The Problem

What we hear from IT ...

"ETL is causing our EDW project to fail" "I wish we could do predictive analytics with our data"

"We failed at our last EDW project"

"We have lots of data today, but we don't know how to do analytics on it"

"We'd like to integrate social media but don't know how"

What we hear from business users ...

"The data warehouse data is nightly, I need real-time data"

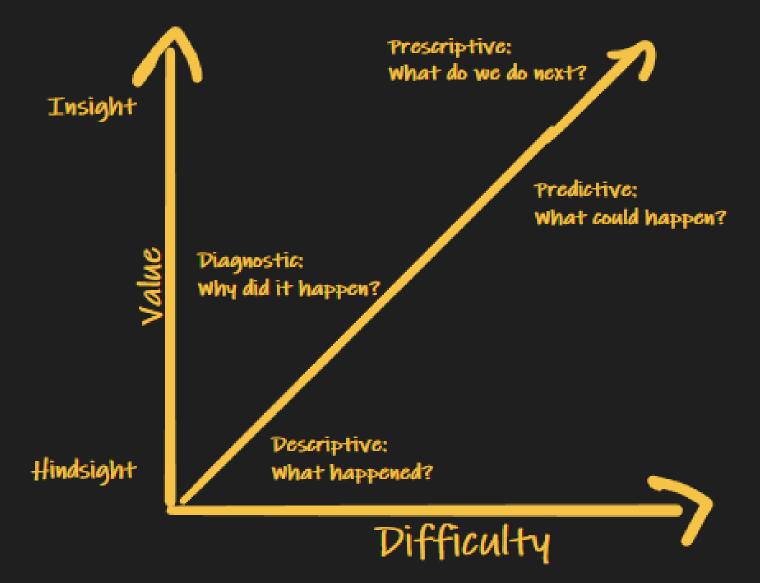
"I want to control my own data"

"Even with the data warehouse and reports, I do analysis in spreadmarts"

"I just want to get my job done"

"The data warehouse doesn't answer my questions"

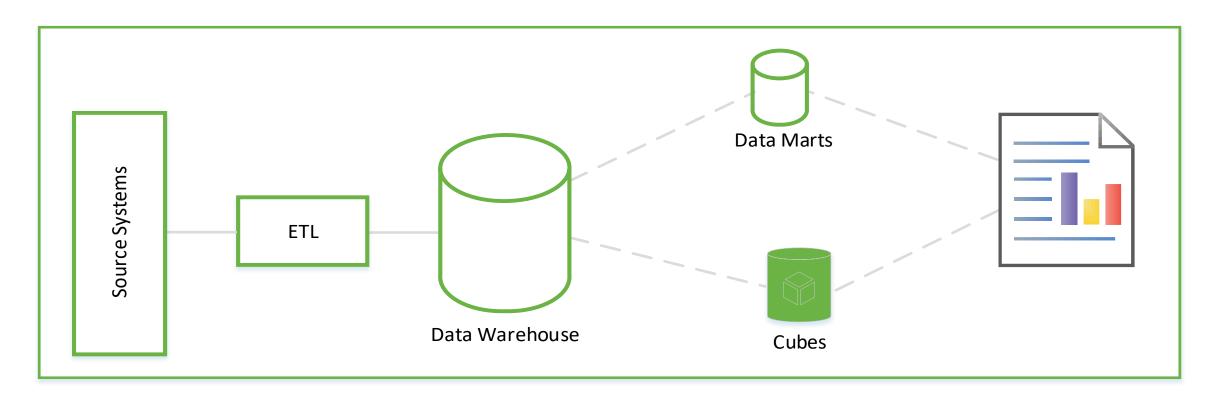
Analytics Maturity Models



Past Solutions

Legacy Thinking

The Philosophy: Model data » Transform data » Load data » Understand data

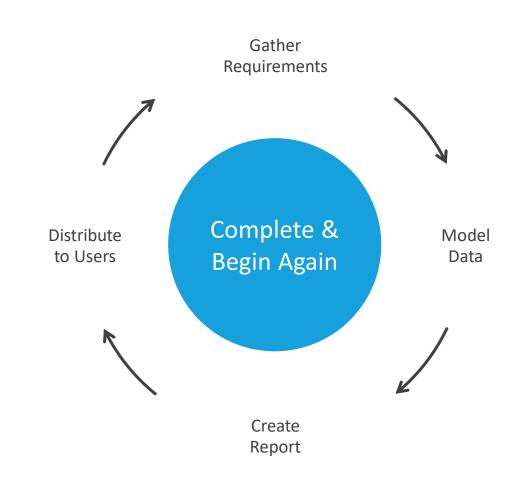


Data Projects have a high fail rate

Too much time is spent in:

- Requirements gathering
- Data modeling
- ETL

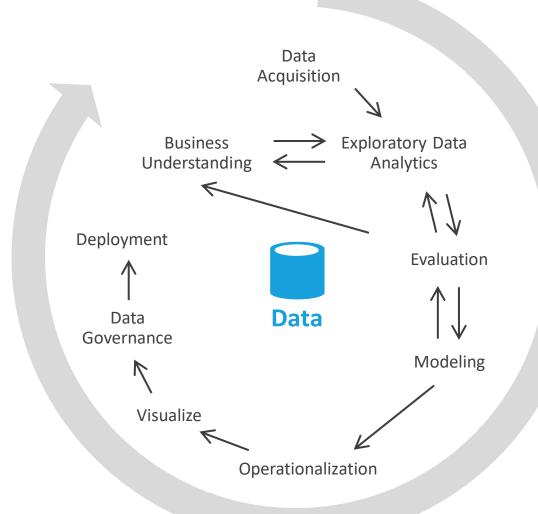
Users only see the fruits of the endeavor after the reports are created



Our Modern Approach

Data Sandboxing

- A robust and well-proven methodology.
- Data science-like.
- Iterative.
- Stresses up-front understanding of data.
- Modeling is done later in the process (schema-on-read).
- ETL might not be needed



Ingest all Data

Extract and Load, NO Transform

Store all data

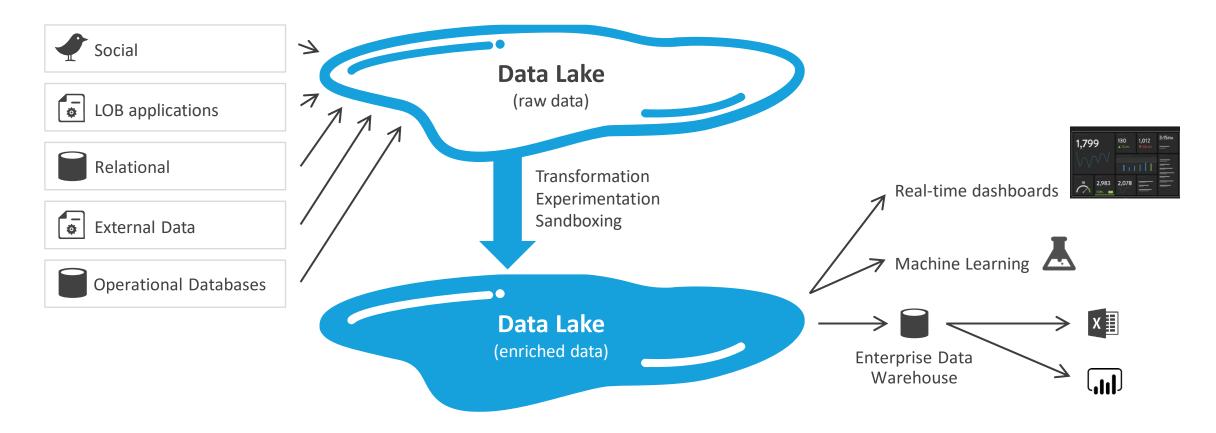
In native format

Do analysis

Using almost any tool

Operationalize

Create schemas and pipelines



Data Catalog



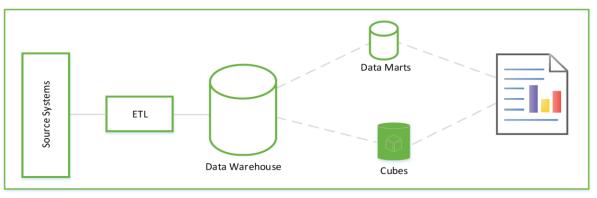
Self-Service Enabler

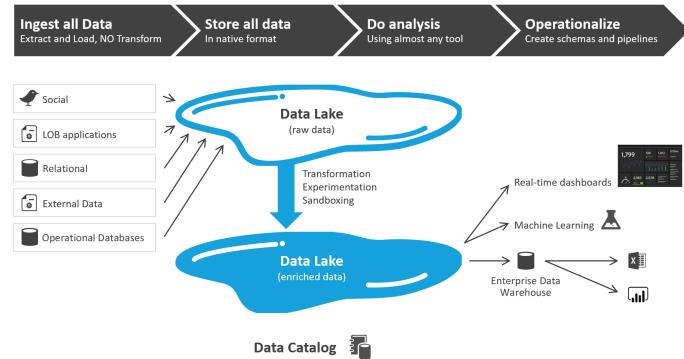
A Data Lake solves 80% of analytical needs

It is not meant to provide operational reporting

- Data ingestion is more real-time, enabling prediction
- The Data Lake, as a source of all data, is built to efficiently feed a data warehouse.
- Fetches all data, no longer have to go back to source systems for minor changes

Real-World Example – Customer 360





Data Lake Design-Folder Structures

A Data Lake Is Just a Folder Structure with Smart Organization and significant processing power

Authentication/Governance

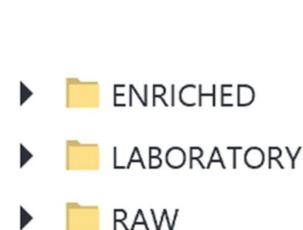
Accounts/Folders/Files

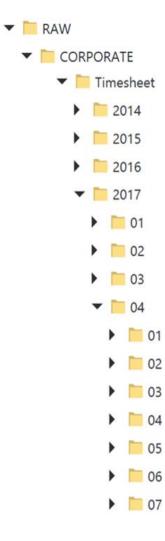
Obvious, Self-documenting Paths

- dev/raw/{datasource}/{object}/YYYYMMDD/
- dev/reject/{datasource}/{object}/YYYYMMDD/
- prod/snapshot/{datasource}/{object}/YYYYMMDD/
- laboratory/jsmith

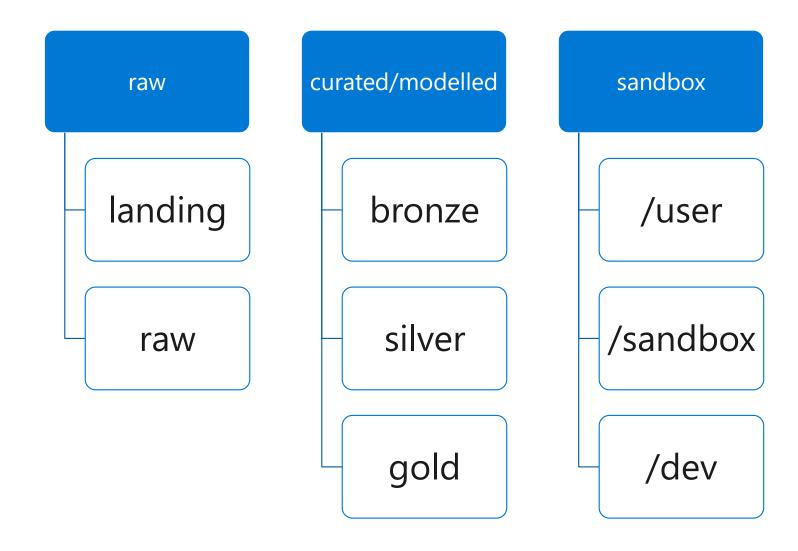
Time partitioning schemes are important

- AVRO/TPFS format
- .../YYYY/MM/DD/HH/MM

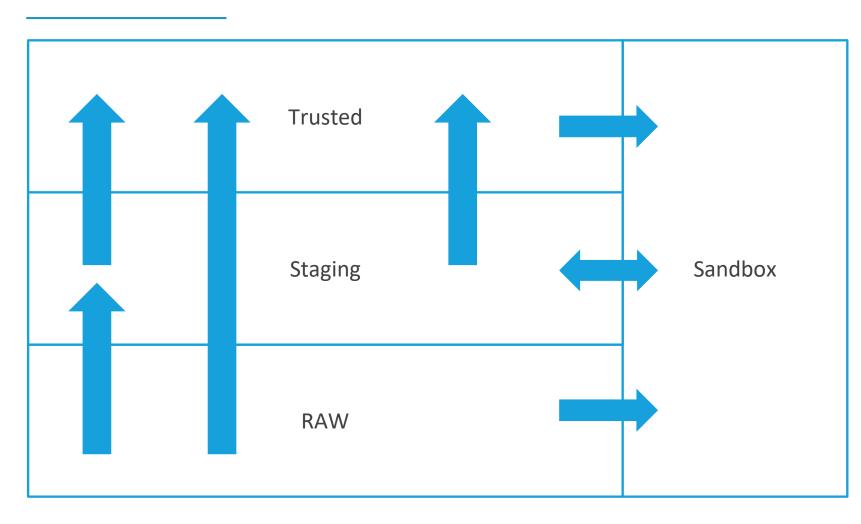




Physical Structure



Analytics workflow



- Development /
 Data Ingestion
- 2 Production
- Data Scientists /
 Business Analysts/
 Developers
- 4 Development
- 5 Production

Data Lake vs. Data Warehouse

Data Lake	Data Warehouse
Complementary to the EDW	Can be sourced from the Data Lake
Load first, understand later	Understand first, load later
Schema-on-read	Schema-on-write
System of Insight	System of Record
Detailed Data	Refined Data
Supports varied data formats	Structured data
Adapts to changing requirements	Difficult to change structure
Optimized for Cost	Optimized for Performance

Microsoft Azure Data Catalog







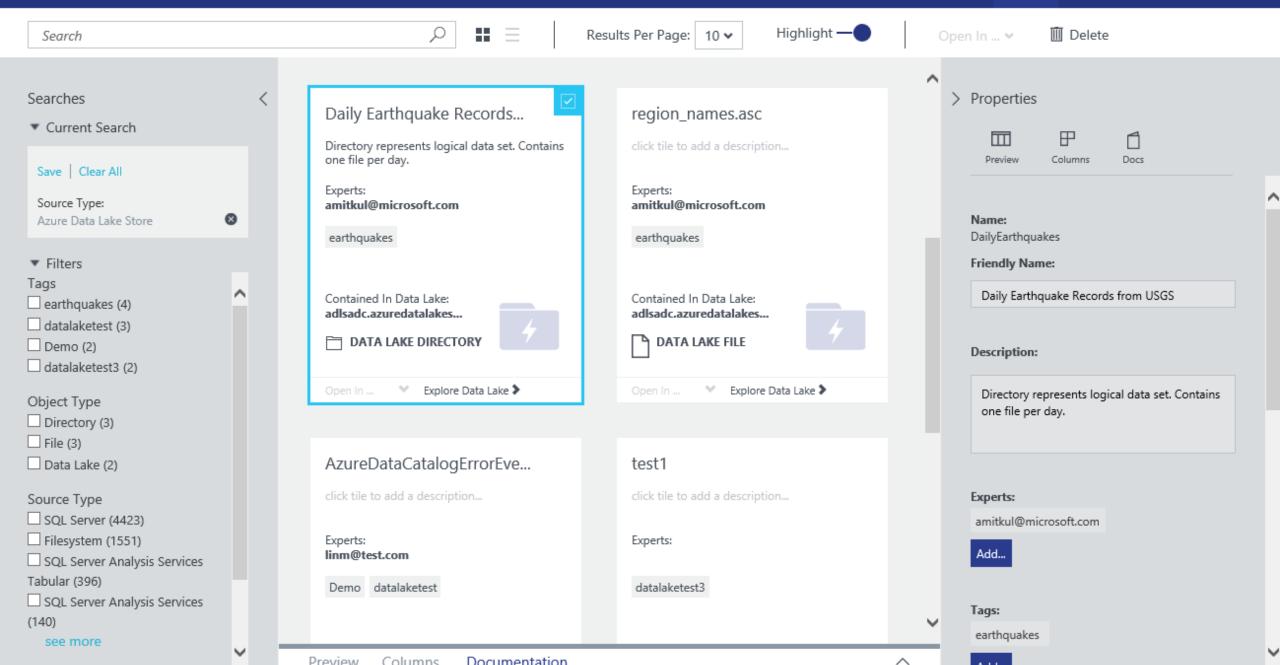
Settings







User



We have a winning idea, help us operationalize it

Ingest all Data

Extract and Load, NO Transform

Store all data

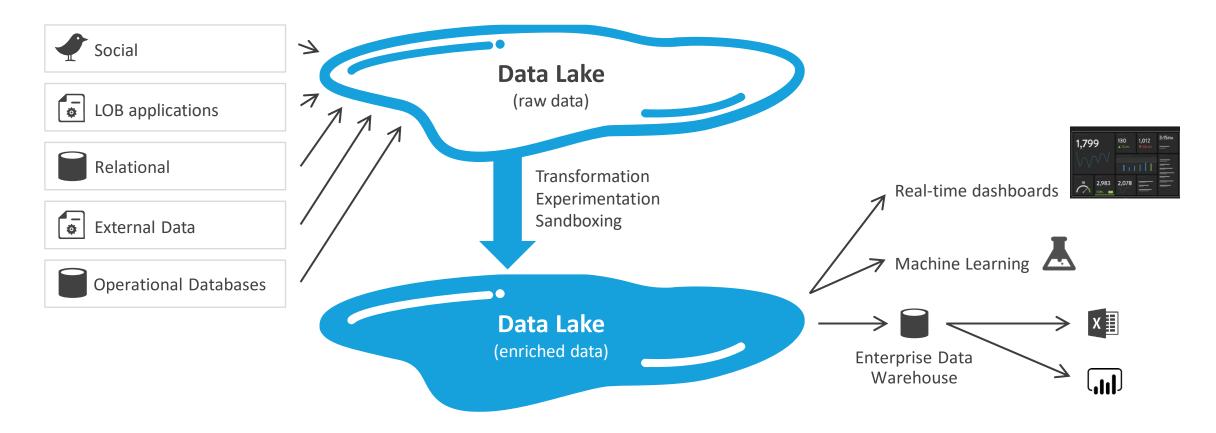
In native format

Do analysis

Using almost any tool

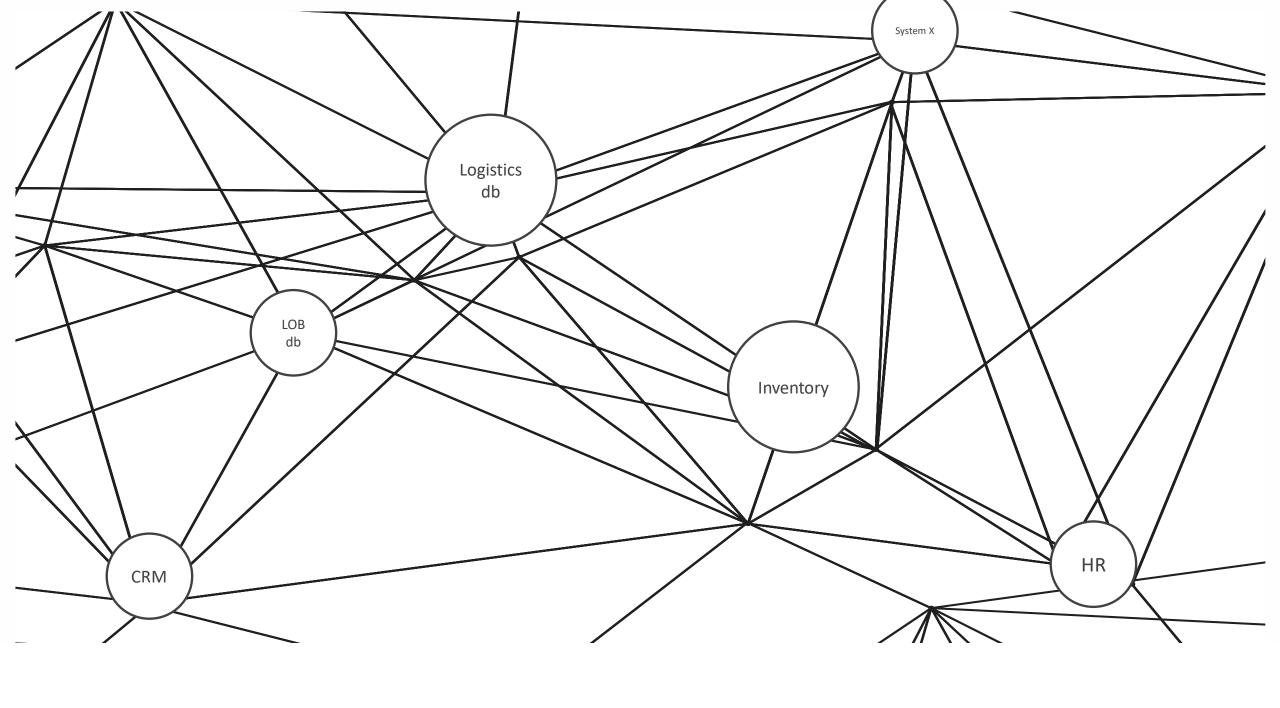
Operationalize

Create schemas and pipelines



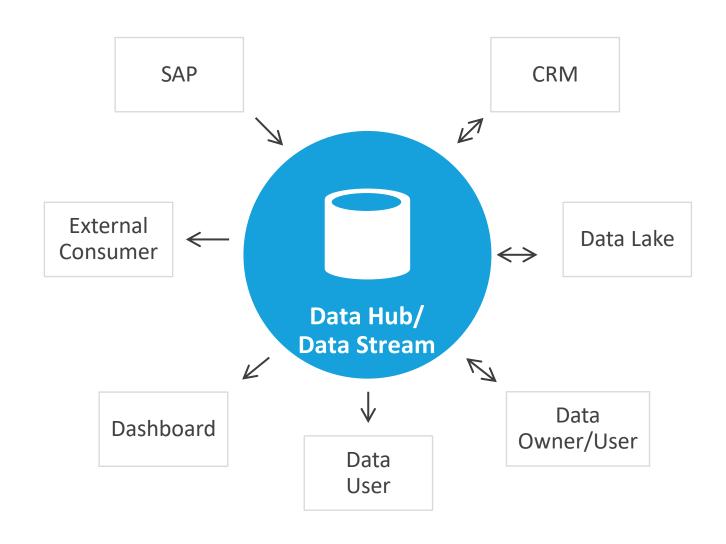
Data Catalog



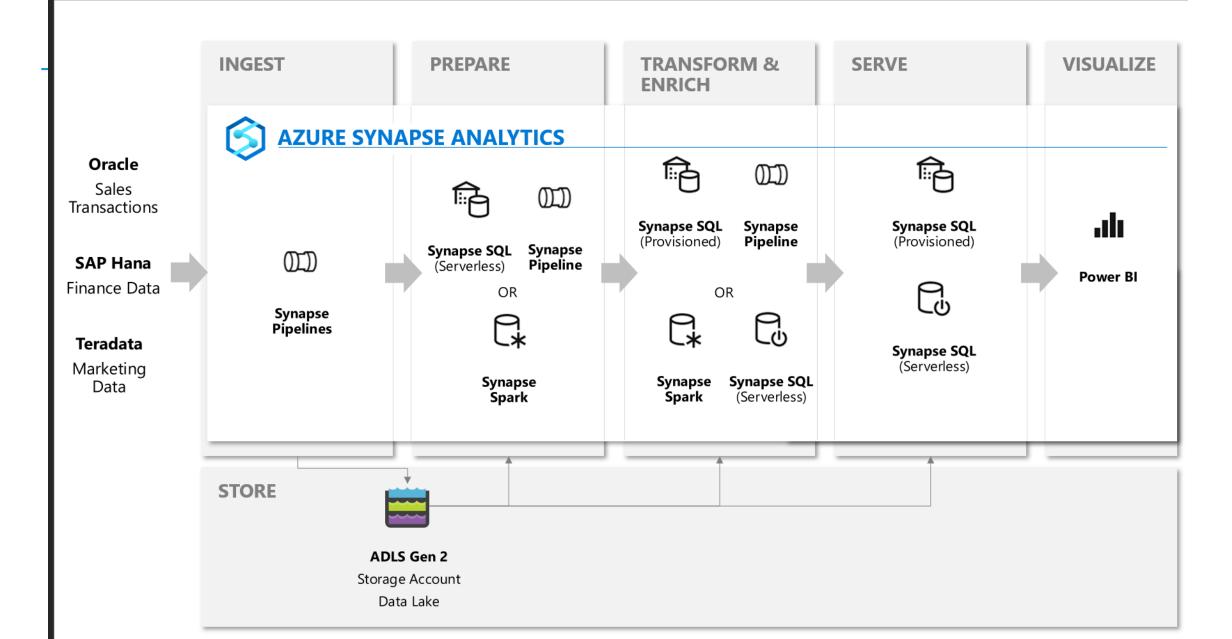


Data Ingestion Factory

(Kappa Architecture)



Modern Data Warehouse



Handling Streaming Data





