

Sales Performance Analysis

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Introduction

The AdventureWorks is a multinational manufacturing company. In this project I've explored and analyzed sales data of its products from 2014 to 2016 to gain insights into the company's revenue trends, product performance and customer preferences.

Using Excel, Alteryx and Power BI, I compared the Sales Budget with the Actual Sales data for 2016. Additionally, I categorized the customers based on their buying habit. Finally, I've discussed the project's limitations and proposed some future directions in the end.

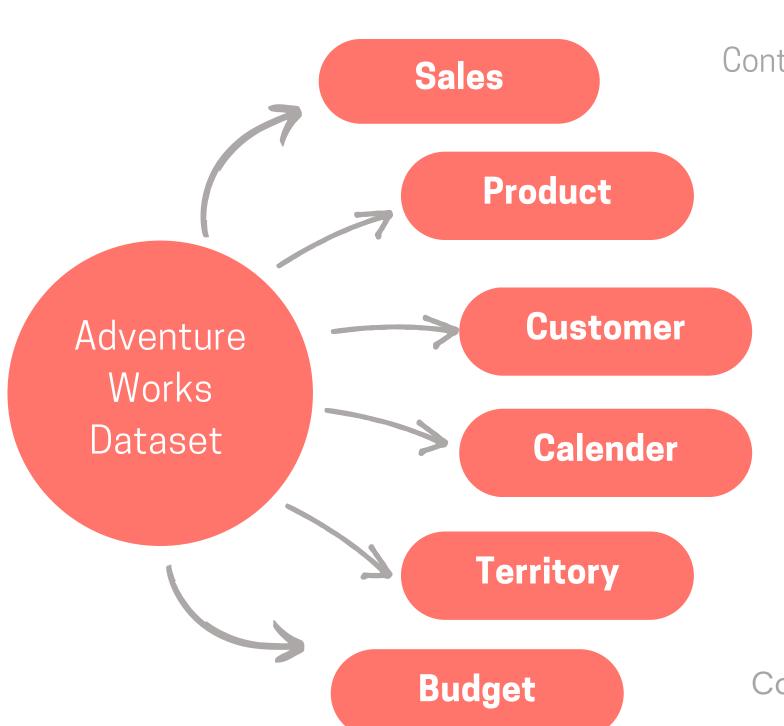
Business Problem



Extract various information such as Sales, Budget, and Variance. You can even compare sales and budgets with various attributes. Extract necessary information about Products and Customers. Make the necessary dashboard with the best you can extract from the data. Use various visualization and features and make the best dashboard. Find key metrics and factors and show the meaningful relationships between attributes. Do your own research and come up with your findings.



Data Sources



Contains daily sales data for various customers and products from 2014 to 2017.

Includes product categories, subcategories, costs, and list prices.

Provides demographic information such as gender, yearly income, education etc.

Covers years, months, and quarters data.

Contains sales country and region information.

Contains monthly budget data for 2016 for various products.

Tools

Methodologies

O1. Excel

02. Alteryx

03. Power BI

04. Google Slides

05. Canva

06. GitHub

01. Data Cleaning

03. Exploratory Data
Analysis

05. RFM Analysis

07. Reporting

02. Data Wrangling

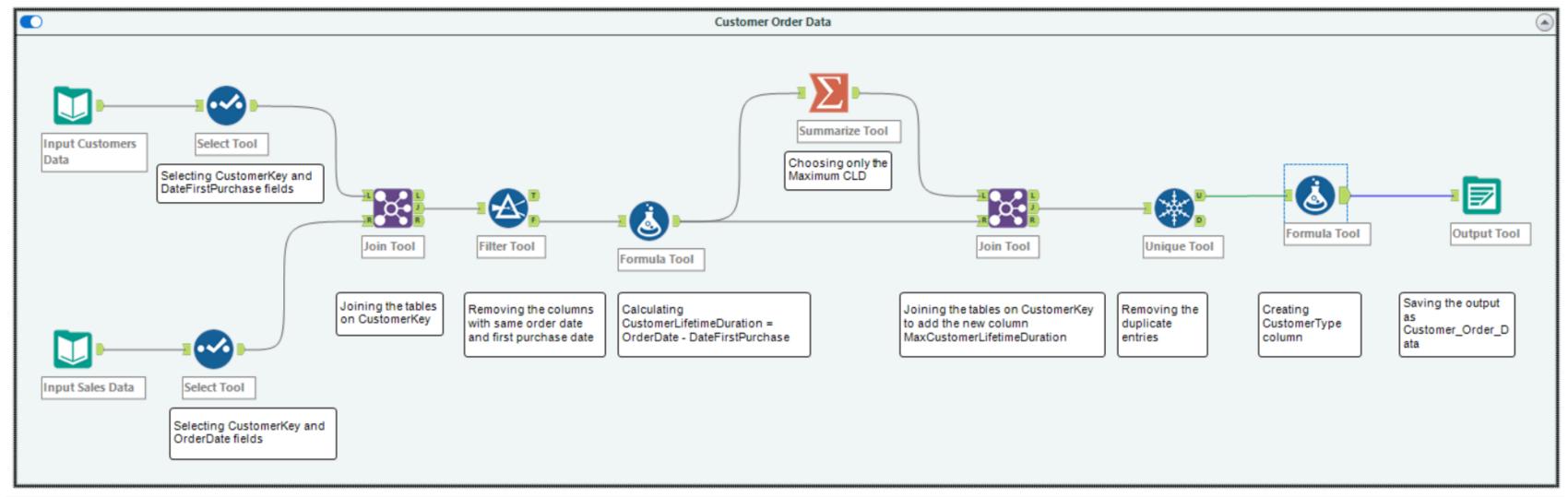
04. Customer Segmentation

06. Data Visualization

I leveraged Alteryx's Data Blending tools to create insightful new variables from existing data columns.

Procedure

3. Creating a new table Customer_Order_details with Customer Lifetime Duration column and CustomerType by categorizing them into 3 distinct groups - High, Medium and Low Value customers.



I have used these metrics to measure the sales performance using various numerical fields.

Sales Revenue

Total revenue generated from sales over a specific period.

Sales Quantity

Total number of units or products sold.

Average Order Value (AOV)

Average revenue earned per order.

Sales Growth Rate

Percentage increase or decrease in sales over a defined period.

Sales by Product Category

Revenue of sales for each product category.

Sales by Customer Segment

Revenue of sales by different customer segments

Metrics

Sales by Time Period

Revenue of sales analyzed over different time periods (monthly, quarterly, annually).

Product Performance

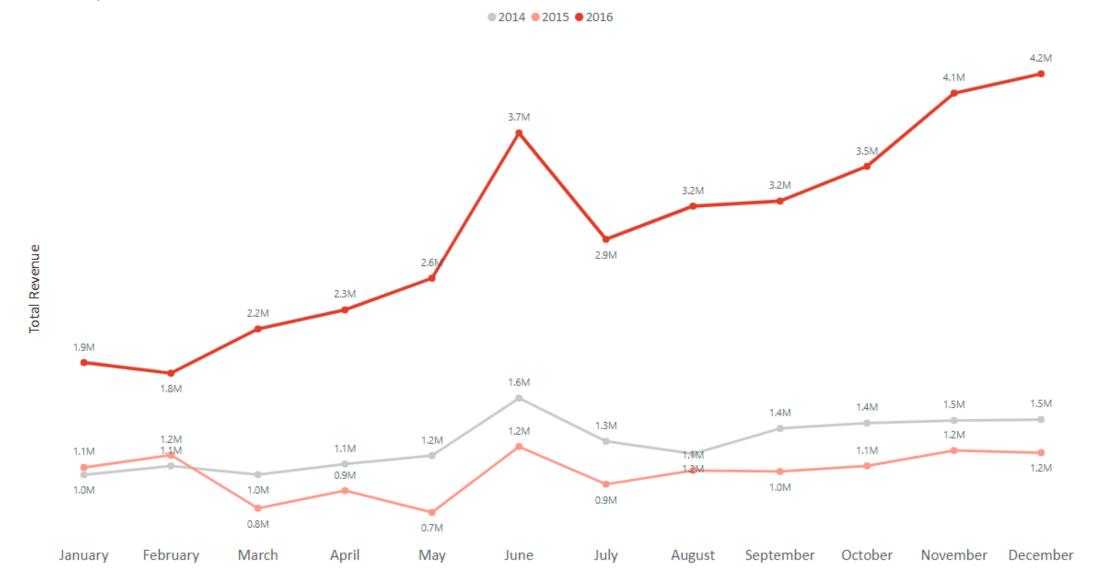
Analysis of how well each product is selling (e.g., top-selling products, slow-moving products).

Sales by Territory

Revenue of sales by geographical territory.

Sales Revenue Trends from 2014 to 2016

Monthly Sales Revenue in Dollars



Sales Overview

2016 has a much higher sales revenue compared to 2014 and 2015. There's a gradual growth apparent for 2016.

Revenue dropped **23**% from 2014 to 2015 but it grew **63**% from 2015 to 2016.

There is a significant spike in **June** for each year. **November** and **December** see an increase in revenue, likely due to holiday shopping and year-end sales events.

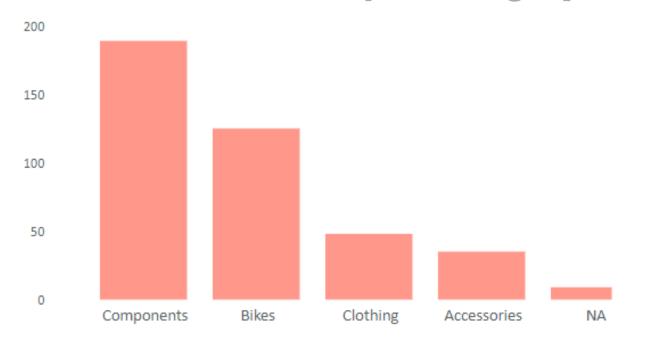
Sales Variance Table (\$)

Variance is the difference between actual and budget sales. The red rows showing negative variance, meaning the **actual sales was lower than the budget**, signifying potential low performing products.

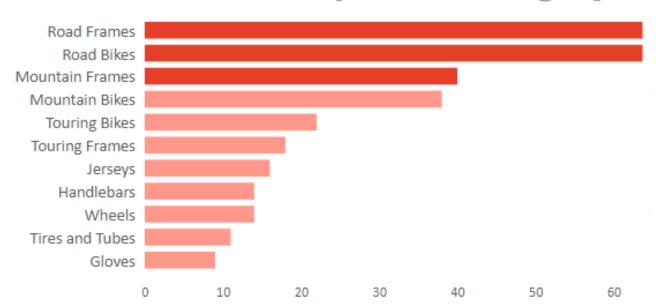
Budget Analysis

ProductName	Jan_Varia	nce Fel	b_Variance Mar	_Variance Ap	r_Variance Ma	y_Variance J	un_Variance	Jul_Variance A	Aug_Variance S	Sep_Variance O	ct_Variance N	Nov_Variance De	c_Variance
All-Purpose Bike Stand		708	3,517	3,657	4,411	4,117	4,317	2,721	3,522	3,437	5,547	4,705	2,322
Classic Vest, S		157	-774	-60	-940	-706	-945	-1,459	-2,071	-1,954	-1,916	-119	-1,793
Bike Wash - Dissolver		183	438	630	598	782	912	768	562	723	689	894	1,194
Half-Finger Gloves, S		-16	-675	-1,158	-272	-650	-374	-700	-1,254	-943	-959	-1,243	-1,227
Patch Kit/8 Patches	-3	312	-17,477	-19,767	-16,725	-20,399	-21,012	-22,684	-21,586	-19,718	-20,401	-22,555	-22,251
Touring-2000 Blue, 60	-115	262	-158,069	-157,294	-258,029	-268,261	-429,526	-283,362	-386,076	-373,211	-457,699	-485,418	-471,206
Sport-100 Helmet, Red	-1	765	-4,128	-5,264	-5,669	-4,960	-7,744	-2,619	-7,032	-2,623	-5,149	-6,676	-5,781
Fender Set - Mountain	1	404	4,250	4,075	3,907	4,006	5,869	5,210	3,598	4,428	4,464	5,895	4,841
Hydration Pack - 70 oz.	1	448	3,374	4,091	3,939	3,460	4,104	3,235	4,316	4,419	2,472	4,517	4,679
Hitch Rack - 4- Bike	1	202	3,845	4,420	3,005	3,584	3,460	2,448	2,934	4,409	2,959	6,223	4,484
Long-Sleeve Logo Jersey, S	-3	655	-8,161	-10,388	-9,110	-9,766	-14,050	-12,851	-8,854	-11,237	-12,033	-12,351	-15,184
AWC Logo Cap		531	1,509	1,451	1,950	1,184	1,937	1,868	1,916	1,851	1,551	2,093	2,470
Water Bottle - 30 oz.		696	-1,440	-412	-1,000	-850	-1,859	-1,045	-1,178	-1,257	-1,250	-866	-1,584

Number of Products per Category



Number of Products per Subcategory

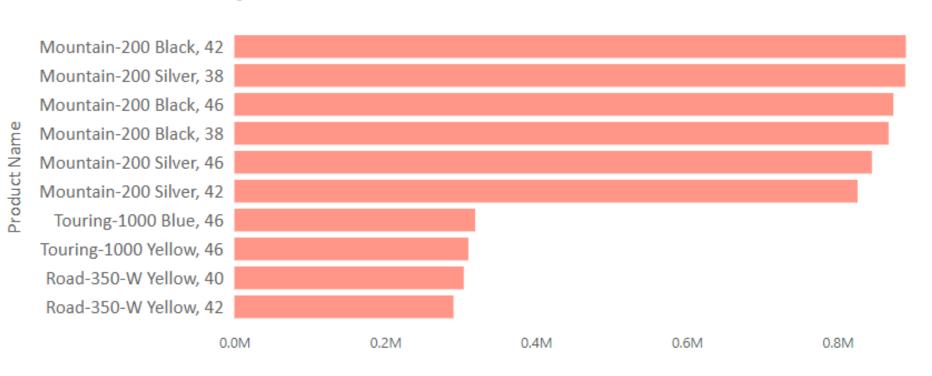


Product Performance

I analyzed the revenue generated by **Product Category**, **Subcategory**, **Product line** and **Color** of **406** products. Here I'm presenting only a few interesting results.

Top 10 Most Profitable Products in 2016

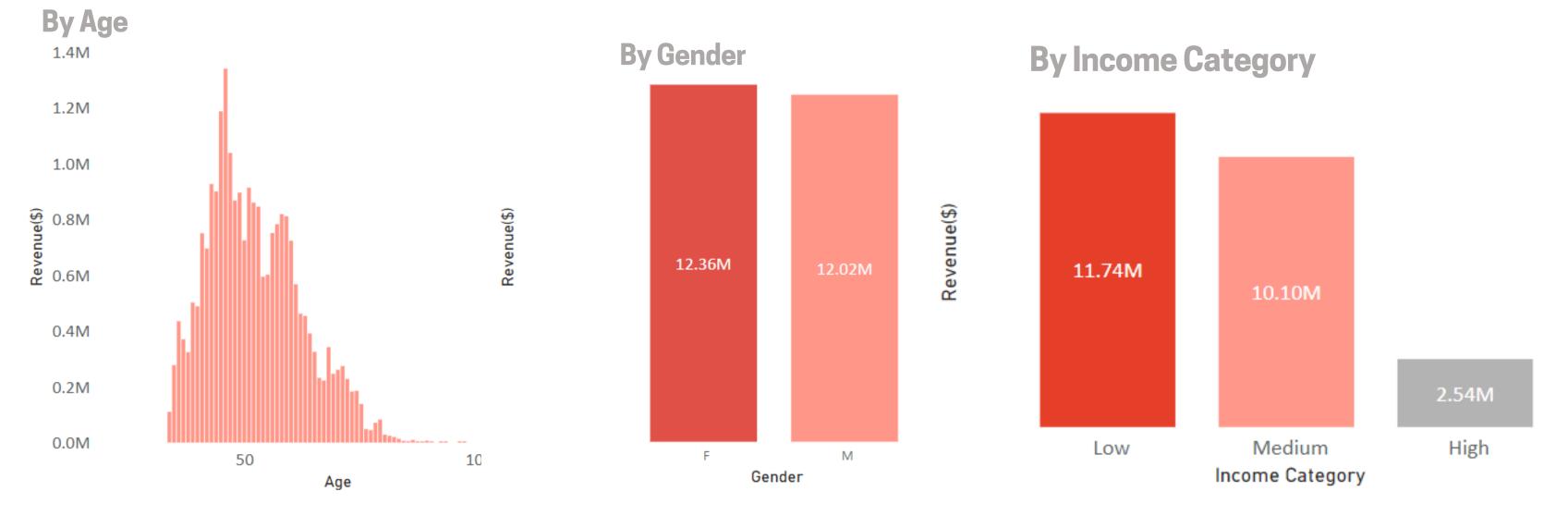
Based on Total Profit generated in Dollars.



Customer Segmentation

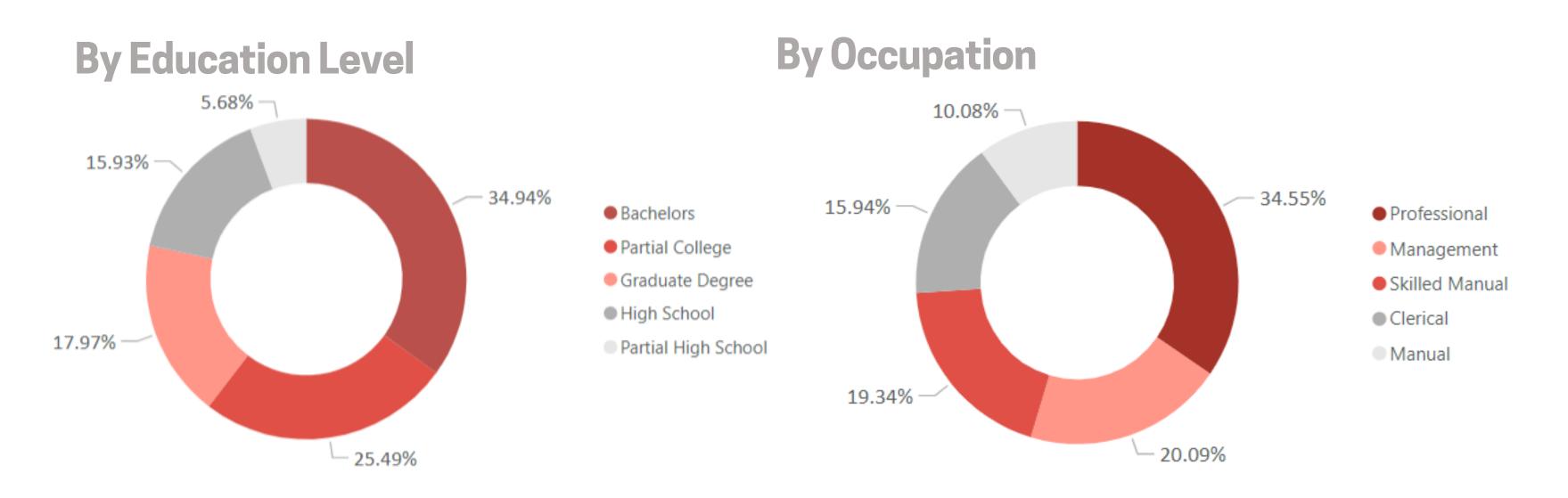
Lused these demographics to segment customers: **Gender, Age, Income Category, Marital Status, Total Children, Number of Children at Home, Education, Occupation, House Ownership, Number of Cars Owned, Country, Commute Distance** and Customer Lifetime Duration (days).

Here I present only a few interesting results.



Customer Segmentation

The largest segment in terms of Education is customers with **Bachelors' degree**, generating approximately **35%** of the total revenue. Customers with a **Professional** background also generating about **35%** of the overall revenue.



RFM Analysis

RFM stands to Recency, Frequency, Monetary value of a customer. RFM Score ranges from 3 - 15.

Recent Customers (3.1%)

- **RFM Score:** 9 12
- Characteristics: High Recency, Low to Moderate Frequency, Moderate Monetary Value
- **Behavior:** Recent buyers who may need encouragement to increase frequency.

Needs Attention (26.3%)

- **RFM Score:** 4 6
- Characteristics: Low Recency, Low Frequency, Low Monetary Value
- Behavior: At risk of churn.

At Risk Customers (2.1%)

- **RFM Score:** 3 4
- Characteristics: Very Low Recency, Very Low Frequency, Very Low Monetary Value
- **Behavior:** High risk of churn or already churned.

Promising New Customers (68.4%)

- **RFM Score**: 6 9
- Characteristics: Moderate Recency, Low Frequency, Low Monetary Value
- **Behavior:** New customers who need nurturing to increase engagement and loyalty.



AdventureWorks Sales Data Analysis (2014-2016) Dashboard

2014 2015 2016

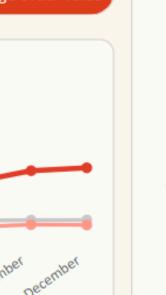
Dashboard Design by Arpita Deb - June 2024

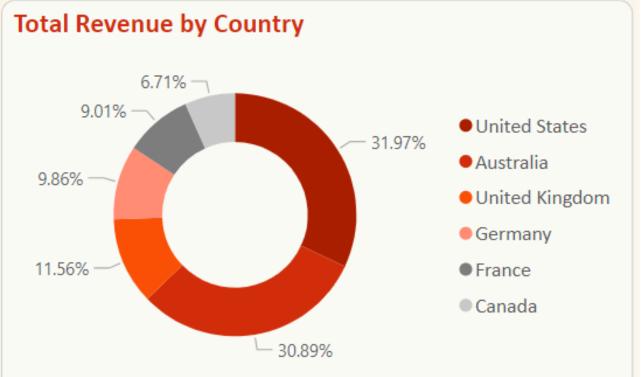
Page 1

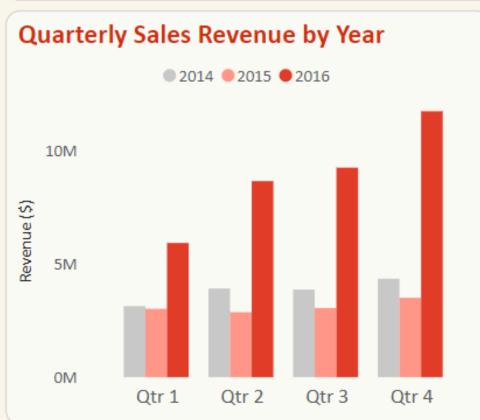
63.30M Sales Revenue

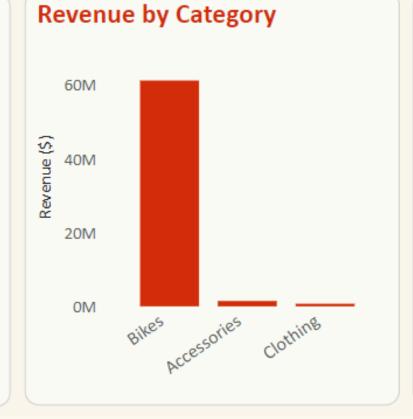
183K **Order Quantity** 24.11M **Total Profit**

543.96 Average Order Value

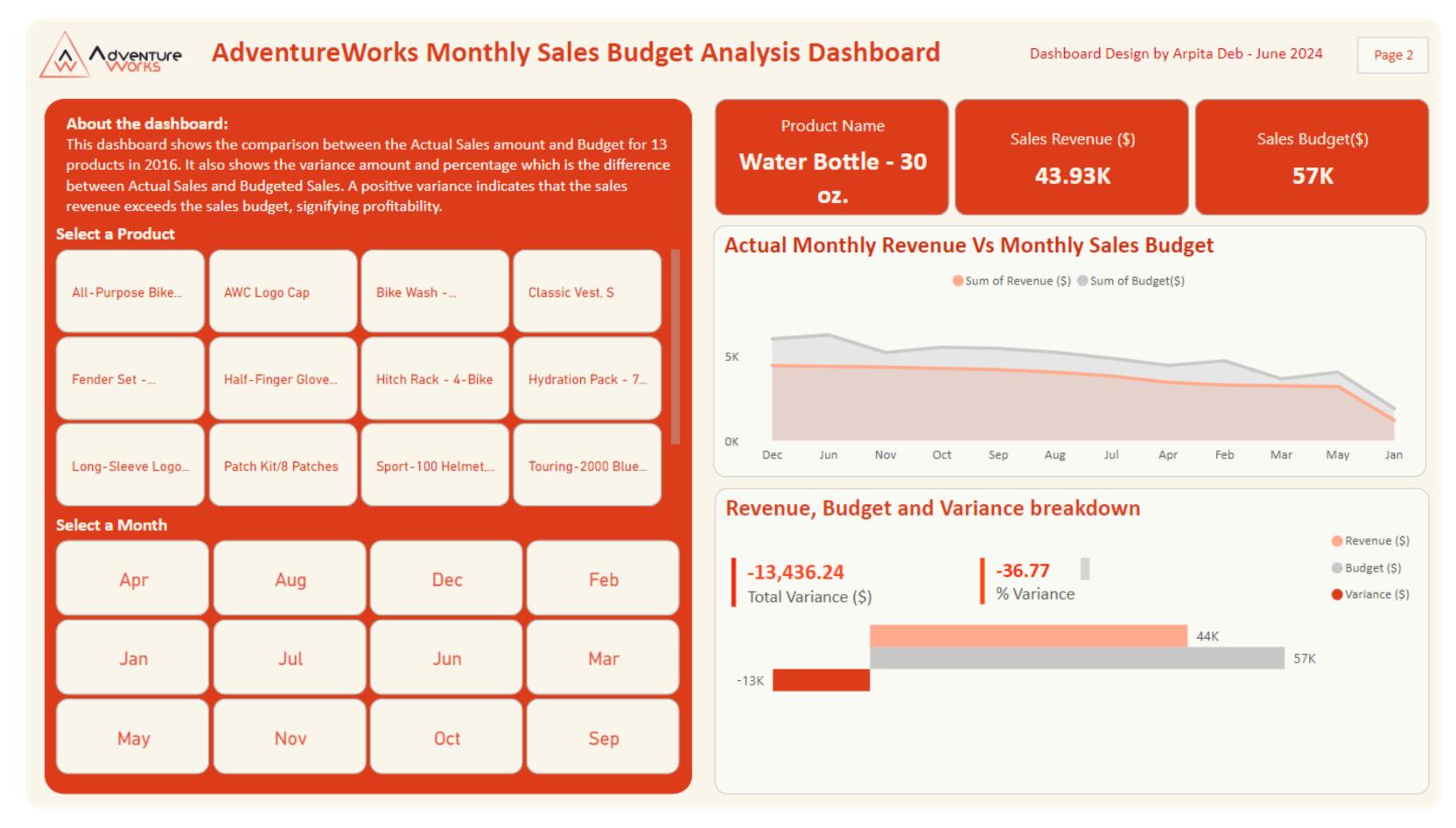








Product Name	Total Profit (\$)				
Mountain-200 Black, 38	11,80,954.92				
Mountain-200 Black, 42	12,43,519.22				
Mountain-200 Black, 46	12,53,243.15				
Mountain-200 Silver, 38	12,21,728.87				
Mountain-200 Silver, 42	11,47,023.88				
Mountain-200 Silver, 46	11,86,980.95				
Road-150 Red, 48	9,48,301.69				
Road-150 Red, 52	8,49,813.38				
Road-150 Red, 56	8,30,115.72				





Adventure Works Customer Segmentation Dashboard

Filter by year







2016

Page 3

About the dashboard: This dashboard is part of the Sales Data **Analysis of Adventure** Works Bike Retailer from 2014 to 2016 showing Revenue generated by various **Customer Segments.**

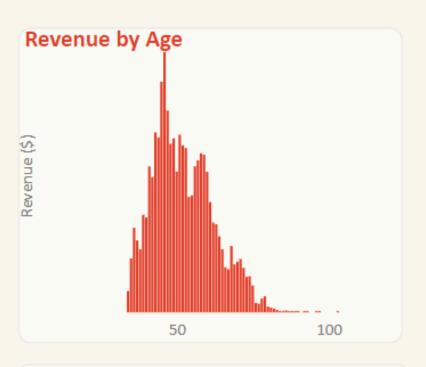
The segments are created based on customer demographics. Select any segment from the Donut or Bar Charts, or any country from the map to see the relevant results.

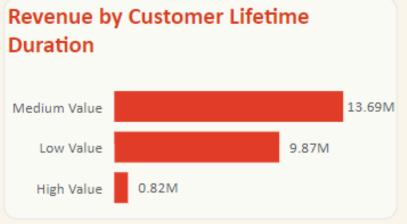
Gender:

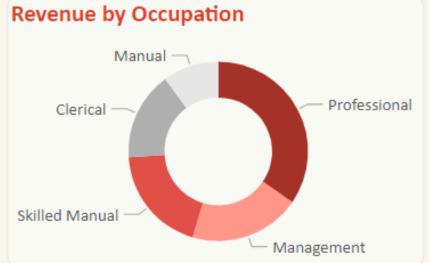
12.36M 12.02M **Marital Status:**

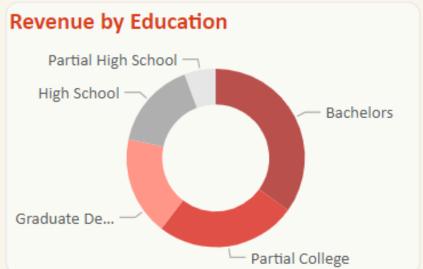
> 12.60M 11.78M

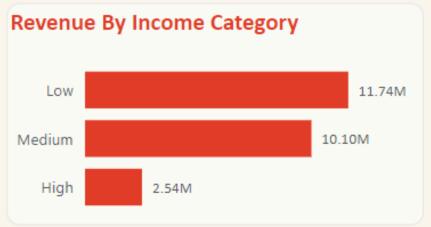
















Conclusion

- The total sales dropped 23% from 2014 to 2015 but grew 63% from 2015 to 2016. June consistently experienced a sales spike.
- 7 out of 13 products had lower actual sales revenue in 2016 than projected. Possible reasons include sales gaps in specific months and pricing differences across countries.
- Majority of customers fell within the 40 to 60 years age range. Gender had minimal impact on revenue. Low-income customers (annual income < \$63,328) are the largest segment in terms of revenue generated.
- Most customers were Promising New Buyers with low frequency and spending. Other segments being recent customers, at-risk churn, and attention-needed.
- Components category had the highest product count compared to bikes, accessories, and clothing.

Recommendations

- Focus marketing efforts on the 40-60 age group. Target High-Potential Segments such as customers with bachelor's degrees or higher education, or professional and management roles.
- 1 Implement retention strategies for skilled manual workers and clerical roles. These segments contribute less to revenue and may be at risk of churning.
- Concentrate efforts in countries with the highest customer concentration (Australia, the United States, Canada, and the UK). Investigate reasons behind lower revenue in Germany and France.
- Adjust pricing strategies based on income levels. Offer budget-friendly options for low-income customers and premium products for high-income customers.
- Since 30% of customers have no children and 60% of customers have no children at home, consider offering family-friendly products.

Limitations

Since the Budget data only contains monthly sales budget for a few products, a detailed Budget Analysis was not possible.

Due to joining of multiple datasets, I lost some amount of data points, which could have enriched the analysis.

A detailed Customer Segmentation has not been performed using Cluster Analysis or other Machine Learning Techniques due to time constraint

The data is outdated.

Next Steps

01

Advanced Customer Segmentation

Perform cluster analysis using machine learning techniques (e.g., k-means, hierarchical clustering) to create more refined customer segments.

02

Time-Series Forecasting

Predict future sales trends
using historical data.
Identify potential growth areas
or challenges.

03

Customer Lifetime Value (CLV) Modeling

Estimate CLV for different customer segments. Understand which segments contribute most to long-term revenue.

References

- AdventureWorks sample databases
- How to Download and Install AdventureWorks
 Database in SQL?
- What Is Average Order Value (AOV)?
 Definition and How to Calculate
- What Is Recency, Frequency, Monetary Value (RFM) in Marketing?
- RFM analysis for Customer Segmentation

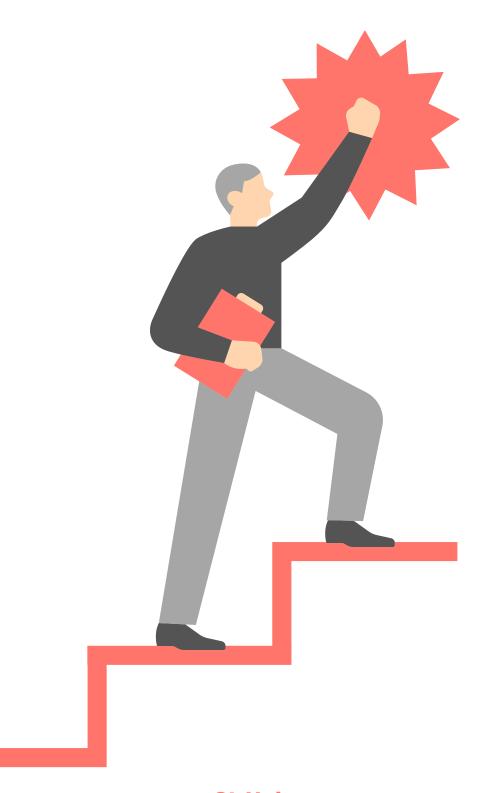
Appendices

- Problem Statement Document
- GitHub Project Documentation
- Data sets used
- <u>Data Cleaning and Transformation Changelog</u>
- Sales Budget Analysis Workflow 1 Data
 Transforming
- Sales Budget Analysis Workflow 2 Data Analyzing
- Power Bl Dashboards



Thank You

FOR YOUR ATTENTION



Presented by

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Dated

June, 2024

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