# Building a Data-Powered Sales Intelligence Platform



Data Products, LinkedIn



### **Durgam Vahia**

Data to the Rescue

Q&A

# Overview

- Data Products At LinkedIn
- Sales 101 Challenges and Opportunities
- **Product Perspectives**

# Data Products @ LinkedIn

# Data Products @ LinkedIn



### Mission

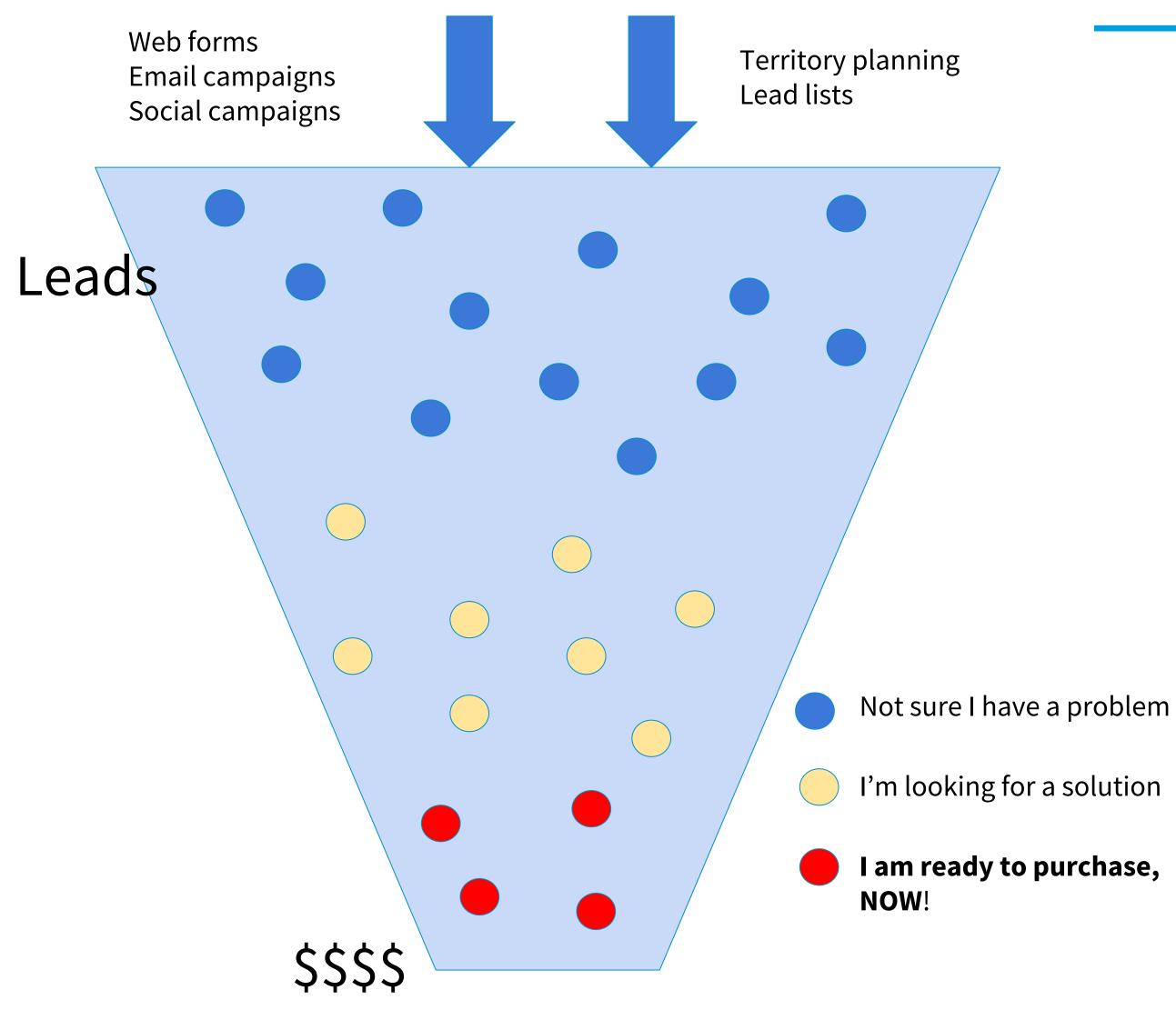
Deliver world class Data Platform that enables employees to make better decisions faster and deliver maximum value to members

### **Areas of focus**

Reporting Search and Discovery Sales Productivity **Developer Productivity** 

Standardization and Knowledge Graph Targeting, Ramping and Experimentation

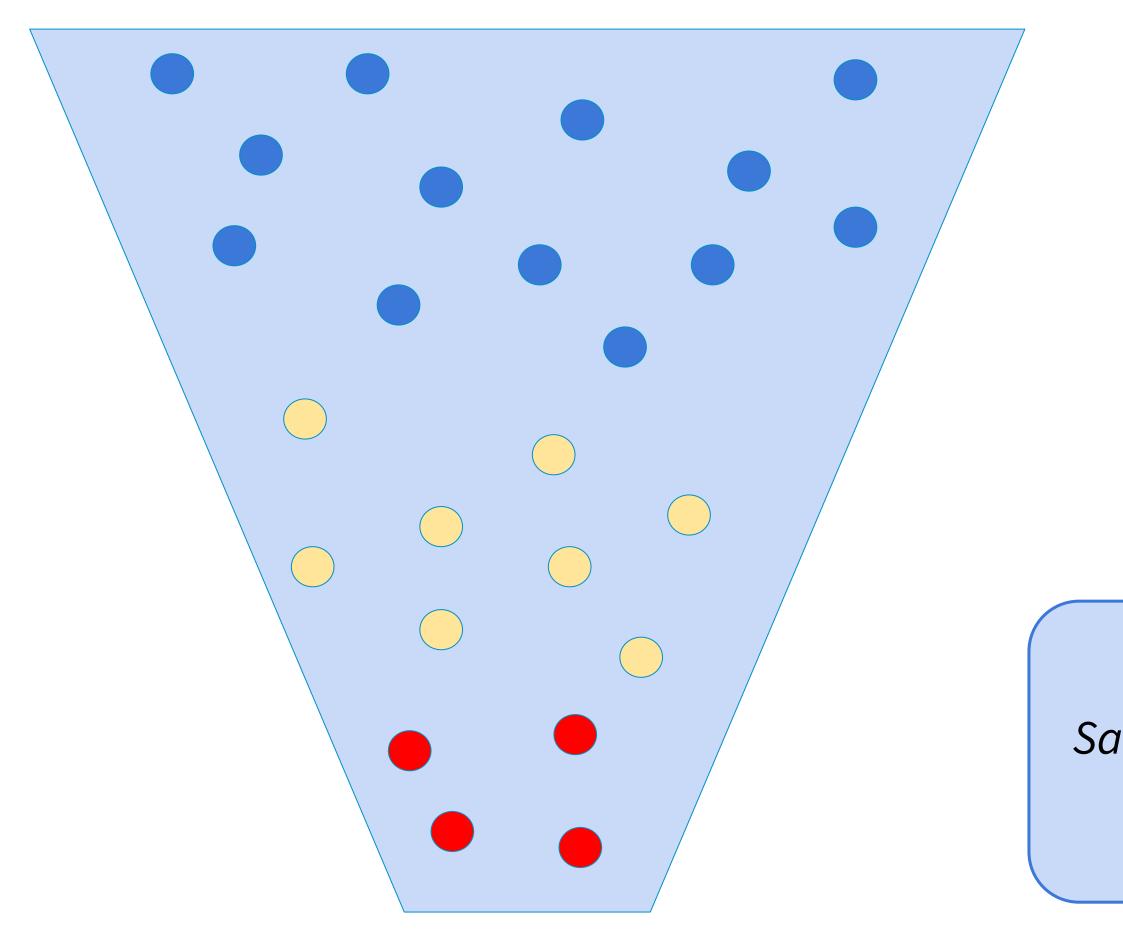
# Sales Intelligence: Challenges and Opportunities





### Job of the Sales rep is to *convert as* many leads as possible into customers, as fast as possible

# Key Challenges in B2B Sales



5-10% of the leads convert to sales

B2B deal takes 2-3 months to complete

2/3rd of all reps miss quota

Sales Velocity =

(# of Leads \* % conversion \* \$/deal)

Avg. Length of the Sales Cycle

# Opportunity

# Can a <u>data</u> product help increase sales velocity?

# Sales Intelligence: Product Perspectives

Start with Empathy (Strategy)	Build fo (Product
Who is my user?	Does the pro my users?
What's the problem?	
	Do the work
Does this problem matter?	interactions
What's better when I'm done?	Are the feed defined?

# Journey of a (Data) Product ..

### for Usability ct-Market Fit)

### **Optimize for Trust** (Scale)

roduct speak to

kflows and s make sense?

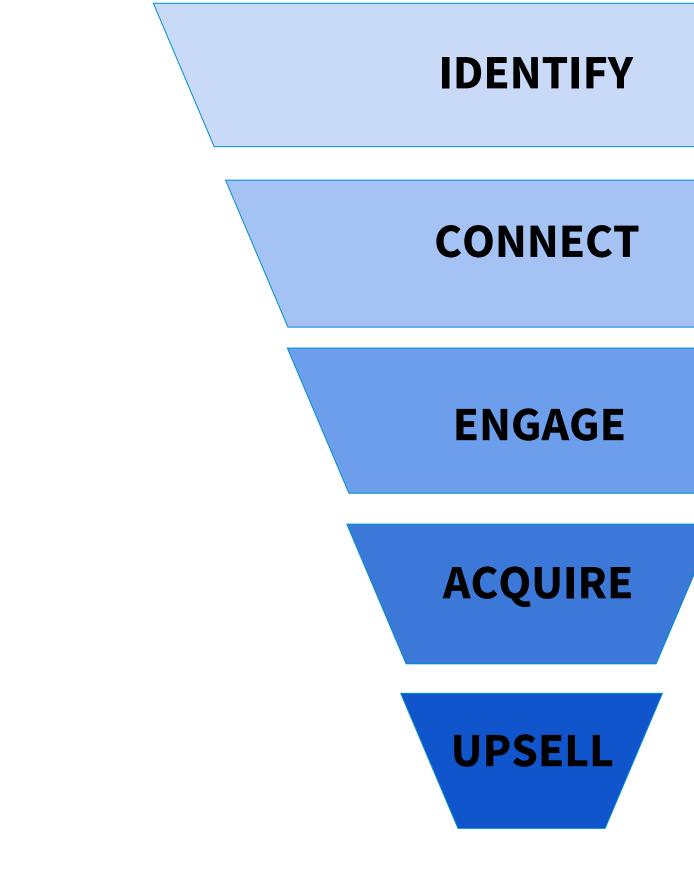
dback loops

Does the broad user base trust the data and recommendations?

Does the product reliably and measurably deliver value?



# **Start with Empathy**





### "I have hundreds of accounts in my book. Which account is most likely to close?"

"There are hundreds of employees I could target, who is a decision maker?"

"I have tons of collateral I could use, which data-story is the most meaningful?"

"I have number of contractual options, which pricing option is the most appropriate?"



*"I have dozens of accounts in my book."* Which account is most likely to Upsell?"







# Identify "Your" Problem Statement



### Relationship Managers manage 10-500 accounts consisting of thousands of users and

- 1. Spend 5 hours/week context switching between different systems, and are
- 2. Unable to construct narratives to engage customers resulting in missed upsell opportunities

### "I have dozens of accounts in my book. Which account is most likely to Upsell?"

# Build the Hypothesis

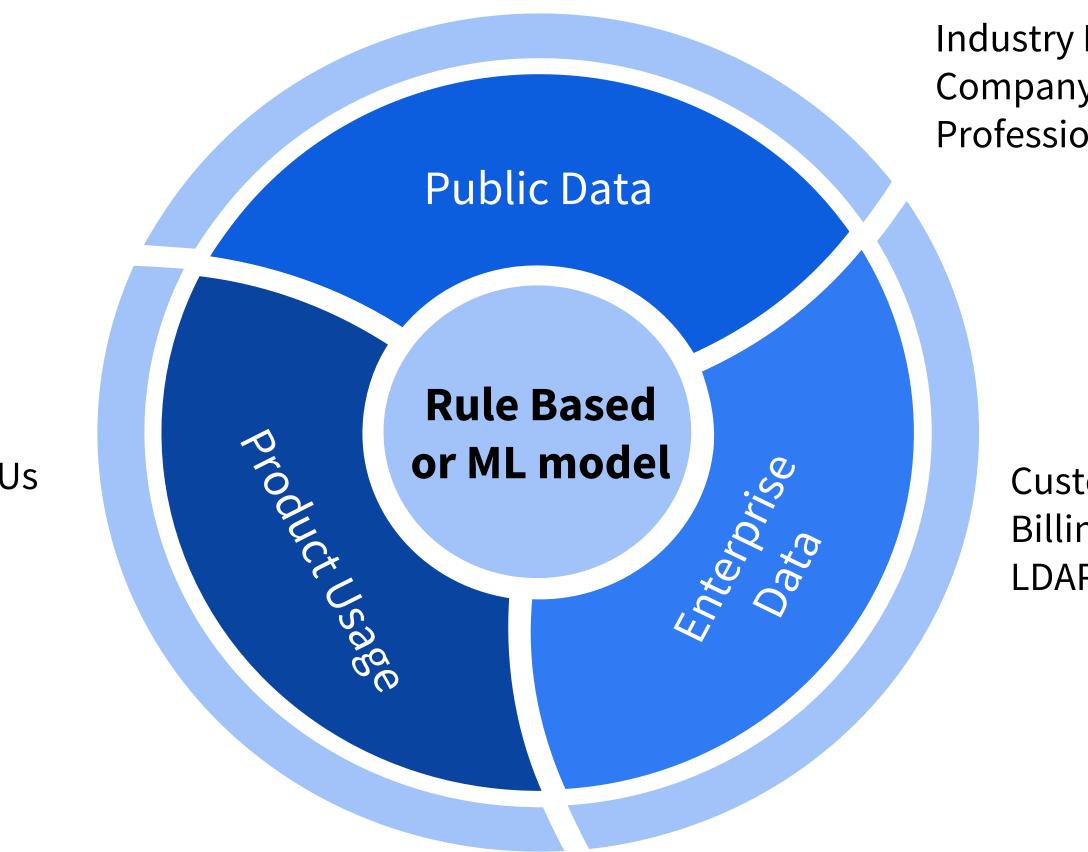
## Opportunity to Upsell if ...

4. Current licenses are well utilized 6. ....



1. The account is in a growing industry 2. The account is growing in revenue/headcount 3. My product has opportunity to grow at the account 5. Users are leveraging key product features

# Identify the Data Sources and build the Modal



Sessions, DAUs/WAUs # of searches # of page views Top users

Industry Reports Company Reports (PR, Web ..) Professional data (e.g. LinkedIn)

> Customer Relationship Mgmt (CRM) Billing and licensing data LDAP and Employee data

Output of the model = Score between 0 (Not likely to Upsell) and 1 (Extremely likely to Upsell)

### Select all 🗄 Save 🛷 Tag 😕 View current employees



### IBM Score 0.8

At IBM, work is more than a job - it's a calling: To build. To design. To code. To consult. To think along with clients and sell. To make ...see all Information Technology and Services · 10,001+ employees · Greater New York City Area



4 connections | 64 senior management hires





In 1939, Bill Hewlett and Dave Packard, college friends turned business partners, started the original Silicon Valley startup in ...see all Information Technology and Services · 10,001+ employees · San Francisco Bay Area



3 connections | 2 senior management hires



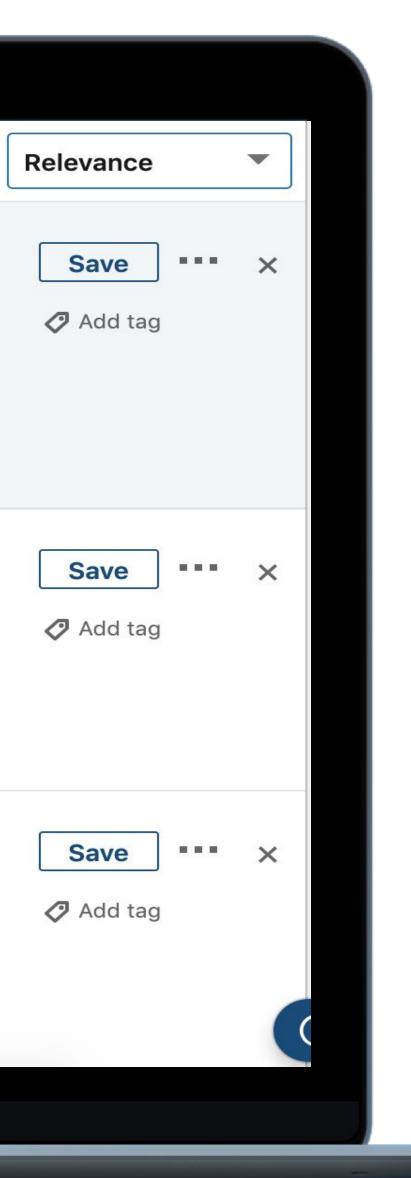
### Score 0.2

Engineering experiences that amaze. We create technology with a purpose: to make life better for everyone, everywhere. Keep R...see all Information Technology and Services · 10,001+ employees · San Francisco Bay Area



HP

1 senior management hire



# Hello World!

### MVP of the prioritization system

### Ranking based on Upsell score



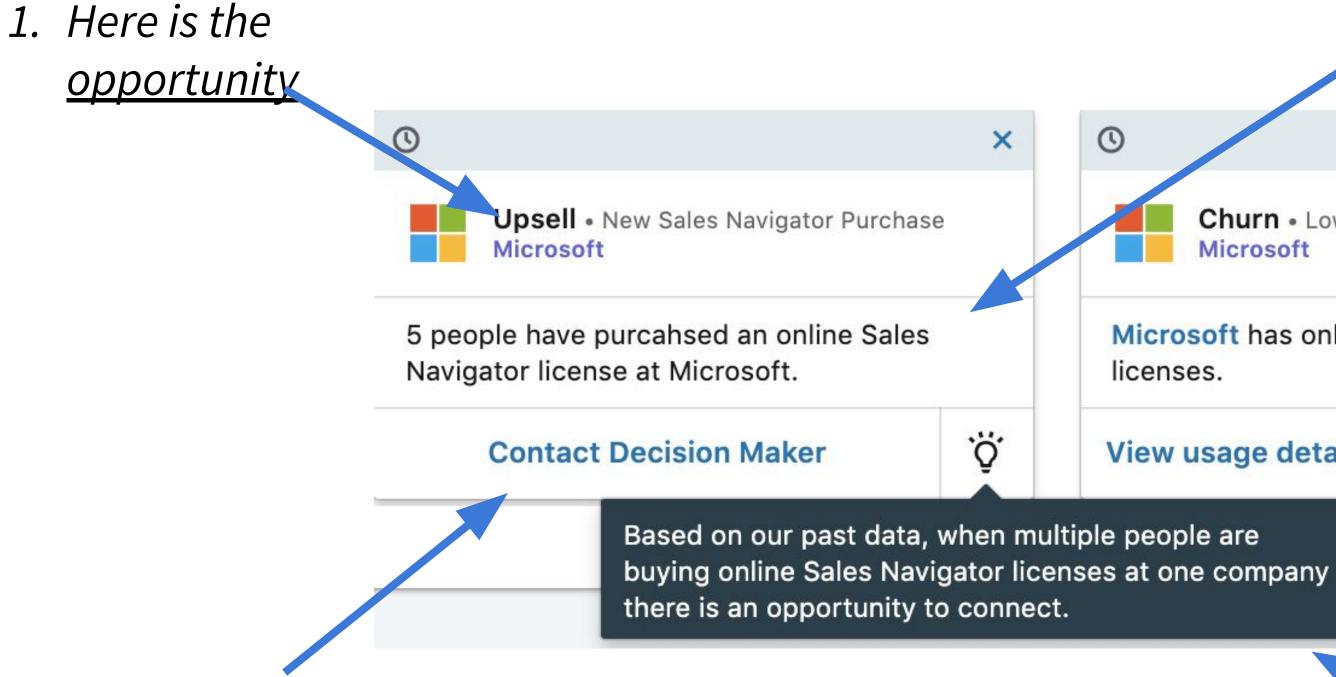


### Net Result: Low adoption and engagement

### "Okay, I kinda get it. How can I use this"

- What is this score? - How is this score calculated? - What does this score mean to me? - What do I do with this recommendation?

# **Build for Usability**



# 4. Here is what you do next

З. т

### Can you bring Serendipity?

2. Here are the <u>signals</u> Churn . Low Pre Microsoft Microsoft has only pr licenses. View usage detail in 3. Here is why these signals

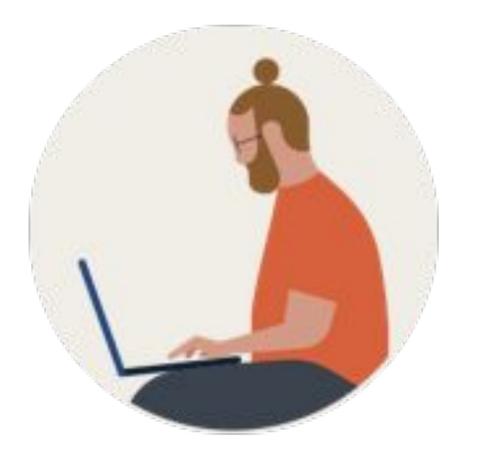
matter

Speak the language of the users

Models must provide narratives - scores are not enough

Sometimes the highest scores are not the most relevant

## "Okay! This makes sense. Can I trust this?"



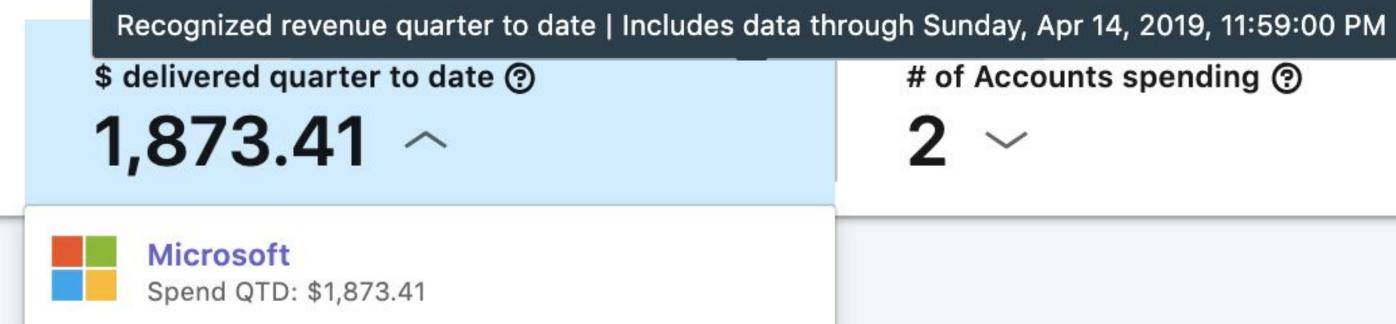
- What are the data sources? - How do I know the data is correct? - Can I provide feedback?

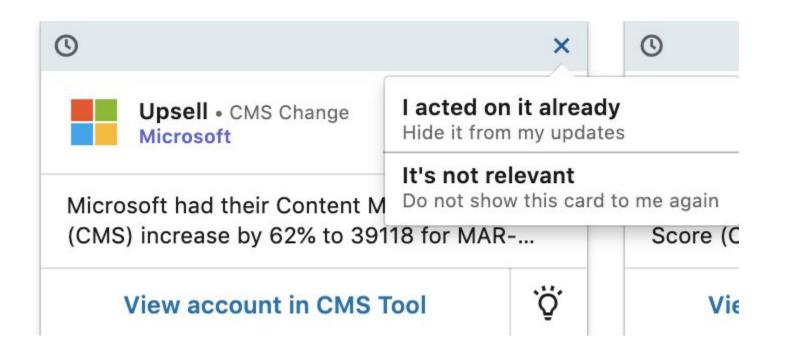


# Scale with Trust

### Welcome Back, Durgam!

Here is a summary of your book of business





### Trust is a key to sustained value

**Personalization:** Understand the individual - LOB, Role, Book ...

**Transparency:** Highlight data sources, refreshes, compliance (GDPR, member-first)

**Metrics that Matter:** Book level, Quota attainment

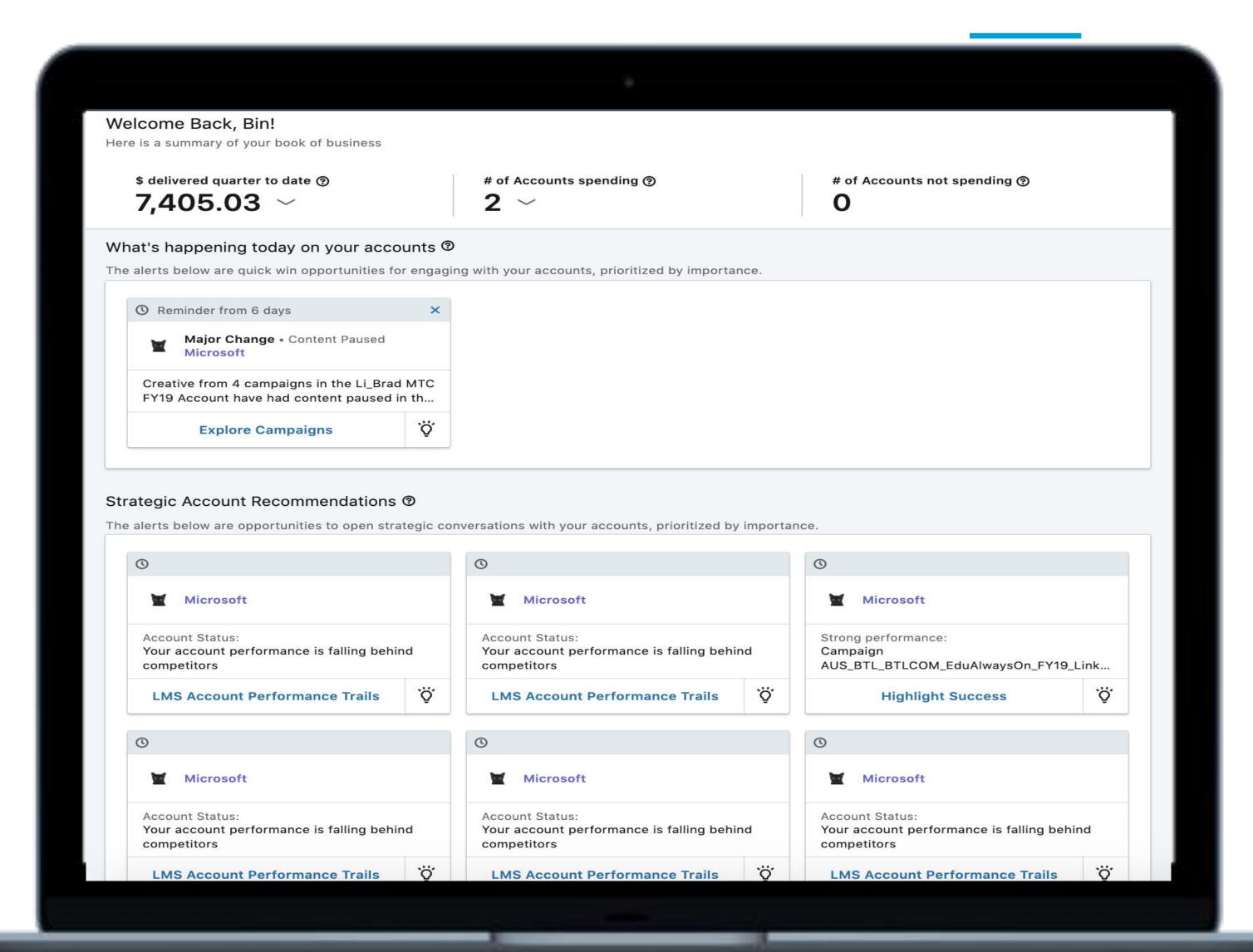
**Drill downs**: Book -> Account -> Subsidiaries

# Sales Intelligence @ LinkedIn



# **Next Best Action**: Deliver personalized and actionable sales intelligence to reps throughout the customer life cycle

# Sales Intelligence @ LinkedIn



Serves all rep personas, all stages of the pipeline

Personalized to an individual

**GDPR** compliant

Success measured by \$ impact, customer experience

Tracking includes DAU/WAU, Impressions, CTRs, Likes





### Basic product principles still hold - Build the right stuff (User Empathy) - Build the right way (Usability)

- Measure and refine

### Perfectly OK to begin with a Rule based model

### Trust is really hard to build

- De-risk the product by solving for value and usability first - Will enable tons of learning and user insights, will help ML feature engineering

- Provide as much data transparency as possible

- Provide feedback mechanism for data/model quality issues



### Life can only be understood backwards; but must be lived forwards

- Soren Kierkegaard

