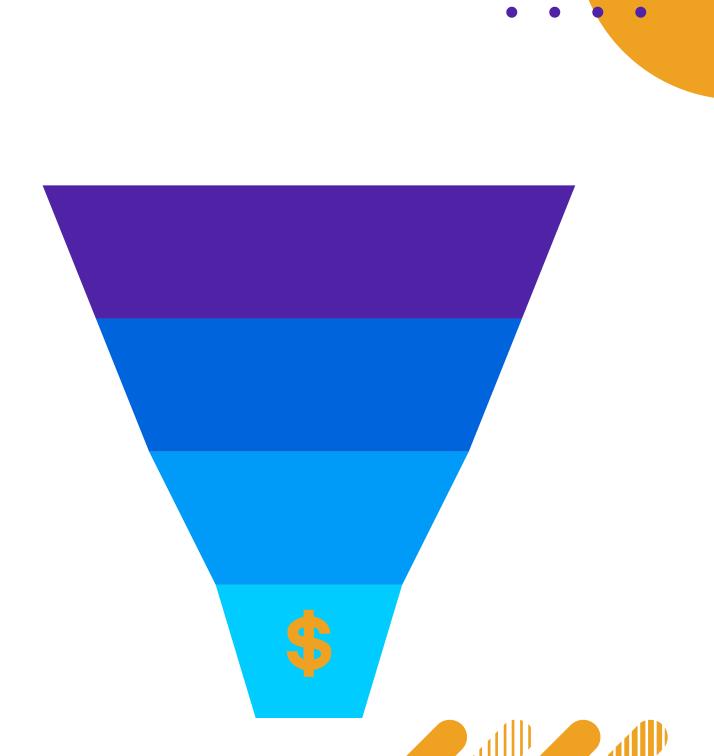
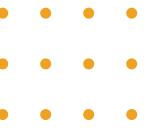


## Sales Forecasting

Sunny, Harsha, Dinesh, Pranjal







#### **Problem Statement**

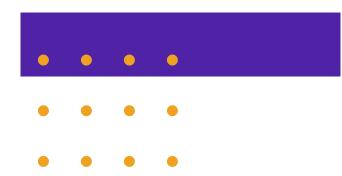
Are we on track to meet our quarterly sales target?

#### **Objective**

Predict the deal outcome early in the quarter to have better revenue predictability and enable insights for timely decision making

#### **Potential Challenges**

- Data threading across stages of the sales funnel
- Data volume will this require distributed ML?
- Feature selection given the potentially wide range of available attributes
- Learn the Art of Data Storytelling present insights in a meaningful and convincing way







#### **Team Members and Tentative Responsibilities**

Sunny Mattas : Data Preparation and Model Building

Harsha Bompalli Mutt: Feature Selection and Model Tuning

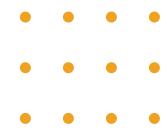
Dinesh Kumar : EDA and Visualization

Pranjal : Model Tuning and Results Interpretation

#### Inspiration

Many companies have been missing their financial targets (one of the many reasons for recent layoffs in the software industry).

Based on our general readings, we realized that companies need more predictability of their revenues and better understand the controllables to prevent revenue leakages.



#### **Activities Involved**

## Data preparation & Data Cleaning

- 6 quarters of sales pipeline data
- All open opportunities as of any given week of the quarter is the starting dataset
- Measurement of the outcome of these opportunities by the end of the quarter
- Data cleanup removing bad data



## Exploratory Data Analysis

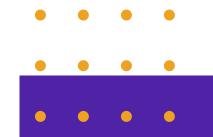
- Visualization
- Examining Relationships
- Examining distributions
- Outlier Detection and Treatment

#### Feature Engineering

- Correlation Analysis
- Feature Selection
- Define binary Class
   Variable as
   opportunity Won Vs
   Not Won (Lost + Open at the end of the quarter)

#### Model Development

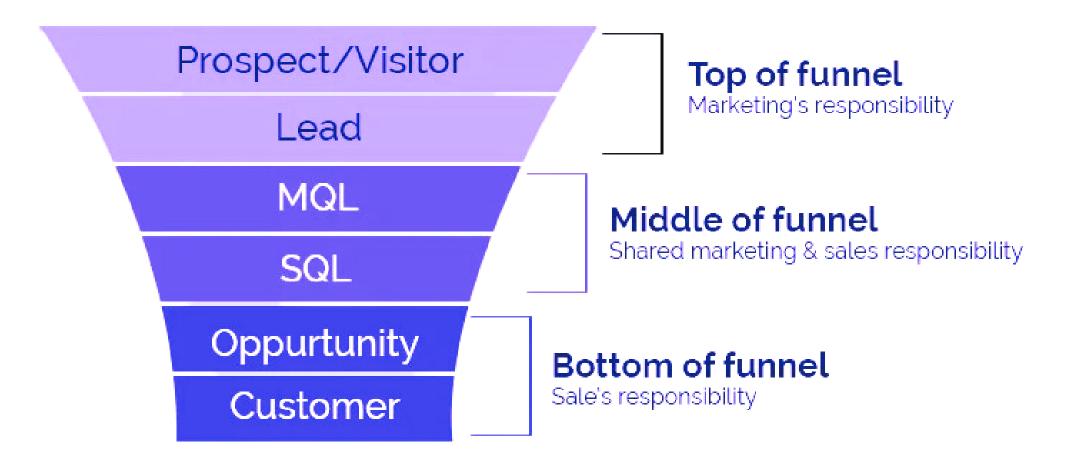
- Divide dataset into training and testing datasets - 5 quarters of training data and most recent quarter of testing data
- Predict the probability of winning opportunity by the end of the quarter
- XGboost

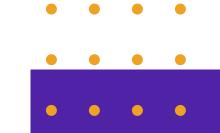


#### **Sales Pipeline Overview**

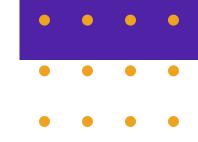
1. Prospect → 2. Discover → 3. Pursue → 4. Prove → 5. Propose & Contract → 6. Closed

#### **MARKETING & SALES FUNNEL**

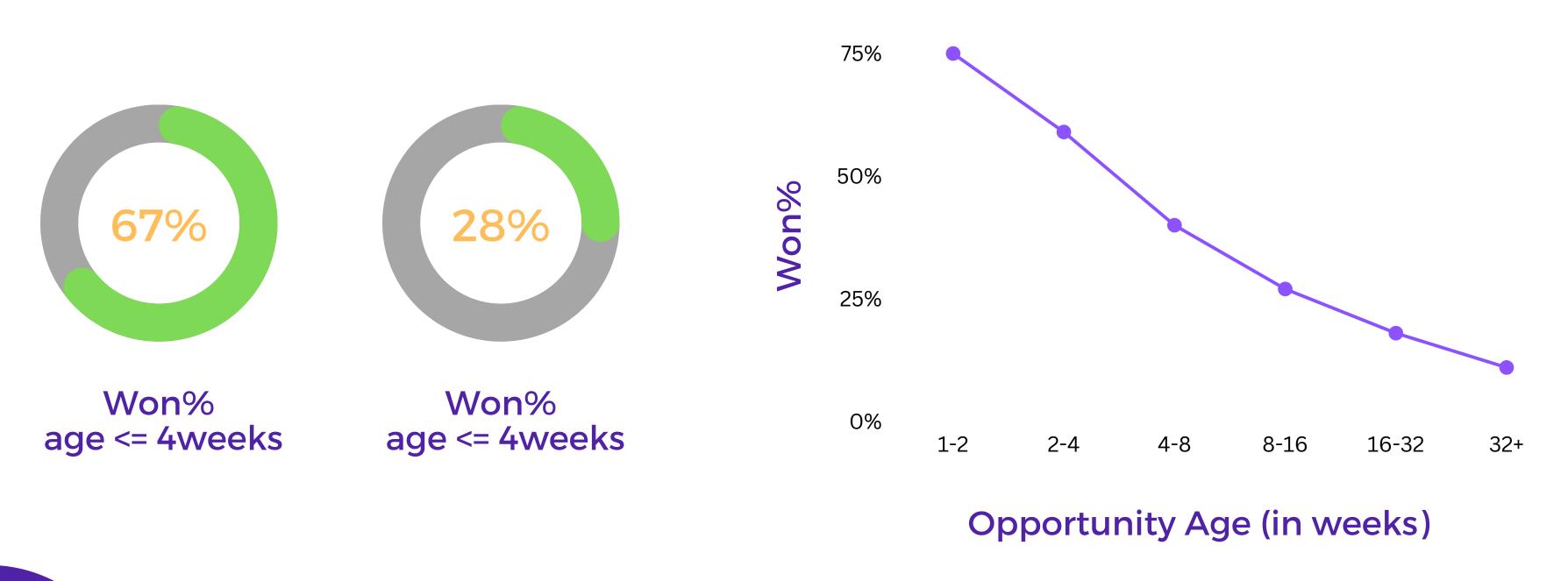


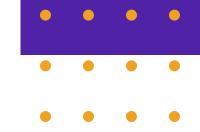






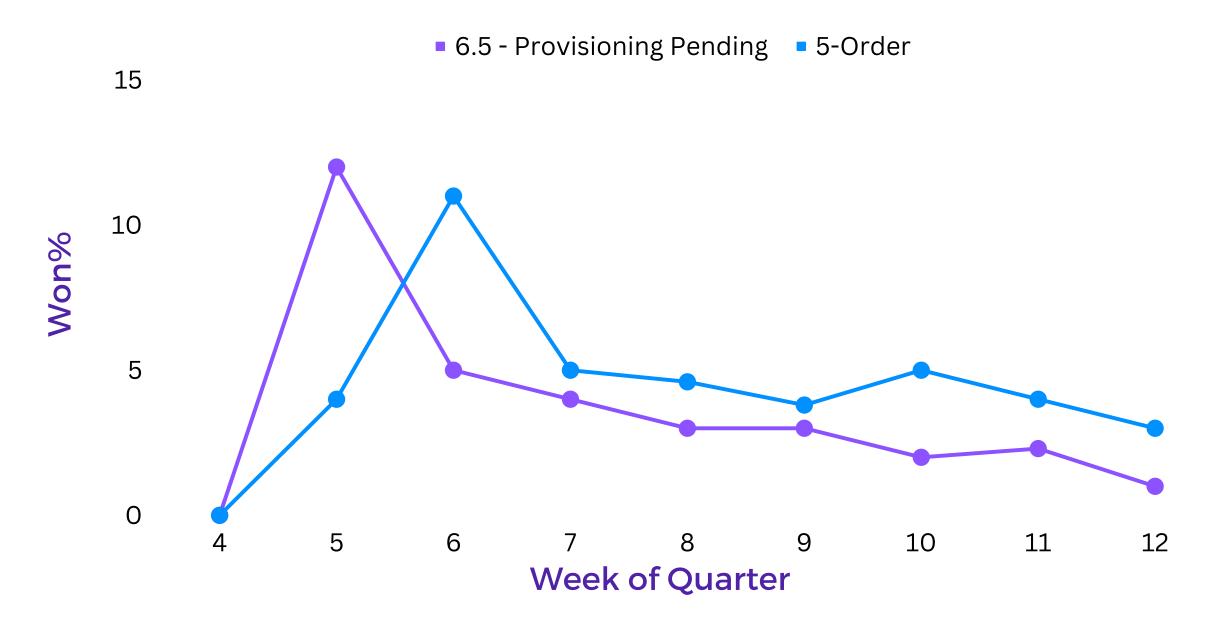
There is 2.5X higher probability for an opportunity to be WON if closed in less than 4 weeks





#### **Key Findings**

Probability of an opportunity to be WON decreases if it stays in the same stage (especially after stage 4-negotiate) longer than ~2 weeks

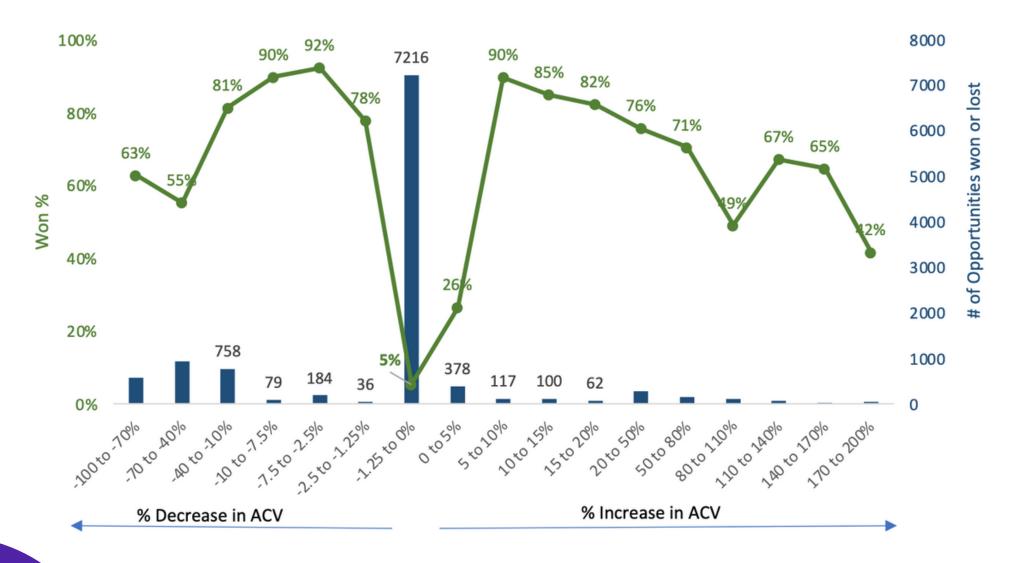


Won % for opportunities open in respective stage as of week 3

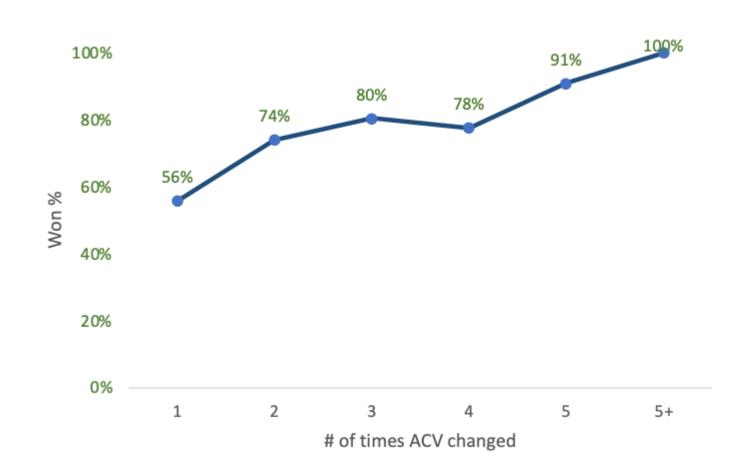


#### Active engagement with customers results in a Win-Win Scenario!!

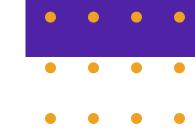
- On an average, 85% opportunities are won if ACV was either increased or decreased within a range of 5-40%.
- Only 5% of the opportunities are won when ACV remained unchanged.



 An opportunity has 31% more chance of winning when it undergoes more than 2 negotiations during its lifecycle







#### **EXISTING** base matters while we pursue NEW logos

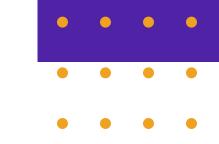
- Opportunities for EXISTING customers has 2.5X higher win-rate compared to NEW customers
- Opportunities with "Product" as part of ELA has 2X higher win-rate than those where it was sold a-la carte

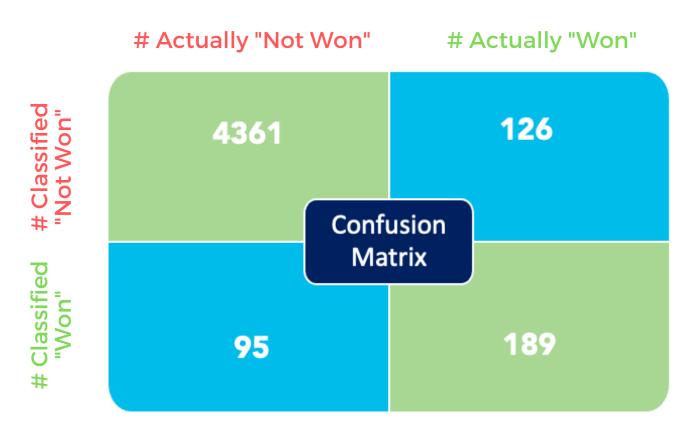


VS









Confusion Matrix based on a classification threshold that maximizes fl score on the test data

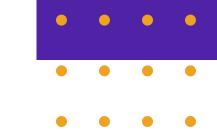
**Model Accuracy Score 95.4%** 



Confusion Matrix based on a classification threshold that maximizes maximizes TPR on test data

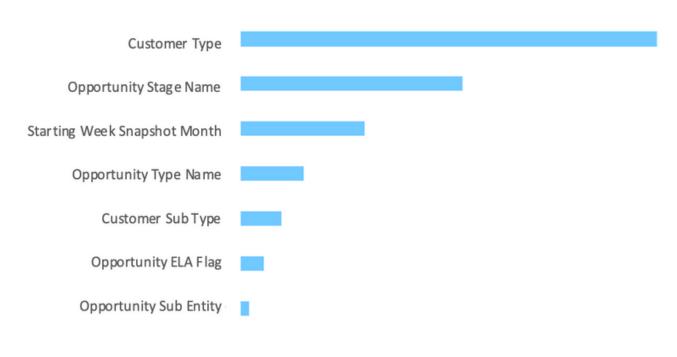
**Model Accuracy Score 92.7%** 



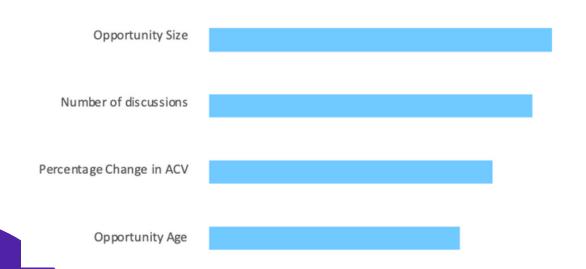


#### Relative Importance of the Features

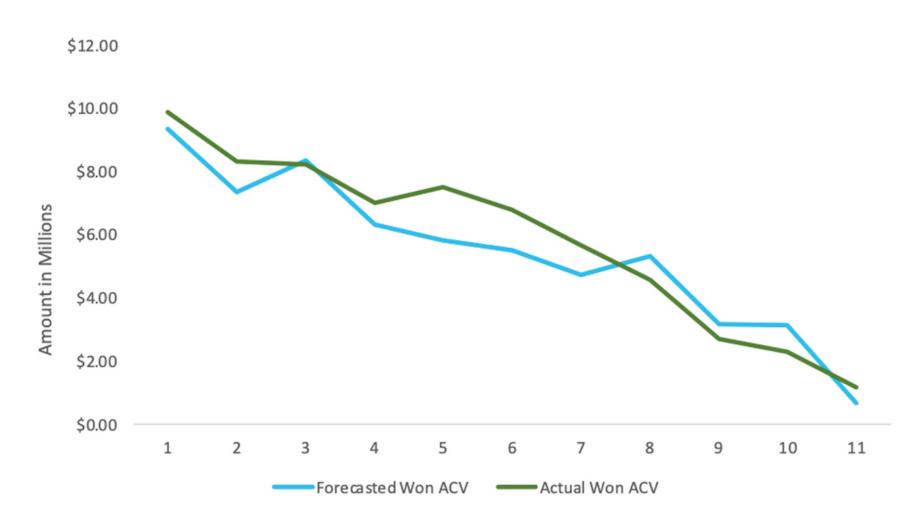




#### Relative Importance (Continuous Features)



#### Total ACV of opportunities forecasted to be Won on Testing Dataset



Model scored as of each starting week in Q1 2023





# Thank You