

Azure Data Lake: What, Why, and How

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Agenda

Azure Data Lake: What, Why, and How

- Data Lake Overview & Use Cases
- Big Data in Azure
- Data Storage in Azure
- Compute in Azure
- Integrating Azure Data Lake in a Multi-Platform Architecture
- Suggestions for Getting Started with a Data Lake Project

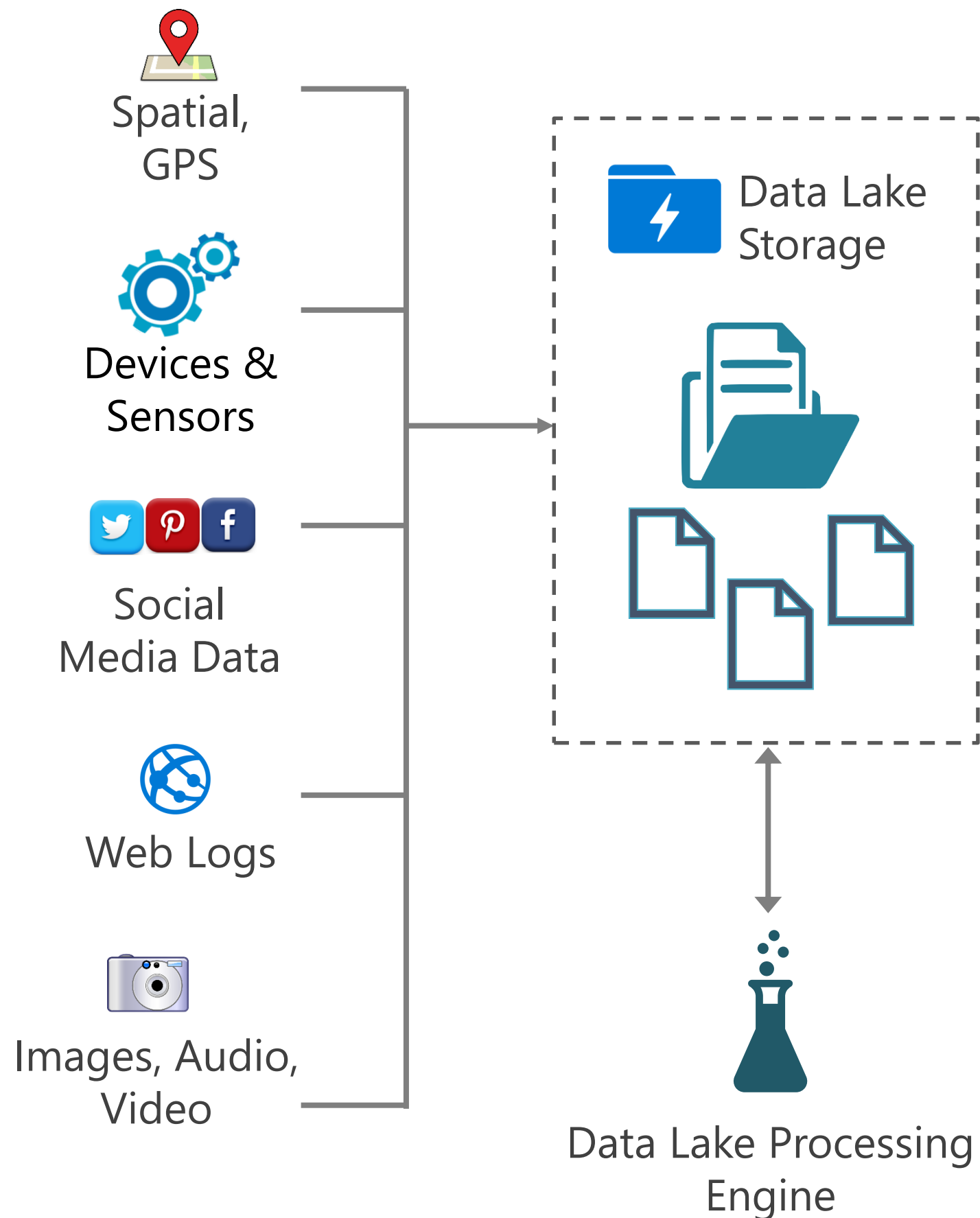
As of
Jan 2019:

Things are changing rapidly. Azure Data Lake Storage Gen2 is in public preview.

Data Lake Overview & Use Cases



What is a Data Lake?

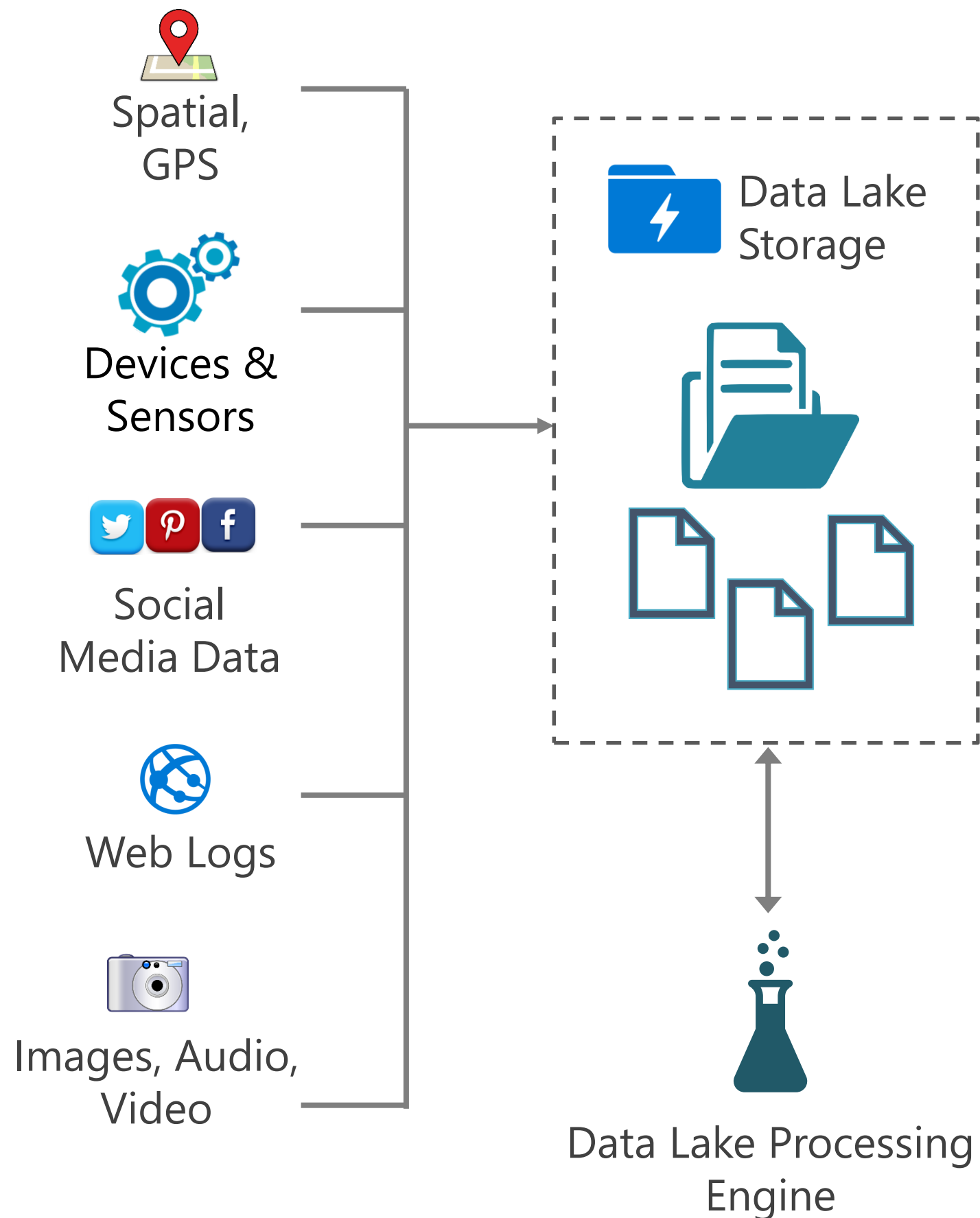


A **repository** for storing large quantities of disparate sources of data in its native format

One **architectural platform** to house all types of data:

- ✓ Machine-generated data (ex: IoT, logs)
- ✓ Human-generated data (ex: tweets, e-mail)
- ✓ Traditional operational data (ex: sales, inventory)

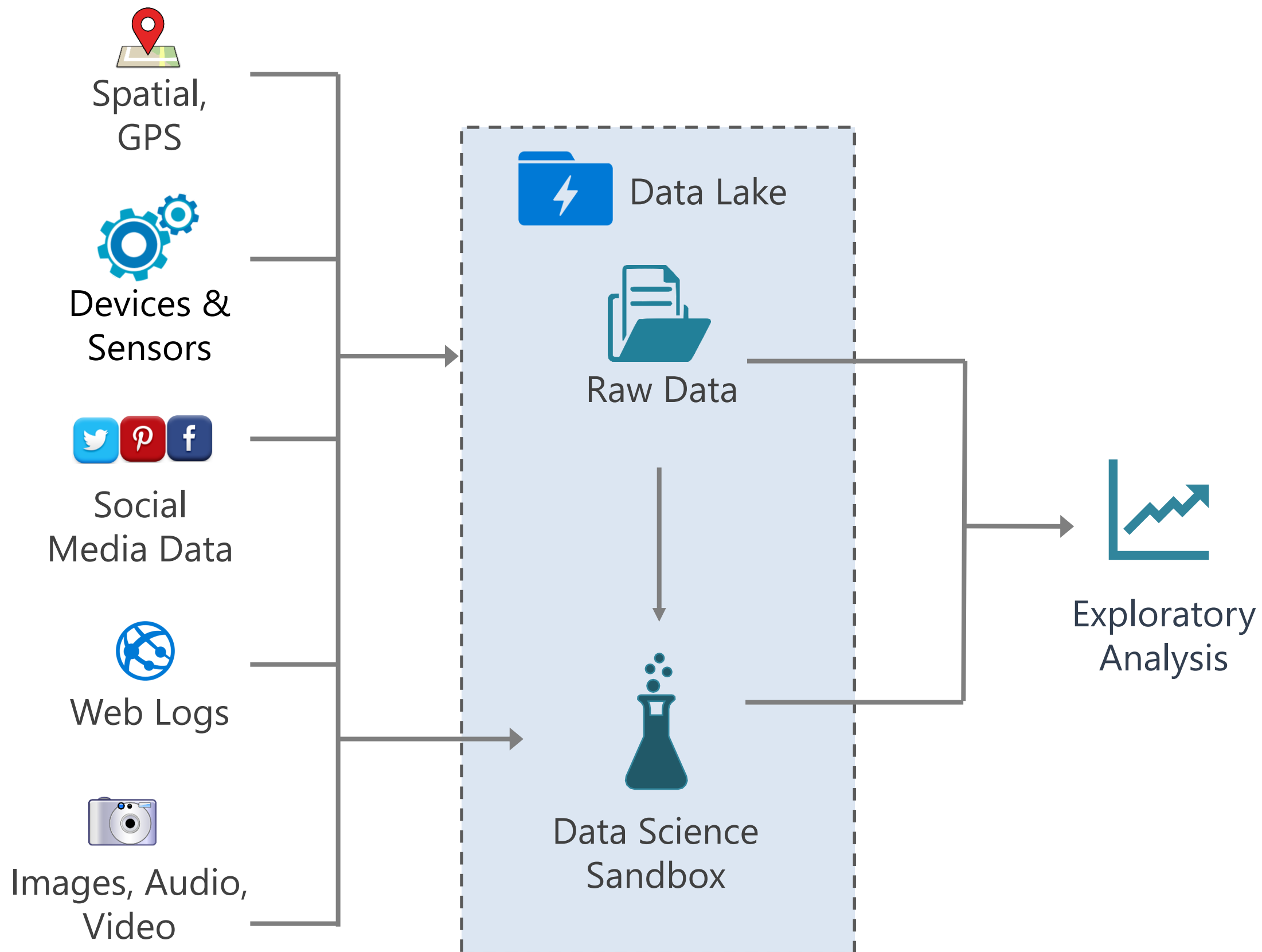
Objectives of a Data Lake



- ✓ Reduce up-front effort to ingest data
- ✓ Defer work to 'schematize' until value is known
- ✓ Allow time for defining business value of the data
- ✓ Store low latency data
- ✓ Access to new data types
- ✓ Facilitate advanced analytics scenarios & new use cases
- ✓ Store large volumes of data cost efficiently

Data Lake Use Cases

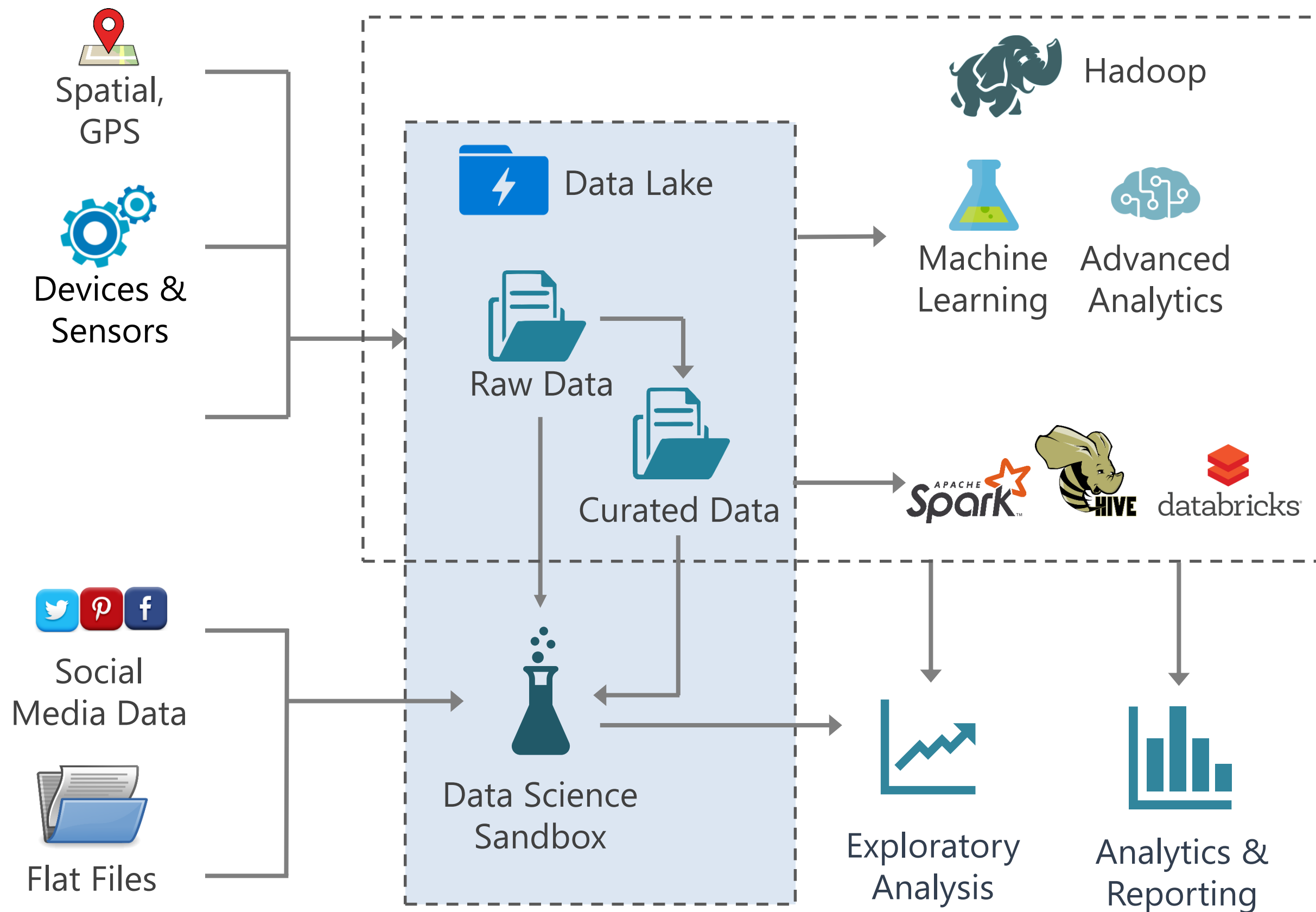
Ingestion of New File Types



- ✓ Preparatory file storage for multi-structured data
- ✓ Exploratory analysis to determine value of new data types & sources
- ✓ Affords additional time for longer-term planning while accumulating data or handling an influx of data

Data Lake Use Cases

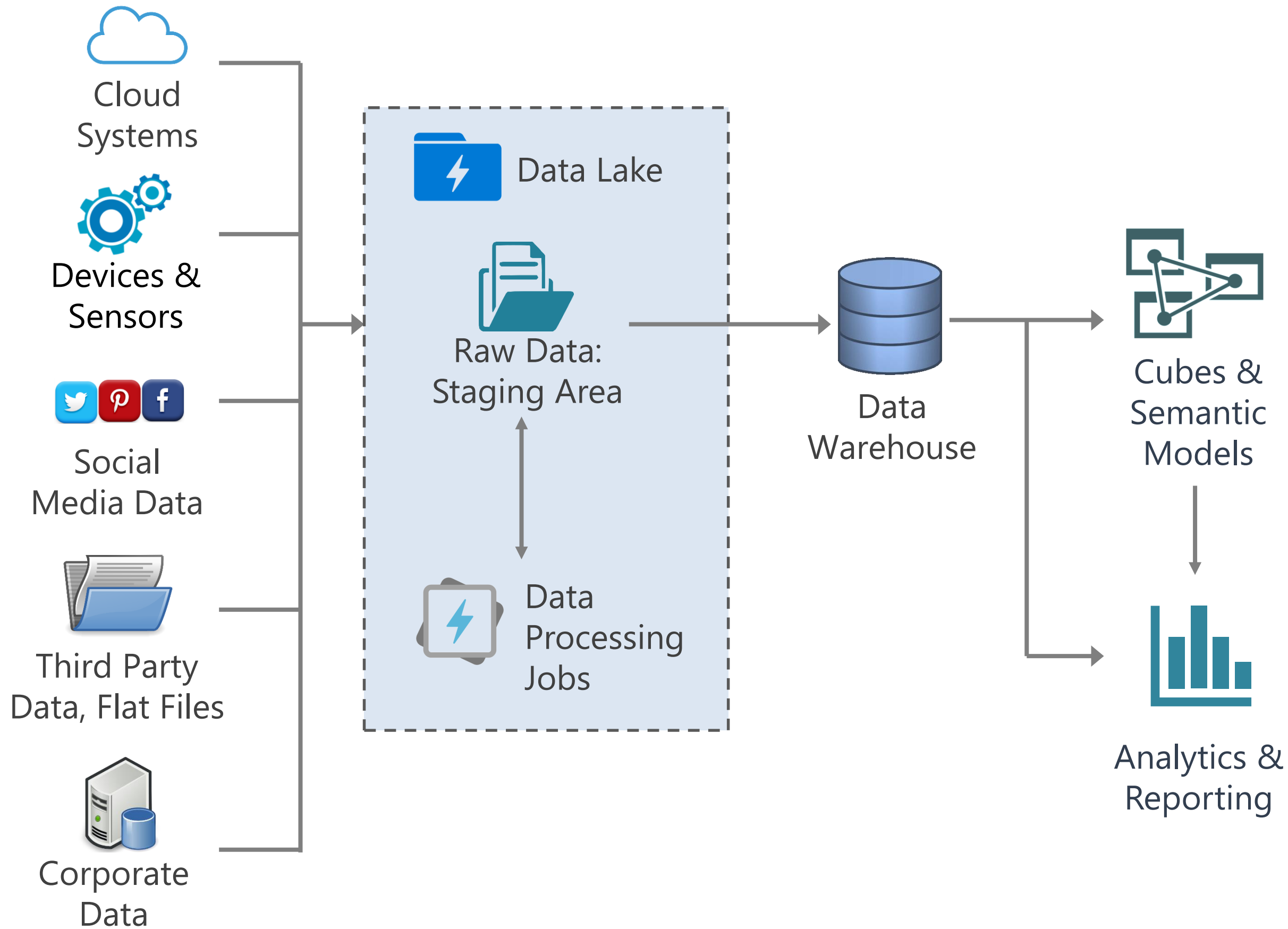
Data Science Experimentation | Hadoop Integration



- ✓ Big data clusters
- ✓ SQL-on-Hadoop solutions
- ✓ Integrate with open source projects such as Hive, Spark, Storm, Kafka, etc.
- ✓ Sandbox solutions for initial data prep, experimentation, and analysis
- ✓ Migrate from proof of concept to operationalized solution

Data Lake Use Cases

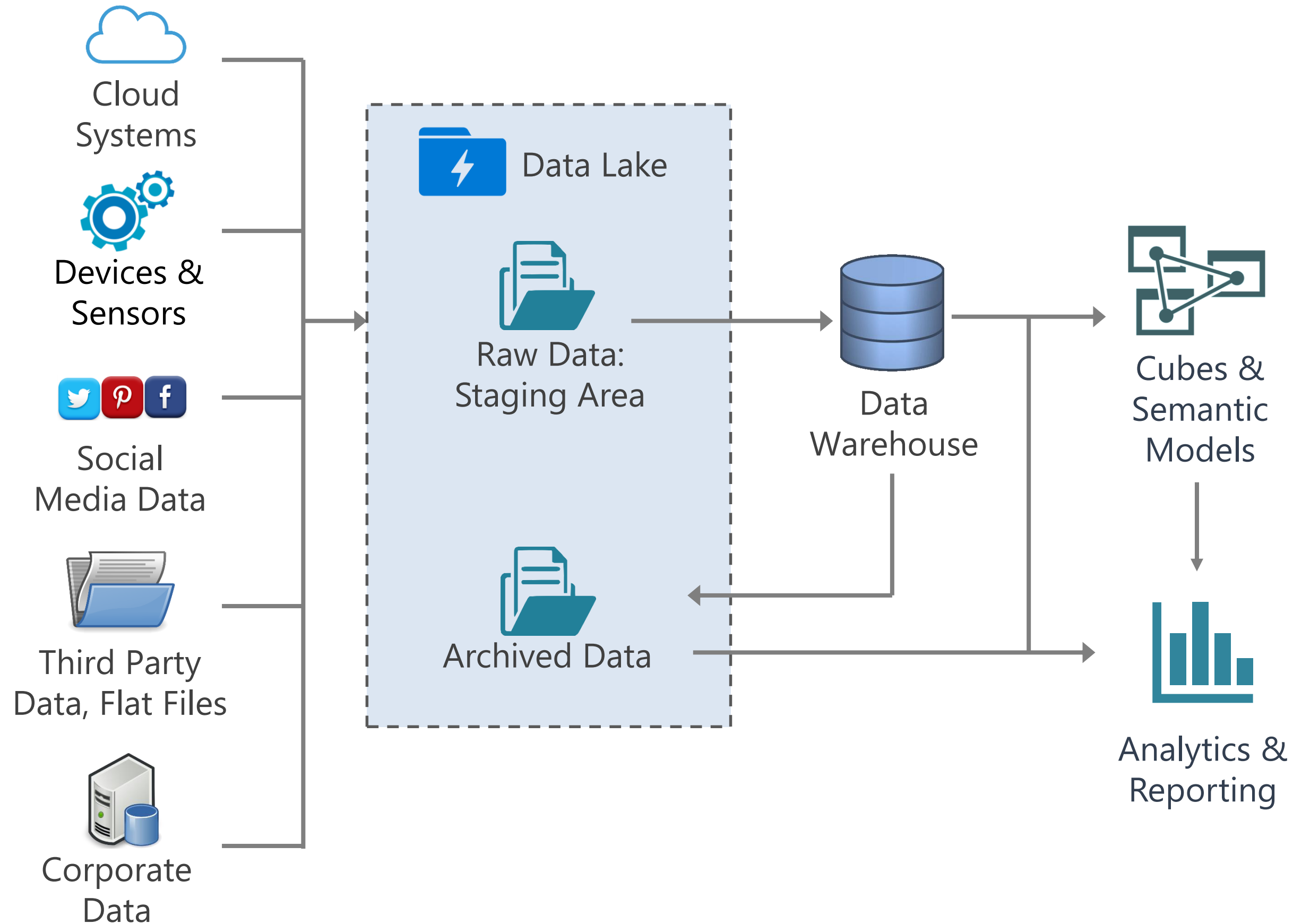
Data Warehouse Staging Area



- ✓ ELT strategy (extract>load>transform)
- ✓ Reduce storage needs in relational platform by using the data lake as landing area
- ✓ Practical use for data stored in the data lake
- ✓ Potentially also handle data transformations in the data lake

Data Lake Use Cases

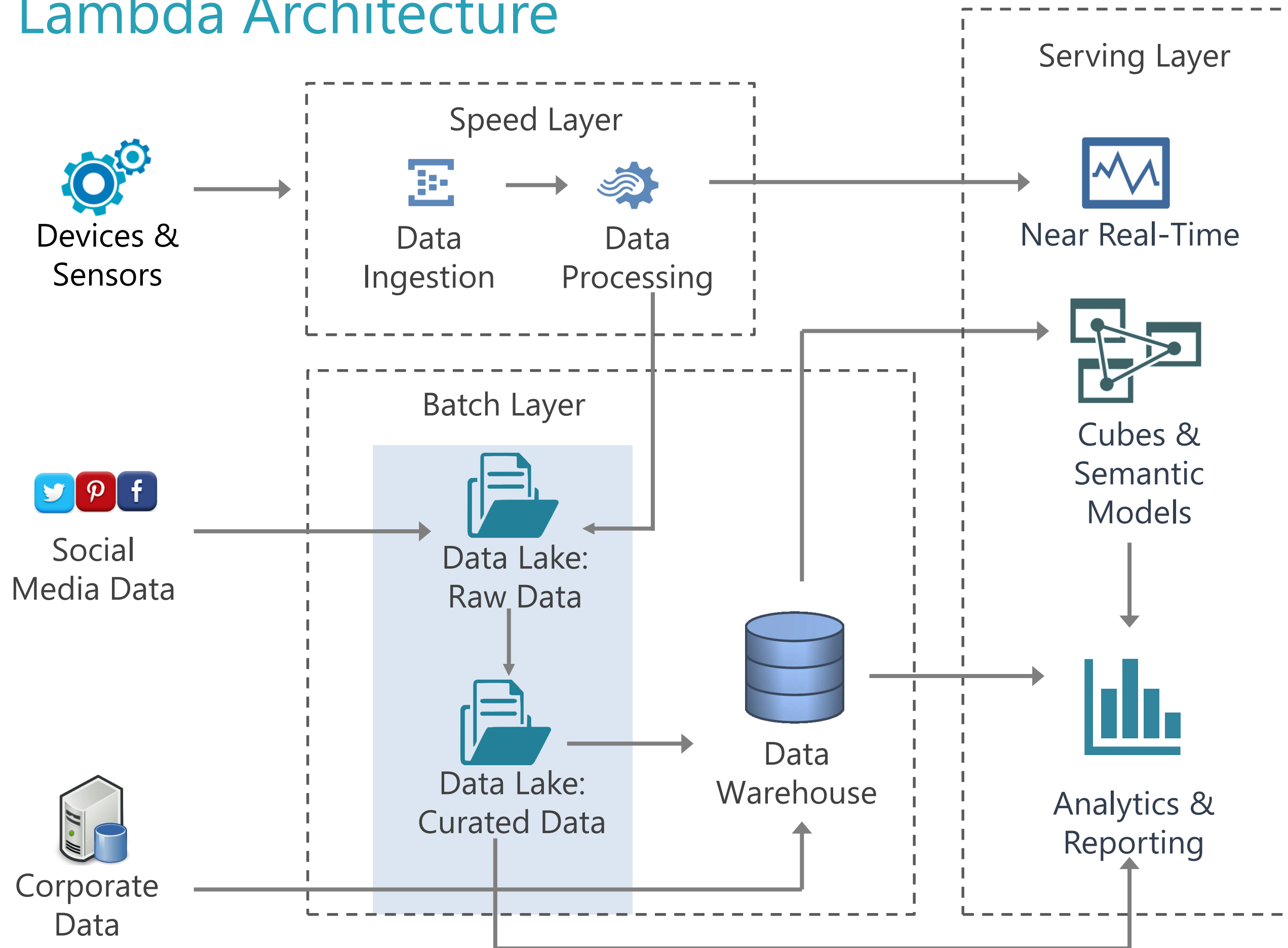
Active Archiving



- ✓ Offload aged data from data warehouse back to the data lake
- ✓ An "active archive" available for querying when needed
- ✓ Federated queries to access:
current data in the DW +
archive data in the data lake

Data Lake Use Cases

Lambda Architecture

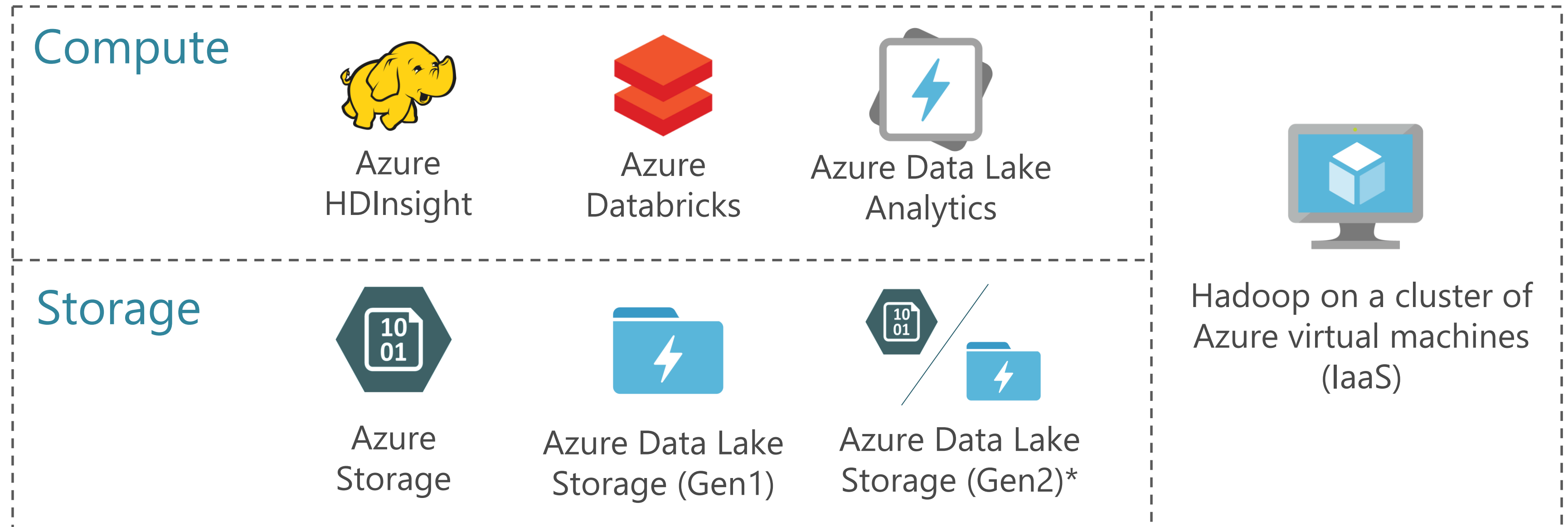


- ✓ Support for low-latency, high-velocity data in near real time
- ✓ Support for batch-oriented operations

Big Data in Azure



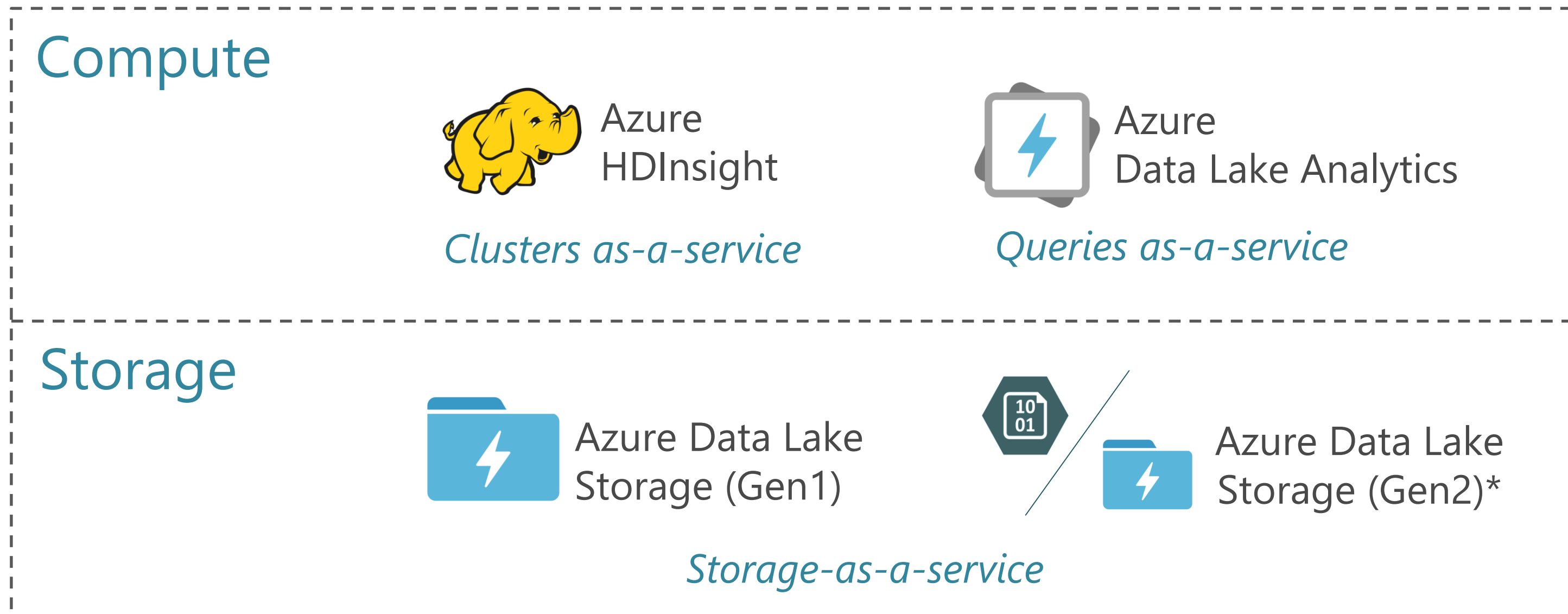
Big Data in Azure



*In Public Preview

Azure Data Lake

Azure Data Lake is a collection of the following services:

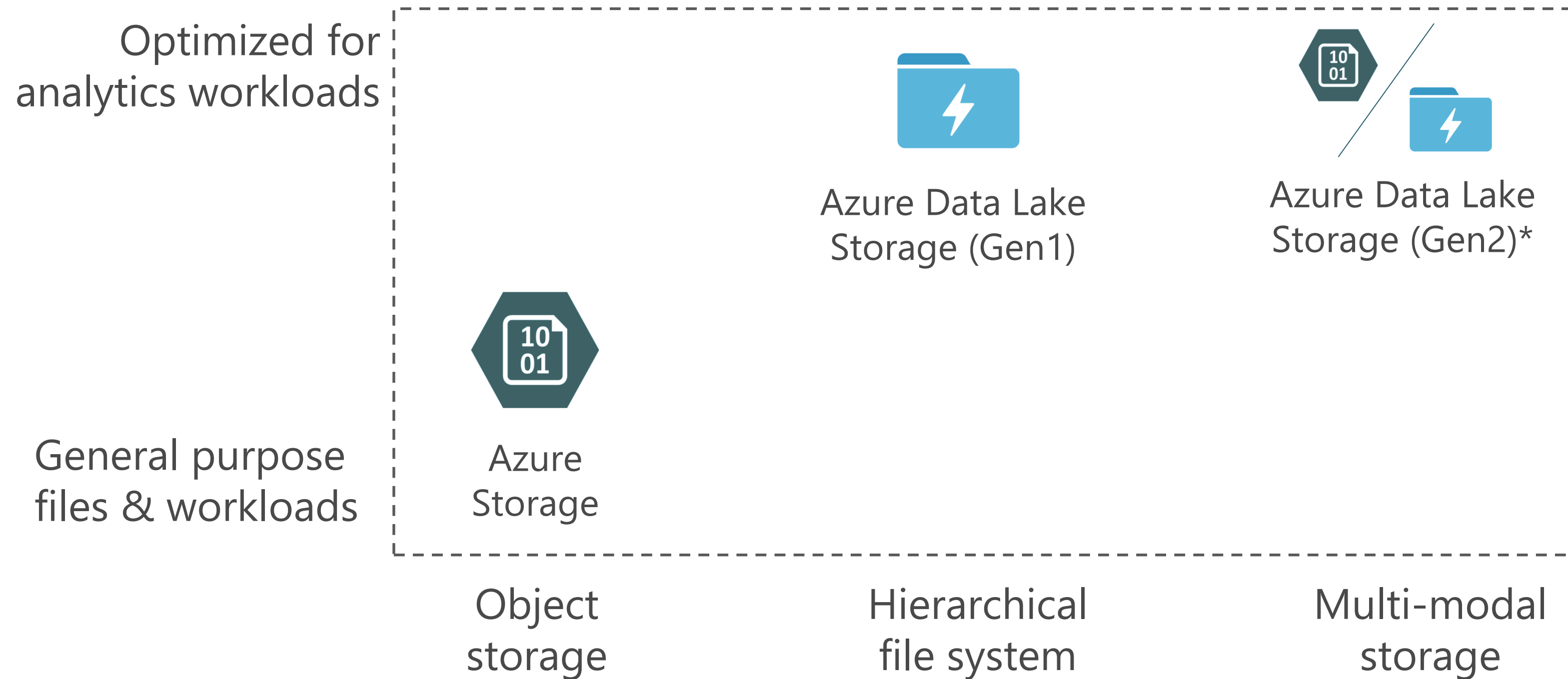


Data Storage in Azure



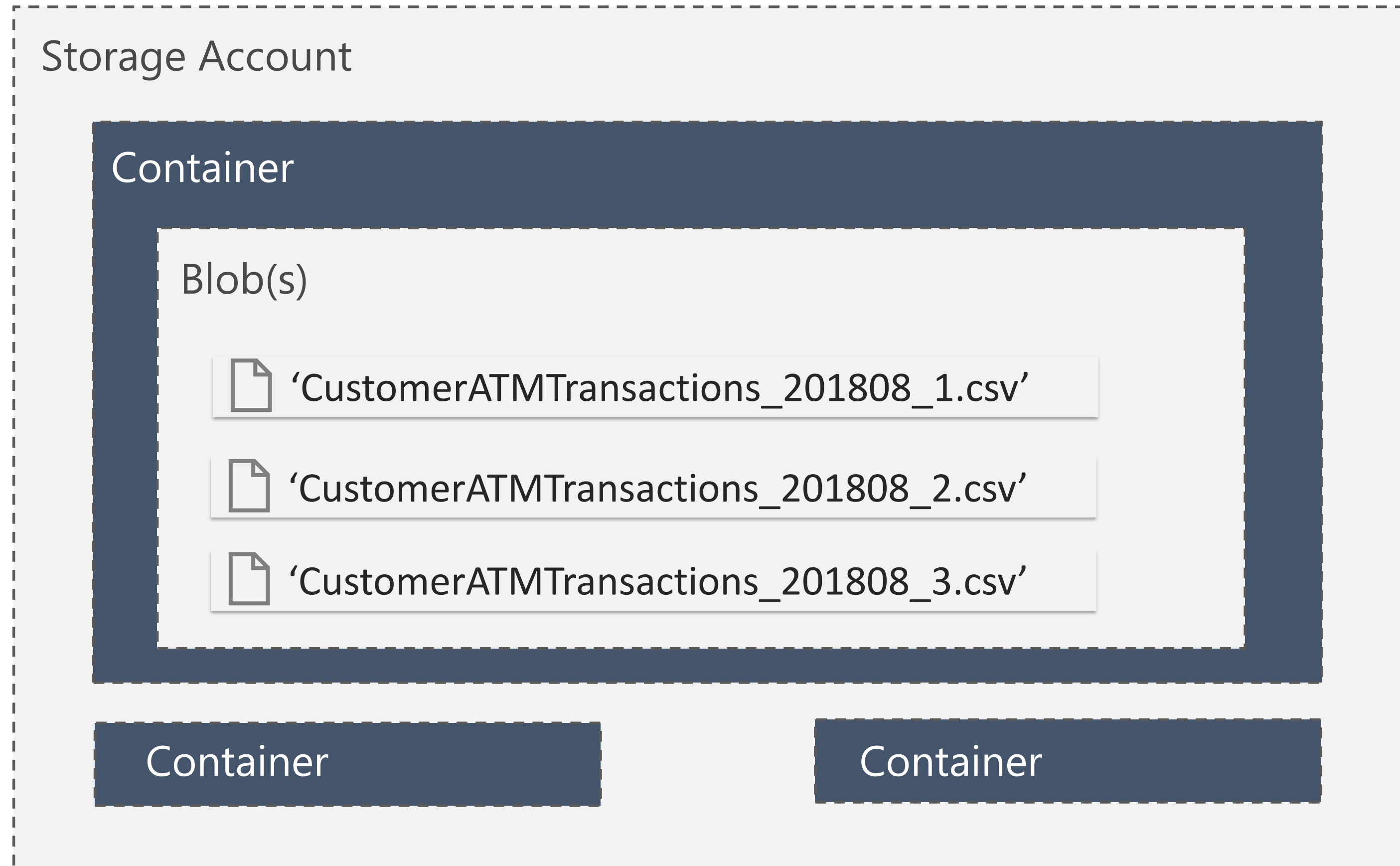
Big Data in Azure: Storage

(Excluding relational and NoSQL data storage options)



*In Public Preview

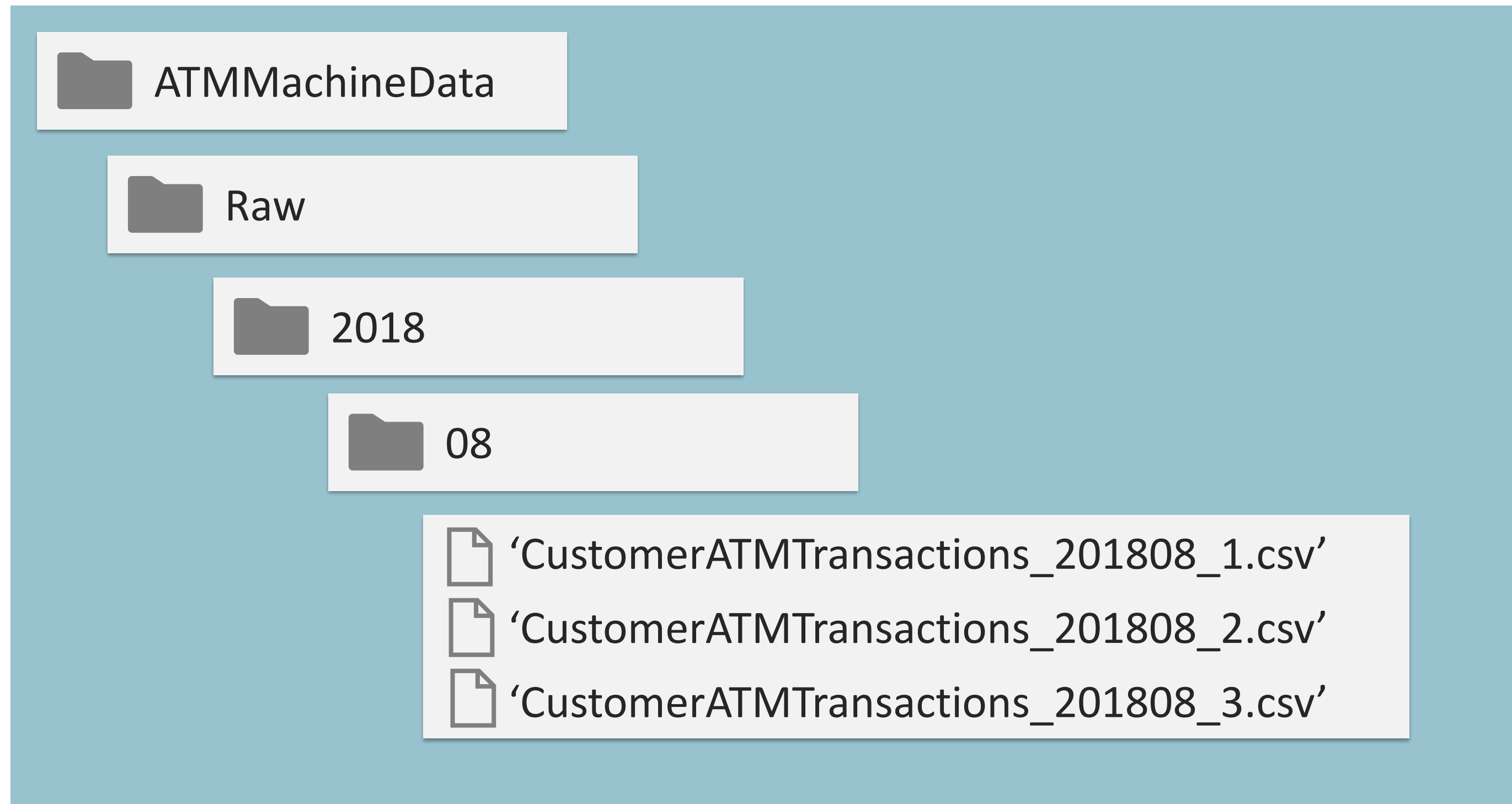
Azure Storage



Object-based storage manages data as discrete units.

Folders are part of the URI, but they're merely simulated. There is no folder-level security, nor folder-specific performance optimizations.

Azure Data Lake Storage Gen 1





Hierarchical file-based storage supports nesting of files within folders.

Folder-level security can be implemented, as well as certain performance optimizations.

Previously – An Either/Or Decision

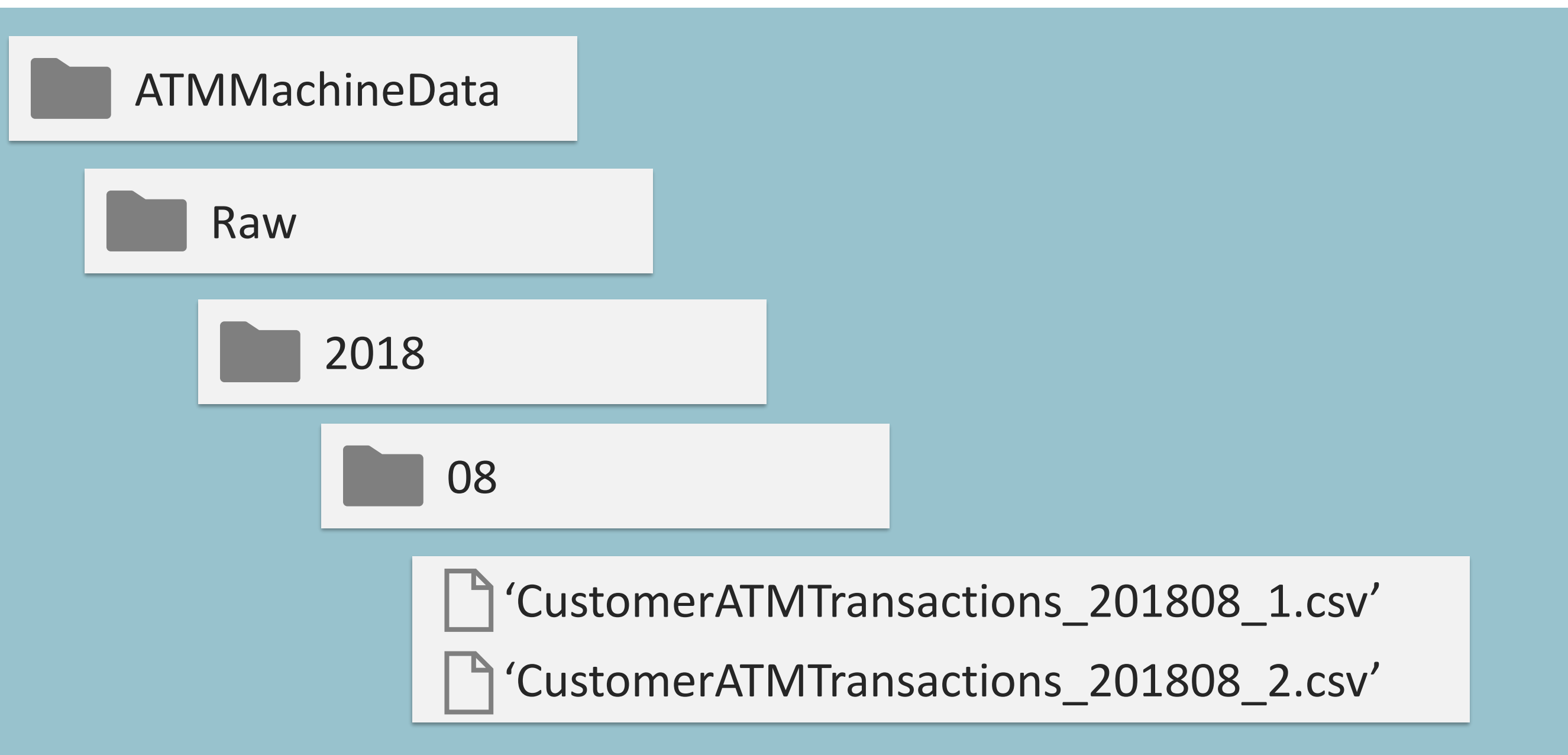
Object
store


 '/ATMMachineData/RawData/2018/08/CustomerATMTransactions_201808_1.csv'
 '/ATMMachineData/RawData/2018/08/CustomerATMTransactions_201808_2.csv'

 Azure
Storage

-OR-

Hierarchical
storage



 Azure
Data Lake
Storage Gen 1

Deciding Between Storage Services

Azure Storage

General purpose object store (containers > blobs)

Additional features not available in ADLS Gen1

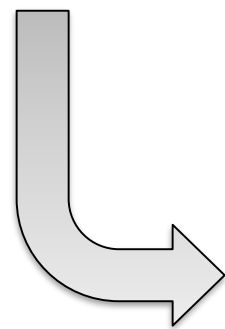
- Data replication and redundancy options
- Available in all regions globally
- Hot/cold/archive tiers
- Lifecycle management (in preview at this time)
- Metadata (key/value pairs)

ADLS (Gen 1)

Hierarchical file system (folders > files)

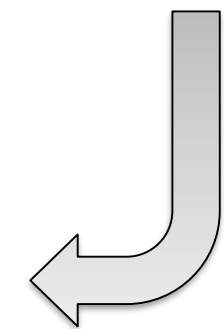
Optimized for analytics workloads

- Hadoop and big data optimizations
- Parallelized reads and writes
- Scaled out over multiple nodes
- Low latency writes with I/O throughput
- Fine-grained security via access control lists



ADLS (Gen 2)

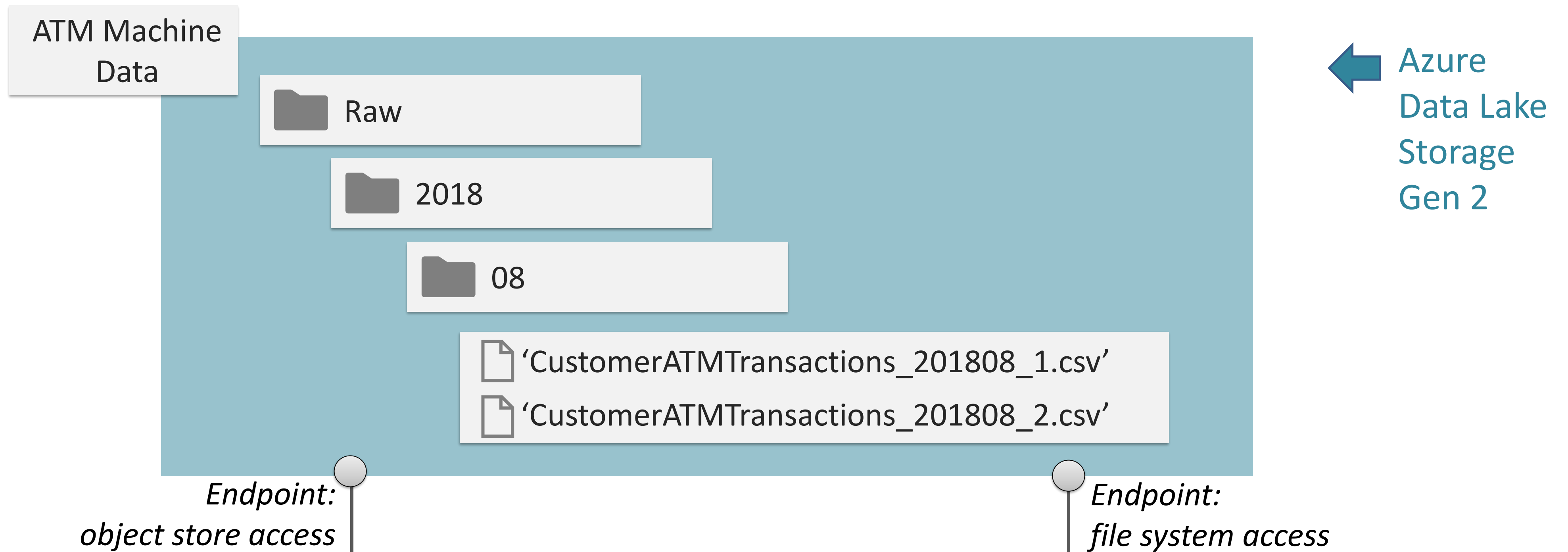
- Multi-modal combining features from both of the above
- Not a separate service: Azure Storage with new features
- Enable the "hierarchical namespace" (HNS) to use



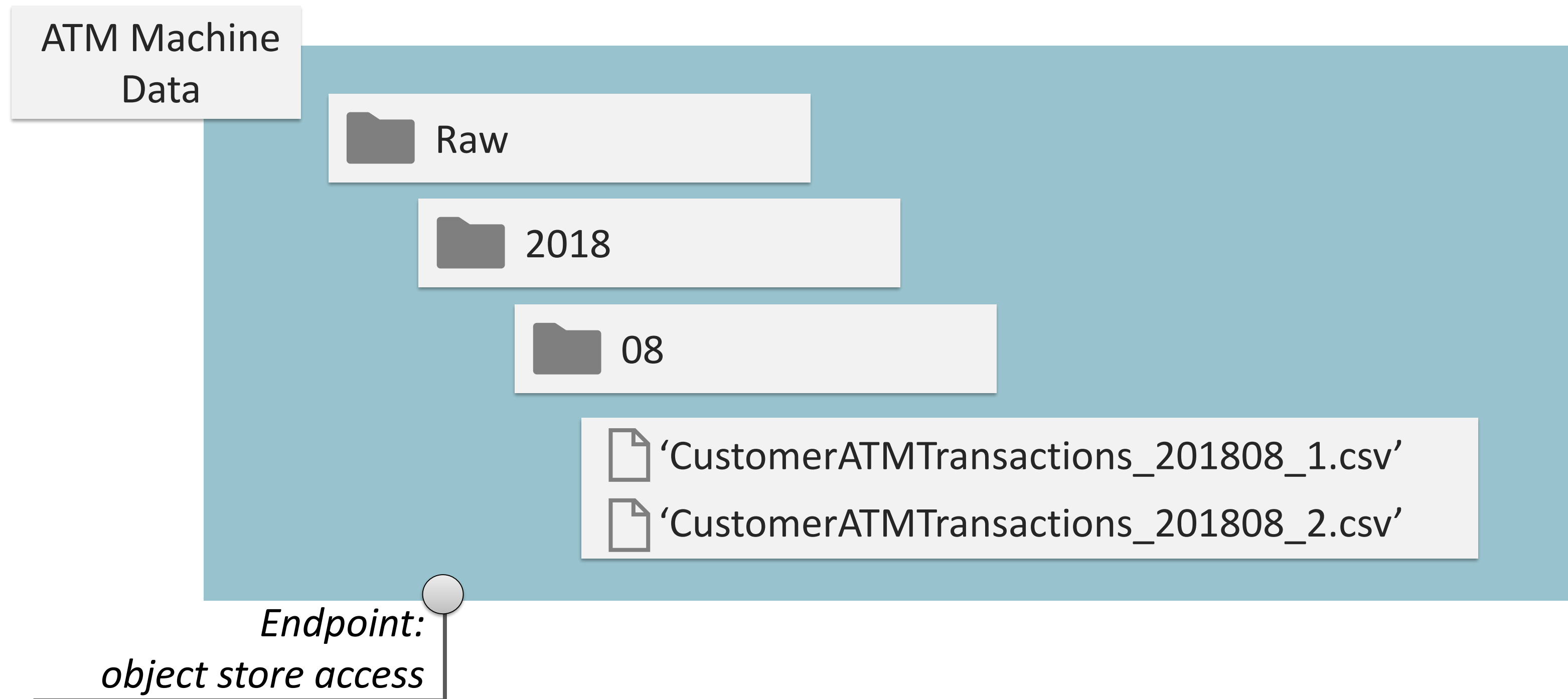
New Multi-Modal Storage Option: ADLS Gen 2

The long-term vision:

The data is stored once, and accessed through either endpoint based on use case / data access pattern. Files & folders are 'first class citizens.'

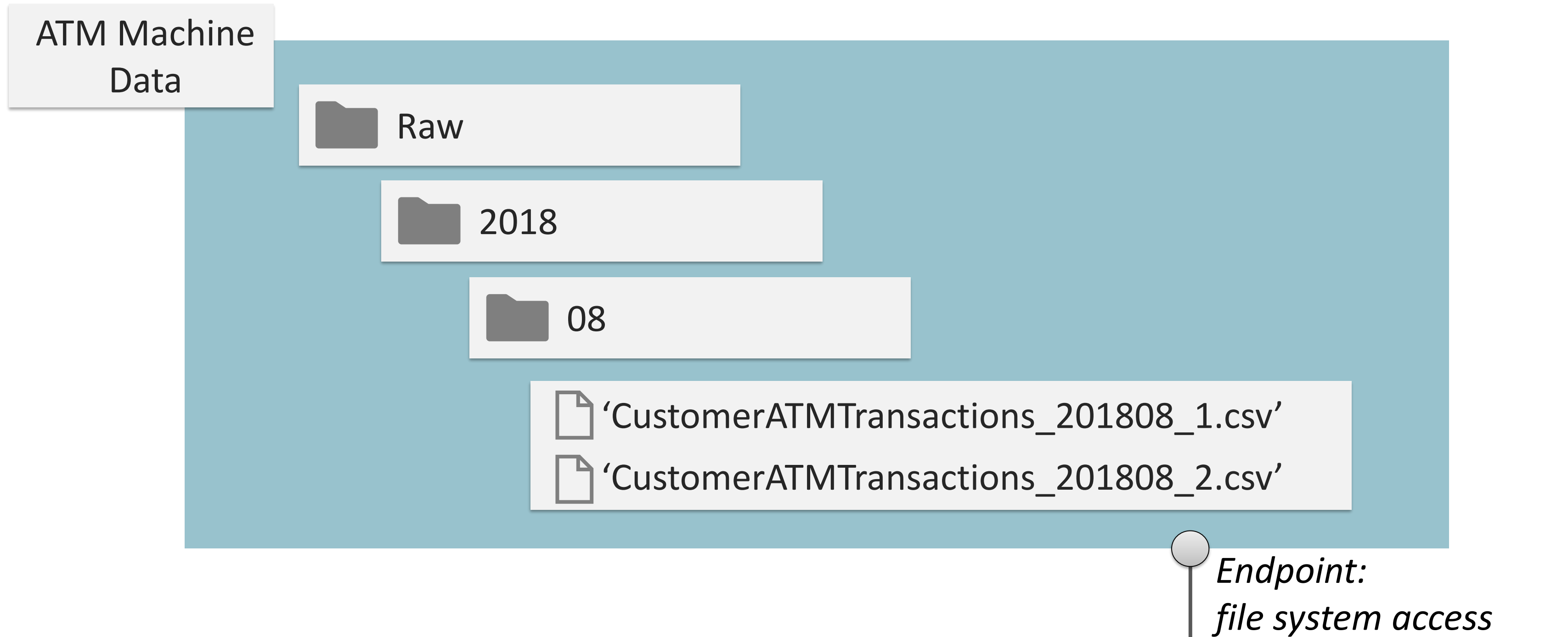


Object Store Endpoint: wasb[s]



```
wasb[s]://containername@accountname.blob.core.windows.net/raw/2018/08/CustomerATMTransactions_2018_1.csv
```

File System Endpoint: abfs[s]



abfs = Azure Blob File System

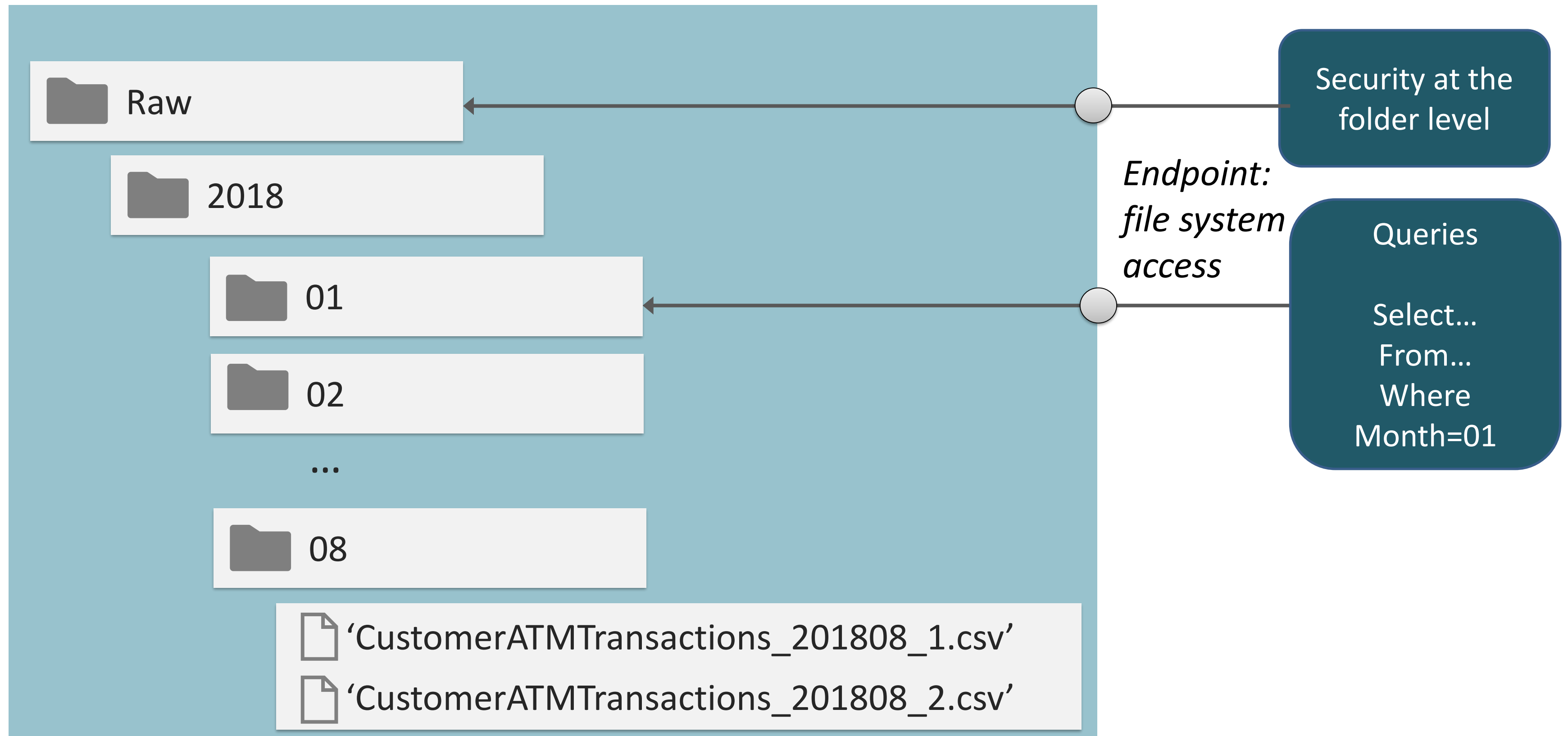
```
abfs[s]://filesystemname@accountname.dfs.core.windows.net/raw/2018/08/CustomerATMTransactions_2018_1.csv
```

↑
abfs is the driver
abfss = SSL

↑
dfs is the endpoint

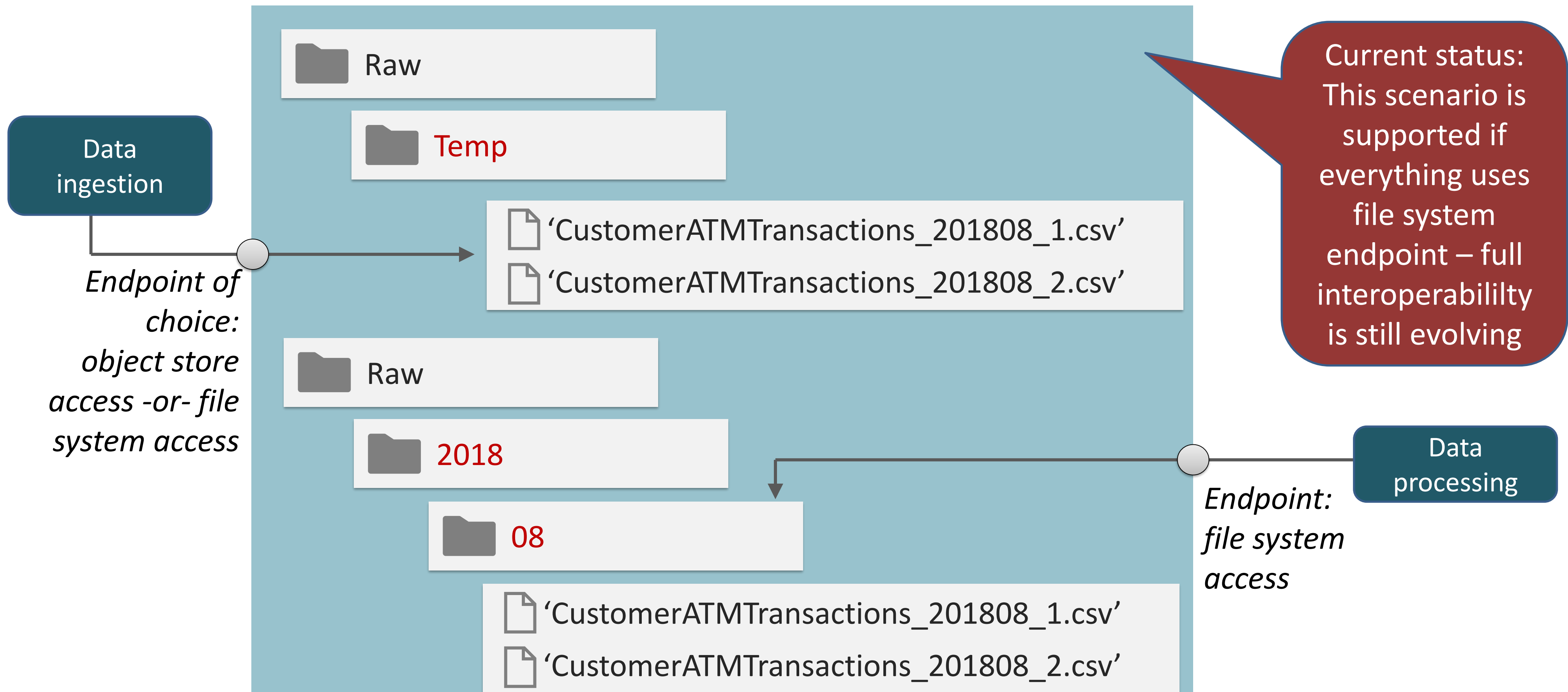
Multi-Modal Advantages with ADLS Gen 2 – Example 1

Leverage partition scans & partition pruning to improve query performance:



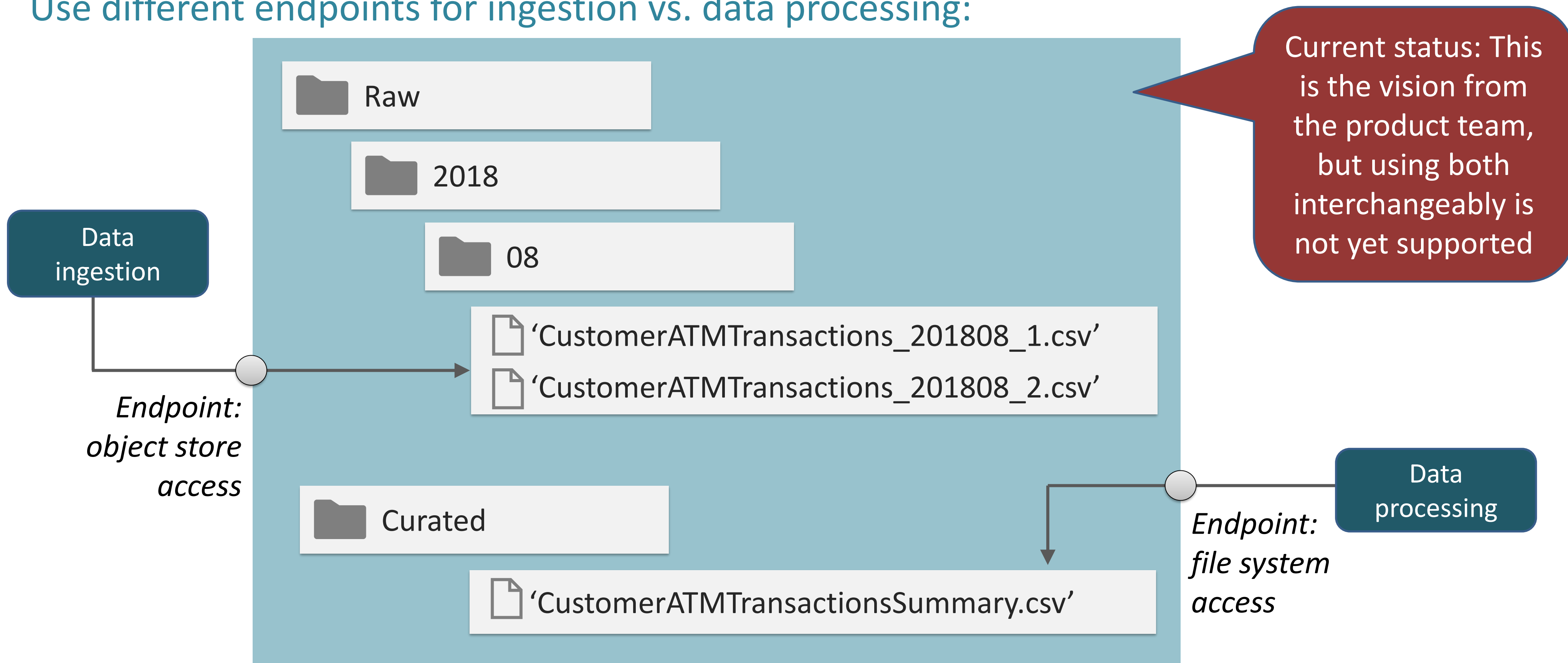
Multi-Modal Advantages with ADLS Gen 2 – Example 2

Metadata-only changes with significantly better performance using the file system endpoint:



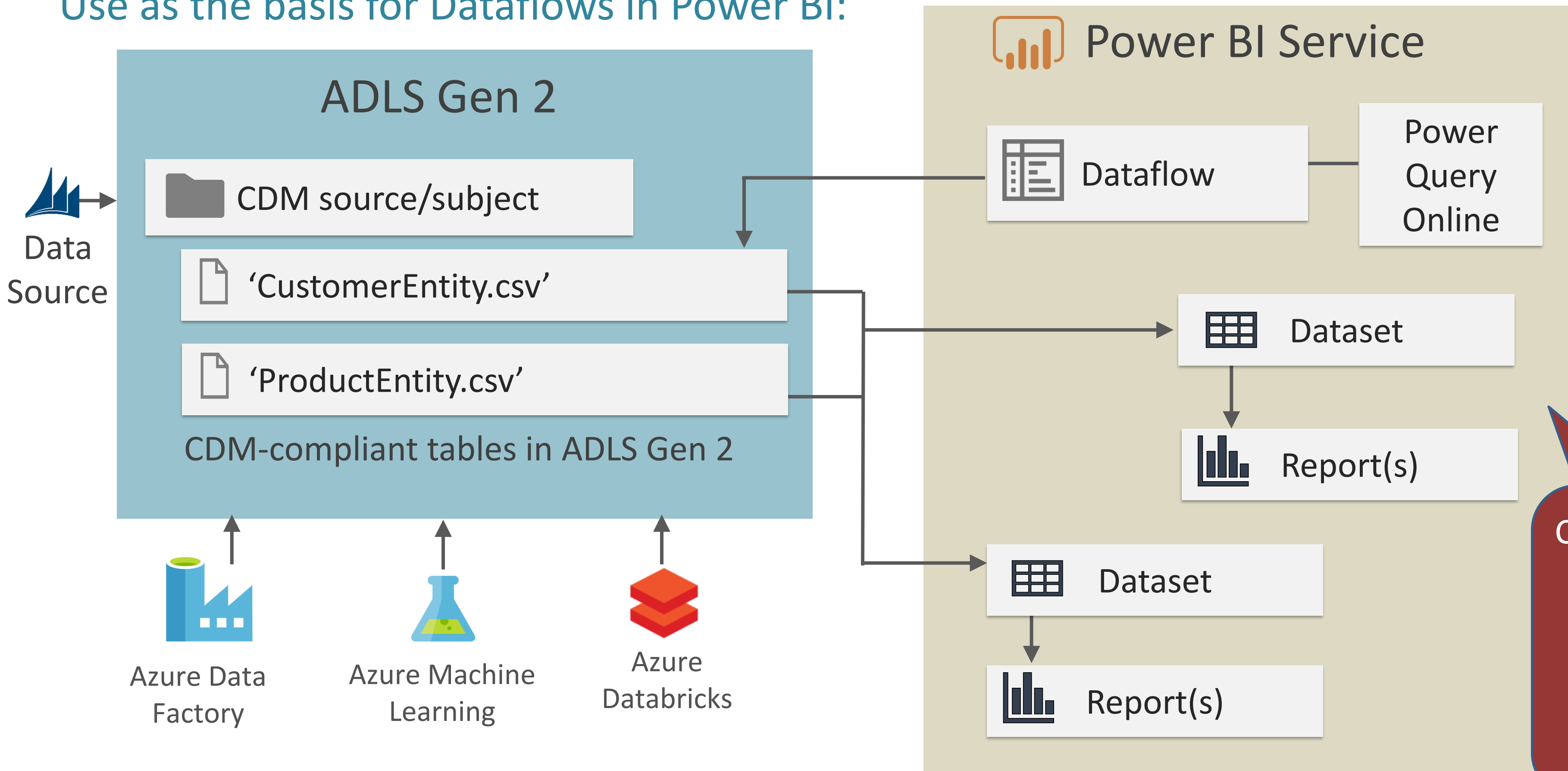
Multi-Modal Advantages with ADLS Gen 2 – Example 3

Use different endpoints for ingestion vs. data processing:



Multi-Modal Advantages with ADLS Gen 2 – Example 4

Use as the basis for Dataflows in Power BI:

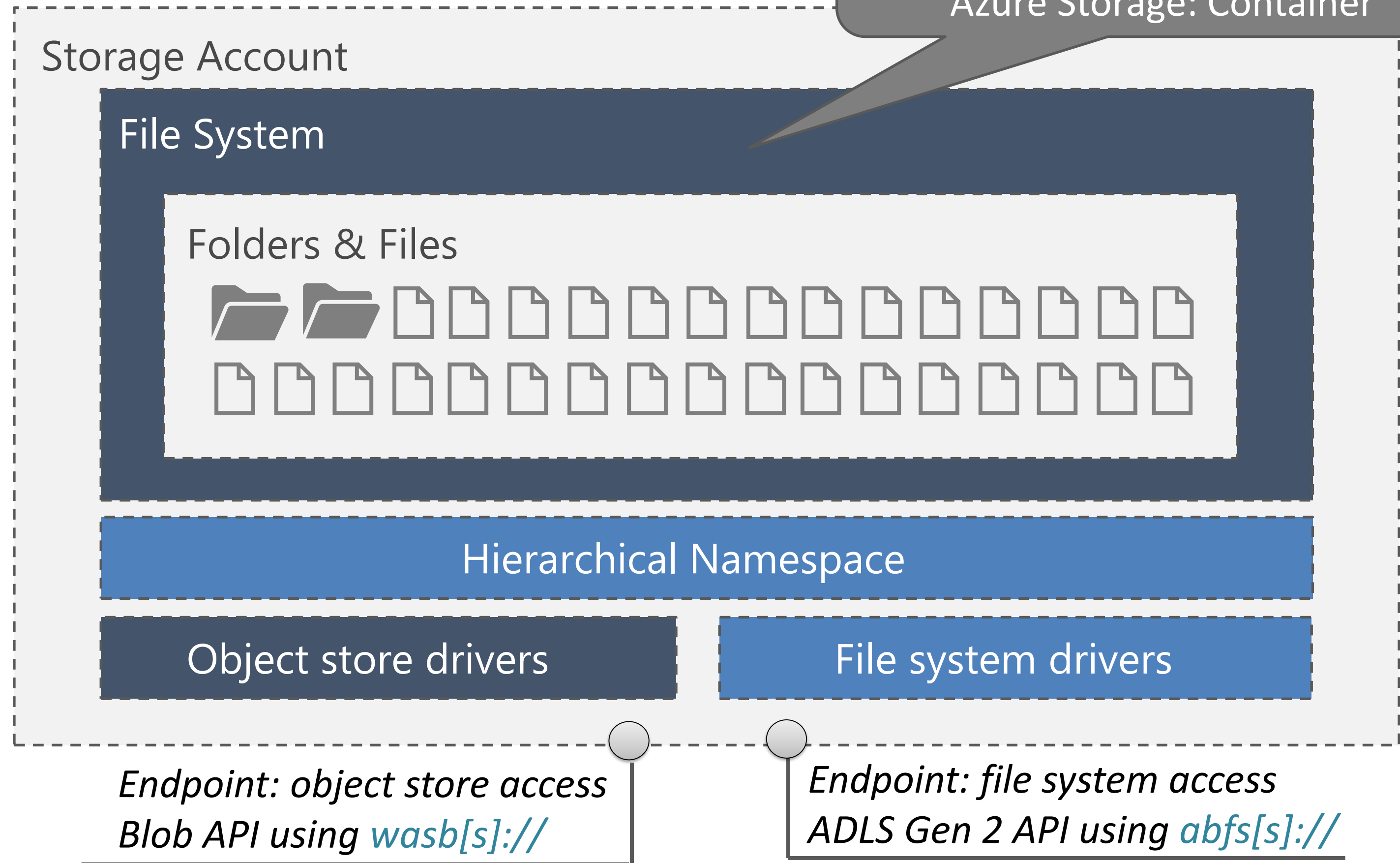


CDM = Common Data Model

Azure Data Lake Storage Gen 2

ADLS Gen 2 =
Azure Storage
with the
Hierarchical
Namespace
(HNS) enabled.

ADLS Gen 2: File System
=
Azure Storage: Container



Adapted from:
<https://azure.microsoft.com/en-us/blog/a-closer-look-at-azure-data-lake-storage-gen2/>

When to ****Disable**** the Hierarchical Namespace?

- ✓ General purpose file storage such as **backups and VHDs**
- ✓ Classic object store **use cases which do not benefit from hierarchical storage** or a high degree of organization (ex: image storage)
- ✓ Custom apps, APIs, or **legacy systems** which only use the Blob API and/or are unaware of file system semantics

The HNS is enabled at the storage account level.
Product team says there's no harm or performance difference
(even for raw I/O) if the HNS is enabled but not used.
However, you'll pay extra cost (~30% extra) on every transaction if HNS is enabled.

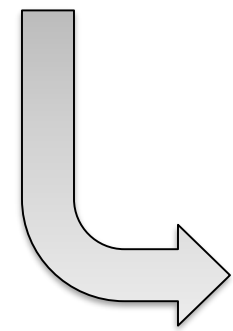
Summary: Goals of ADLS Gen 2

- *Unify* the data lake story on Azure
- Take advantage of the *best of both feature sets* (object storage & hierarchical storage)
- *Multiple protocol endpoints* to allow flexibility for use cases
- *Avoid duplicating data* for specific use cases or tools ('islands of data')
- *Overcome limitations of object storage* (ex: metadata only operations)
- Improved *performance for big data analytics*
- Implement *full data lifecycle* and *data policies*
- *Low cost* with high-performing *throughput*
- Integrate with the new '*dataflows*' functionality in Power BI

Summary: Current State of the Storage Options

Azure Storage

- ✓ Still a very valid option for object store workloads

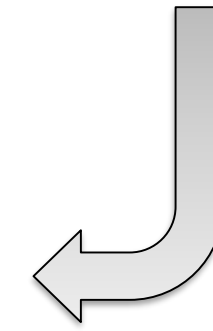


/ ADLS (Gen 2)

- ✓ In public preview
- ✓ Feature support is evolving over time

ADLS (Gen 1)

- ✓ Fully supported in existing regions
- ✓ No new features



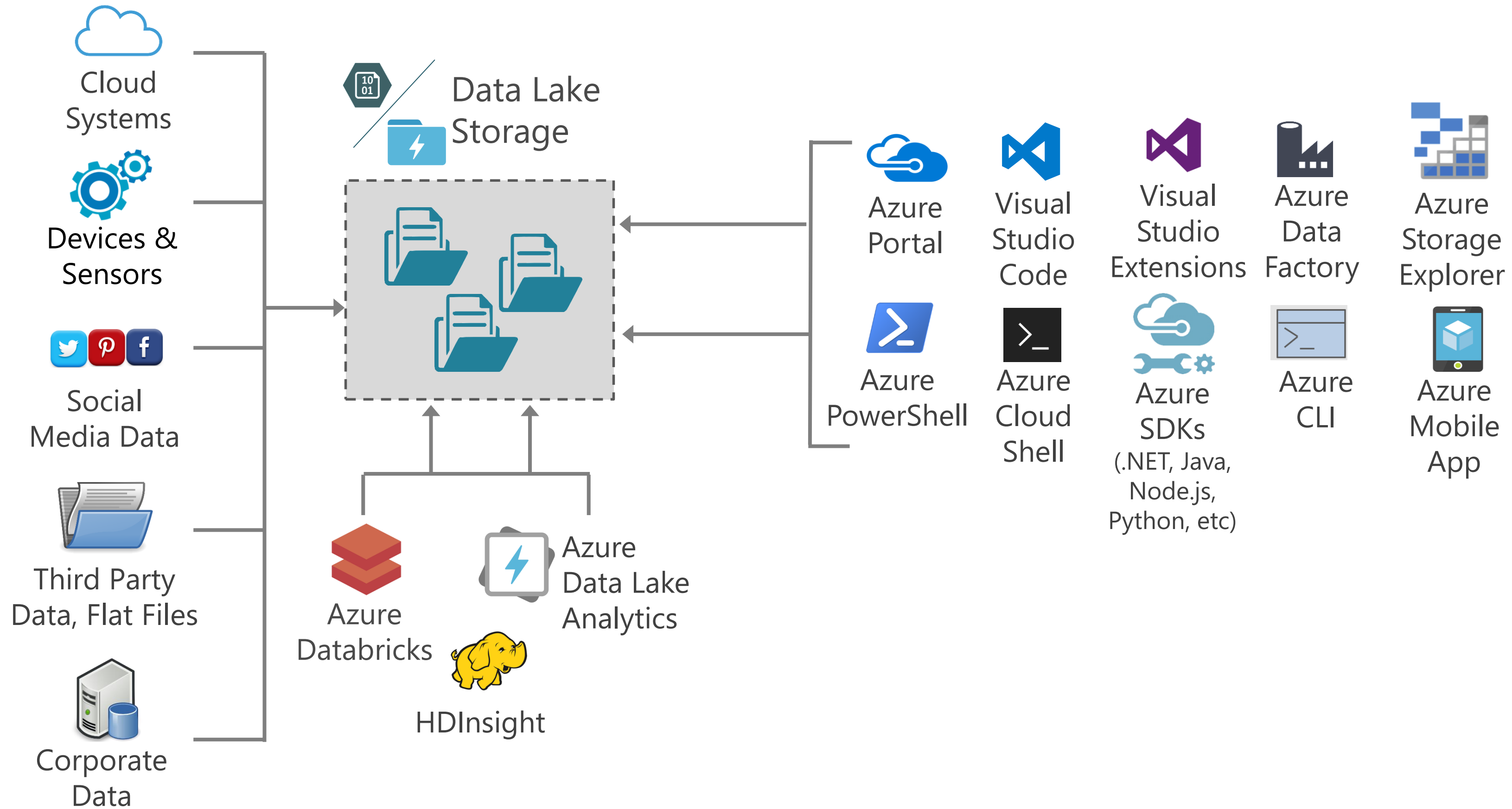
Demo: ADLS Gen 2



Compute in Azure



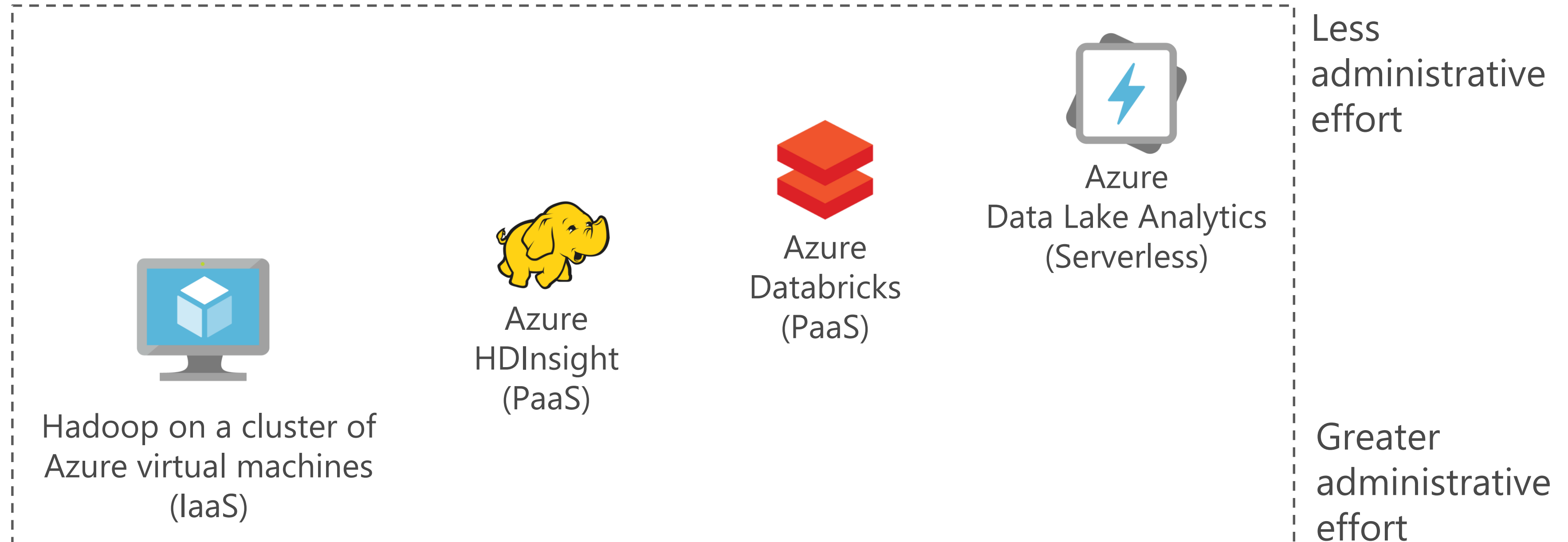
Compute Services and Data Management



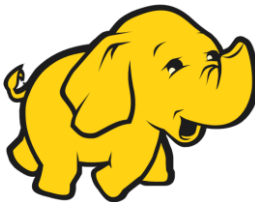
Big Data in Azure: Compute

Higher level of complexity,
control, & customization


Easiest entry point
to get started



Hadoop on a cluster of
Azure virtual machines
(IaaS)


Azure
HDInsight
(PaaS)

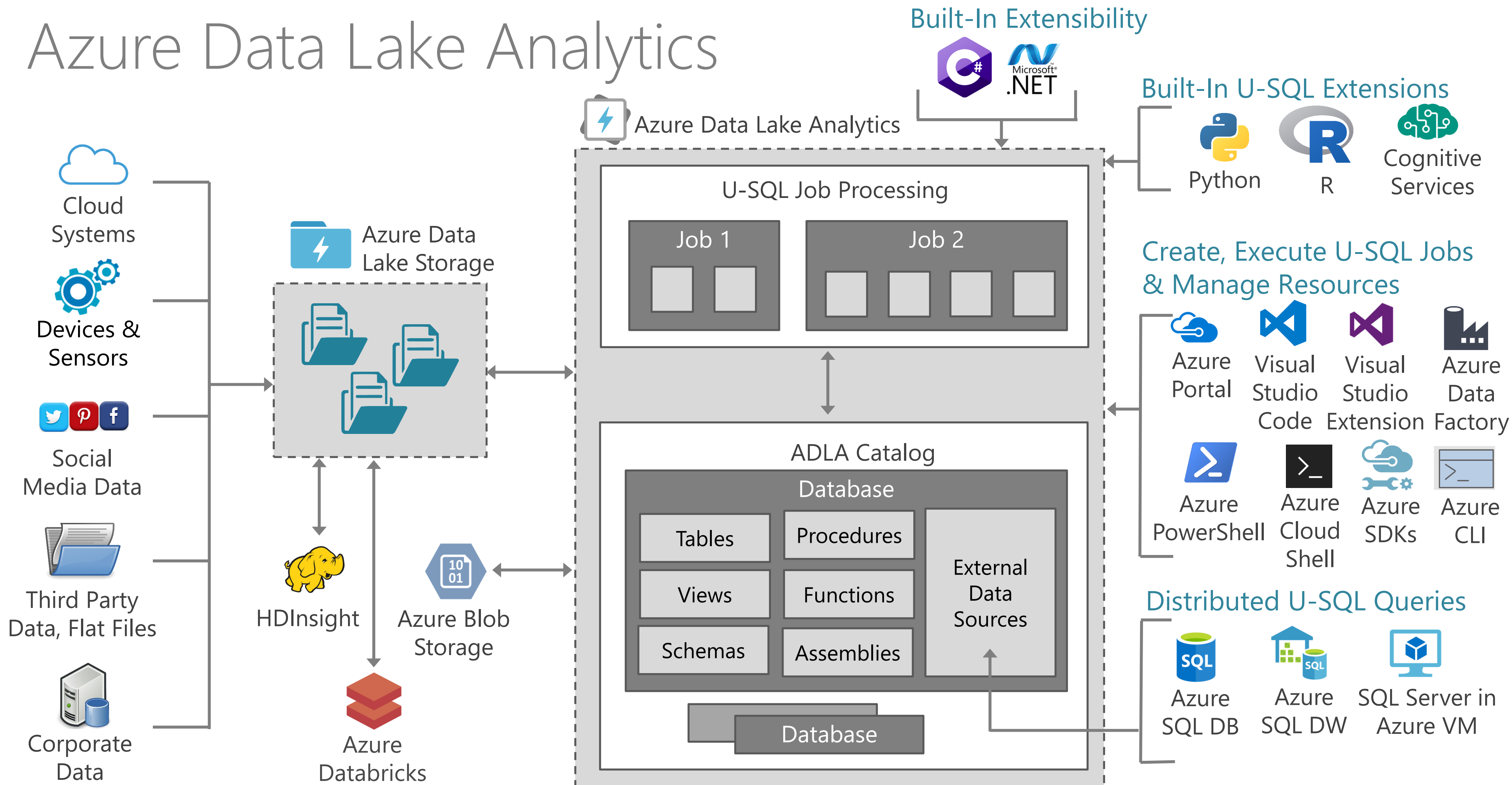

Azure
Databricks
(PaaS)


Azure
Data Lake Analytics
(Serverless)

Greater integration
with various Apache
projects

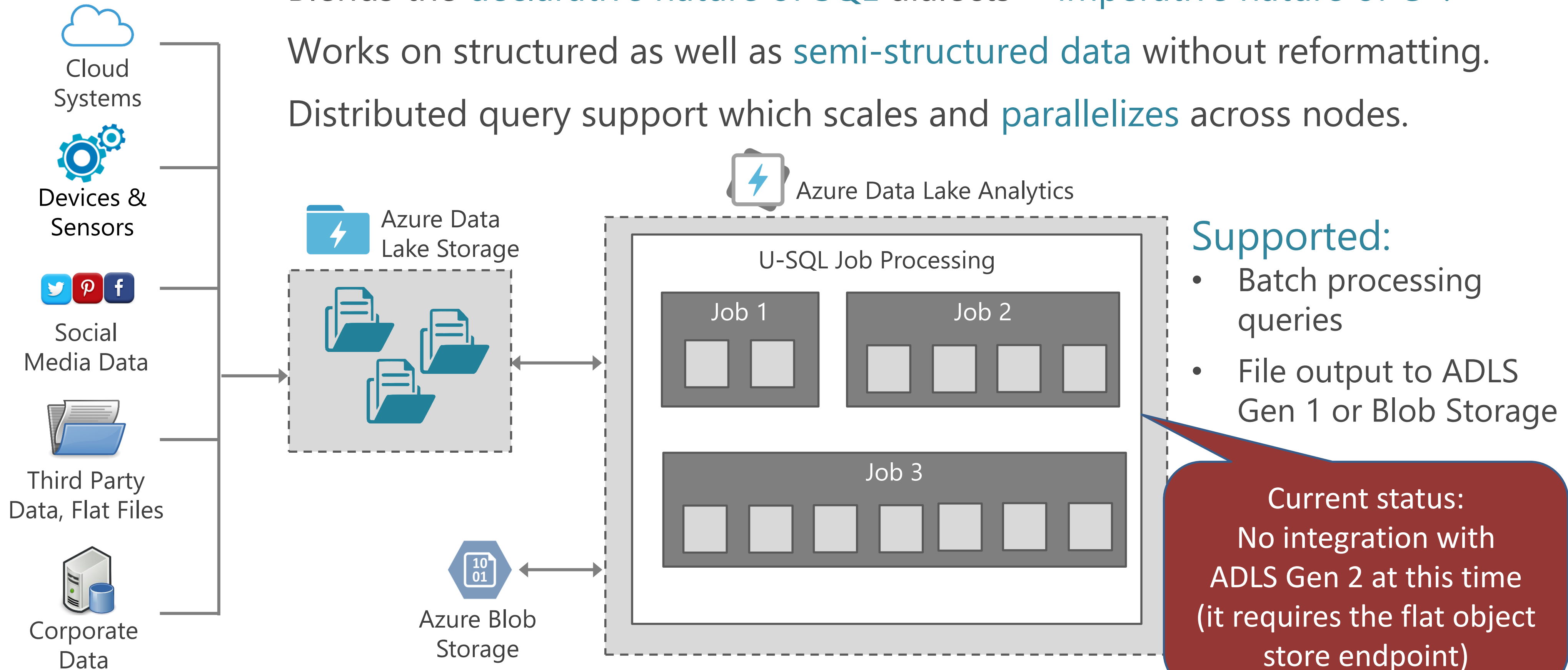
Less integration
with various Apache
projects

Azure Data Lake Analytics



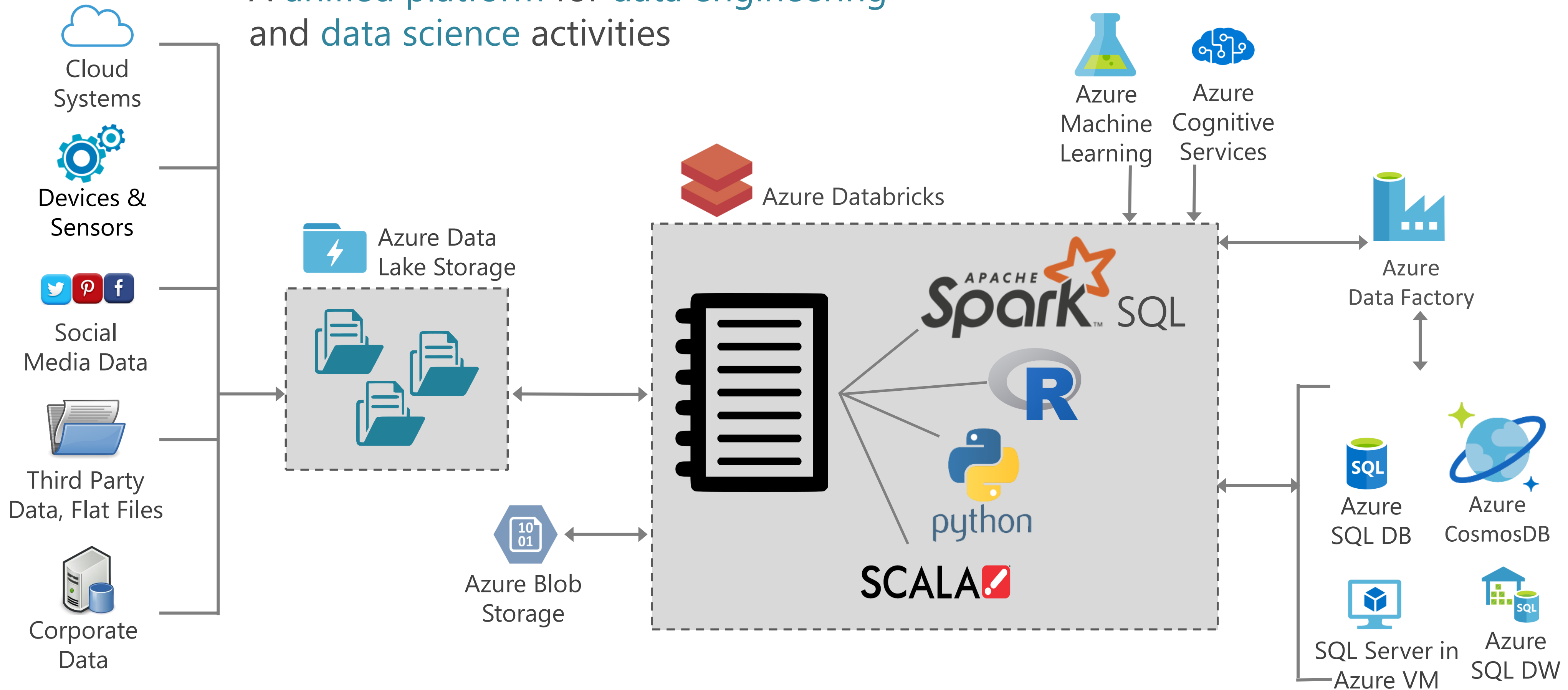
U-SQL: Unified SQL

Blends the **declarative nature of SQL** dialects + **imperative nature of C#**.
Works on structured as well as **semi-structured data** without reformatting.
Distributed query support which scales and **parallelizes** across nodes.



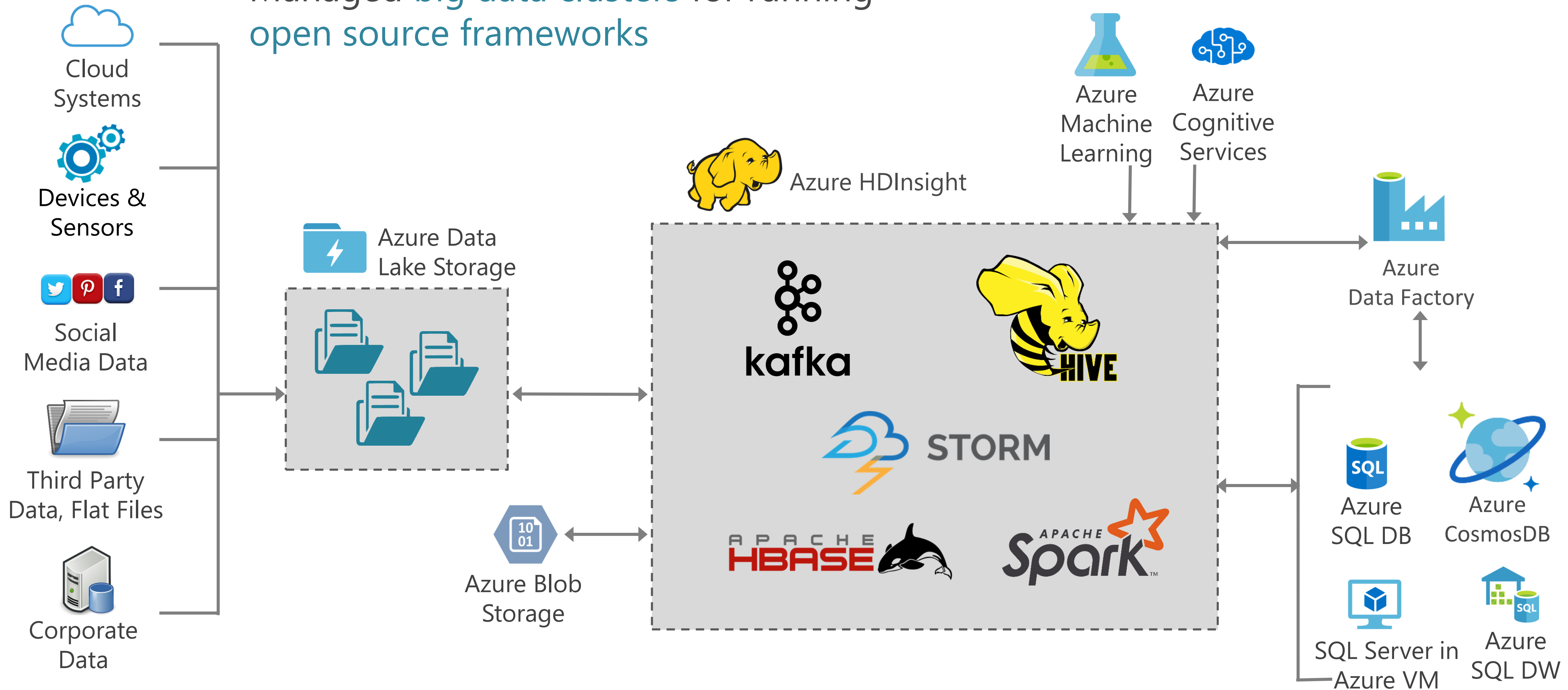
Azure Databricks

A unified platform for data engineering and data science activities



Azure HDInsight

Managed big data clusters for running open source frameworks



Deciding Between Compute Services

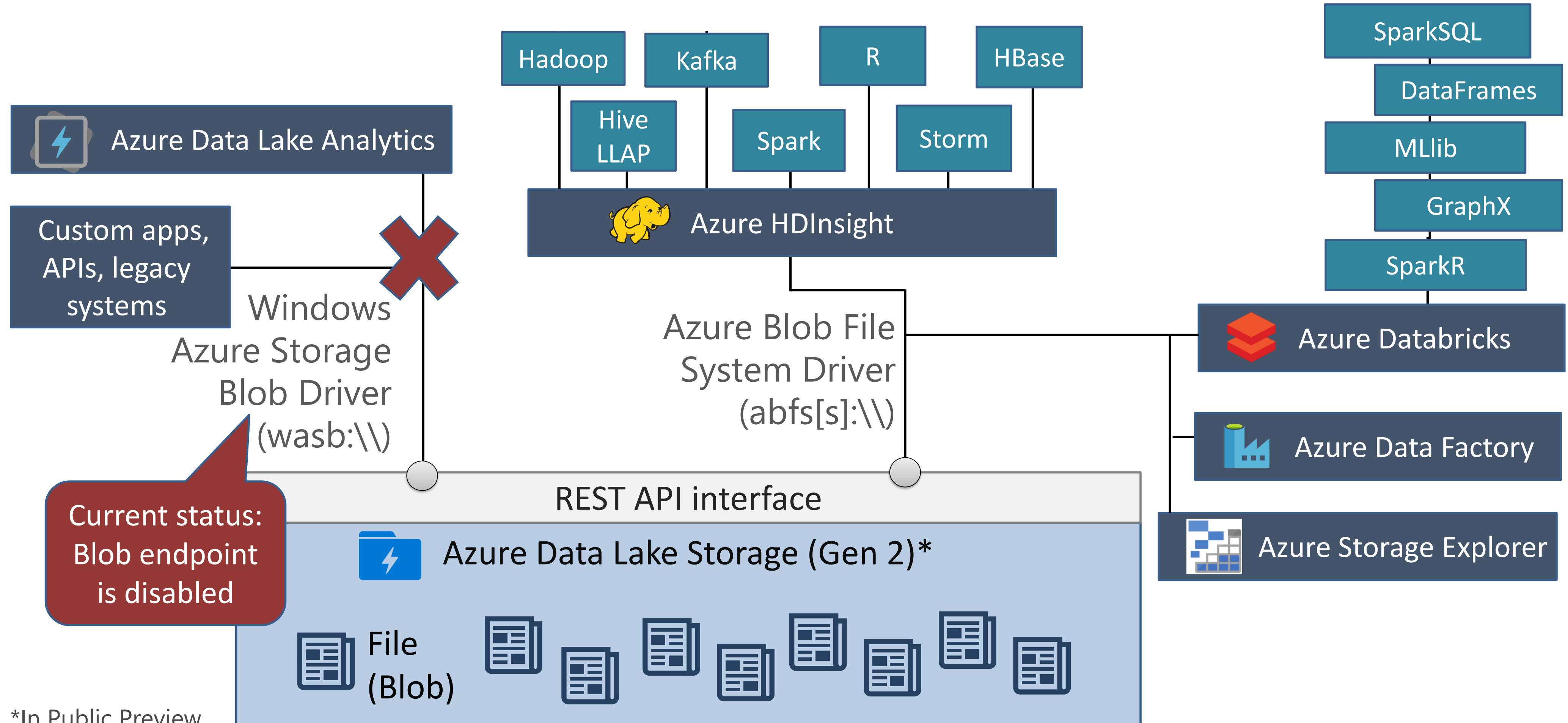
 Hadoop VM HDInsight Databricks ADLA

Type:	IaaS	PaaS	PaaS	Serverless
Purpose:	Running your own cluster of Hadoop virtual machines	Running a managed big data cluster	Running a managed, optimized Spark framework	Running U-SQL batch jobs
Suitable for:	Full control over everything; investment in distributions such as Hortonworks, Cloudera, MapR	Integration with open source Apache projects and/or greater control over clusters	Collaborative notebooks; easier deployments; utilizing Spark in a variety of ways	Focus on running individual jobs (scripts) rather than managing a cluster

1st choice

Caution

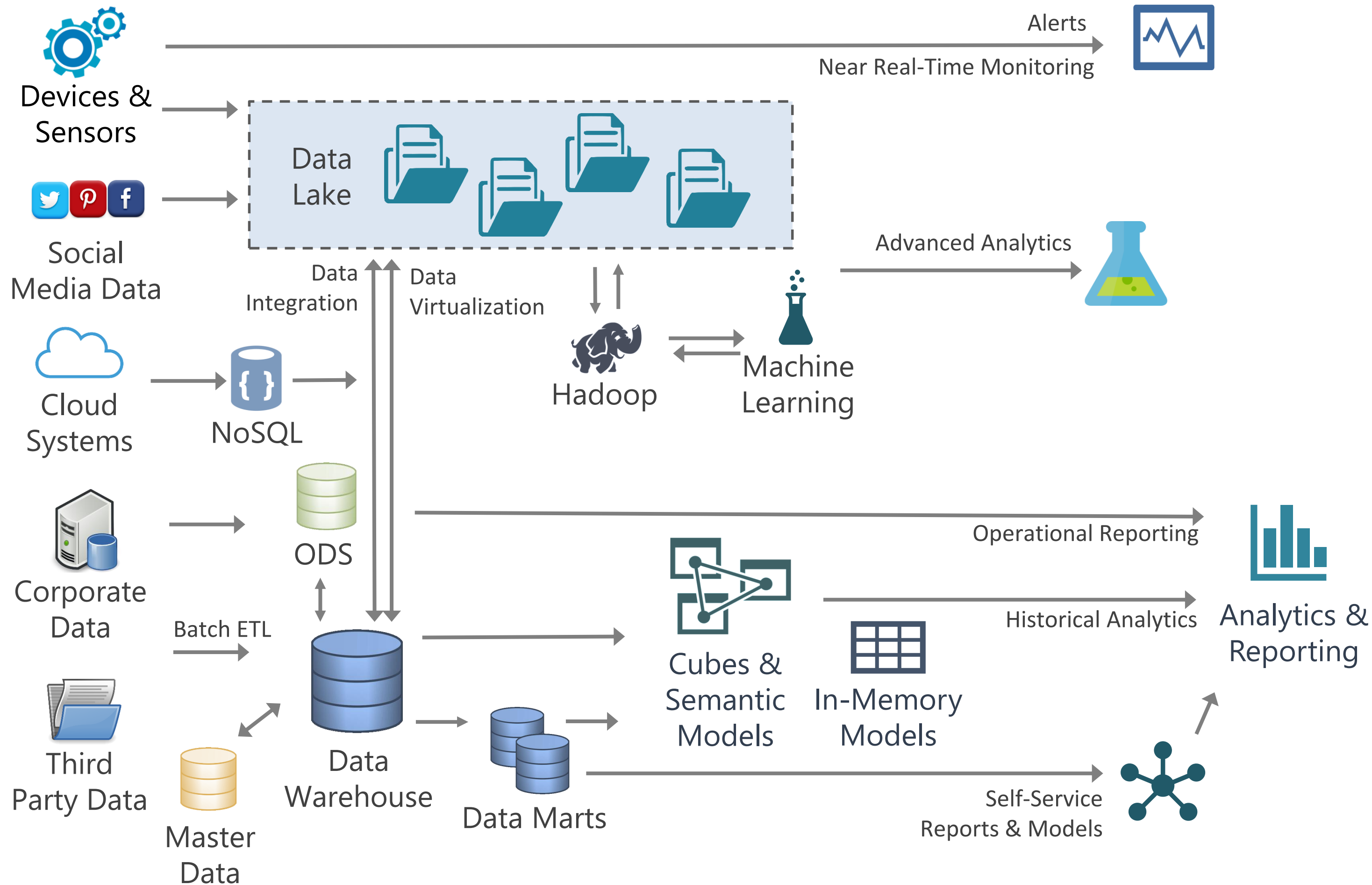
Interacting with ADLS Gen 2



Integrating Azure Data Lake in a Multi-Platform Architecture



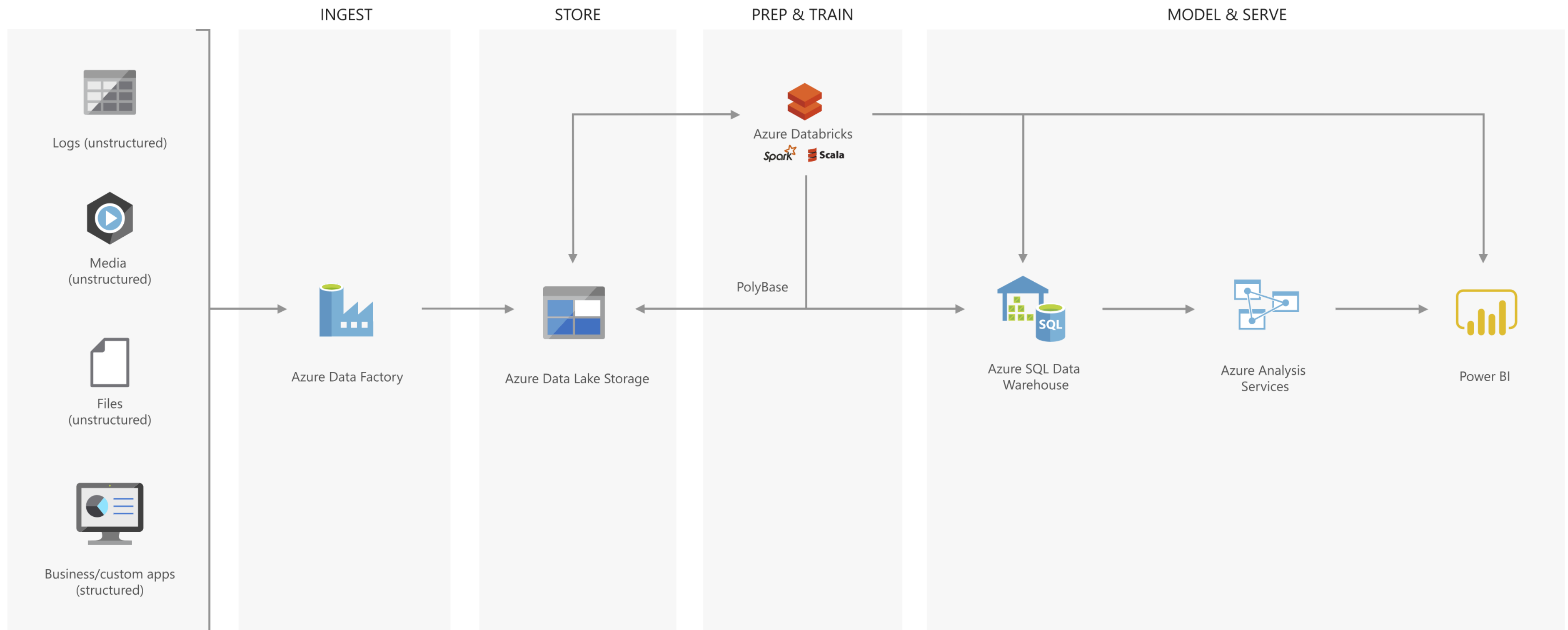
Multi-Platform Architecture



- ✓ Handle a variety of data types & sources
- ✓ Larger data volumes at lower latency
- ✓ Bimodal: self-service + corporate BI to support all types of users
- ✓ Newer cloud services
- ✓ Advanced analytics scenarios
- ✓ Balance data integration & data virtualization

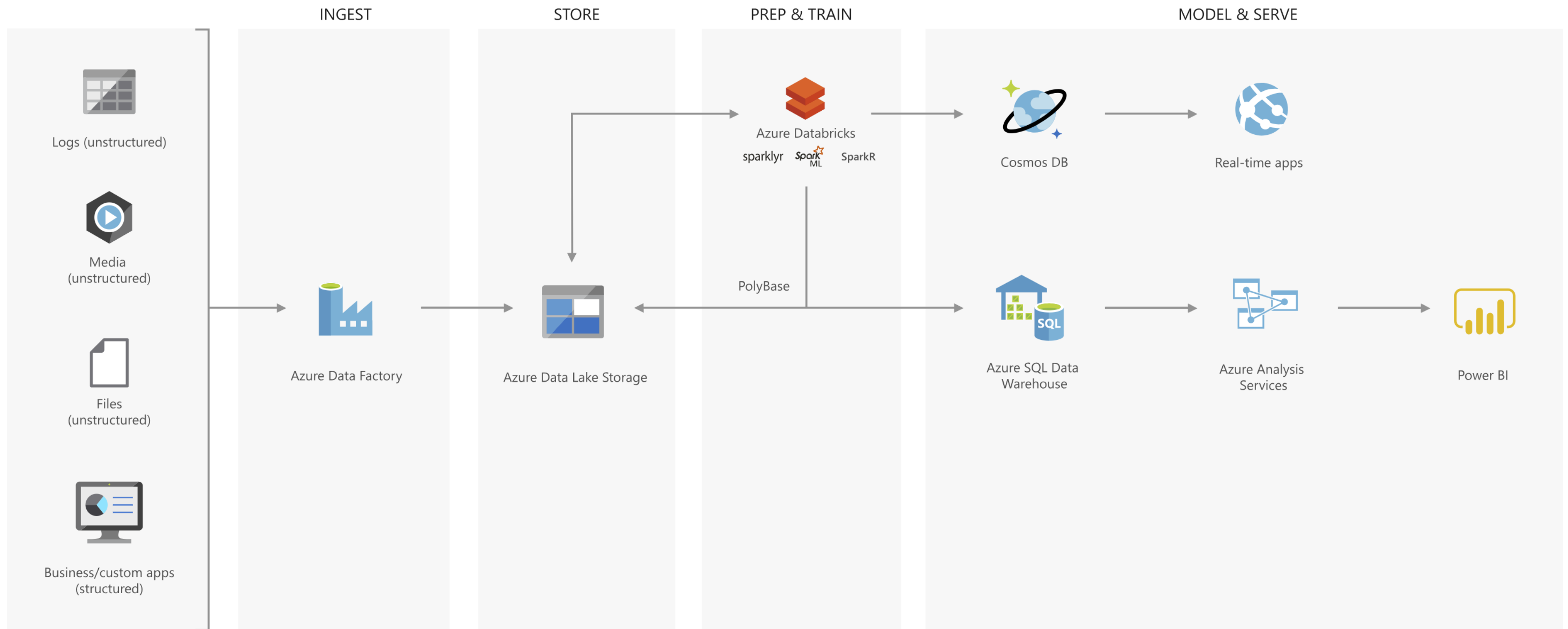
Azure Data Lake Implementation Options

Modern data warehouse



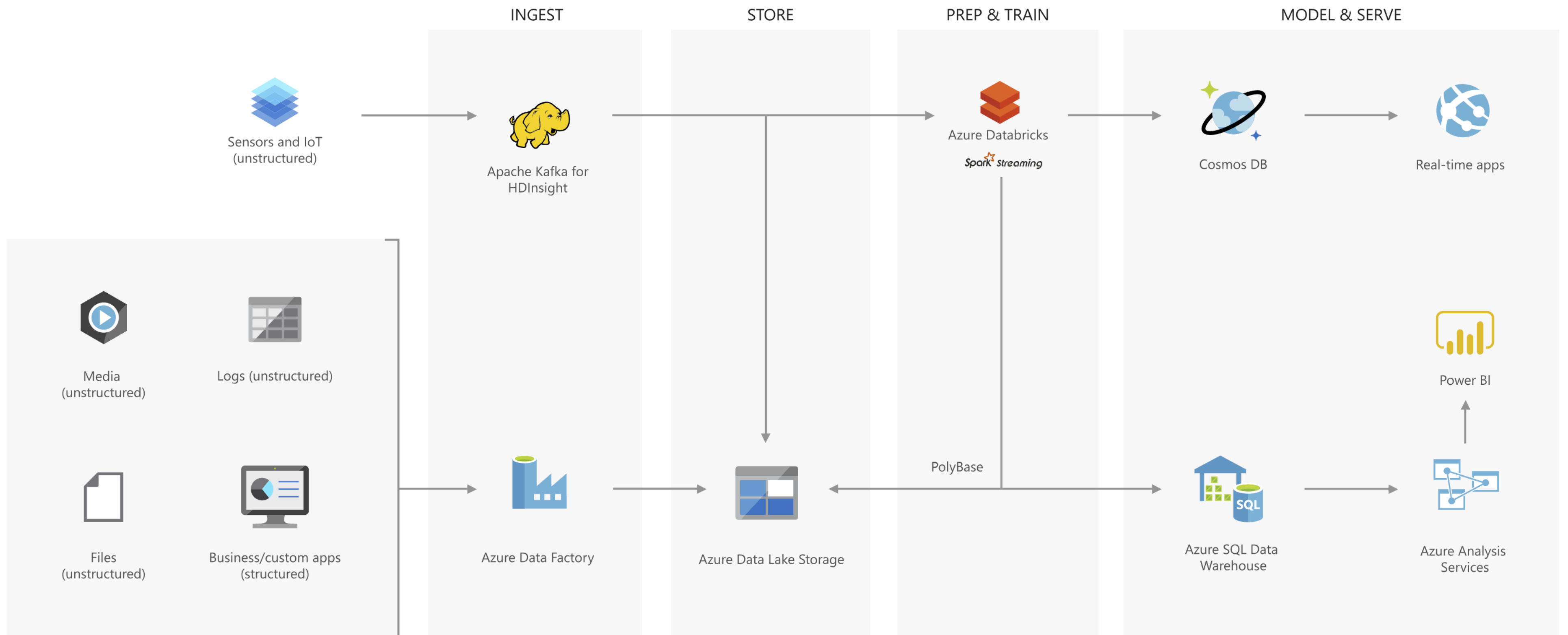
Azure Data Lake Implementation Options

Advanced analytics on big data



Azure Data Lake Implementation Options

Real time analytics



Suggestions for Getting Started with a Data Lake Project



Getting Started With a Data Lake Project

Is a Data Lake the Right Choice?

Do you have *non-relational* data?

Various data types, various sources,
whether it's "big data" or not

Do you have *IoT* type of data?

Streaming or
micro-batch
pipelines

Do you have *advanced analytics scenarios* on unusual datasets?

Do you need to *offload ETL processing (ELT)* and/or *archive data* from a data warehouse or other systems to low-cost storage?

Readiness:

Are you ready willing to learn *different development patterns* and/or *new technologies*?

Are you ready to handle the *trade-offs of 'schema on read'* vs 'schema on write'?

Getting Started With a Data Lake Project

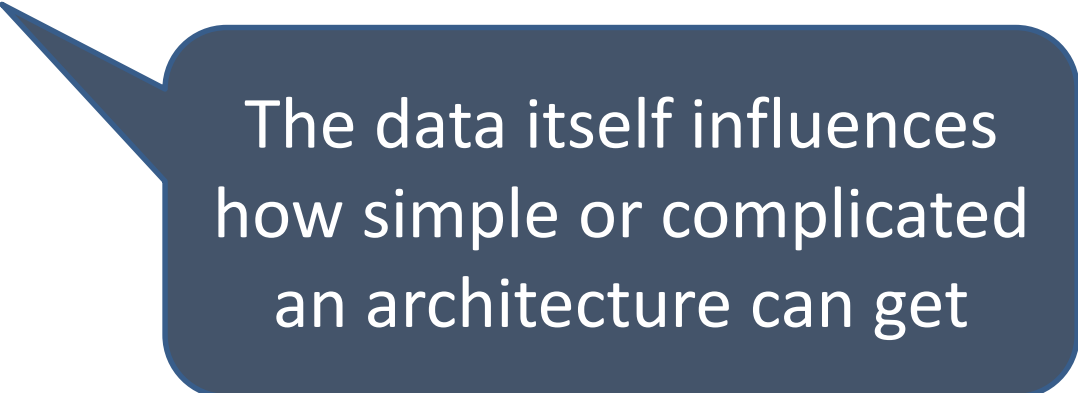
Data

What types of **data ingestion pipelines** do you have, at what frequency?

- Batch
- Micro-batch
- Streaming

What are the current + anticipated **data size volumes**, and in what format?

- Structured data
- Semi-structured data
- Unstructured data
- Geospatial data



The data itself influences how simple or complicated an architecture can get

To what extent does semi-structured data need to be **integrated** with the structured data?

Getting Started With a Data Lake Project

Data Movement & Storage

What level of **data integration** (ETL or ELT) vs. **data virtualization** provides optimal data access?

- Data movement can be expensive
- Data might be too large to practically move
- Time window for data processing may be small
- Latency (freshness) of data varies

How much data movement are you willing to do?

Which do you value more?

- Polyglot persistence strategy ("best fit engineering")
- Architectural simplicity

A multi-platform architecture is more appealing if you subscribe to a polyglot persistence strategy. Success very much depends on staff skills.

Getting Started With a Data Lake Project

Information Delivery

What are the expectations + needs of your **user population**?

- Casual users
- Data analysts
- Data scientists
- IT, BI specialists, big data engineers

What type of **data consumption** do you support?

- Centralized reporting & analytics
- Decentralized self-service models
- Departmental or subject-specific data marts
- Application integration

The user base translates into expectations for how the information is to be delivered, which translates into technology choices

The more diverse your user population is, the more likely you will have a multi-platform architecture with both schema-on-read and schema-on-write

Getting Started With a Data Lake Project

Organizing the Data Lake

- ✓ Based on optimal data retrieval & security boundaries
- ✓ Avoid a chaotic, unorganized data swamp
- ✓ Take advantage of data pruning optimizations (esp year/month/day) when running queries

Common ways to organize and/or tag the data:

Time Partitioning

Year/Month/Day/Hour/Minute

Subject Area

Security Boundaries

Department
Business unit
etc...

Downstream App/Purpose

Data Retention Policy

Temporary data
Permanent data
Applicable period (ex: project lifetime)
etc...

Business Impact / Criticality

High (HBI)
Medium (MBI)
Low (LBI)
etc...

Owner / Steward / SME

Probability of Data Access

Recent/current data
Historical data
etc...

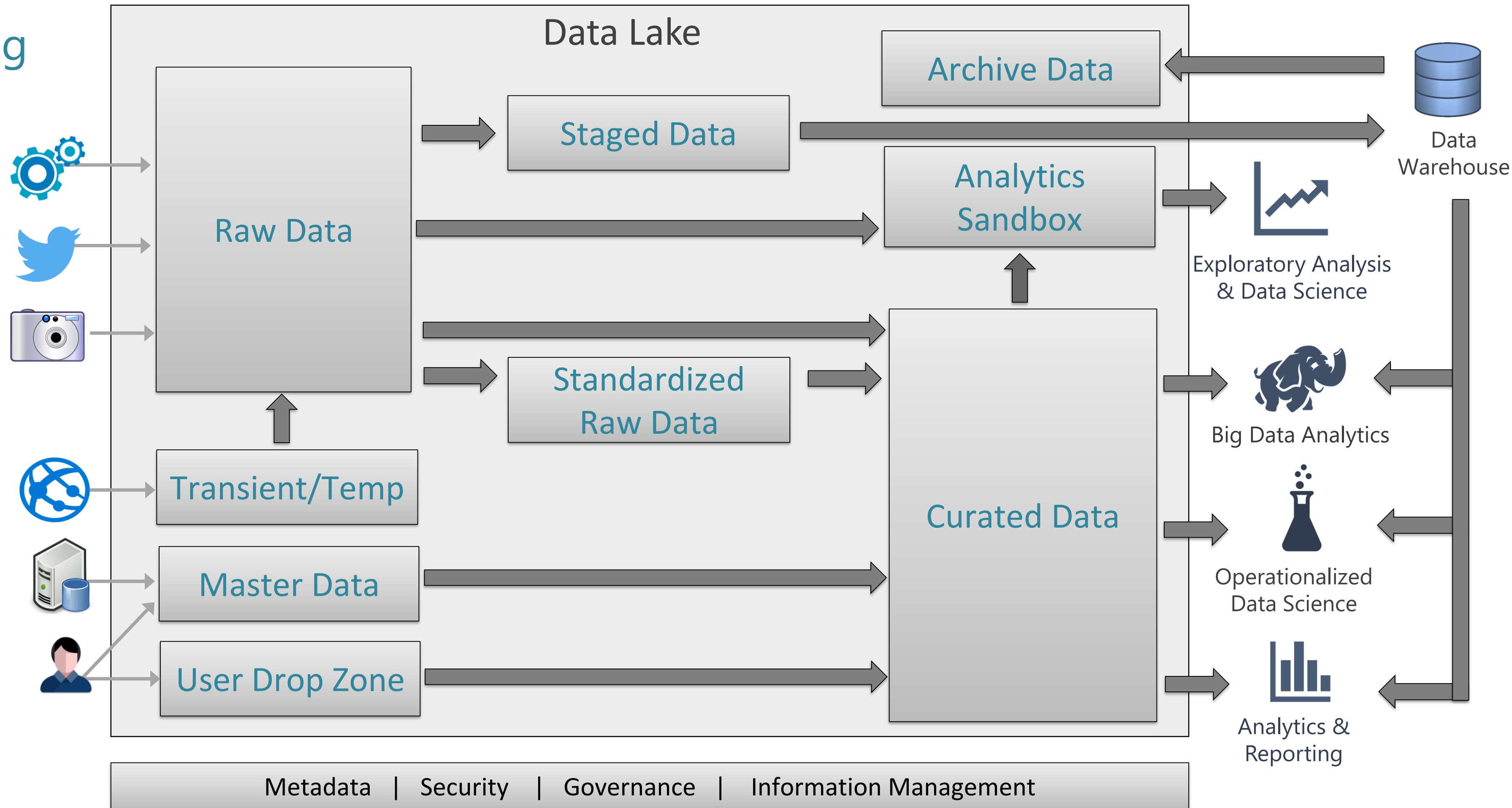
Confidential Classification

Public information
Internal use only
Supplier/partner confidential
Personally identifiable information (PII)
Sensitive – financial
Sensitive – intellectual property
etc...

Getting Started With a Data Lake Project

Organizing the Data Lake

Although a data lake emphasizes getting started quickly, there is still up-front planning



Getting Started With a Data Lake Project

Data Lake Challenges

Technology	Process	People	Data
<ul style="list-style-type: none">✓ Complex, multi-layered architecture✓ Unknown storage & scalability✓ Data retrieval✓ Working with uncurated data✓ Performance✓ Change management✓ Evolving, maturing tech	<ul style="list-style-type: none">✓ Right balance of deferred work vs. up-front work to minimize chaos✓ Ignoring established best practices for data mgmt✓ Data quality✓ Governance✓ Security✓ Disaster recovery for large solutions	<ul style="list-style-type: none">✓ Expectations & trust✓ Data stewardship✓ Redundant effort✓ Data engineering skillsets✓ Ownership changes between teams to operationalize solutions	<ul style="list-style-type: none">✓ Data volumes✓ Read & write performance✓ Relating disparate data✓ Schema changes over time✓ Diversity of file formats & types

Getting Started With a Data Lake Project

Always do a **proof of concept** before making a big commitment, including **file management, data access & security**.

Data tagging & cataloging is critical. Capture **metadata** whenever possible.

Consider the **experience level** of your staff, and the ability to support a complex solution.

Assess the impact of **open source** technologies vs. **proprietary** technologies.



Getting Started With a Data Lake Project

Cloud offerings/features/functionality are constantly changing. It is very challenging to **keep up**.

Your goals for going to the cloud will consistently involve **trade-offs: cost, complexity, security**.

Though traditional data warehousing is evolving, the concept of curated, cleansed, user-friendly data structures are still extremely relevant & needed.



Getting Started With a Data Lake Project

Organize data by **time series** whenever possible.

Consider **data access patterns** when designing the folder structure. "Pruning" of data can happen when data is set up in a well-designed hierarchy.

Make careful decisions about **file formats** you select. Every format has tradeoffs.

File sizes:

- There is a 4.77 TB file limit in Gen 2 (there was no specific file limit in Gen1).
- The traditional Hadoop 'small file size' problem still exists (though technology is evolving to help).
- The ideal, practical file size is still ~250MB - ~2GB.



Azure Data Lake: What, Why, and How

Melissa Coates
Solution Architect, BlueGranite

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Appendix A

Suggestions for Continued Learning

Suggestions for Continued Learning

(1/4)

Azure Data Lake Developer Center: <http://azure.github.io/AzureDataLake/> ← tons of helpful links

U-SQL Center: <http://usql.io/>

U-SQL Tutorial: <https://saveenr.gitbooks.io/usql-tutorial/content/>

Azure Data Lake Samples & Documentation: <https://github.com/Azure/AzureDataLake/>

U-SQL Samples & Documentation: <https://github.com/Azure/USQL>

Azure Data Lake Release Notes: https://github.com/Azure/AzureDataLake/tree/master/docs/Release_Notes

Contains new features, fixes, deprecations, and breaking changes

Azure Data Lake Team Blog: <https://blogs.msdn.microsoft.com/azuredatalake/>

Azure Data Lake Extensions for Visual Studio: <https://www.microsoft.com/en-us/download/details.aspx?id=49504>

Azure Data Lake Course from Microsoft Virtual Academy: https://mva.microsoft.com/en-US/training-courses/introducing-azure-data-lake-17795?l=SugmcFt9D_3111787171 ← highly recommended

ADL Twitter account: <https://twitter.com/azuredatalake>

ADL team member Twitter accounts: <https://twitter.com/MikeDoesBigData> + <https://twitter.com/saveenr>

Suggestions for Continued Learning

(2/4)

ADLS Best Practices: <https://docs.microsoft.com/en-us/azure/data-lake-store/data-lake-store-best-practices>

ADLS Performance Tuning Guidance: <https://docs.microsoft.com/en-us/azure/data-lake-store/data-lake-store-performance-tuning-guidance>

ADLA Saving Money & Controlling Costs:

<https://onedrive.live.com/?authkey=%21AHceDeuGX5PKbVw&id=3BDE3286AB2E59F7%211005&cid=3BDE3286AB2E59F7>

Azure Data Architecture Guide: <https://docs.microsoft.com/en-us/azure/architecture/data-guide/>

Creating External Data Sources for PolyBase and Elastic Queries: <https://docs.microsoft.com/en-us/sql/t-sql/statements/create-external-data-source-transact-sql>

Azure Big Data Blog: <https://azure.microsoft.com/en-us/blog/topics/big-data/>

Suggestions for Continued Learning

(3/4)

Blog Posts

Querying Data in Azure Data Lake Store with Power BI:

<https://www.sqlchick.com/entries/2018/5/6/querying-data-in-azure-data-lake-store-with-power-bi>

Zones in a Data Lake: <https://www.sqlchick.com/entries/2017/12/30/zones-in-a-data-lake>

Granting Permissions in Azure Data Lake: <https://www.sqlchick.com/entries/2018/3/16/granting-permissions-azure-data-lake>

Running U-SQL on a Schedule with Azure Data Factory to Populate Azure Data Lake:

<https://www.sqlchick.com/entries/2017/10/8/running-u-sql-on-a-schedule-with-azure-data-factory>

Querying Multi-Structured JSON Files with U-SQL in Azure Data Lake: <https://www.sqlchick.com/entries/2017/9/4/querying-multi-structured-json-files-with-u-sql-in-azure-data-lake>

Handling Row Headers in U-SQL: <https://www.sqlchick.com/entries/2017/7/27/handling-row-headers-in-u-sql>

Two Ways to Approach Federated Queries with U-SQL and ADLA:

<https://www.sqlchick.com/entries/2017/10/29/two-ways-to-approach-federated-queries-with-u-sql-and-azure-data-lake-analytics>

Suggestions for Continued Learning

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Video

What's New with Azure Data Lake Storage Gen 2

<https://www.youtube.com/watch?v=DJkFSpis2B0>

E-Book

Data Lakes in a Modern Data Architecture

<https://www.blue-granite.com/data-lakes-in-a-modern-data-architecture-ebook>

Appendix B

Terms & Definitions

Definitions

(1/3)

Data Warehouse

Repository of data from multiple sources, cleansed & enriched for reporting; generally 'schema on write'

Data Lake

Repository of data for multi-structured data; generally 'schema on read'

Hadoop

(1) Data storage via HDFS (Hadoop Distributed File System), and
(2) Set of Apache projects for data processing and analytics

Lambda Architecture

Data processing & storage with batch, speed, and serving layers

ETL

Extract > Transform > Load: traditional paradigm associated with data warehousing and 'schema on write'

ELT

Extract > Load > Transform: newer paradigm associated with data lakes & 'schema on read'

Semantic Model

User-friendly interface for users on top of a data warehouse and/or data lake

Definitions

(2/3)

Data Integration

Physically moving data to integrate multiple sources together

Data Virtualization

Access to one or more distributed data sources without requiring the data to be physically materialized in another data structure

Federated Query

A type of data virtualization: access & consolidate data from multiple distributed data sources

Polyglot Persistence

A multi-platform strategy which values using the most effective technology based on the data itself ("best fit engineering")

Schema on Write

Data structure is applied at design time, requiring additional up-front effort to formulate a data model (relational DBs)

Schema on Read

Data structure is applied at query time rather than when the data is initially stored (data lakes, NoSQL)

Definitions

(3/3)

