

From Metadata to Lakehouse

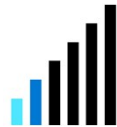
Andreas Bergstedt
Data and AI Global Black Belt



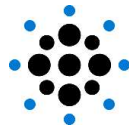
Introduction / Agenda



The Evolution of ETL



Why Lakehouse



Metadata Driven
Engineering



Driving at Scale



Key take-aways

Introduction / Agenda



The Evolution of ETL

```
for /%a in (
copy /y C:\my
d.
copy /y C:\my
d.
copy /y C:\myfile1.tx
d.
```

SQL Server Enterprise Manager - [DTS Package: Address Book]

Package Edit Connection Task Workflow

Connection

Task

Package.dtsx [Design]*

Control Flow Data Flow Event Handlers

Data Flow Task: Data Flow Task

Flat File Source

testPatientAPI_web.json

Schema: <No Schema Selected>

Activities

- Move & transform
- Azure Data Explorer
- Azure Function
- Batch Service
- Databricks
- Data Lake Analytics
- General
- HDInsight
- Iteration & conditionals
- Machine Learning

Map Transformation

Input Fields

Output Fields

Calculated Fields

Generated Fields

Initial

Logins

Reset Password

Logins

Ready

Completed Successfully

File Edit View Search Diagram DDI Tools Window Run Help

Des... Opera... Topol... Pop.TRG_CUSTOMER

Projects

Demo

Sales Administration

Packages

Mappings

Archive Customer

Pop.TRG_CITY

Pop.TRG_COUNTRY

Scenarios

Uses

Used by

SOURCE_GROUP

FILE_DEMO_SRC_UNIT

SRC_SALES_PERSON

SRC_AGE_GROUP

ODI_DEMO_SRC_UNIT

TARGET_GROUP

ODI_DEMO_TRG_UNIT

SRC_SALE

SRC_AGE

LOOKUP

JOIN

TRG_CUSTOMER

Copy data

Location_HTTP

Customer_Salesforce

Products_SAP

Wrangling Data Flow (Preview)

SalesDataPrep

SalesAnalytics

Machine Learning Execute Pipeline

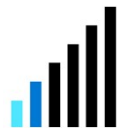
FeedbackLoopML

Parameters Variables Output

Pipeline run ID: fb1ef562-5b23-46b8-b243-ee012e5a043

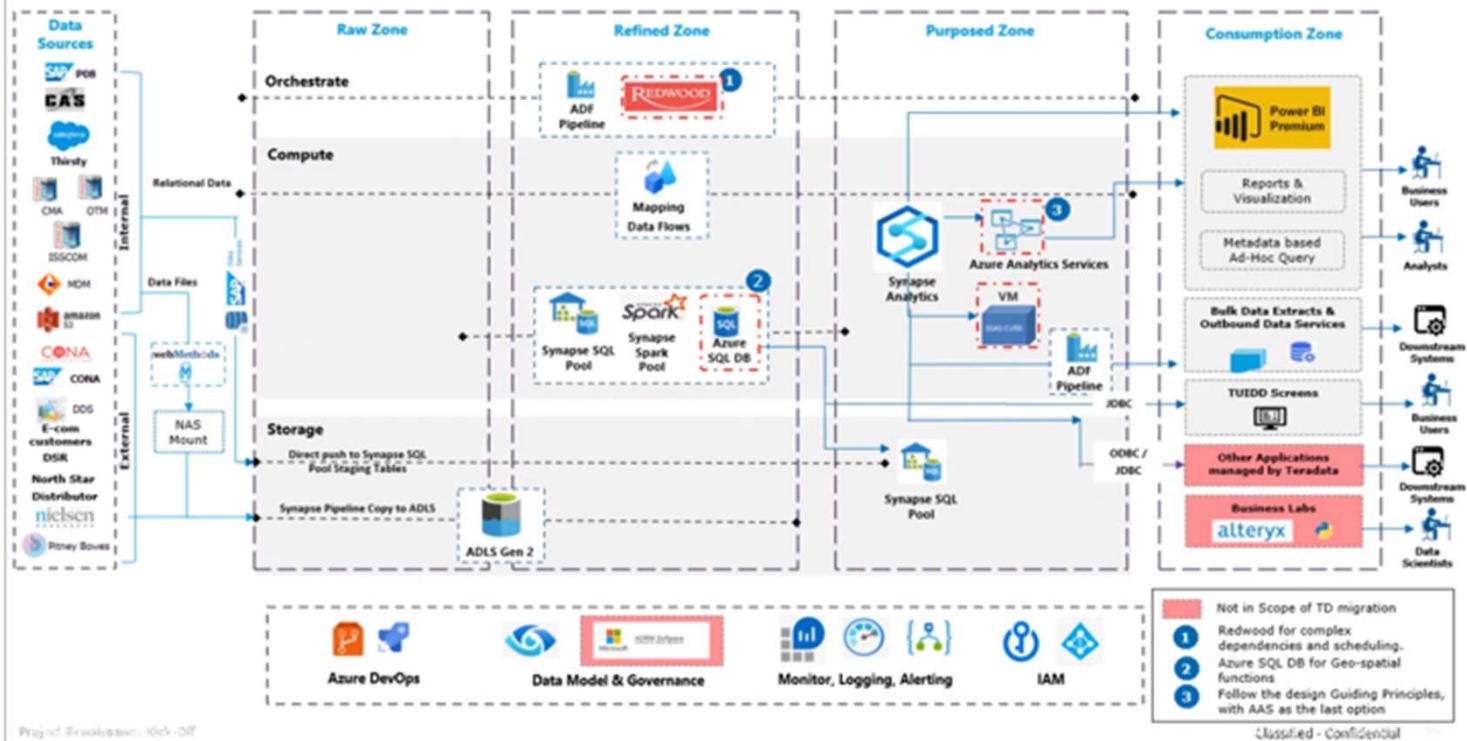
Name	Type	Run start	Duration	Status	Integration runtime	Run ID
Location_HTTP	Copy	2020-08-28T18:10:13.633	00:00:06	Succeeded	DefaultIntegrationRuntime (East US 2)	fd86f9a-770c-4229-ad7b
Customer_Salesforce	Copy	2020-08-28T18:10:13.289	00:00:10	Succeeded	DefaultIntegrationRuntime (East US 2)	03cf907b-a760-4693-b0c
Products_SAP	Copy	2020-08-28T18:10:13.289	00:00:08	Succeeded	DefaultIntegrationRuntime (East US)	d9a3395a-59b5-416d-a1c
POS_SQL	Copy	2020-08-28T18:10:13.258	00:00:55	Failed	DefaultIntegrationRuntime (East US)	eb991e07-7462-4680-9e6
Clickstream_S3	Copy	2020-08-28T18:10:13.274	00:00:20	Failed	DefaultIntegrationRuntime (East US)	9d1110c9-0507-4d39-a05

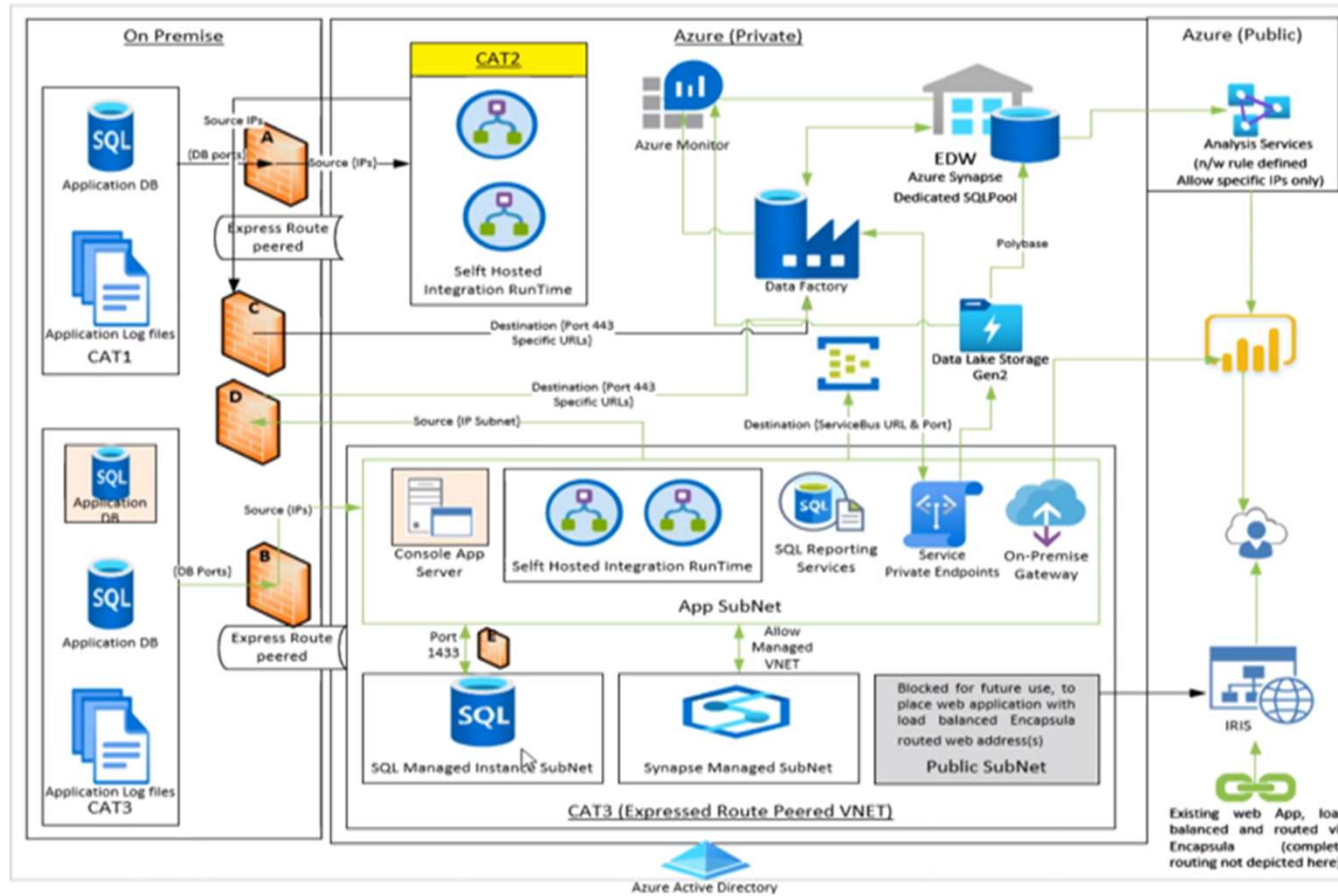




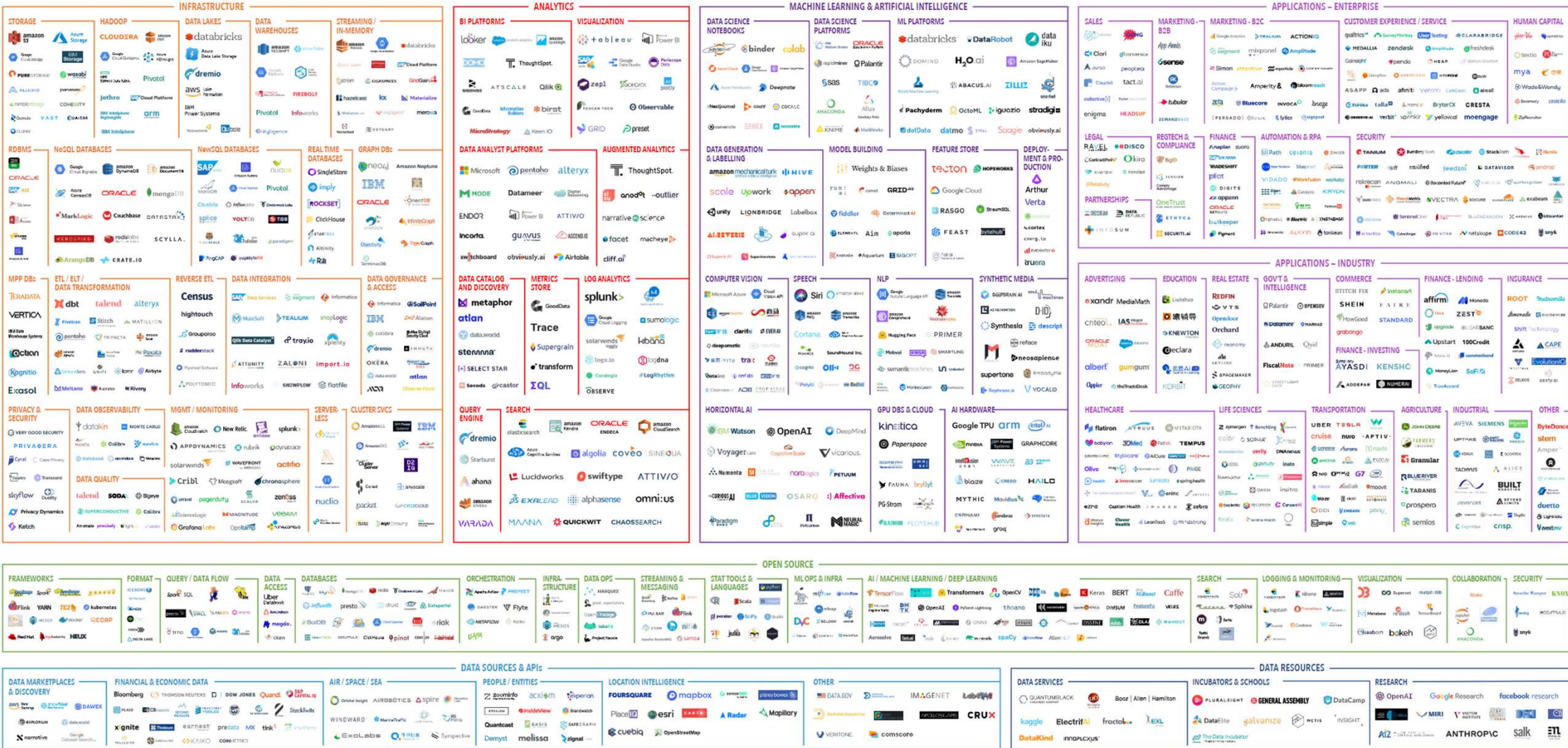
Why Lakehouse

Target Architecture





MACHINE LEARNING, ARTIFICIAL INTELLIGENCE, AND DATA (MAD) LANDSCAPE 2021



 Microsoft Azure



Azure Synapse Analytics



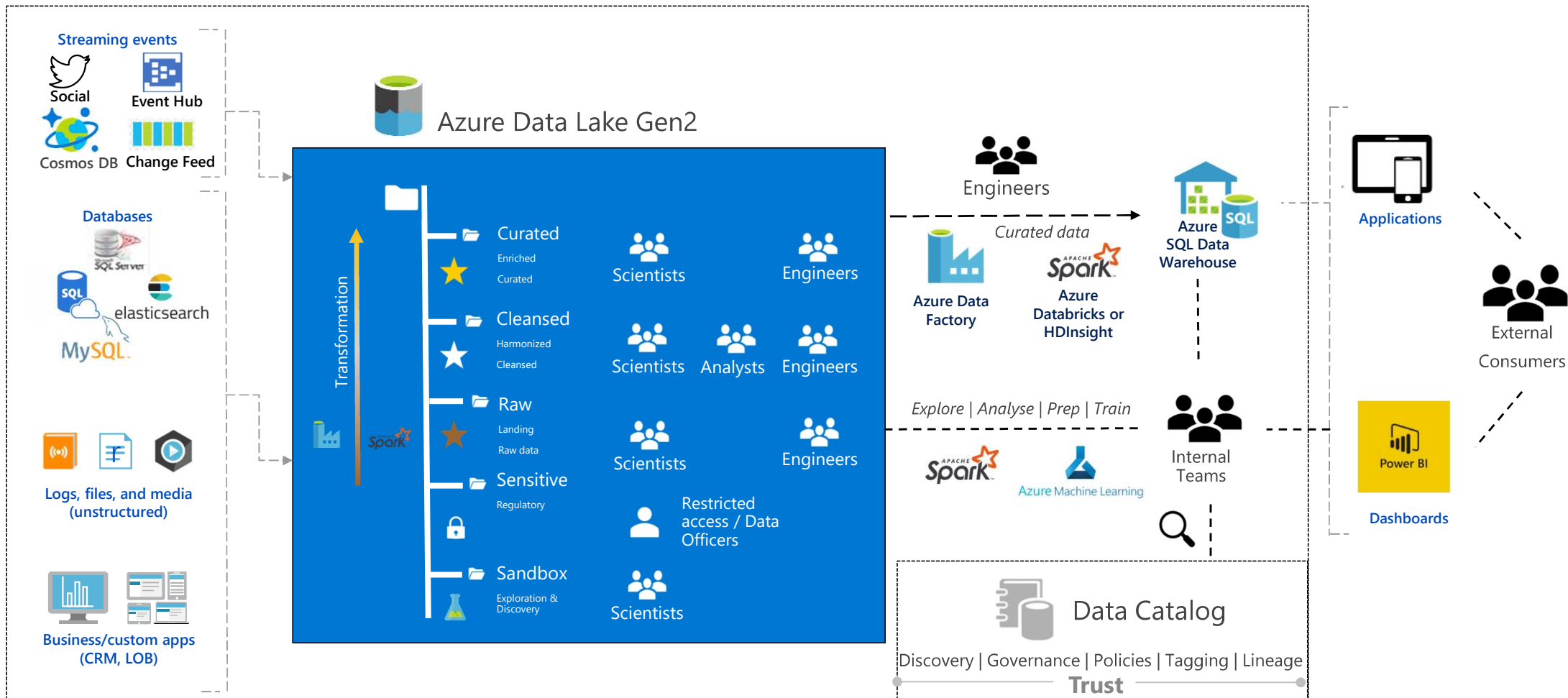
DELTA LAKE



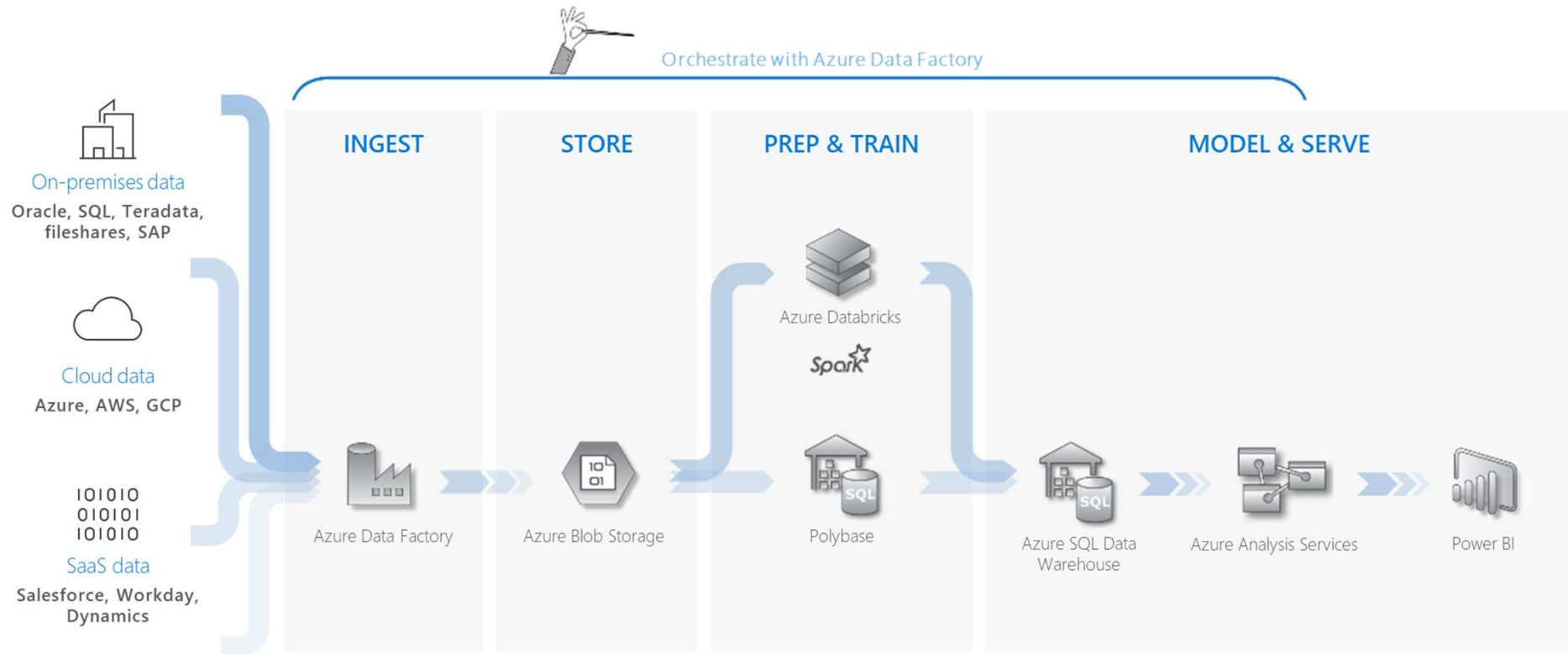
databricks

ICEBERG 

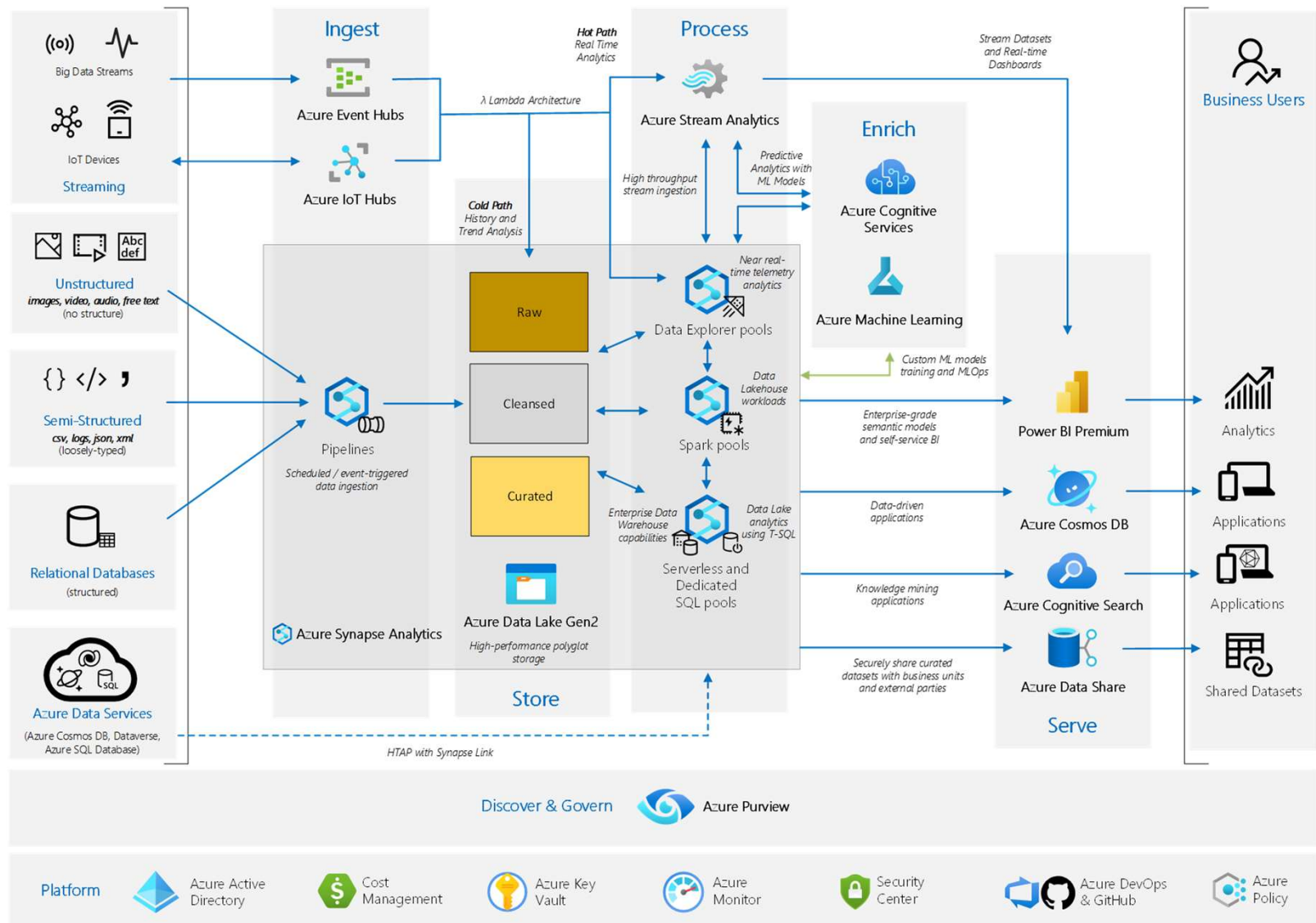
Data Lake Architecture – Concepts, Tools & Process



Modernize your enterprise data warehouse at scale



Microsoft Azure also supports other **Big Data** services like **Azure HDInsight**, **Azure SQL Database** and **Azure Data Lake** to allow customers to tailor the above architecture to meet their unique needs.



Medallion architecture

Although the 3-layered design is common and well-known, there are many discussions on the scope, purpose, and best practices on each of these layers.



Bronze layer

Typically raw, "as-is"

- Maintains the raw state in the structure "as-is"
- Data is immutable (read-only)
- Delivery-based partitioned tables, i.e., YYYYMMDD
- Mostly Parquet. Sometimes other formats
- Can be any combination of streaming and batch transactions
- May include extra metadata (schema)
- May be fed from a "mediation layer"
- Used for debugging, testing



Silver layer

Matched and conformed

- Uses data quality rules for validation
- Usually only functional data
- Historization is merged (SCD2)
- Efficient storage format; Delta
- Versioning for rolling back
- Handles missing or incorrect data
- Usually enriched with reference data
- Source-oriented, although queryable and cluttered around subject areas
- Usually used by operational analytical teams

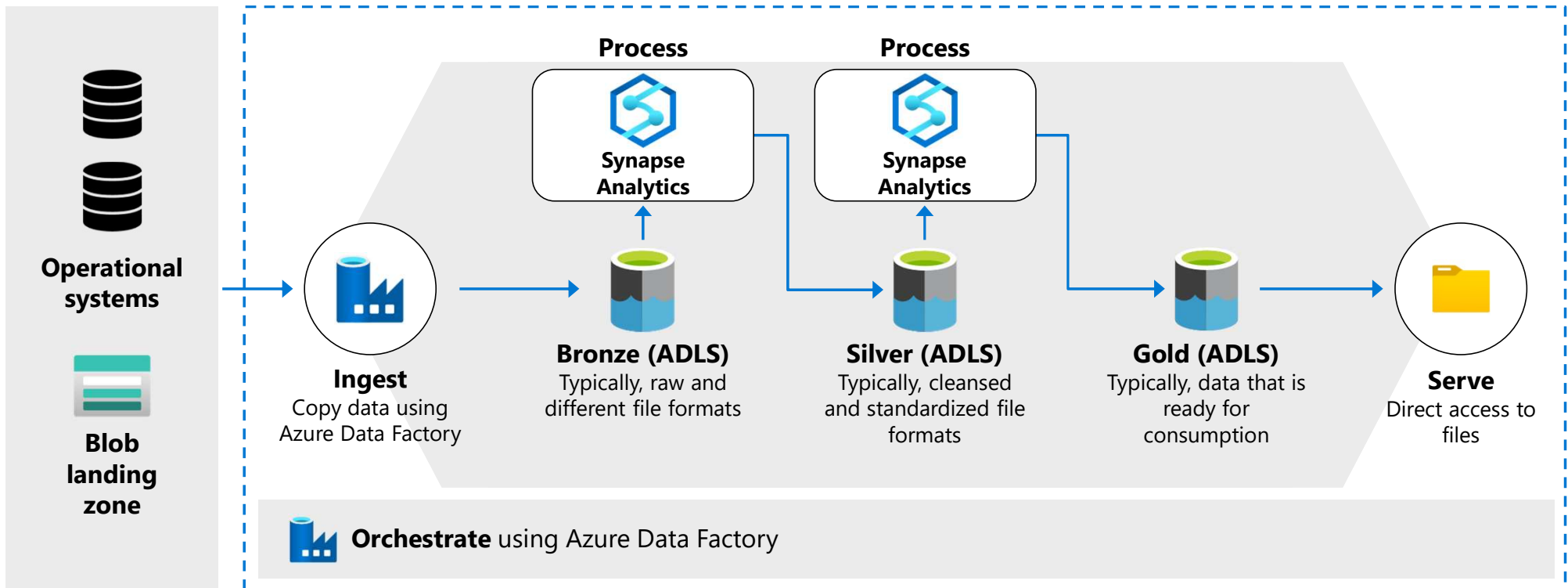


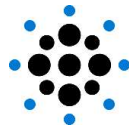
Gold layer

Refined business-level

- What enterprises call data products: consumer-ready / user-friendly data
- Data is highly governed and well-documented
- Historization is applied only for the set of use cases or consumers
- Contains complex business rules, such as calculations and enrichments
- Efficient storage format; Delta
- Versioning for rolling back
- Might contain additional sub layers for sharing or distributing data

Basic Lakehouse Architecture





Metadata Driven
Engineering

Available Frameworks

The screenshot shows the GitHub repository for **SynapseMetadataEngine** by **Andreas-bergsedt**. The repository is public and has 3 branches and 1 tag. The file list includes:

- `.vs`: Synchronized Code to Git Library (2 years ago)
- `SynapseSQLPool`: Added fresh VS DW project (2 years ago)
- `SynapseWorkspace`: Updated to consolidated sp version (2 years ago)
- `_include`: Create head.html (last year)
- `LICENSE`: Initial commit (2 years ago)
- `README.md`: Update README.md (2 years ago)
- `test.md`: Update test.md (last year)

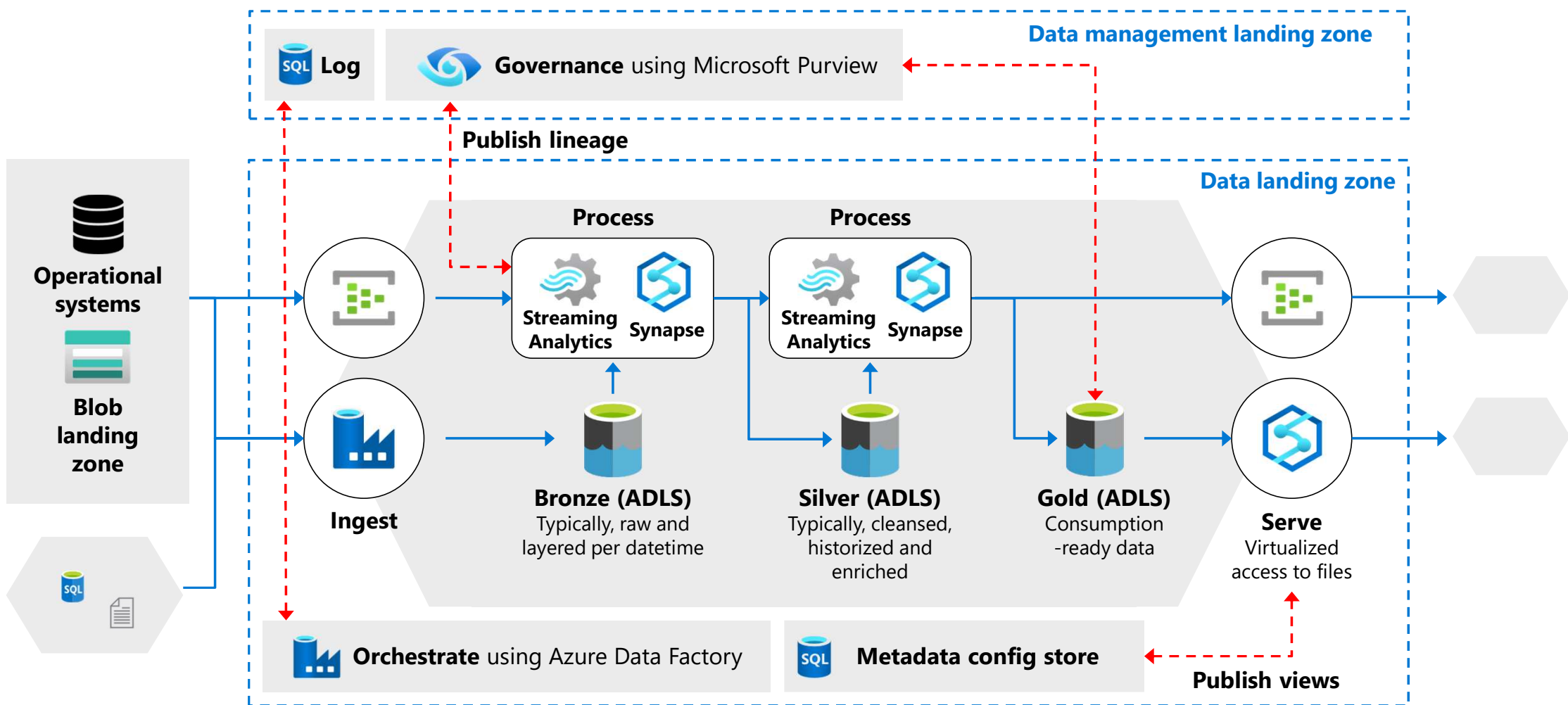
The README section is titled "About the Synapse Metadata Engine project" and describes it as a generic metadata driven framework for building data pipelines.

The screenshot shows GitHub search results for the query **metadata etl**. There are 142 results. The results are filtered by **Repositories** (142), **Issues** (2k), **Pull requests** (4k), **Discussions** (96), and **Users** (2). The languages section lists: **Jupyter Notebook**, **Python**, **Java**, **C#**, **Scala**, **JavaScript**, **Shell**, **Go**, **PLSQL**, **Ruby**, and **More languages...**. The advanced filters show **Owner** and **Size**.

The search results list includes:

- data-solution-automation-engine/data-warehouse-automation-metadata-schema**: Generic interface exchange format for Data Warehouse Automation and ETL generation. (C# · 28 · Updated 3 days ago)
- dagster-io/dagster**: An orchestration platform for the development, production, and observation of data assets. (Python · 7.4k · Updated 13 minutes ago)
- recap-build/recap**: Recap tracks and transform schemas across your whole application. (Java · 239 · Updated 12 days ago)
- dswarm/dswarm** (Public archive): an open-source data management platform for knowledge workers (<https://github.com/dswarm/dswarm-documentation/wiki>) (Java · 57 · Updated on 19 Dec 2017)
- Datavault-UK/automate-dv**: A free to use dbt package for creating and loading Data Vault 2.0 compliant Data Warehouses (powered by dbt, an open source data engineer...) (metadata sql etl snowflake datawarehousing)

The Lakehouse Pattern in Azure

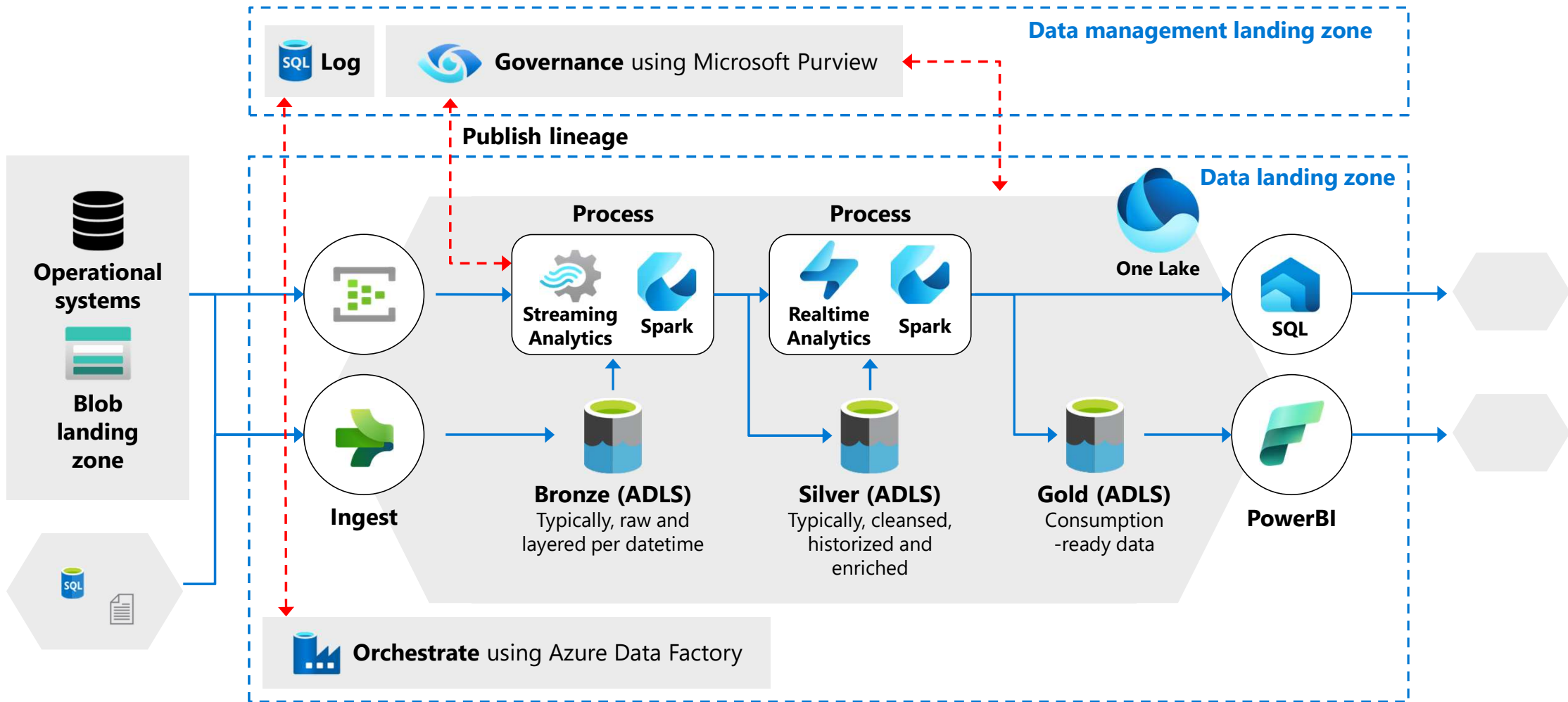


Metadata Driven Engineering Examples

The screenshot displays the Synapse Studio interface, illustrating a metadata-driven engineering workflow. The interface is divided into several key sections:

- Integrate Panel (Left):** Contains a search bar and a list of pipelines. The pipelines are organized into folders:
 - childPipeline:** Includes `executeDWProcedure`, `parquet2Silver`, `ProcessBronze2Silver`, and `sql2parquet`.
 - controlPipeline:** Includes `getsqlMetadadata`, `Process_Raw_Jobs`, `Process_Source_Jobs`, `Process_Silver_Jobs`, and `RunRawJobs`.
 - orchestrationPipeline:** Includes `dailyBatch` and `hourlyBatch`.
- Activities Panel (Center):** Features a search bar and a list of activities categorized by Synapse, Move & transform, Azure Data Explorer, Azure Function, Batch Service, Databricks, Data Lake Analytics, General, HDInsight, Iteration & conditionals, and Machine Learning.
- Main Canvas:** Displays a workflow diagram with three main stages, each represented by an 'Execute Pipeline' task:
 - Stage 1:** `ValidateMetadata` (task: `getsqlMetadadata`).
 - Stage 2:** `ProcessRawJobs` (task: `Process_Raw_Jobs`).
 - Stage 3:** `ProcessSilver2Gold` (task: `Process_Silver_Jobs`).

The Lakehouse Pattern in Microsoft Fabric

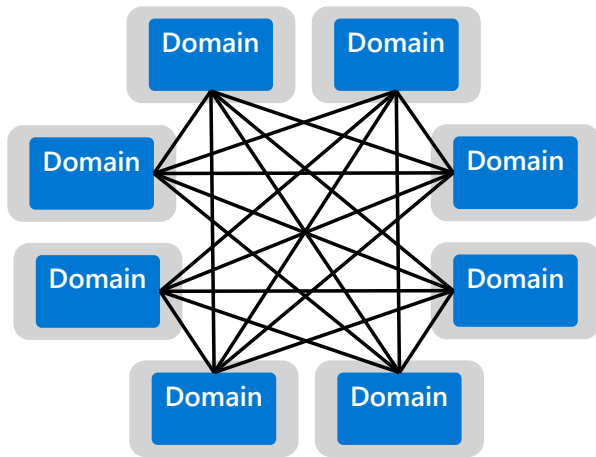




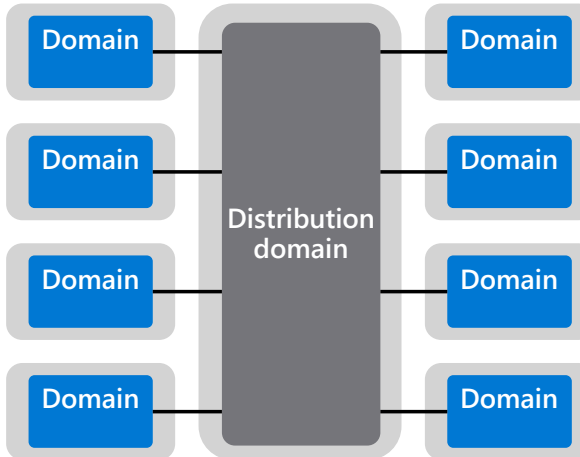
Driving at Scale

 = team independency

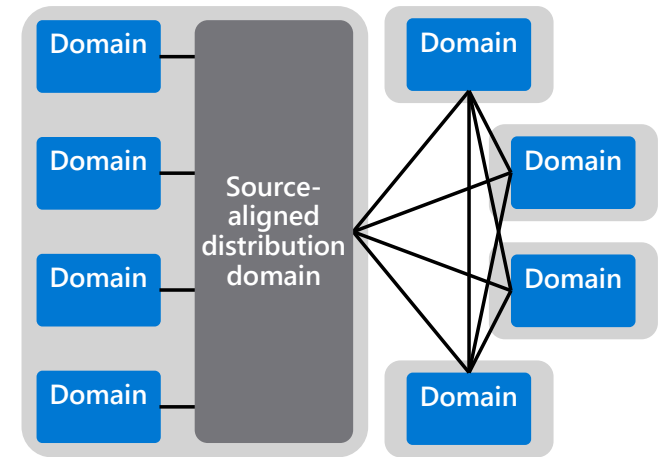
Fine-grained fully federated mesh



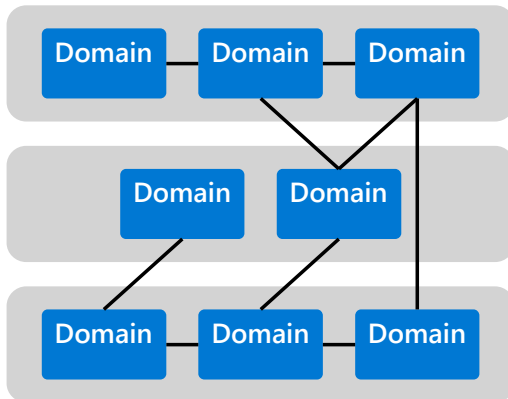
Fine-grained and fully governed mesh



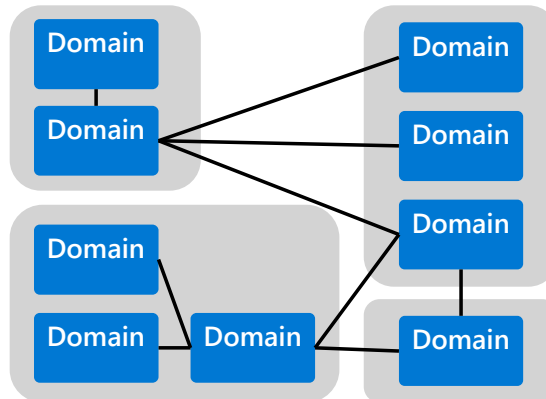
Hybrid federated mesh



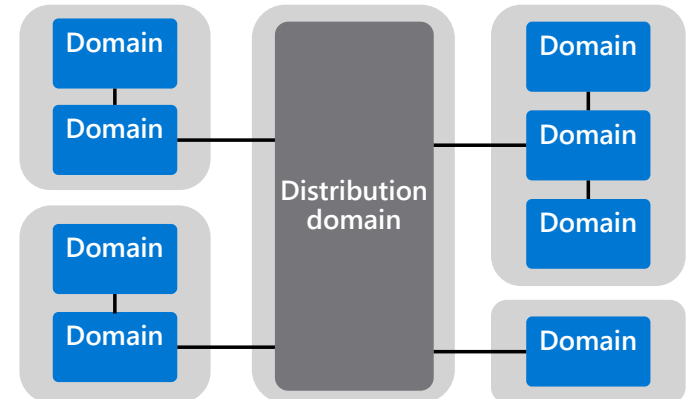
Value chain-aligned mesh



Coarse grained aligned mesh

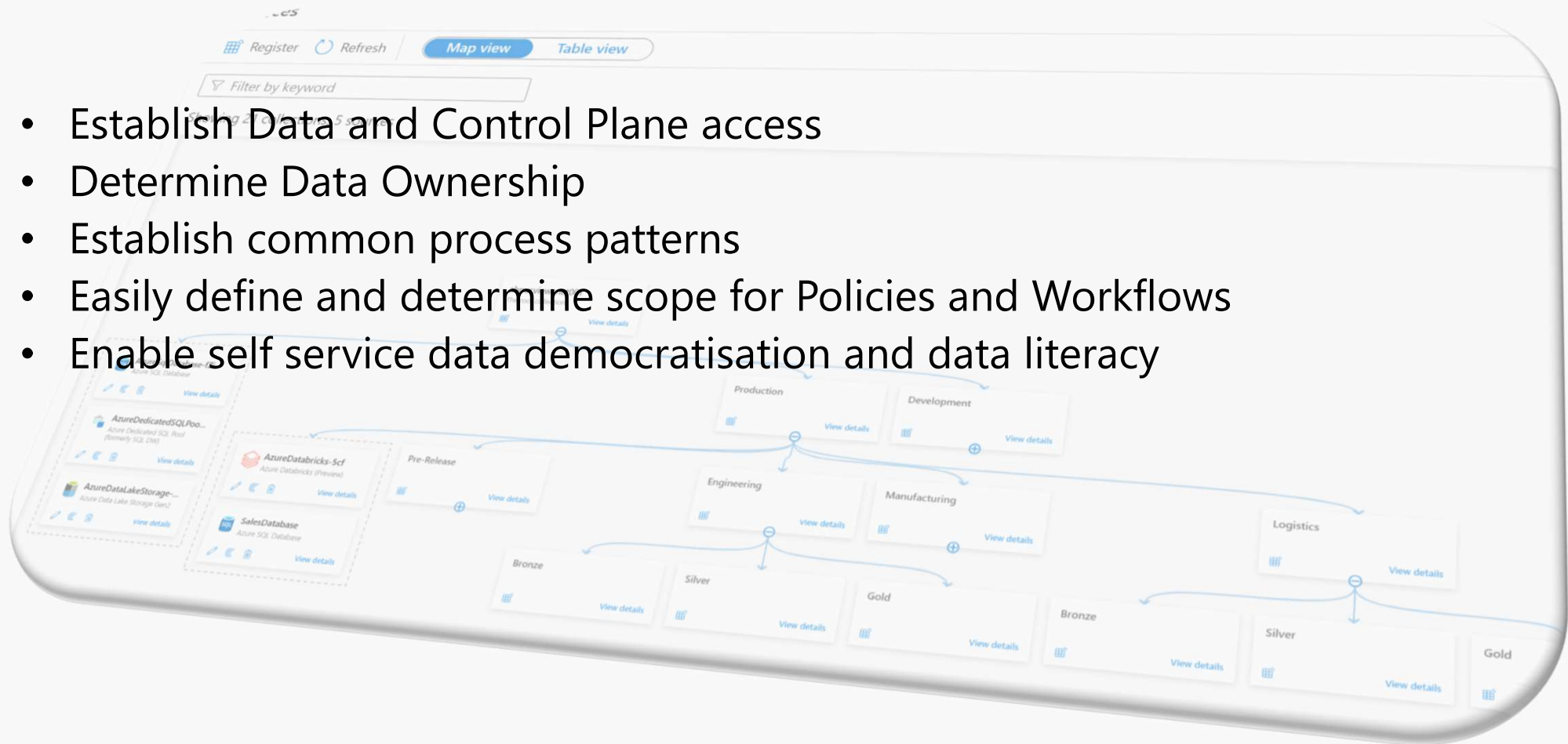


Coarse grained and governed mesh

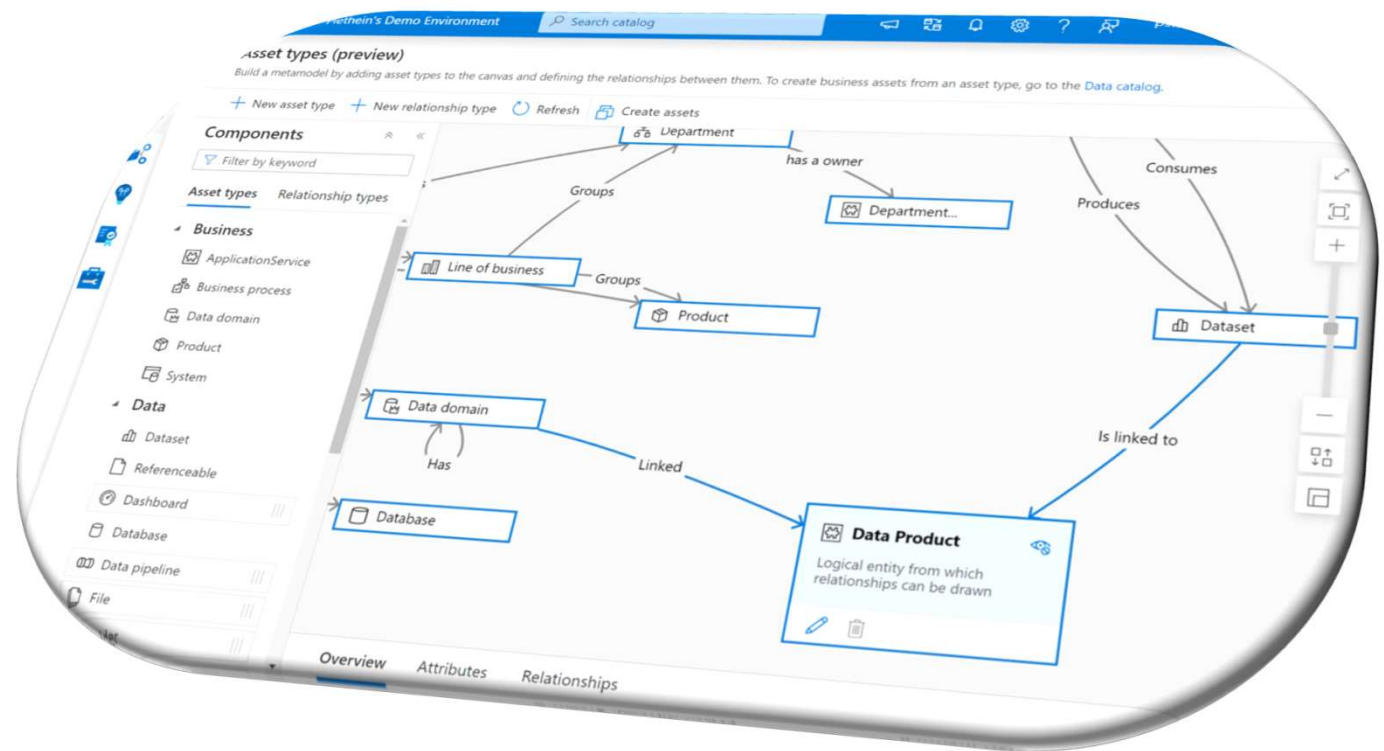


Define the Lakehouse in Purview

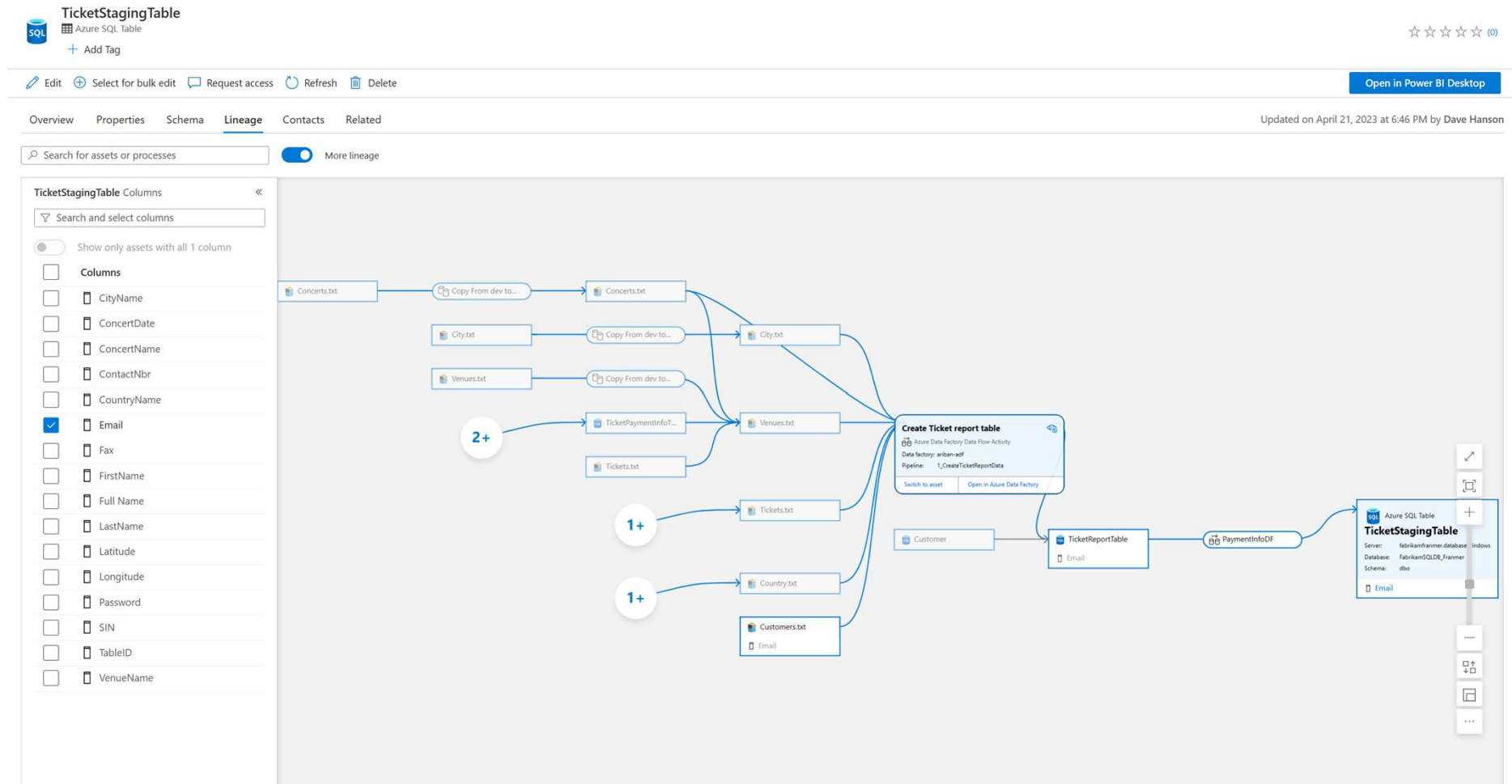
- Establish Data and Control Plane access
- Determine Data Ownership
- Establish common process patterns
- Easily define and determine scope for Policies and Workflows
- Enable self service data democratisation and data literacy



Encourage and practice good stewardship



Use native capabilities to enable visibility



- The Lakehouse pattern is a scalable long term supportable concept
- Metadata driven data engineering provides clear data boundaries at scale
- Data engineering goes hand in hand with data governance

<https://github.com/Andreas-berstedt/Metdata2Lakehouse>



Key take-aways

