0 Western US 23 Distinct Southeastern Asia 0 Unique South America Empty string 0 Eastern US Min Canada Southern Asia Max Western... Western Asia Eastern Asia Central US Northern Europe Southern Europe Caribbean Southern US

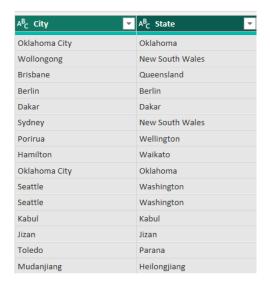


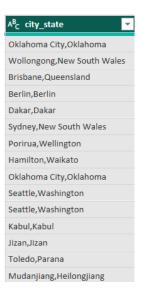
Merging Columns for Enhanced Data Representation

Certain columns were merged to create more comprehensive fields, improving data readability and streamlining analysis. Specifically, the City and State columns were combined into one column, enhancing the clarity of location-based insights.



Before After





Feature Engineering: Creating New Columns

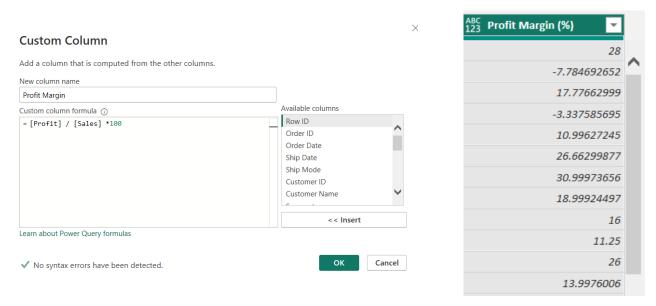
Feature engineering plays a crucial role in enhancing data analysis by deriving new insights from existing variables. In this project, new calculated columns were created using DAX (Data Analysis Expressions) in Power BI to improve the analytical model's performance. Two key features were introduced:

1. Profit Margin Calculation

The Profit Margin column was created to analyze the profitability of each order. It is calculated using the formula:

Profit Margin = (Profit / Sales) × 100

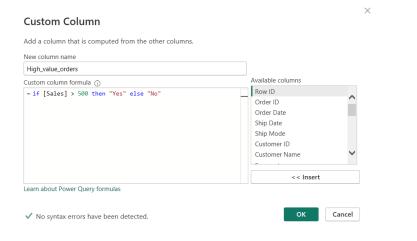
- A higher profit margin indicates better profitability.
- If the profit margin is negative, the order was sold at a loss.
- This metric helps assess how efficiently the company generates profit from sales.

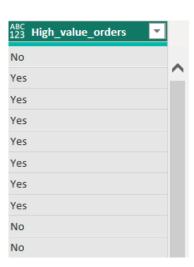


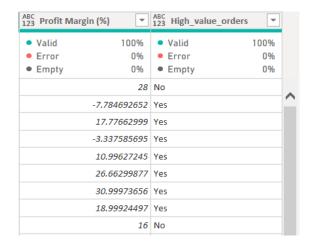
2. Identifying High-Value Orders

A new column was created to classify orders as "High-Value" if the total sales amount exceeded \$500. This feature helps in:

- Identifying premium customers who place large orders.
- Analyzing purchasing trends of high-spending customers.







This pre-processing ensures that data is clean, structured, and ready for analysis in Power BI. Each step verifies data integrity, making it suitable for visualization and decision-making.

Dashboard 1: Sales & Profit Analysis

Dashboard Overview

The Sales & Profit Analysis Dashboard is designed to help businesses track their sales performance, profit trends, and high-value orders. By analyzing key performance indicators (KPIs), the dashboard provides insights into revenue generation, profitability, and product performance.

Problem Statement

Businesses often face challenges in understanding:

- Which products and categories contribute the most to profit.
- How sales and profit fluctuate over time.
- The impact of high-value orders on business revenue.
- The profitability of different categories and sub-categories.

How to optimize product sales to maximize overall profit. To address these concerns, the dashboard visualizes sales and profit metrics, enabling data-driven decision-making to improve business performance.

Key Performance Indicators (KPIs)

Total Sales: The total revenue generated from all sales.

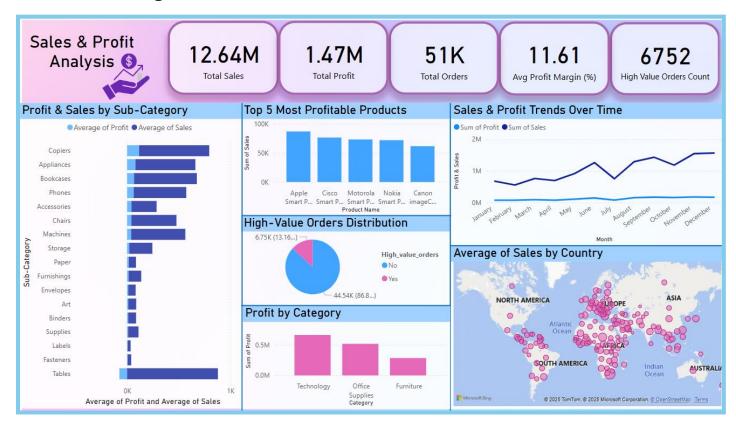
Total Profit: The overall profit after deducting costs.

Total Orders: total number of orders placed.

Average Profit Margin (%): (Total Profit / Total Sales) × 100, measuring profitability efficiency percentage.

High-Value Orders Count: Identifies orders above a certain price threshold.

Dashboard Image



Data Visualization Explanation

1. Sales & Profit Trend Chart

Chart Type: Line Chart

Metrics:

Sum of Sales (Dark Blue Line)

Sum of Profit (Light Blue Line)

Findings:

This line chart displays the trend of total sales and profit over 12 months (January to December).

- Sales show an upward trend, peaking in June, September, and December, indicating revenue growth.
- Profit remains stable with minor fluctuations, highlighting controlled margins.
- o The gap between sales and profit reflects operational costs and discounting impacts.
- Seasonal trends suggest higher demand mid-year and year-end, likely due to promotions or holidays.

Use Case: Helps identify sales and profit trends over time to optimize business strategy and financial planning.

2. Top 5 Most Profitable Products

Chart Type: Clustered column chart

Metric: Sum of Sales

Findings:

- The Apple Smart Phone has the highest sales among all products.
- Other high-performing products include Cisco Smart Phone, Motorola Smart Phone, Nokia Smart Phone, and Canon imageClass
- The sales volume among these top products is relatively consistent, indicating steady demand across high-value items.

Use Case: Helps in identifying best-selling products to optimize inventory and marketing strategies.

3. High-Value Orders Distribution

Chart Type: Pie Chart

Metric: High-Value Orders (Yes/No)

Findings:

- 86.8% of total orders are not high-value.
- Only 13.16% of orders qualify as high-value orders.
- Even though high-value orders are a smaller percentage, they might contribute significantly to overall revenue.

Use Case: Understanding the distribution of high-value orders can help optimize pricing strategies, promotions, and targeted customer engagement.

4. Profit by Category

Chart Type: Clustered column chart

Metric: Sum of Profit

Findings:

- Technology is the most profitable category.
- o Office Supplies Category follows, with Furniture being the least profitable.
- The profit distribution varies significantly, highlighting the need for category-specific pricing and inventory strategies.

Use Case: Helps in profit optimization by identifying high-margin categories to focus on while addressing low-margin areas.

5. Profit & Sales by Sub-Category

Chart Type: Stacked bar chart

Metric: Average Profit (light blue), Average Sales (dark blue)

Findings:

- o Tables have high sales but low negative profit, indicating pricing inefficiencies.
- Copiers, Appliances, and Machines balance sales and profit well.
- Fasteners, Labels, and Art have minimal sales and profit and need better marketing.

Use Case: Identifies profitable and underperforming sub-categories for better pricing and inventory decisions.

6. Average of Sales by Country

Chart Type: Map bubble size

Metric: Average Sales (bubble size & density)

Findings:

- North America, Europe, and Asia lead in sales, while Africa and South America show growth potential.
- Larger bubbles indicate major sales hubs, guiding resource allocation.

Use Case: Helps identify high-demand regions for market expansion and sales optimization.

This dashboard offers key insights into sales, profitability, product trends, and regional performance. It helps prioritize high-performing products, adjust low-profit items, and optimize inventory and marketing based on seasonal trends. Overall, it supports data-driven decisions for growth and efficiency.

Dashboard 2: Shipping Cost & Discount Insights

Dashboard Overview

This dashboard provides key insights into shipping costs and discount distribution and their impact on order fulfillment and profitability. It helps identify trends, inefficiencies, and optimization opportunities in logistics and discounts.

Problem Statements

- Analyzing shipping expenses over time to identify cost-heavy periods and inefficiencies.
- Examining the impact of discounts on profitability to optimize pricing strategies.
- Evaluating fulfillment times across different shipping methods to enhance efficiency.
- Identifying high-cost regional order volume improves logistics and reduces shipping expenses.

Key Performance Indicators (KPIs)

Total Shipping Cost: Provides an overview of overall shipping expenses

Average Shipping Cost: Provides the average cost per order, helping assess logistics efficiency.

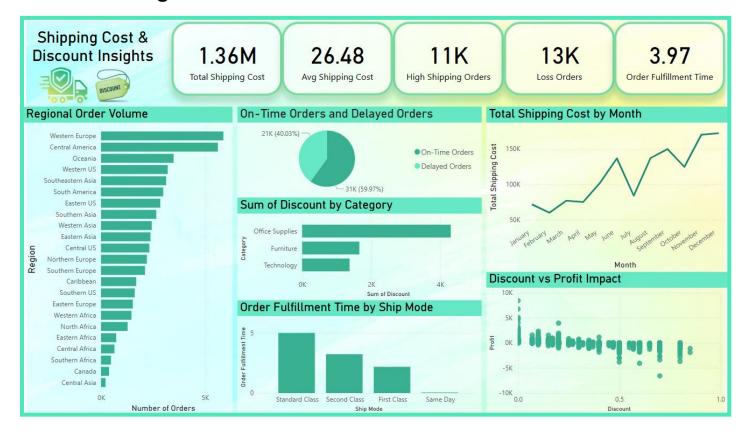
High Shipping Orders: Identifies orders with excessive shipping costs, highlighting potential

areas for cost optimization.

Loss Orders: Represents orders where costs outweigh revenue, often due to high shipping fees or large discounts.

Order Fulfillment Time: Measures the average time taken to process and deliver orders, influencing customer experience and efficiency.

Dashboard Image



Data Visualization Explanation

1. Regional Order Volume

Chart Type: Bar chart

Metric: Number of orders per region.

Findings:

- Western Europe and Central America have the highest order volumes.
- Central Asia and Canada have the lowest order volumes.
- Order volume distribution suggests a strong market presence in certain regions, while others may need better logistics or marketing strategies.

Use Case: Helps in identifying key markets for expansion and optimizing regional shipping strategies.

2. On-Time Orders and Delayed Orders

Chart Type: Pie Chart

Metric: Percentage of on-time vs. delayed orders.

Findings:

- o 59.97% of orders are delayed, highlighting potential logistical inefficiencies.
- o Only 40.03% of orders are delivered on time, which could impact customer satisfaction.
- A significant portion of delayed orders suggests a need for improved supply chain management.

Use Case: Helps in improving delivery performance by addressing bottlenecks in supply chain operations.

3. Sum of Discount by Category

Chart Type: Bar chart

Metric: Total discount provided across product categories.

Findings:

- o Office Supplies receive the highest discounts, suggesting a competitive pricing strategy.
- Furniture receives moderate discounts, indicating controlled price adjustments.
- Technology has the lowest discounts, possibly due to higher margins or less price competition.

Use Case: Helps evaluate discounting strategies and assess profitability by category.

4. Total Shipping Cost by Month

Chart Type: Line Chart

Metric: Monthly total shipping cost trend.

Findings:

- Shipping costs increase steadily throughout the year, with peaks in June, September, and December.
- o The rise in December suggests higher demand during the holiday season.
- Some months show a dip, indicating possible operational cost optimizations or seasonal fluctuations.

Use Case: Assists in budgeting and forecasting shipping expenses for peak periods.

5. Order Fulfillment Time by Ship Mode

Chart Type: Column chart

Metric: Average time taken to fulfill orders based on shipping mode.

Findings:

- Standard Class takes the longest time for fulfillment, impacting delivery speed.
- Same-day shipping is the fastest but likely incurs higher costs.
- Faster shipping modes (First Class, Second Class) provide a balance between speed and cost.

Use Case: Helps in optimizing shipping methods to improve delivery efficiency.

6. Discount vs. Profit Impact

Chart Type: Scatter plot

Metric: Relationship between discount amounts and profit.

Findings:

- Higher discounts tend to reduce profitability, with some transactions showing negative profits.
- o Some low-discount orders still result in losses, indicating other cost factors.
- A balanced discounting approach is needed to maintain profitability without compromising sales.

Use Case: Helps in balancing discounting strategies to maximize revenue while maintaining profitability.

The dashboard highlights trends in fulfillment time, helping to streamline delivery processes for better customer satisfaction. By analyzing regional shipping cost variations, businesses can refine logistics strategies. These insights enable cost-effective decision-making while balancing efficiency and profitability.

Star Schema: Data Organization

Fact Table: SalesFact — Central transactional data (sales, shipping, discounts, profits)

Dimension Tables & Relationships: (one to many)

SalesFact → CustomerData

Linked by Customer ID (Customer details)

SalesFact → ProductData

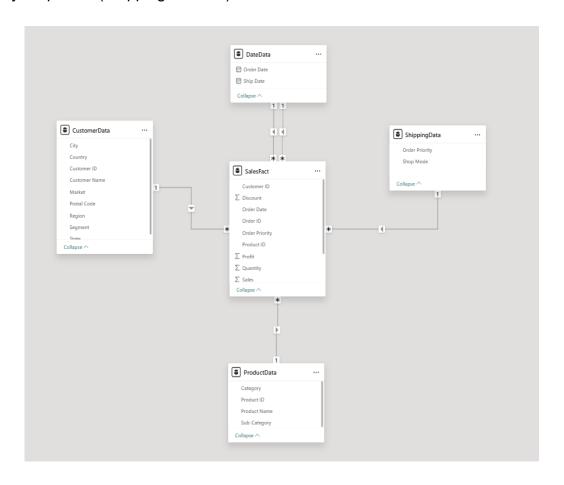
Linked by Product ID (Product information)

SalesFact → ShippingData

Linked by Ship Mode (Shipping method details)

SalesFact → DateData

- Linked by Order Date (Order timeline)
- Linked by Ship Date (Shipping timeline)



Five OLAP Operations

1. Roll-up Analysis: Yearly Summary of Sales and Profit

(Chart: "Sum of Profit and Sum of Sales by Year")

→ Aggregates data from lower levels (e.g., months) to a higher time level (years).

2. Drill-down Analysis: Monthly Breakdown of Sales and Profit

(Chart: "Sum of Profit and Sum of Sales by Month")

→ Explores deeper time granularity within the year for detailed trend analysis.

3. Slicing Operation: Sales in Technology Category

(Chart: "Sum of Sales by Category")

→ Filters the entire dataset to focus only on the Technology category.

4. Dicing Operation: Profit for Appliances in Germany

(Chart: "Sum of Profit by Country and Sub-Category")

→ A narrow data slice filtered by Country = Germany and Sub-Category = Appliances.

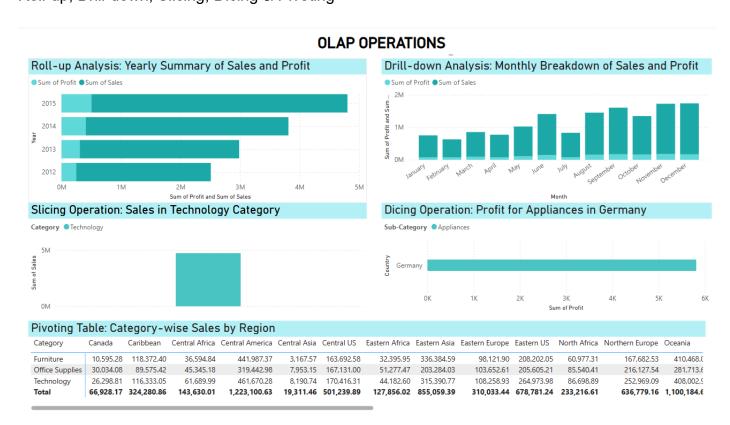
5. Pivoting Table: Category-wise Sales by Region

(Table at the bottom)

→ Rearranges data to compare multiple metrics across Categories and Regions.

Visualizations of OLAP Operations

Roll-up, Drill-down, Slicing, Dicing & Pivoting



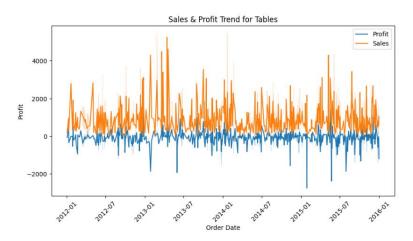
Analyzing and Solving Profitability Issues in Table A Hypothesis-Driven Approach

1. Identifying the Problem

During our analysis of Table's sales and profitability data, we observed an unexpected trend: under specific conditions, Table is experiencing negative profits. This is a significant issue because profitability is a key measure of financial sustainability. If losses continue, it could impact the long-term viability of the business.

To better understand the root causes of this issue, we conducted a detailed hypothesis-driven analysis focusing on three key factors that could be influencing profitability:

- High Discounts: Discounts are meant to attract customers, but they might be reducing profits too much.
- Shipping Costs: Higher shipping costs could be eating into profit margins.
- High-Value Orders and Shipping Costs: High-value orders are expected to generate more profit, but they might be incurring disproportionately high shipping costs.



2. Hypothesis Testing and Findings

Hypothesis 1: Impact of Discounts on Profitability Hypothesis Statement:

- Null Hypothesis (H₀): Discounts do not significantly reduce profit.
- Alternative Hypothesis (H₁): Discounts significantly reduce profit.

Analysis:

We categorized the Table data into two groups:

- High-discount orders (Discount > 30%)
- Low-discount orders (Discount ≤ 30%)

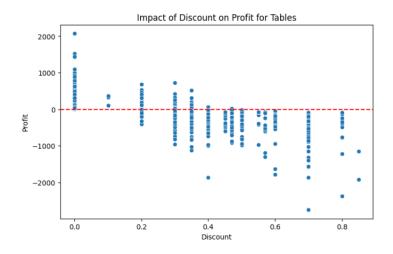
A t-test was performed to compare the profits of these two groups.

Results: T-statistic: -18.54 and P-value: 3.44e-61 (very small)

Conclusion: Since the p-value is far below 0.05, we reject H_0 and conclude that discounts significantly reduce profit.

Interpretation:

While discounts are a common strategy to boost sales, the data shows that high discounts are leading to losses rather than increasing profitability. This suggests that Table's discounting strategy needs to be revised.



Hypothesis 2: Effect of High Shipping Costs on Profitability Hypothesis Statement:

- Null Hypothesis (H₀): High shipping costs do not significantly reduce profit.
- Alternative Hypothesis (H₁): High shipping costs significantly reduce profit.

Analysis:

We divided the Table data into:

- High-shipping-cost orders (Above median shipping cost)
- Low-shipping-cost orders (Below median shipping cost)

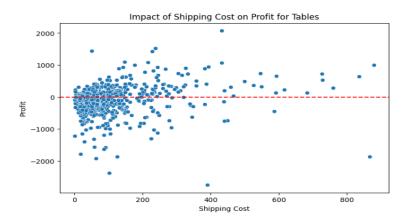
A t-test was conducted to compare profits between these two groups.

Results: T-statistic: 2.69 and P-value: 0.0074

Conclusion: Since the p-value is below 0.05, we reject H₀ and conclude that high shipping costs significantly reduce profit.

Interpretation for Table:

Shipping costs are significantly impacting profitability, which means that Table needs to optimize logistics expenses to avoid unnecessary costs.



Hypothesis 3: Relationship Between High-Value Orders and Shipping Costs Hypothesis Statement:

- Null Hypothesis (H₀): There is no significant difference in shipping costs between high-value and low-value orders.
- Alternative Hypothesis (H₁): High-value orders have significantly higher shipping costs.

Analysis:

- High-value orders were defined as those in the top 25% of total sales.
- We performed a t-test comparing shipping costs for high-value vs. low-value orders.

Results:

T-Statistic: 88.20 and P-value: 0.0

Conclusion: Since the p-value is extremely small, we reject H₀ and confirm that high-value orders incur significantly higher shipping costs.

Interpretation:

Shipping costs should ideally scale proportionally with order value, but the data suggests that Table's high-value orders are disproportionately expensive to ship. This indicates inefficiencies in shipping cost allocation.



3. Proposed Solutions to Improve Table's Profitability

Optimizing Discount Strategies:

- Implement a tiered discounting model, where discounts are based on total spending rather than high discounts across all products.
- Shift towards loyalty-based discounts instead of excessive markdowns, ensuring that only repeat customers receive high-value incentives.

Reducing Shipping Costs:

- Negotiate better rates with shipping providers to lower logistics expenses.
- Introduce a minimum order value for free shipping to offset costs.

Optimize packaging methods to reduce dimensional weight charges.

Controlling High Shipping Costs for High-value Orders:

- Establish regional distribution centers to minimize long-distance shipping expenses.
- Implement batch shipping where multiple orders are consolidated, reducing per-order shipping costs.

Conclusion

Our analysis of Table's profitability issues revealed that high discounts and excessive shipping costs are major contributors to negative profits. Through hypothesis testing, we quantified these impacts and identified key areas for improvement.

By optimizing discount strategies and improving shipping cost management, Table can effectively mitigate losses and enhance profitability. These data-driven recommendations ensure that Table's pricing and logistics strategies are aligned with sustainable business growth.