# Quantifying and Analyzing Operational Volatility

## Dell: Team 4 Capstone Project Sum/Fall 2024

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## Abstract

This project explores methodologies to quantify and analyze operational volatility within Dell Technologies' business processes. Operational volatility, defined as the degree of variability in deal closures, revenue changes, and supply chain efficiency, has significant implications for cost management and customer satisfaction. The project leverages data-driven approaches, combining advanced data preprocessing techniques, feature engineering, Data analysis techniques like descriptive analysis, exploratory data analysis (EDA), statistical analysis, prescriptive analysis, time-series analysis, anomaly detection, and data visualization to develop actionable insights and metrics. Using three-quarters of operational data, key steps included data integration, z-score normalization, and the construction of a comprehensive volatility score.

The findings highlight cyclic volatility trends, driven primarily by closure date fluctuations. The report also identifies key high volatility segments and offers strategic recommendations for process optimization, including the incorporation of real-time data for predictive modeling. This work demonstrates the feasibility of leveraging advanced analytics to enhance decision-making and operational efficiency at Dell Technologies.

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# 1. Business Understanding

## 1.1 Company Overview

Dell Technologies is a global technology leader offering innovative solutions in IT infrastructure, data storage, PCs, and servers. Established in 1984 by Michael Dell, the company has grown to operate in over 180 countries, boasting a workforce that spans across diverse geographies. Dell's comprehensive portfolio includes hardware solutions like laptops and servers, software for data analytics and security, and cloud-based services tailored for modern businesses. The company's operations are underpinned by a strong commitment to sustainability, exemplified by efforts in reducing electronic waste and advancing energy-efficient technologies. With a focus on customer-centric solutions, Dell prioritizes adaptability and innovation, maintaining its position as a trusted partner in digital transformation for businesses of all scales worldwide.

## 1.2 Project Scope and Objectives

This project aims to systematically assess and address operational volatility within Dell's business ecosystem.

The key objectives include:

- Designing a structured methodology to quantify deal volatility by identifying, analyzing, and leveraging critical operational metrics. These metrics serve as the foundation for understanding and managing fluctuations in business processes.

- Establishing and monitoring Key Performance Indicators (KPIs) to evaluate the effectiveness of volatility management strategies. By tracking these metrics, the project aims to support continuous improvement and ensure alignment with Dell's strategic goals. These KPIs provide actionable insights, enabling real-time adjustments and long-term optimizations in supply chain, sales and deal management processes.

- Extracting actionable & meaningful insights from the data & aid stakeholders in identifying high-risk scenarios and implementing targeted strategies to mitigate the potential financial and operational impacts of unpredictability.

# 2. Methodology

## 2.1 Data Understanding and Preprocessing

The data analysis was conducted using Python within the Jupyter Notebook environment, offering an interactive and iterative platform for data exploration and analysis. The methodology followed the industry-standard CRISP-DM (Cross-Industry Standard Process for Data Mining) framework.

The dataset included three quarters of operational data from Dell's current fiscal year, which begins on the last Saturday of February each year. It is comprised of two components: one file for each quarter, detailing deals week over week and another file containing all booked deals during these three quarters. These files provided insights into deal closures, products, quantities, revenue, and time-to-close metrics.

Key preprocessing steps included:

- **Data Cleaning**:
    1. Addressed missing values using the back-fill method to maintain data continuity.
    2. Removed duplicate records to prevent redundancies and ensure data accuracy.
    3. Standardized formats (e.g., date, currency, and categorical variables) to ensure consistency and improve dataset reliability.

- **Data Integration**:
    1. Merged quarterly datasets using Deal IDs and Session as primary keys, effectively resolving schema inconsistencies.
    2. Ensured a consolidated and unified data structure by aligning data types, column names, and hierarchical relationships across sources.

- **Data Transformation**:
    3. Applied Z-score normalization to standardize numerical variables and log transformation to handle skewed distributions, optimizing the dataset for subsequent analysis.

- **Data Reduction & Validation**:
    1. Selected the most relevant features using correlation analysis to eliminate redundancies and reduce dimensionality.
    2. Performed comprehensive validation checks to identify and address outliers and anomalies, enhancing overall data integrity and accuracy.

By employing these preprocessing techniques, the analysis achieved a comprehensive understanding of Dell's operational trends. This robust data foundation facilitated the identification of patterns, extraction of key metrics, and generation of actionable insights to inform strategic decision-making and process optimization within Dell's business ecosystem

## 2.2 Feature Engineering

Feature engineering in Jupyter Notebook played a pivotal role in getting to a dataset where we can easily calculate a final score. Key steps included:

- **New Feature Creation**: Using Python, features such as "time-to-close", "session week number," and "revenue changes" were programmatically extracted from given data to encapsulate critical business metrics that would then be used for a scoring methodology.

- **Normalization**: Z-score normalization was applied to scale key variables, such as "Closed Date Change" & "Log Revenue Change" and log normalization was used to create "Log Revenue Change" From "Revenue change". This process utilized Python functions to standardize these metrics, bringing us closer to our goals of creating a framework.

By utilizing Jupyter Notebook's capabilities, feature engineering steps not only improved data quality but also provided a framework for extracting actionable insights and enhancing the interpretability of Dell's operational trends.

## 2.3 Developing a Volatility Score

The Final Volatility Score serves as a cornerstone metric to quantify and monitor fluctuations in Dell's business operations. This metric was developed using a systematic approach facilitated by the computational and visualization capabilities of Jupyter Notebook.

Key steps in constructing the Final Volatility Score included:

- **Weighted Volatility Calculation**: This step calculates immediate variability in deal operations. Revenue changes and time-to-close fluctuations were scaled and weighed equally. We also provided the functionality to adjust these weights as per the business domain knowledge in our interactive Power BI dashboard, to Dell.

- **Running Volatility Analysis**: To capture temporal patterns, a rolling window mechanism was implemented using Python's data processing libraries. Specifically, a 4-week rolling window was applied to evaluate trends in revenue and closure date changes. This approach highlighted cyclical trends and smoothed short-term variability. Again, the flexibility to adjust this 4-week window was provided to the sponsors through Dashboard.

- **Integration into a Composite Volatility Score**: The weighted and running volatility measures were combined into a unified metric. This composite score offered a holistic view of operational stability, reflecting both immediate and temporal variations in business processes. The integration step utilized Python's mathematical functions to ensure accuracy and reproducibility.

The resulting Final Volatility Score provided a nuanced understanding of Dell's operational dynamics, enabling the identification of high-risk segments and the formulation of data-driven strategies to enhance process stability. Jupyter Notebook's visualization tools further supported this analysis by illustrating trends and variations in the Volatility Score across different timeframes and deal types, aiding stakeholders in strategic decision-making.

## 3. Descriptive Analytics

The descriptive analytics analysis provided key insights into the patterns and drivers of volatility within Dell's operations:

**Volatility in Booked vs. Non-Booked Deals:**
- The average volatility score for booked deals was 0.8016, significantly higher than 0.3689 for non-booked deals.
- This highlights that booked deals exhibit considerably more volatility compared to non-booked deals.

**Volatility Based on Time to Close:**
- Deals with shorter timelines of 0–3 weeks showed the highest volatility score at 0.9850, while those taking 11–15 weeks to close had a much lower score of 0.3035.

- Shorter deal cycles experience more fluctuations, indicating instability in processes requiring urgent closures.

**Quarterly Volatility Analysis:**
- The second quarter recorded the highest booked volatility at 1.12, compared to 0.89 in Q3 and 0.51 in Q1.
- Non-booked deals displayed relatively stable volatility across the quarters, peaking at 0.50 in Q2.
- These trends emphasize the cyclical nature of volatility, particularly at quarter-ends.

**Product-Specific Volatility:**
- Analysis of the top 10 products revealed significant variations in volatility.
- Products with higher deal volumes, such as Product 3 (0.32) and Product 12 (0.41), exhibited moderate volatility.
- Products with fewer deals experienced higher volatility, with Product 2 recording the highest score at 0.96.
- This indicates that demand concentration and limited deal volume increase volatility for specific products.

**Weekly Volatility Trends:**
- Week 22 saw significant volatility peaks, correlating with changes in time-to-close and revenue volatility.
- Post-peak weeks observed a gradual decline, suggesting seasonal or periodic trends in volatility.

**Distribution of Volatility:**
- 87.58% of deals fell into the medium volatility category, while 10.01% were classified as highly volatile.
- Only 12.61% of high-volatility deals were booked, highlighting the challenges high volatility poses to deal closure.

These findings provide a detailed understanding of operational volatility patterns and their drivers, forming the basis for targeted strategies to enhance stability and efficiency in Dell's operations.

# 4. Prescriptive Analytics

To address operational volatility, the following strategies can optimize product performance, improve deal timelines, and stabilize operations:

**Demand Prediction and Resource Alignment:**

- Historical trends should be analyzed to predict demand more accurately.
- Proactive adjustments in stock flexibility and pricing can dynamically align inventory and resources with volatility, ensuring better management and profitability during fluctuating demand periods.

**Targeted Deal Management:**

- Deals with short timelines (under three weeks) and extended timelines (over 27 weeks) were the most volatile.
- Segmenting deals based on timelines enables tailored interventions, enhancing operational visibility and reducing variability.

**Booking Process Standardization:**

- Booked deals exhibited a higher volatility score (0.8019) than non-booked deals (0.3689), indicating a need to address drivers like late-stage changes, inconsistent customer demands, and pricing issues.
- Standardizing and expediting the booking process is critical to reducing delays, improving deal closure efficiency, and minimizing disruptions.

By implementing these strategies, Dell can achieve greater stability, enhance process efficiency, and effectively respond to volatility within its operations.

## 5. Recommendations

To enhance Dell Technologies' ability to manage operational volatility and achieve greater efficiency, the following strategies are recommended:

**Implement Real-Time Dashboards and Predictive Models:**
- Develop real-time dashboards to forecast volatility trends and detect early signs of instability.
- Use predictive analytics to optimize inventory turnover based on volatility patterns, preventing overstocking or stockouts.
- Enable proactive decision-making by providing stakeholders with up-to-date insights into operational metrics.

**Automate Workflows and Service Level Agreements (SLAs):**
- Automate workflows and SLAs to streamline operations and reduce manual intervention.
- Leverage historical volatility data to align staffing and resources with peak periods of deal activity and closure times.
- This approach minimizes delays and inefficiencies, ensuring resources are optimally allocated to meet business demands effectively.

**Standardize Deal Closure Processes and Enhance Customer Communication:**
- Use volatility insights and tools like Salesforce Forecasting to improve the accuracy of delivery time predictions.
- Standardize deal closure processes to reduce delays and boost operational efficiency.
- Enhance customer communication with transparent updates, improving satisfaction and fostering trust.

**Strengthen Supplier Partnerships:**

- Establish contracts with suppliers that are adaptive to volatility-driven demand fluctuations.
- Foster seamless collaboration and stronger partnerships by ensuring flexibility in supply chain agreements.
- This strategy strengthens resilience and ensures consistent operational performance during periods of fluctuating demand.

## 6. Challenges

The project faced several limitations that highlight areas for improvement and opportunities for future exploration. A key challenge was the **lack of industry-specific adjustments**, where the model did not account for critical factors such as regional or seasonal variations, limiting its applicability in specialized domains. Additionally, the reliance on **simplistic assumptions in weighting and scoring** treated all components as equally important, overlooking the varying impacts of short-term and long-term fluctuations.

Another limitation was the **absence of external factors** such as market trends, customer behavior, and macroeconomic conditions, which could provide a more comprehensive understanding of volatility. Furthermore, the project adopted a **static analysis approach**, focusing only on historical data without leveraging predictive techniques to forecast future outcomes.

Additional challenges included the **simplicity of the data** provided, which was highly streamlined and cleaned compared to real-world datasets that typically require extensive preprocessing. While the dataset included three quarters' worth of data and 2.4 million records, there were **limited data features** available, which constrained the scope of analytics and reduced opportunities for deeper insights. Finally, transitioning from **Tableau to Power BI** presented an initial challenge as the team had to adapt to the new tool. But considering BI costs 7 times less than its counterpart, it is highly capable.

## 7. Concluding Remarks & Future Directions

**Operational Insights:**
> This project successfully quantified and analyzed operational volatility at Dell Technologies, identifying key patterns and drivers such as cyclic trends, shorter deal timelines, and product-specific demand fluctuations. These insights enable a more nuanced understanding of volatility, forming the foundation for strategic decision-making and process improvements.

**Impact on Business Efficiency:**
> By developing a Final Volatility Score and integrating it into real-time dashboards, the project provides actionable metrics for enhancing operational stability, improving resource management, and aligning inventory and pricing strategies with demand variability.

**Enhanced Predictive Modeling:**
> Future research could incorporate machine learning models with industry-specific adjustments, such as accounting for regional, seasonal, and macroeconomic factors, to improve the accuracy of volatility predictions and optimize strategic planning.

**Broader Application of Volatility Metrics:**

Expanding the analysis to other business units, including supply chain and customer service, can identify additional volatility drivers, allowing Dell to implement comprehensive, cross-functional strategies for operational resilience and efficiency.

## 8. Acknowledgments

First, we would like to express our sincere appreciation to our sponsor Mr. Joseph Collins at Dell Technologies / EMC for providing us with the project and working with us, it was an absolute pleasure. We greatly appreciate his time, guidance & insights, without which we would never have achieved this milestone.

Next, we would like to say thank you to our professors at the Manning School of Business, University of Massachusetts Lowell. Professor Tao Gao, our mentor, helped unravel some of the complexities of this capstone during the execution of the project and Professor Luvai Motiwalla, our project coordinator, who provided work and time frames for the project. We appreciate their efforts in the success of our capstone.

Overall working on this capstone project has been a great experience and helped us in learning more about supply chain management, team & time management. The journey of completing this project, though filled with challenges, has been profoundly enlightening. Each challenge became an opportunity to enhance our problem-solving abilities and adapt to dynamic situations. Working with Dell's supply chain data offered invaluable insights into the immense potential of leveraging data-driven decision-making. This capstone project has been a transformative and educational experience, bridging the gap between theoretical knowledge and practical application. We are deeply grateful for the opportunity to tackle real-world challenges and look forward to applying these invaluable lessons in our future endeavors.
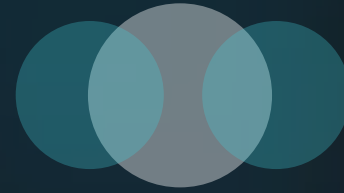
# Quantifying and Analyzing Operational Volatility

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- **Company Overview**: Dell Technologies is a global technology leader offering innovative products and services, including PCs, servers, data storage, and IT infrastructure.

- **Founding**: Established by Michael Dell in 1984.

- **Global Reach**: Operates in over 180 countries.

- **Core Strengths**: Focuses on customer-centric solutions, sustainability, and technological innovation.

- **Reputation**: Trusted partner for digital transformation by businesses and individuals globally.

# PROJECT OVERVIEW & GOALS

- Develop a methodology to **quantify deal volatility** by identifying key metrics.

- **Provide actionable insights** into deals stability through the Volatility Score, helping stakeholders identify high-risk deals because volatility or unpredictability can be expensive.

- **Establish KPIs** to track the effectiveness of volatility management and provide actionable insights for continuous improvement.
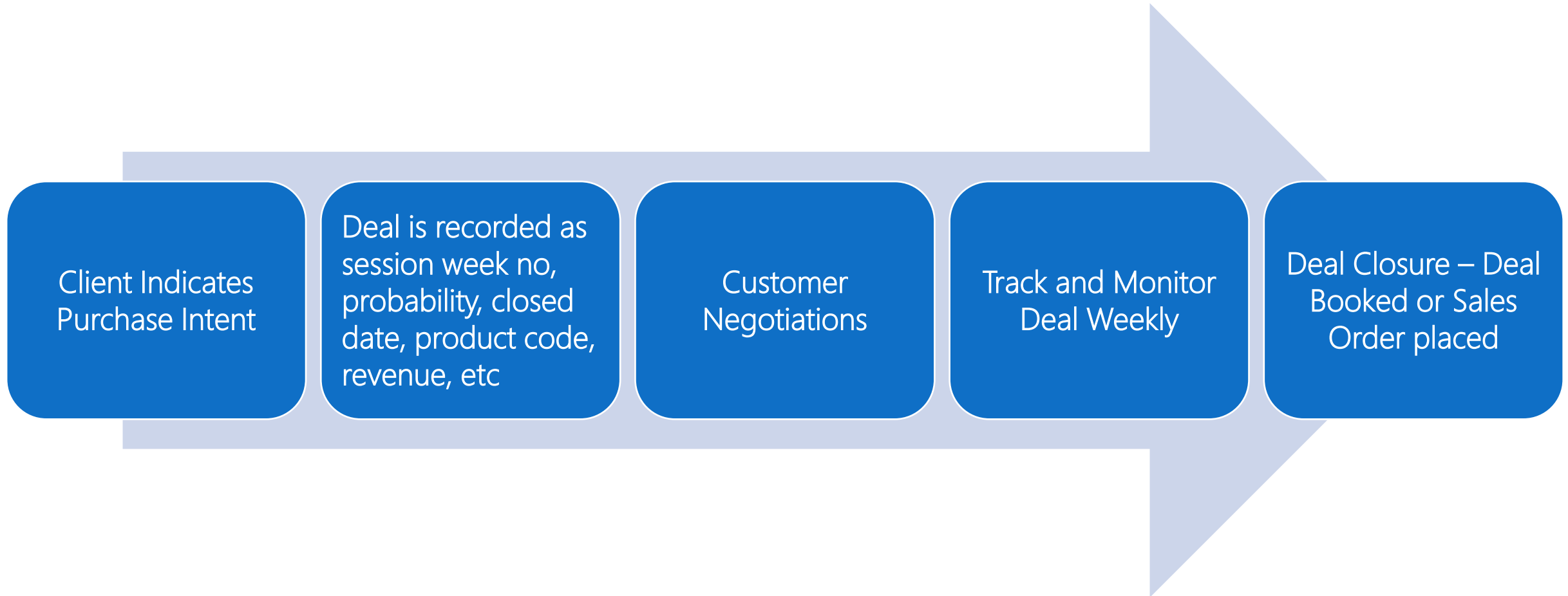
# METHODOLOGY

Cross Industry Standard Process for Data Mining (CRISP-DM)



**Business Understanding**
- Goals & Objectives
- Business Process

**Data Understanding**
- Data Exploration
- Data Quality, patterns & Anomalies

**Data Preprocessing**
- Data Cleaning using Python
- Transformation
- Feature Engineering

**Evaluation**
- Volatility Scoring
- Trend Analysis
- Data Validation & Accuracy
- Business Impact

**Deployment**
- PowerBI Reports & Dashboards
- Data Driven Insights
- Actionable Recommendations

DATA

4

# BUSINESS PROCESS



Client Indicates Purchase Intent

Deal is recorded as session week no, probability, closed date, product code, revenue, etc

Customer Negotiations

Track and Monitor Deal Weekly

Deal Closure – Deal Booked or Sales Order placed

# DATA PREPROCESSING

## DATA COLLECTION

Received 3 quarters worth of data, one file for each quarter and a file for booked deals

## DATA INTEGRATION

Merge/Appended quarterly data based on Deal ID's.
Resolve Schema Differences, Standardize column names, data types, and formats.

## DATA REDUCTION

Feature Selection: Identified and retained only the most relevant features. Reduction using techniques like aggregations to reduce the number of features while retaining essential information.

## DATA CLEANING

Handled Missing Data, Imputed missing values using back-fill method, Identified and eliminated duplicate records. Corrected Inaccuracies by standardizing formats.

## DATA TRANSFORMATION

Normalization and Standardization based on z score.
Feature Engineering: Created new features and modified existing ones to capture important patterns by calculating time to close, closed date week no., session week no.

## DATA VALIDATION

Verified data consistency, accuracy, and completeness. Checked for outliers or anomalies that may affect analysis.
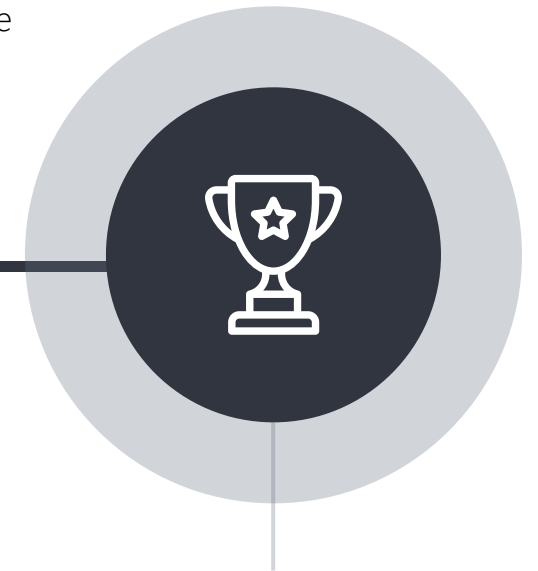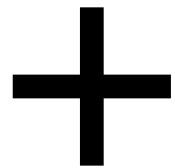
# VOLATILITY SCORE

The degree of variation or fluctuation in a value over time. In the context of business, it measures how much metrics like revenue, deal closure time, or customer demand change. High volatility means frequent or large changes, while low volatility indicates more stability and consistency.

## RUNNING VOLATILITY

Measures the rolling standard deviation of scaled closed date changes and revenue changes over a 4-week window, capturing the fluctuation for each deal.

**+**

## WEIGHTED VOLATILITY

Combines adjusted z-scores of scaled closed date changes and revenue changes using equal weights to assess immediate deal variability with respect to all the deals.

**→**

## FINAL VOLATILITY

The sum of running volatility and weighted volatility, providing a comprehensive measure of overall deal volatility.

# DESCRIPTIVE ANALYTICS

## 0.8016
Average Volatility Score for Booked Deals.

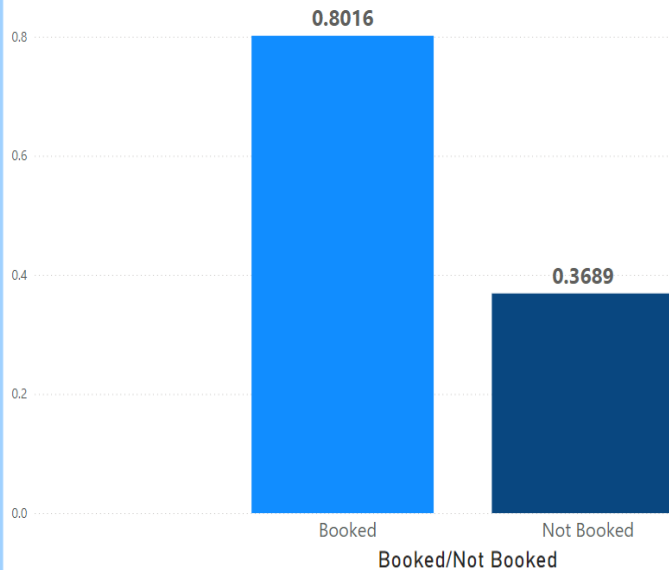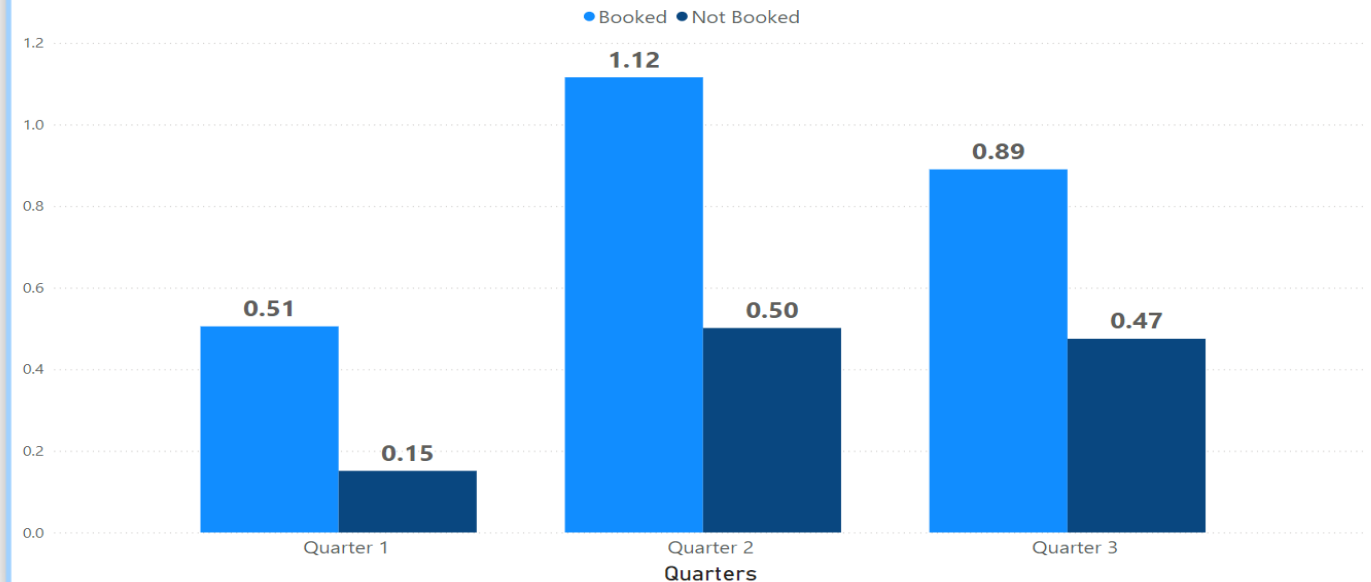| Time to close | Volatility Score |
|---|---|
| 0-3 | 0.9850 |
| 3-7 | 0.7544 |
| 7-11 | 0.5312 |
| 11-15 | 0.3035 |

## 0.3205
Average Volatility for the Most In-Demand Product (3).

**Average Volatility for Booked and Not Booked Deals**

- Booked: 0.8016
- Not Booked: 0.3689

**Quarterly Volatility**
- ● Booked  ● Not Booked

| | Booked | Not Booked |
|---|---|---|
| Quarter 1 | 0.51 | 0.15 |
| Quarter 2 | 1.12 | 0.50 |
| Quarter 3 | 0.89 | 0.47 |

Top 10 Products' Volatility
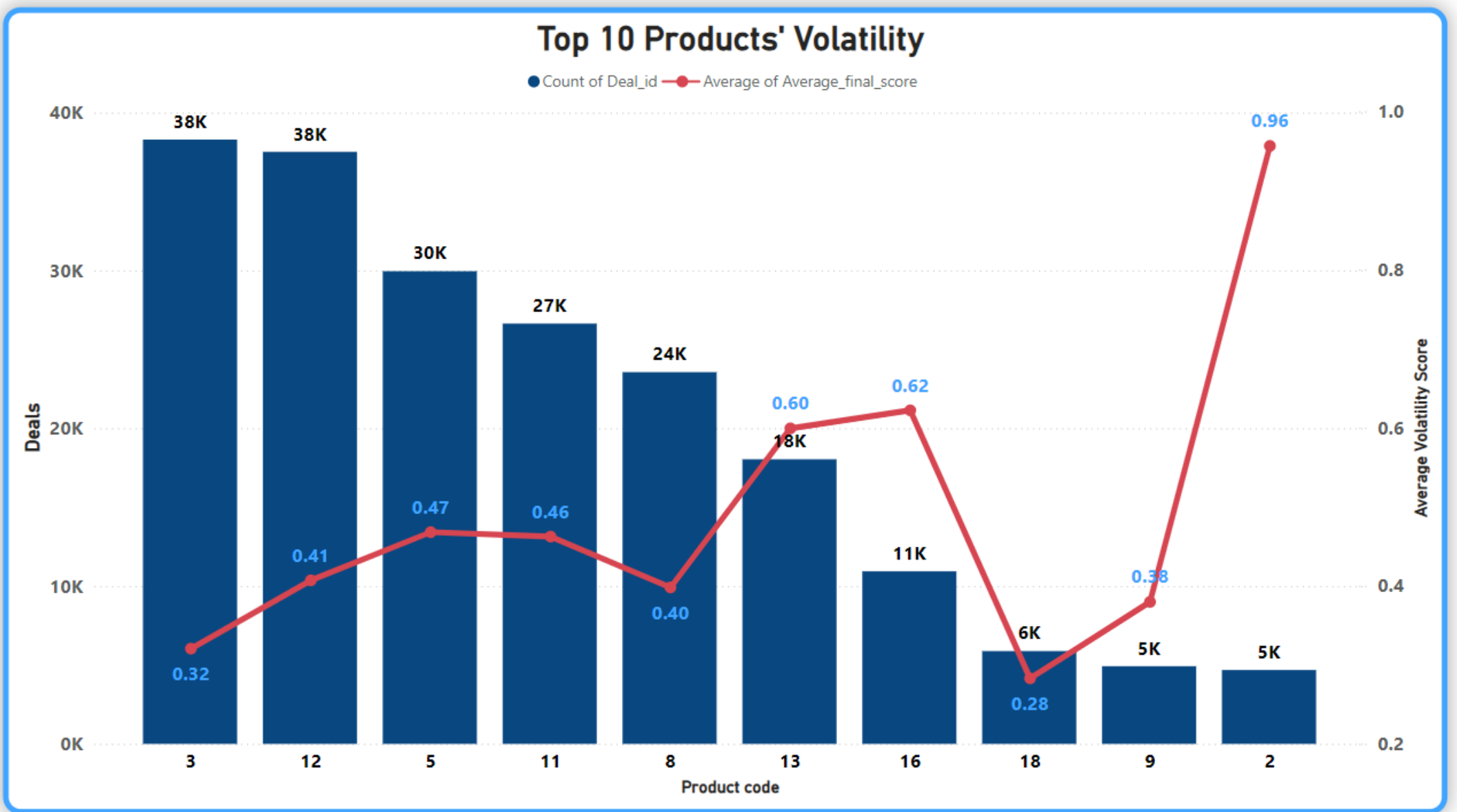
# DATA DRIVEN INSIGHTS

**67%**
VOLATILITY

Booked deals exhibited 67% higher volatility compared to not booked deals.

**89.90%**
NOT BOOKED DEALS

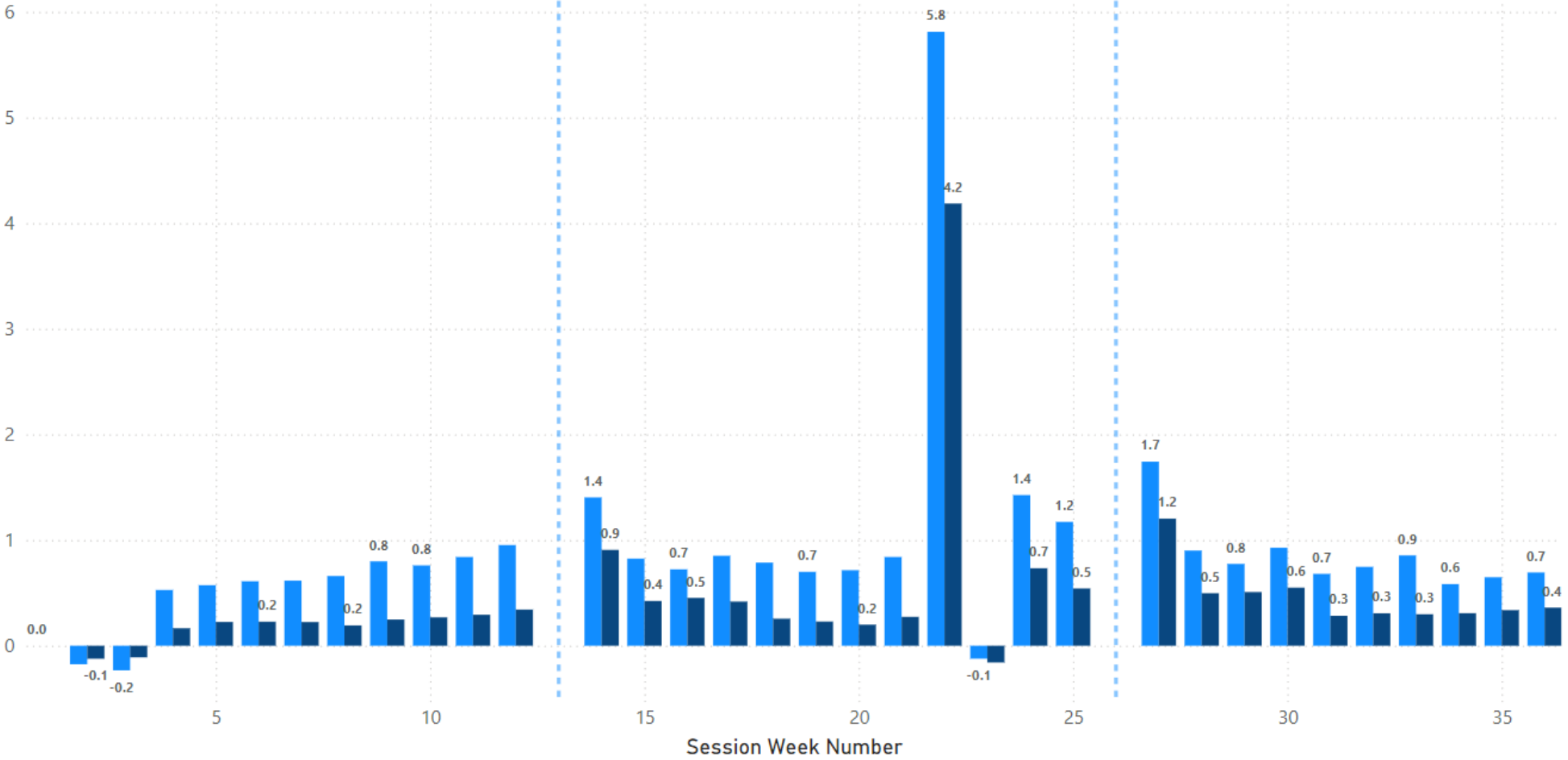89.9% of the deals were not booked over the last three quarters.

**0.96**
PRODUCT VOLATILITY

In the last 3 quarters, Product 2 showed greater volatility than the other top 10 products.
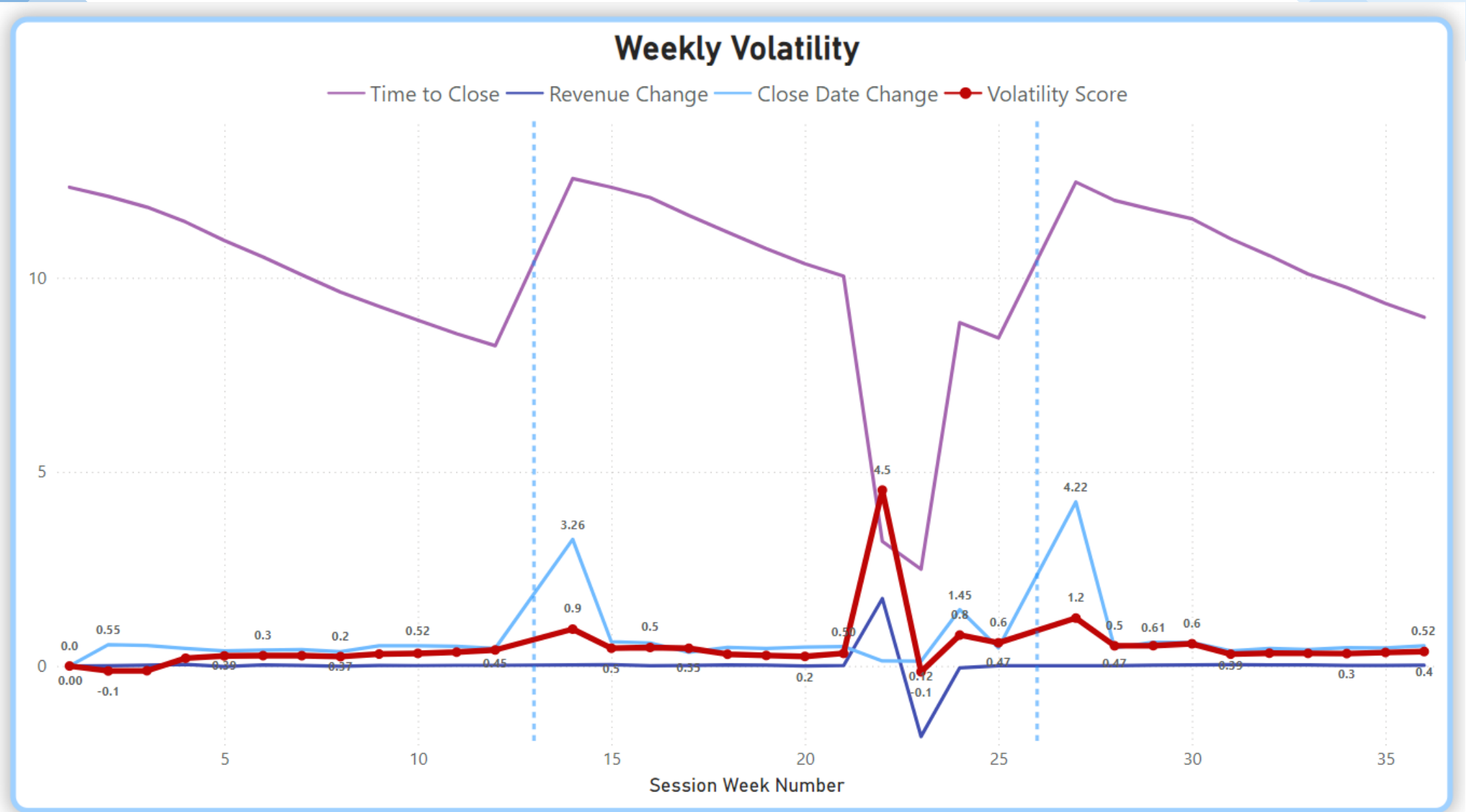


Weekly Volatility for Booked and Not Booked Deals

● Booked ● Not Booked

# DATA DRIVEN

- Significant peaks occur at the end of quarters, showing the highest fluctuation.

- Closed Date Change in correlation to Time to close are the primary contributor to volatility.

- Time to Close and Revenue Changes remain relatively stable, indicating that volatility is largely driven by deal closure dates.

- Around Week 22, a significant change in Time to Close led to an increase in revenue volatility.

- Volatility decreases after the peaks, suggesting the fluctuations may be seasonal or periodic.



Weekly Volatility

Distribution of Volatility for Booked and Not Booked Deals
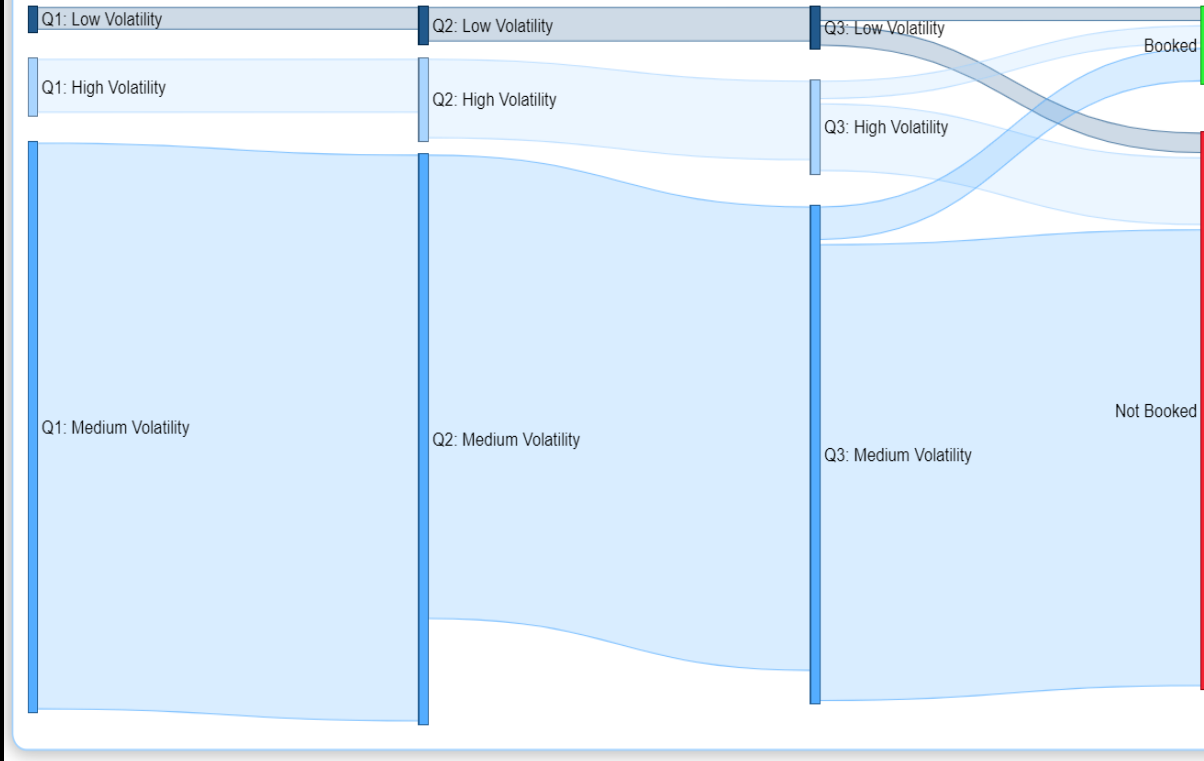
**Low**
👥 2.42%

**Medium**
👥 87.58%

**High**
🧑 10.01%

- There are 2.42% low volatile deals, 87.58% Medium Volatile Deals and 10.01% Highly Volatile deals
- Out of 10.01% High Volatile Deals 12.61% were Booked
- Volatility decreases as the initial time to close increases, with the highest volatility (0.9850) in 0-3 weeks and the lowest (0.0350) in 23-27 weeks.
- Most deals (3-15 weeks) have moderate volatility, with the peak volume in the 3–7-week range.

Initial Time to Close For Booked and Not Booked Deals
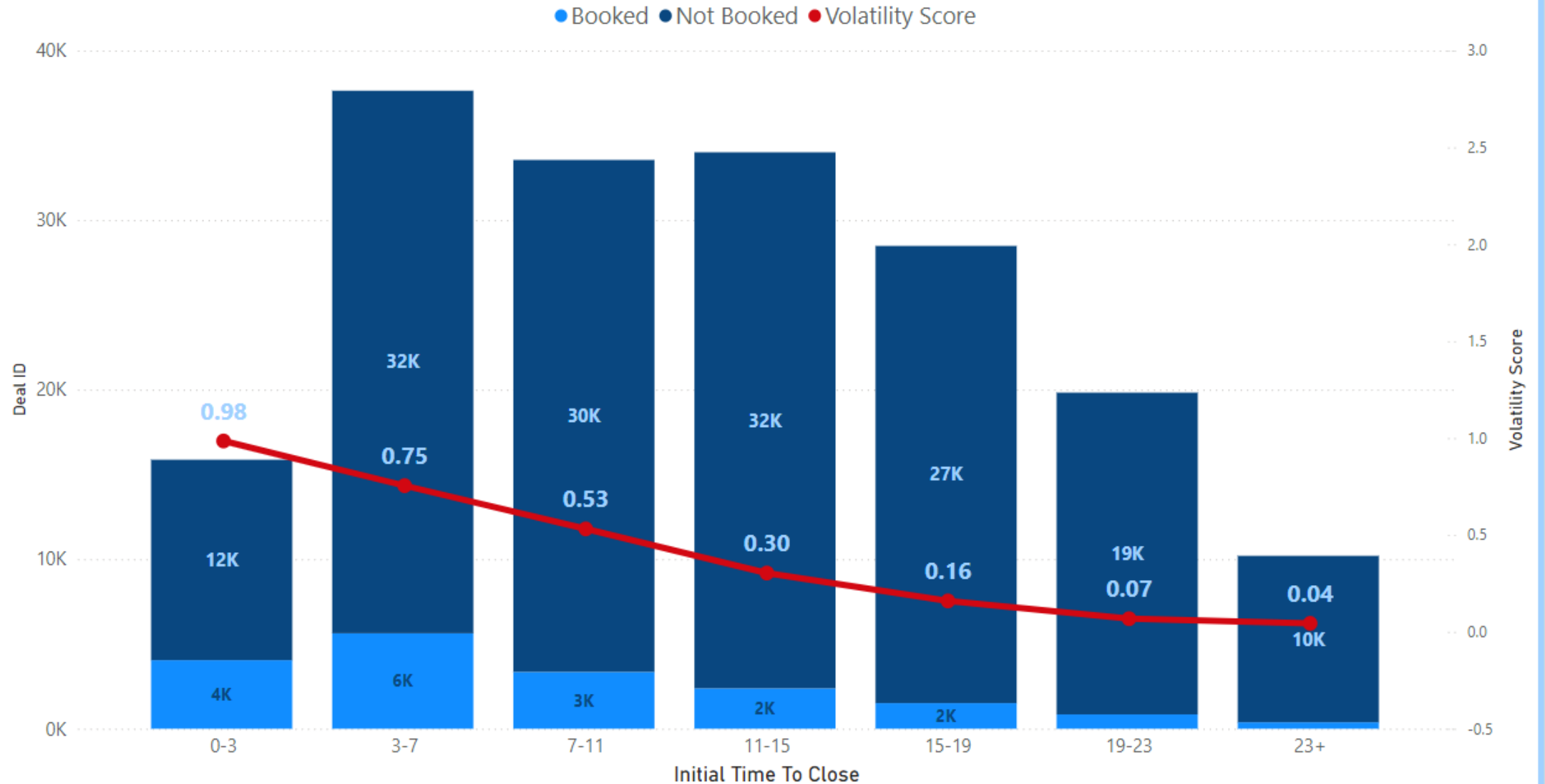
# PRESCRIPTIVE ANALYTICS

## OPTIMIZE PRODUCT VOLATILITY

- **Leverage historical trends** and volatility to predict demand
- **Increase Stock Flexibility:** Implement agile restocking during peak demand.
- **Dynamic Pricing:** Adjust prices based on demand and volatility to maximize profitability.

## OPTIMIZE VOLATILITY ASSOCIATED WITH DEAL VISIBILITY

- **Optimize Deal Timelines:** Streamline processes to enhance efficiency, focusing on managing low(<3 weeks initial time to close) and high(>27 weeks initial time to close) deals which are prone to be more volatile.
- **Tiered Deal Management:** Categorize deals by time to close and apply tailored strategies.

## ANALYZE VOLATILITY WITHIN BOOKED AND NOT BOOKED

- **Volatility score for not booked deals are 0.3689 and booked deals are 0.8019**
- **Identify Contributing Factors:** Analyze the key drivers of volatility in booked deals, such as late-stage changes, customer demands, or pricing fluctuations.
- **Streamline Booking Process:** Standardize and expedite the booking process to minimize delays and reduce volatility in deal closures.

# RECOMMENDATIONS

**Implement real-time dashboard and predictive models**

Forecast volatility trends and detect early signs of instability. Focus on optimizing inventory turnover based on volatility patterns to prevent overstocking or stockouts.

**Automate workflows and Service Level Agreements (SLA)**

Leveraging historical volatility data to align staffing and resources with peak periods of deal activity and close times, reducing delays and inefficiencies.

**Standardize deal closure and customer communication**

Use volatility insights and customer-centric tools like Salesforce Forecasting to accurately predict delivery times and reduce delays.

**Strengthen Supplier Partnership**

Establish volatility-driven contracts that adapt to demand fluctuations, fostering seamless collaboration and stronger partnerships.

# LIMITATIONS

1. **Lack of Industry-Specific Adjustments:** Model fails to account critical industry-specific factors like regions or seasons, limiting its practical utility in specialized domains.

2. **Simplistic Assumptions in Weighting and Scoring:** Equal weights are used for key components like Weighted Volatility and Historical Volatility to calculate the Final Volatility Score. This assumes equal importance of short-term and long-term fluctuations without empirical validation.

3. **No Consideration of External Factors:** Does not incorporate external factors like market trends, customer behavior, supplier behavior, or macroeconomic conditions.

4. **Static Analysis Approach:** This project relies on current fiscal year volatility metrics but does not use predictive modeling techniques to forecast future deal outcomes or changes in volatility.

# THANK YOU

Sponsor: Mr. Joseph Collins

Faculty: Prof. Luvai Motiwalla

Mentor: Prof. Tao (Tony) Gao

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# QUANTIFYING & ANALYZING OPERATIONAL VOLATILITY

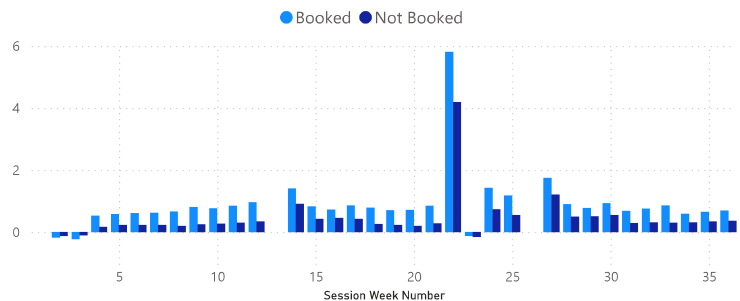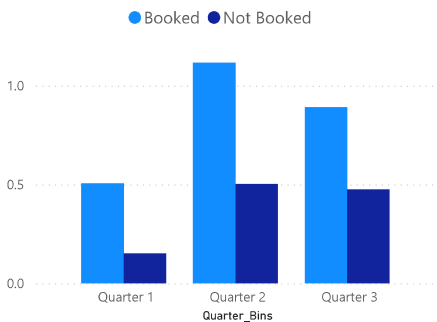| Session Week No. | Deal ID | Close Date Change | | Booked/Not Booked | Time to Close | |
|---|---|---|---|---|---|---|
| All | All | -62 | 209 | All | 0 | 261 |

## Weekly Volatility

Time to Close — Revenue Change — Close Date Change — Volatility Score



Session Week Number

## Weekly Volatility for Booked and Not Booked Deals

Booked — Not Booked



Session Week Number

## Quarterly Volatility

Booked — Not Booked



Quarter_Bins

## Initial Time to Close For Booked and Not Booked Deals

Booked — Not Booked — Average of Final_Volatility_Score



Initial Time To Close

## Top 10 Products' Volatility

Deal Id — Volatility Score



Product Code

## Distribution of Volatility for Booked and Not Booked Deals