



PROCUREMENT  
& SUPPLY CHAIN

# The Value of the Lakehouse

HOW T-MOBILE ARTICULATED THE BENEFIT  
OF A MODERN DATA PLATFORM



# Why You're Here

To learn how we convinced non-technical leadership that the Lakehouse architecture was the right path forward for T-Mobile data.

## Agenda

Intro

Pain Points to Strategy

Strategy to Architecture

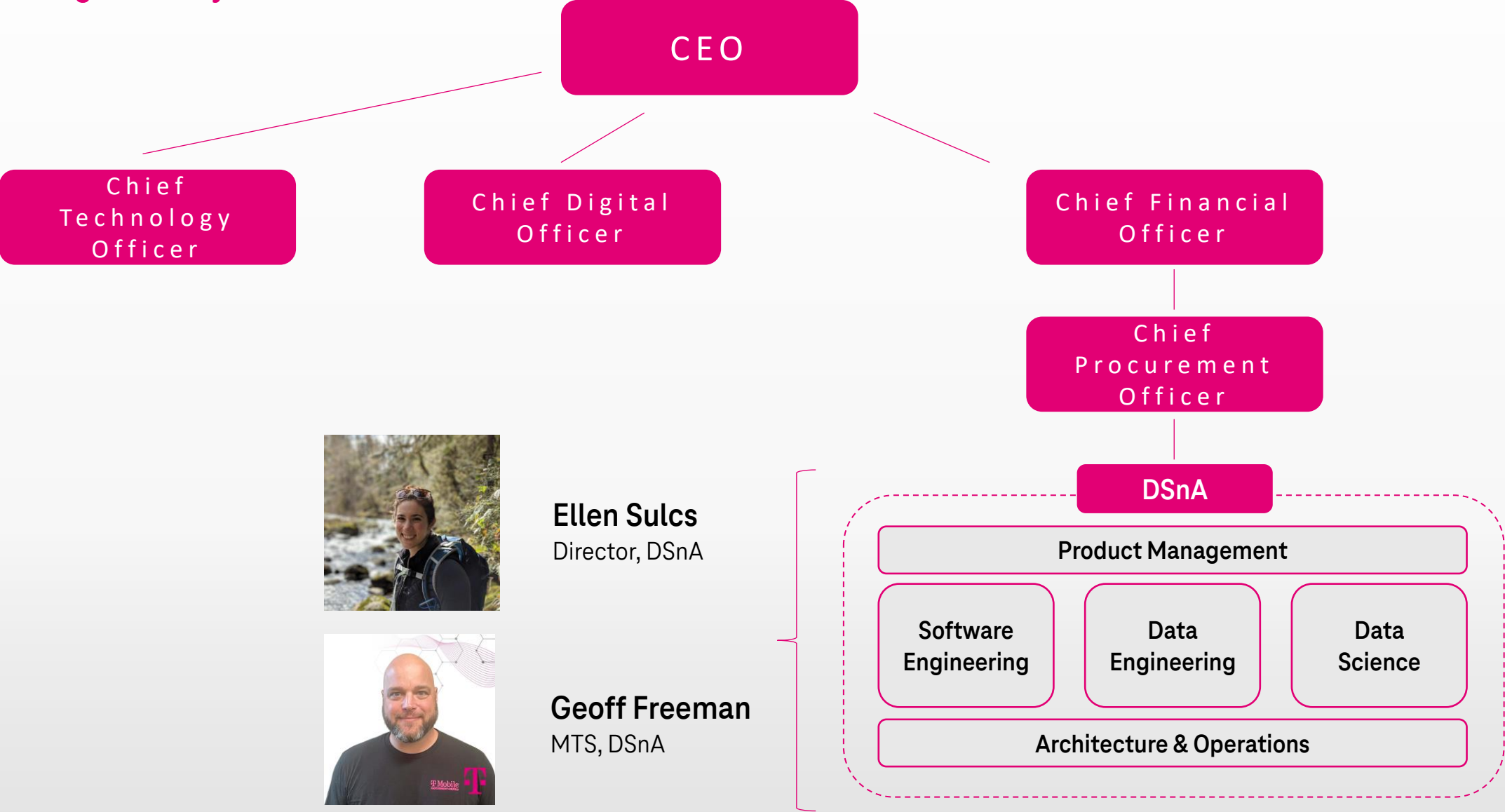
Managing Concerns and Expectations

Lessons Learned



# Who We Are

Technologists in a business focused organization with a mandate to manage data for Procurement & Supply Chain reporting and analytics.



# Pain Points to Strategy



# The Common Pain Points – Analyst to Exec

I just onboarded a new vendor and I need data from their system

My reports are out of date and won't refresh

Your data looks different than what I see in the system

I need to join my spend data with data from the network build team


Our cloud spend is getting out of hand


I use this data for operational decisions and it's not up to date


I need to share the output of my algorithm with my 3PL


My query keeps failing


# Organizing the Pain Points into Platform Strategy


 Your data looks different than what I see in the system


 I use this data for operational decisions and it's not up to date





 **Data Latency**


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 **Data Mobility**


 My reports are out of date and won't refresh


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 **Predictable Performance**

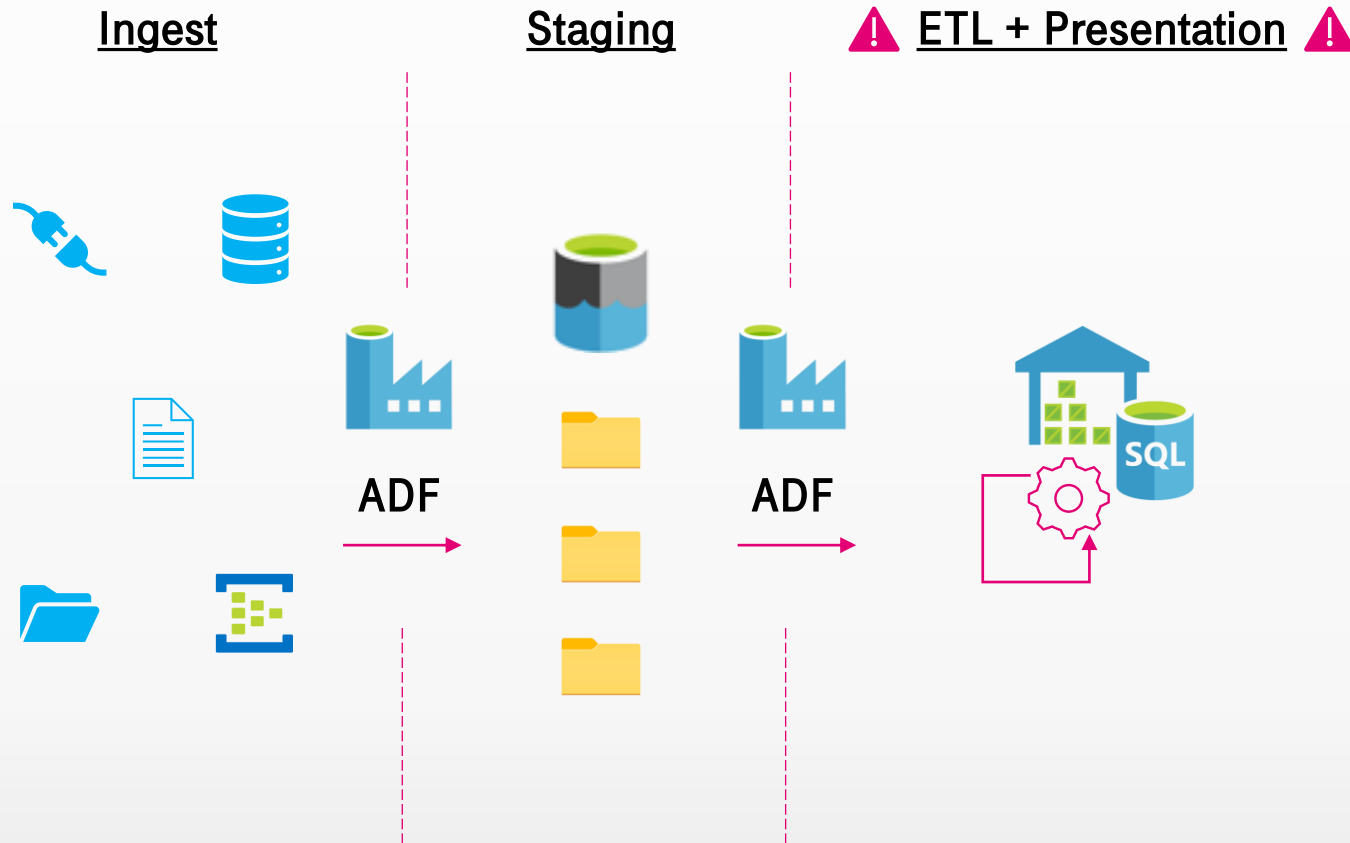
 Our Azure spend is getting out of hand



 **Economies of Scale**



# The OG Architecture – Why the Pain Points Hurt



## DW appliance is the problem:



**Predictable performance** impacted by shared and constrained storage/compute resources create noisy neighbor issues



**Latency** caused by performance issues blocking ETLs



**Mobility** is directly limited by the siloed nature of the DW; sharing requires redundancy.



**Economies of scale** limited by ADW hardware – bigger hardware is exponentially more expensive

# The Data Strategy

**Platform Vision:** Scalable data management that sets the enterprise standard



## Low Data Latency

The **most up to date raw and transformed data** is made available for all consumers and use cases.



Refresh Cadence



ETL Processing Time



## High Data Mobility

Data is easily **consumed where it resides** and **democratized**. Various organizations and solutions can consume and **share data seamlessly**.



# of Data Connections (in and out)



Time to Onboard Data



## Predictable Performance

**Workload isolation** through serverless workloads ensures **consistent query experience**, eliminating “noisy neighbor” problem.



Data Availability



Query Failures



## Economies of Scale

Ensure a **cost-effective management** of Lakehouse architecture and realize **enterprise data management efficiencies**.



Infrastructure Spend



Resource Efficiency



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**75%** Average refresh cadence from 6hrs to 1.5hrs



ETL Processing Time



**88%** Complex build times reduced from 4hrs to 30mins



## Predictable Performance

Workload isolation through serverless workloads ensures **consistent query experience**, eliminating “noisy neighbor” problem.



Data Availability



**1%** Data availability increase from 98% to 99%



Query Failures



**60%** Reduction in query failures from 1.5% to 0.6%



## High Data Mobility

Data is easily **consumed where it resides** and **democratized**. Various organizations and solutions can consume and **share data seamlessly**.



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**30%** 55 connection points (41 in / 14 out)



Time to Onboard Data



**80%** Data onboarding time from 5 days to 1 day



## Economies of Scale

Ensure a **cost-effective management** of Lakehouse architecture and realize **enterprise data management efficiencies**.



Infrastructure Spend



**30%** \$120k Azure spend saved per month



Resource Efficiency



**30%** 10 sources added with equivalent staff

# Strategy to Architecture



# Mapping the Architecture to the Strategy

Lakehouse principle of true separation of compute and storage addresses all pillars of our strategy.



High speed storage – noisy neighbors – data movement = faster data



Query data where it resides – if it's in the data lake, it can be queried. No more silos.



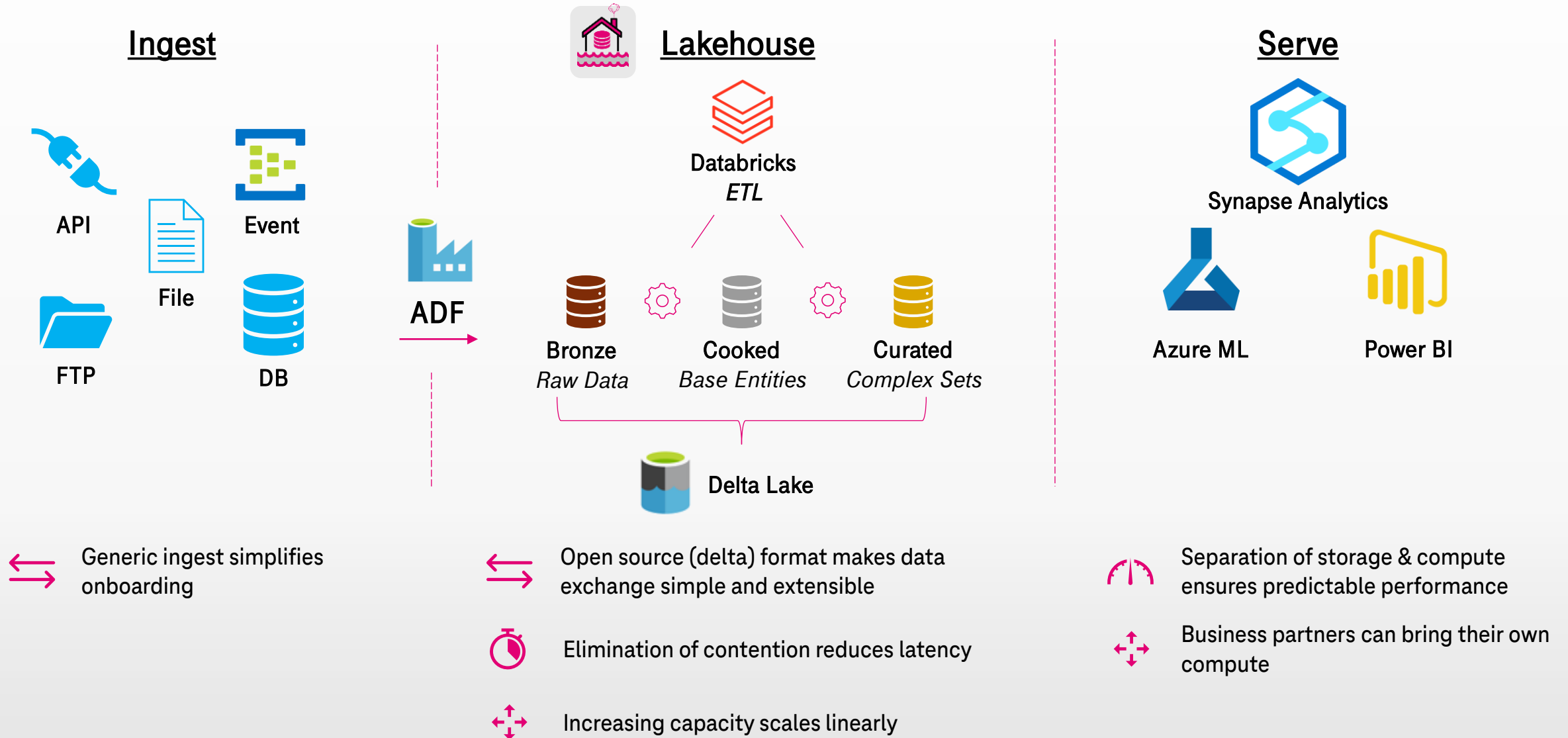
No more noisy neighbors hogging resources, no more waiting on locked objects.



Commodity hardware is cheap to add, reduction in data duplication reduces TCO of data.

# Lakehouse Deep Dive

Lakehouse provides data ingest, exchange, centralization and transformation for both operational and analytics needs.



# Managing Concerns and Expectations



# Feedback on the Strategy



## How much time?

### Not a relevant concern

Every ETL migrated reduces cost and improves performance. Even if we never finish, we'll see benefits.

*We gave ourselves 1yr to migrate 30+ systems (2k+ tables) of data to Lakehouse. Primarily lift and shift.*



## How much money?

### No resource costs/additions

Rolled into the reserved 30% capacity for tech debt reduction

### Minor, initial infrastructure increase

Planned for small spike (~15%) prior to migration of first ETL chunk

*We never saw an initial spike in cost.*



## What are the user impacts?

### A change in the database connection used

Serverless endpoints ensure compatibility across existing suite of reporting tools (SSMS, Alteryx, PBI, ...)



## What are the security implications?


### Improved security with process driven access barriers

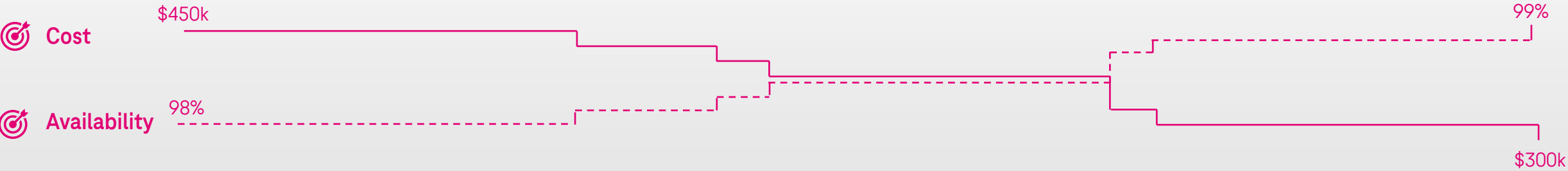
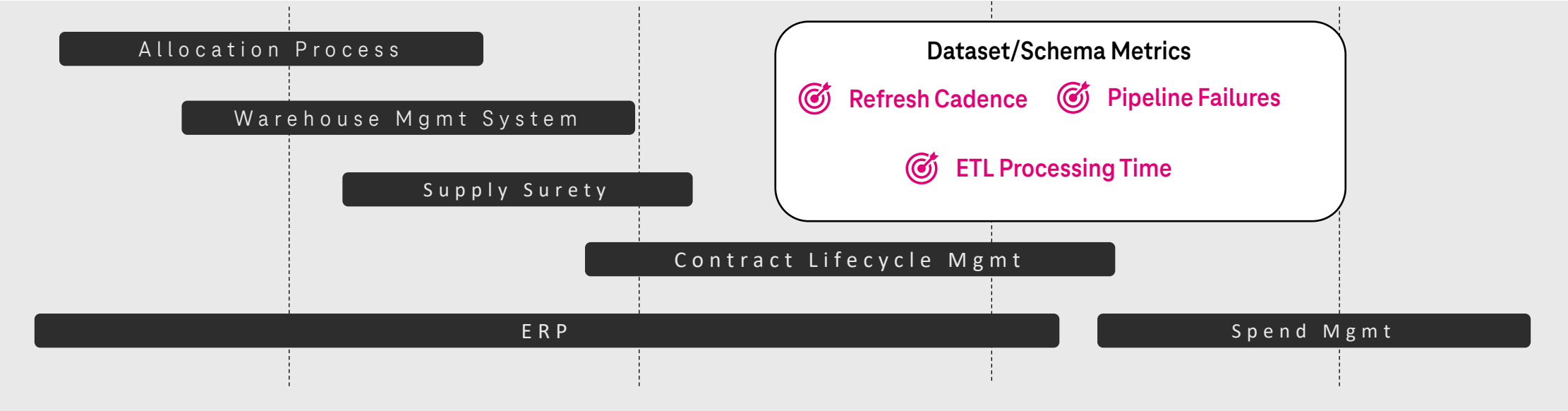
Use of AAD groups (or IAM roles in AWS, etc) is industry best practice. Using the same groups across all fabrics makes it easy to understand, audit, and maintain.

*Make barriers to access process driven, instead of technology driven*

# Supporting the Migration

Prioritize based on metrics and business value to demonstrate immediate incremental value.

 **Top Priority: 5G Build Allocation Process**  
\$20M loss per allocation miss

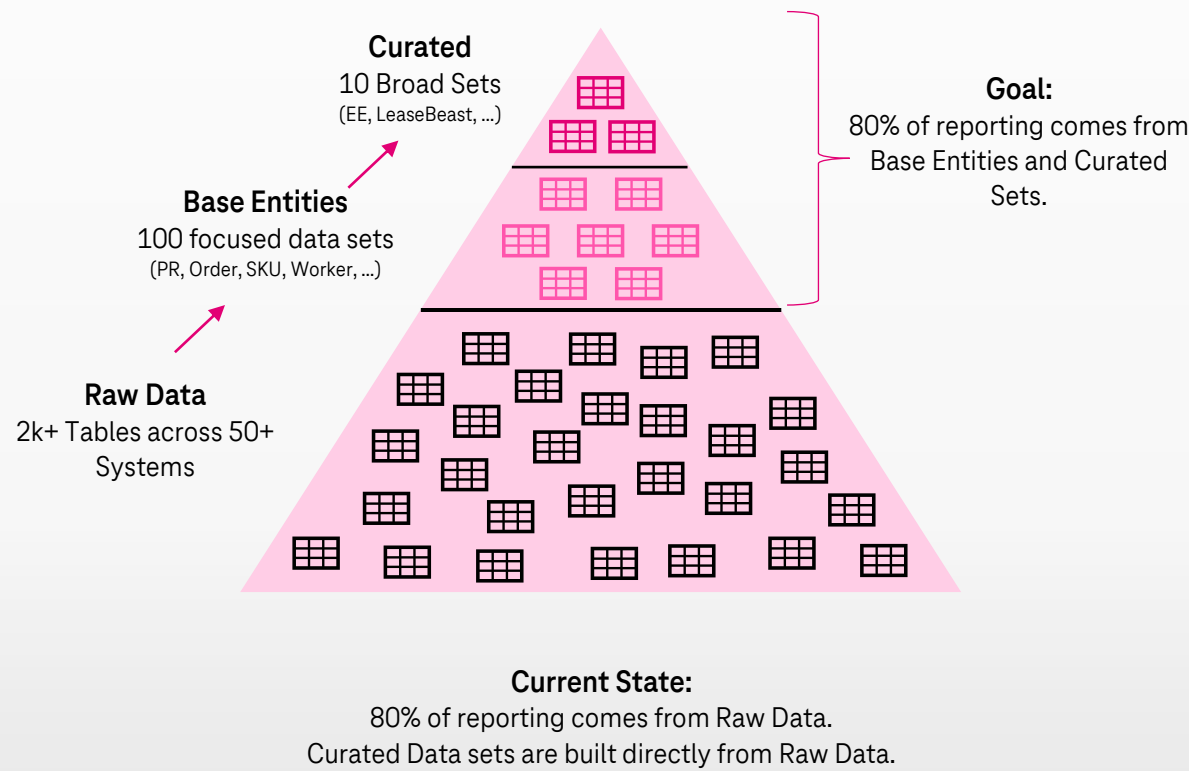




# Where We Are Today

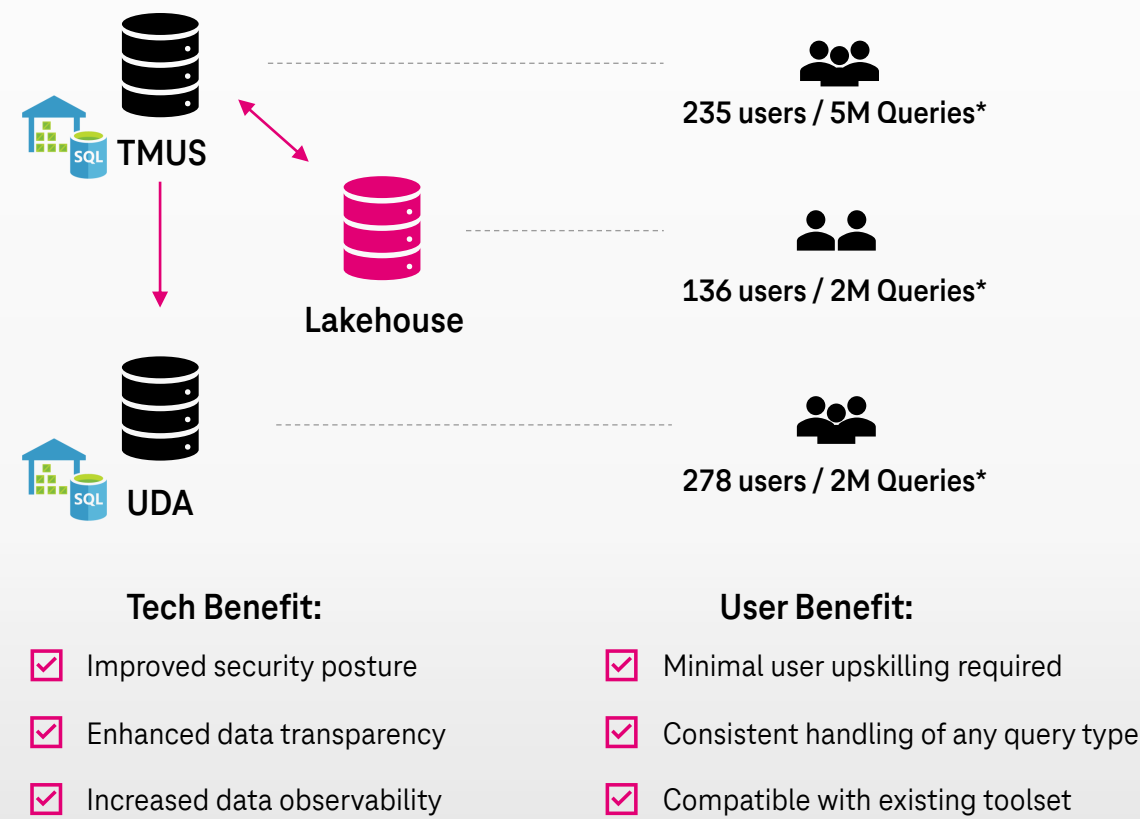
In 2022, we used the Lakehouse to reliably contain the chaos. In 2023, we seek to simplify what exists. Here's what we're doing internally:

## Focus on Data Modeling



## Unify Compute Experience

Goal: Push it all to the Lakehouse



\*Users & Query Volume from Q1 '23

# Lessons Learned



# Quantifying the Lakehouse Success

The Lakehouse exceeded our expectations.



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## **Enterprise tools and applications are the limiting factor in data latency and cost management.**

Business platforms struggle to exchange large data with NRT capability. Change data capture is not the norm.



## **A data model is still valuable.**

Data models aren't just for performance improvements. We can reliably contain an infinite amount of data, but a model that directs users to tried and true sources and curated sets is invaluable.



## **Old habits die hard.**

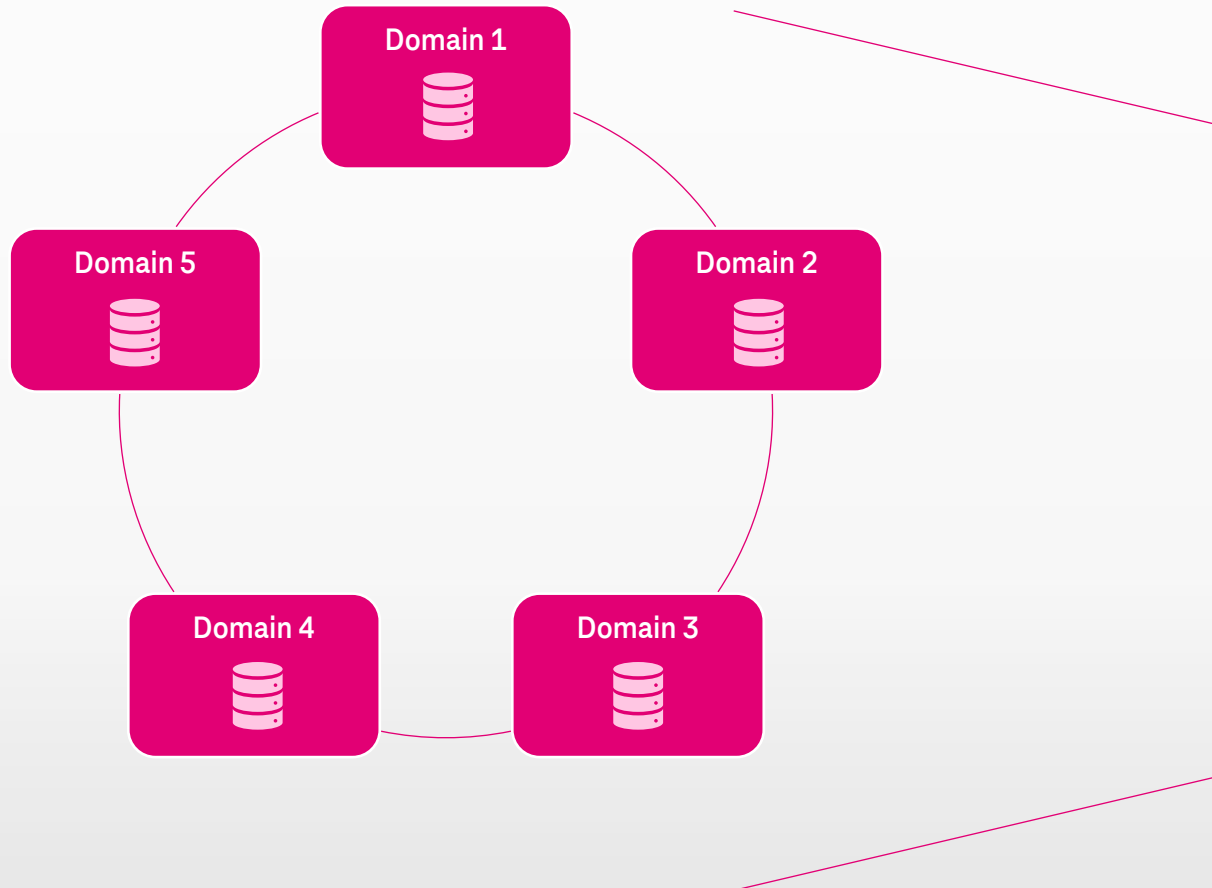
Don't underestimate the culture investment and the eventual need for a top-down push to migrate users to the Lakehouse.

# Now What?

Taking the Lakehouse architecture enterprise wide: networked Lakehouses with shared governance platforms.

## Networked Lakehouses

Domain data managed by BU-embedded data teams



## Shared Governance Toolset

Simple data discovery and access experience



### **Automated Access Management**

Custom service over AAD API



### **MDM/RDM Capabilities**

Through standard SaaS solutions



### **Data Catalog**

User-friendly data discovery and context



### **Data Quality**

Centralized health monitoring



### **End to End Lineage**

Leveraging openlineage.io and Unity Catalog

# Interested in the Specifics?

Check out some of our publications:

## Articles/Publications:

Why a Delta Lakehouse?

<https://medium.com/tmobile-dsna/why-delta-lakehouse-27d325c18c42>

What is a Lakehouse?

<https://medium.com/tmobile-dsna/what-is-a-data-lakehouse-8bdad0f7a67a>

Case Study – How Databricks Photon Saves us 25% on Compute

<https://medium.com/tmobile-dsna/how-databricks-photon-saves-us-25-on-compute-9941f178d7b7>

Case Study – Connecting Millions of Customers with the Largest 5G Network

<https://www.databricks.com/customers/t-mobile>

Case Study – T-Mobile Supports 5G Rollout with Azure Synapse Analytics and Power BI

<https://customers.microsoft.com/en-us/story/1581506762107713013-t-mobile-telecommunications-power-bi>

## Recordings:

D3L2: Implementing a Data Lakehouse for Improved Data Science and Analytics

[https://www.youtube.com/watch?v=IJlvIWa\\_X1o](https://www.youtube.com/watch?v=IJlvIWa_X1o)

D3L2: Why Did We Migrate to a Data Lakehouse on Delta Lake

[https://www.youtube.com/watch?v=1V\\_u9hbol-Q](https://www.youtube.com/watch?v=1V_u9hbol-Q)

# Q&A

