

Supplemental Framework

Slide Show Presentation

Problem Solving Framework



FACE – Framing of Problem

| Customer | Inventory | Marketing | On-Offline | Technology |
|---|--|--|--|---|
| <ul style="list-style-type: none">• Declining in-store customers• Focus on converting underperforming stores to fulfillment hubs for online growth | <ul style="list-style-type: none">• Inventory carry-over is higher• Inventory discount is hurting margins | <ul style="list-style-type: none">• Lack of customer centric promotions• Customer profile data mining is virtually non-existent | <ul style="list-style-type: none">• Lack of Customer experience• Virtually zero personalization• Degrading instore shopping experience• Ordinary mobile app feature cadence | <ul style="list-style-type: none">• Scattered use of Google Cloud• Slower delivery & deployment• Lack of benchmarking from digitally native competitor• Slower processing of customer data |

EACE – Framing of Problem

- **Stakeholders**

CEO – Jeffery Gennette

CTO – Naveen Krishna

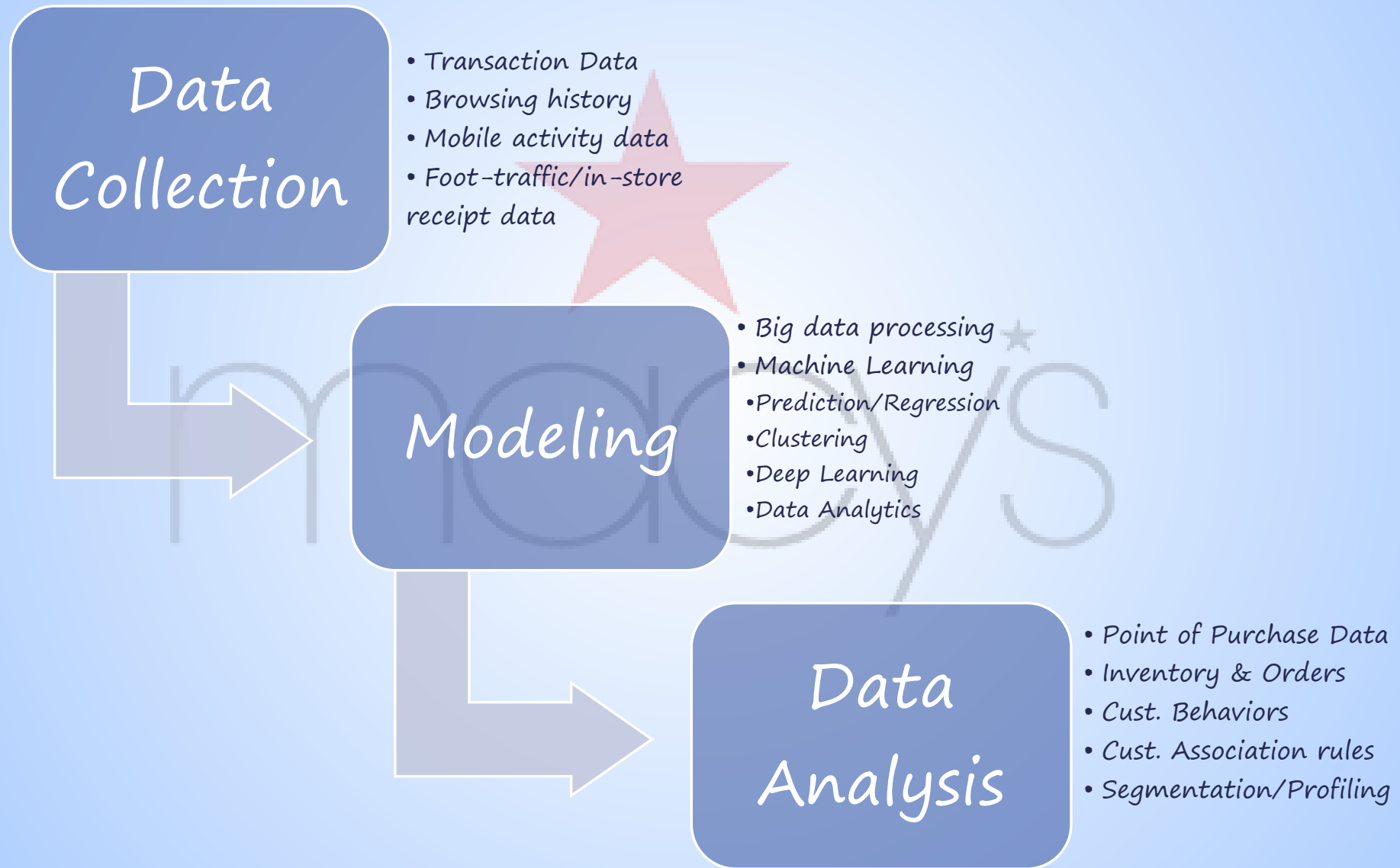
- **Challenges**

- Changing consumer behavior
- Changing business model i.e. fashion --> Consumer Preferences & Personalization
- Effectively managing multiple touchpoints i.e. in-store, online, mobile, tablet
- Everchanging technology and preferences of NextGen consumers
- Habitual customers as a result of promotions/markdowns driven by inventory carry-overs

- **Nature of Problems**

- **Deterministic**
 - Inventory management/assortment based on prediction
 - Value chain optimization
- **Stochastic**
 - Customer spending forecast over time
 - Personalized recommendation based on past buying habit and future prediction

FACE – Analysis to solve the problem



FACE – Analysis to solve the problem

- **Modeling Approach**

- Mainly predictive and interpretation model to analyze each customer's
 - Spending
 - Browsing history trends/patterns
 - Usage of promotion, prediction on future usage
 - Segmentation of customer based on
 - Spending
 - Likelihood of buying certain styles
 - Leverage AutoML modeling techniques for speed, scale and insight

- **Data**

- In-store customer point of sale data
- Customer shared image data on Macy's social media channels – FB, Instagram, Twitter etc.
- Any third-party data that can provide customer insight/profile
- Geospatial data for demographic cluster insight
- Customer activity data

FACE – Communication

Communication Medium

- Presentation
- Graphics & Visualization of KPI and findings

Customer Conversion

- % Online visitors
- % Anonymous vs. registered
- % Engagement vs. shopping

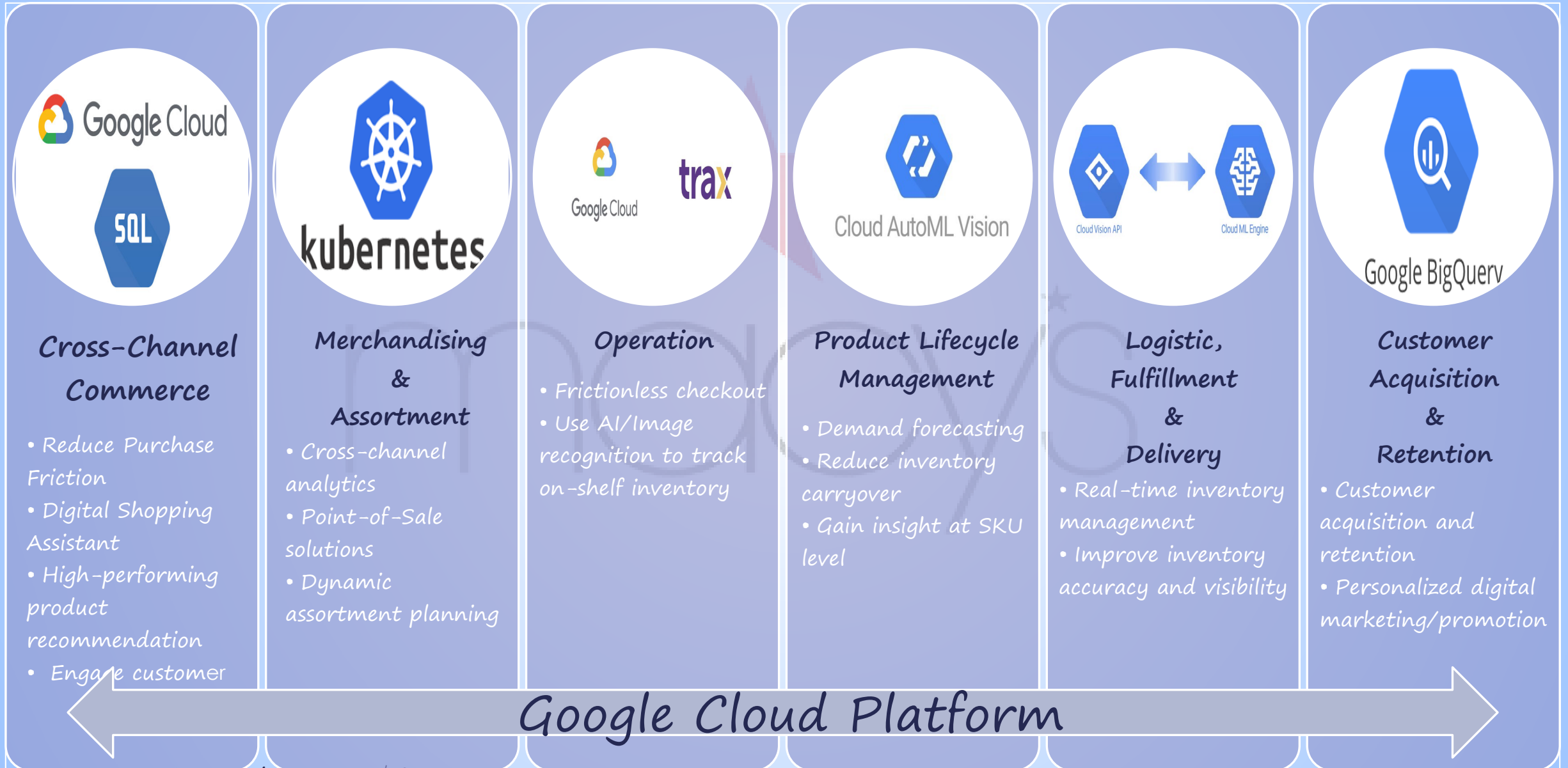
Customer Experience

- In-store shopping
- Online recommendation

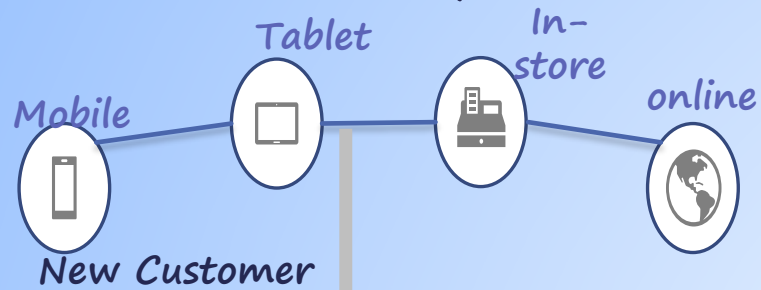
Customer Segmentation

- RFM (Recency, Frequency, Monetary) analysis
- Group by transaction
- Identify in-store customer

FACE – Embed



Example Use Case



Customer Application

- Messages
- Personalized Promotions
- Order Confirmations
- Order pick-up



Serverless Customer Messaging

/Users/{uid}/customers/{id}



Personalized Promotions



Win-Win

Technology Savings



Increased foot-traffic



Increased loyalty



Higher Customer Savings



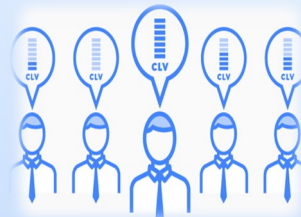
event-driven scaling



Pay only for what you use



Low CLV:CAC



Most Profitable Customer



Intimate Customer Relationship



No Servers to manage



Quick deployment

FACE – The Pachinko Machine

Analytical Solution Approach

- Retail Fashion Industry
- Macy's – A fashion retailer in departmental/ecommerce space
- Inventory and markdown has impacted revenue.
- Customer engagement and conversion is less than 2%
- In-store customer experience is declining and zero visibility in customer profile/preferences



Problem

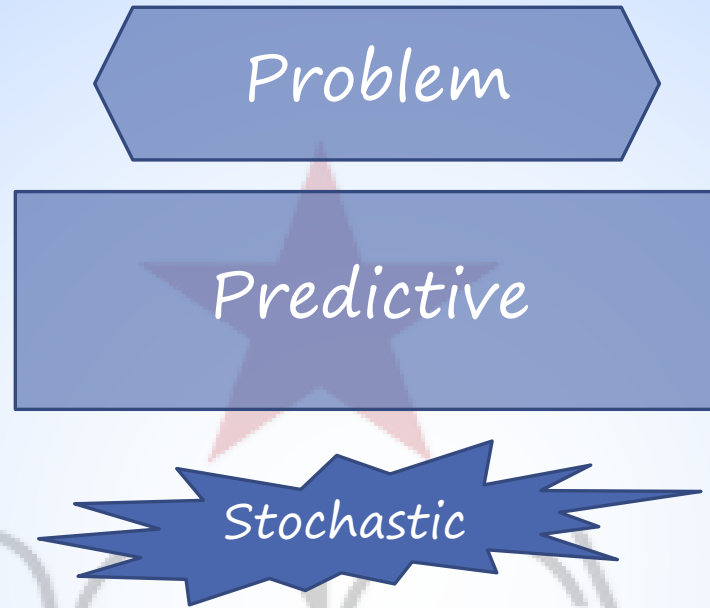
Descriptive
WHAT?

- How many unique visitors to Macys.com
- % of total number of customer engage/buy?

Diagnostic
WHY?

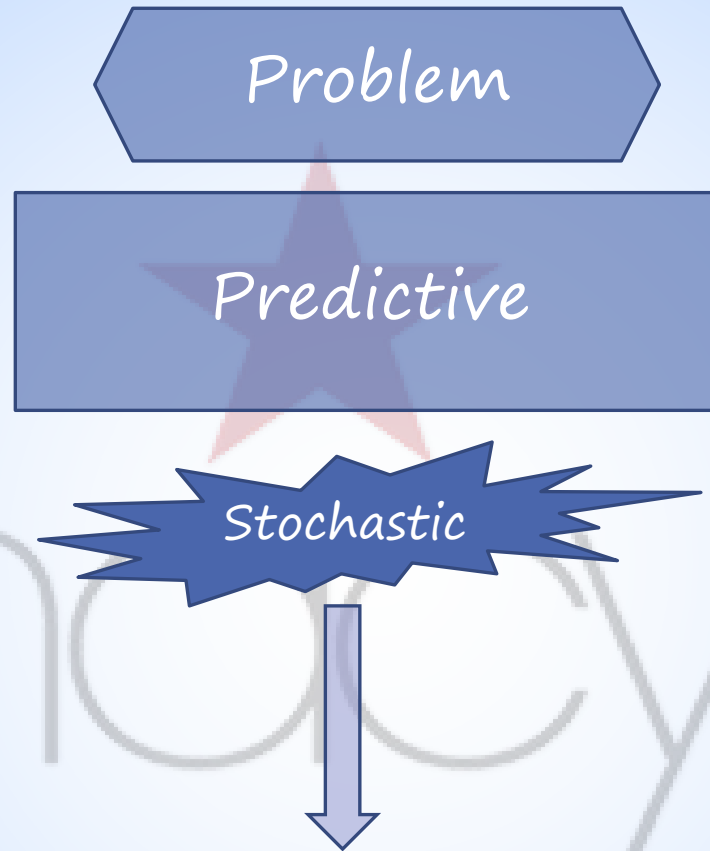
- What factors cause to remain anonymous?
- % of total number of customer sign-up?

Analytical Solution Approach



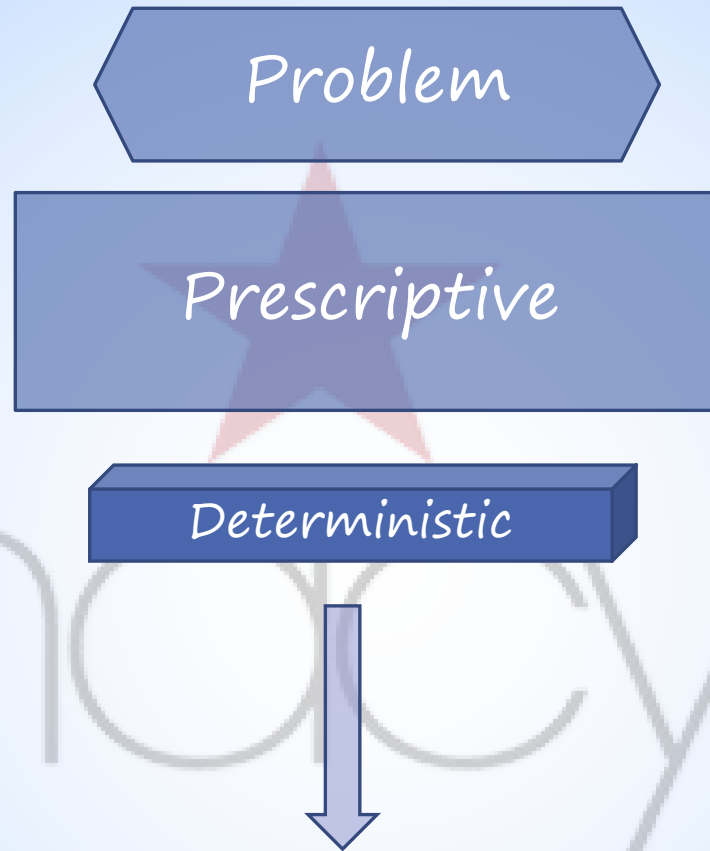
- Faster/real-time data acquisition, processing for prompt customer insights on
 - Behavior
 - Shopping habits & preferences
- Identify customers with high buying probability
- Offer personalized style recommendation before checking out online
- How to encourage customer to signup and engage?

Analytical Solution Approach



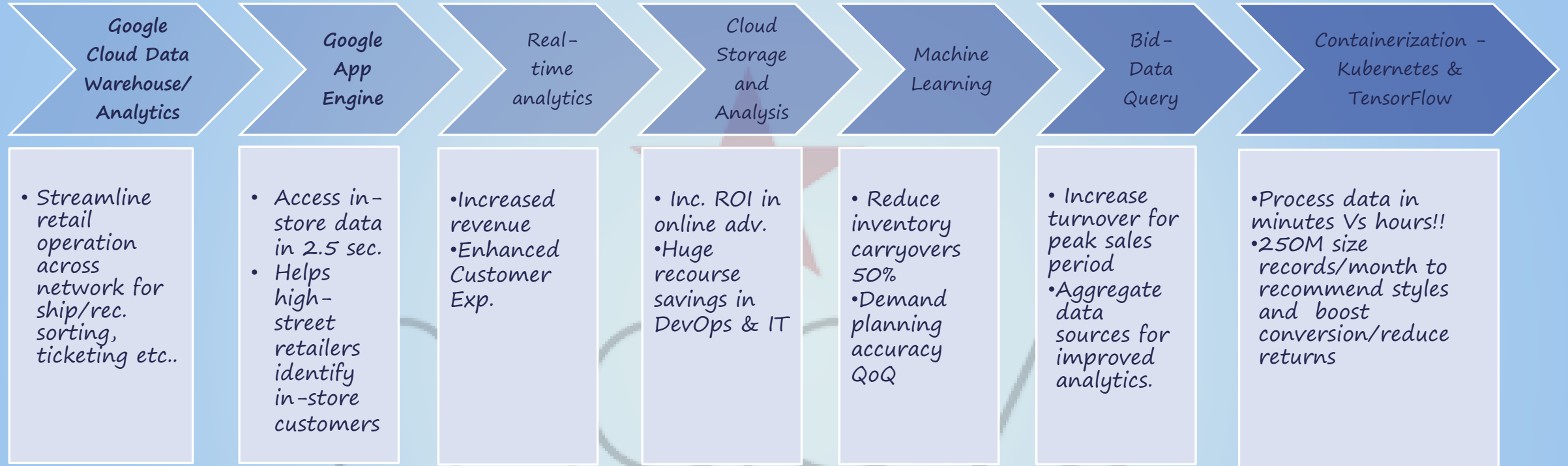
- Potential Modeling algorithms
 - Logistic Regression
 - Random Forest
 - Deep Learning
 - Clustering

Communication & Actions



- Minimize inventory carryover
 - Decision variables to optimize inventory assortment and management
 - Machine learning with Google Cloud Platform

Embedding final models



Existing Google Cloud Native Architecture Platform – 10x faster deployment

Proposed Architecture

