Data Lake Overview

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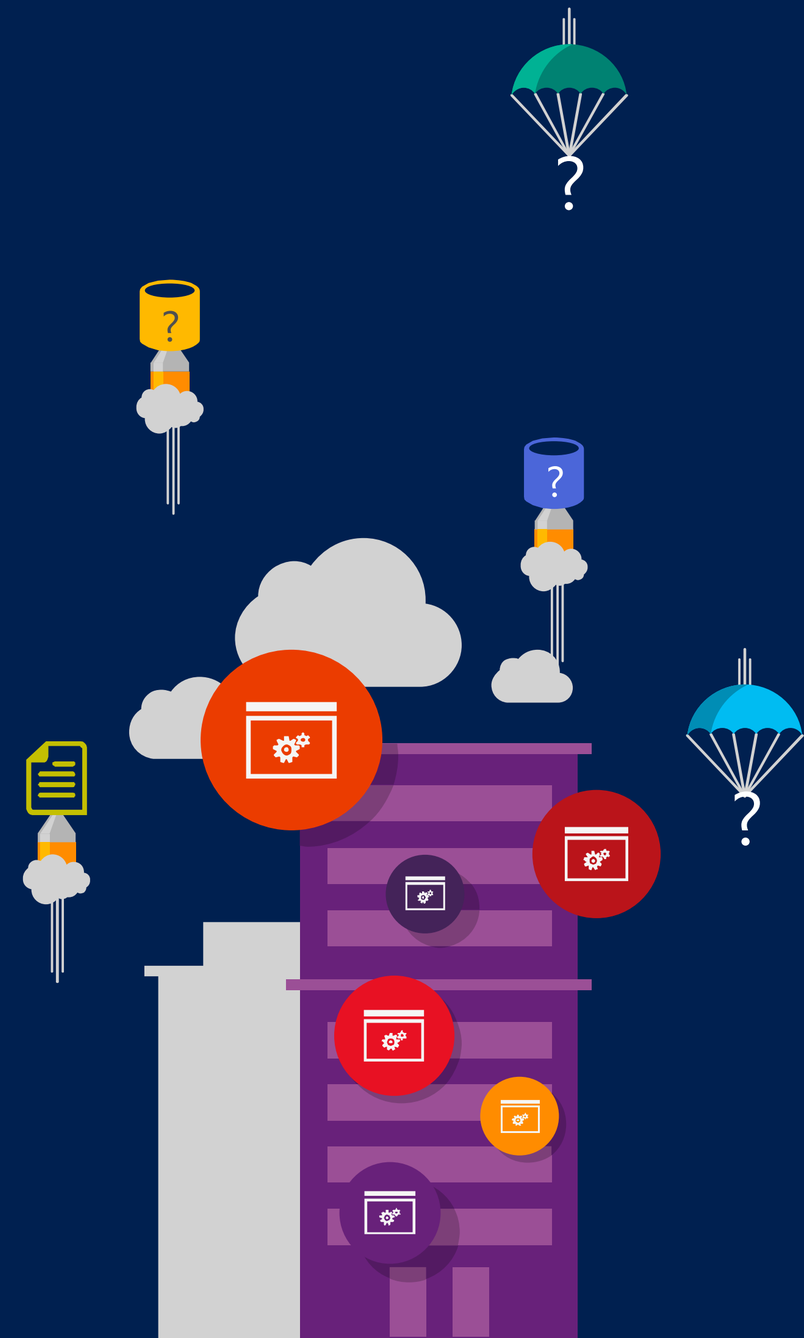
About Me

- Microsoft, Big Data Evangelist
- In IT for 30 years, worked on many BI and DW projects
- Worked as desktop/web/database developer, DBA, BI and DW architect and developer, MDM architect, PDW/APS developer
- Been perm employee, contractor, consultant, business owner
- Presenter at PASS Business Analytics Conference, PASS Summit, Enterprise Data World conference
- Certifications: MCSE: Data Platform, Business Intelligence; MS: Architecting Microsoft Azure Solutions, Design and Implement Big Data Analytics Solutions, Design and Implement Cloud Data Platform Solutions
- Blog at JamesSerra.com
- Former SQL Server MVP
- Author of book “Reporting with Microsoft SQL Server 2012”

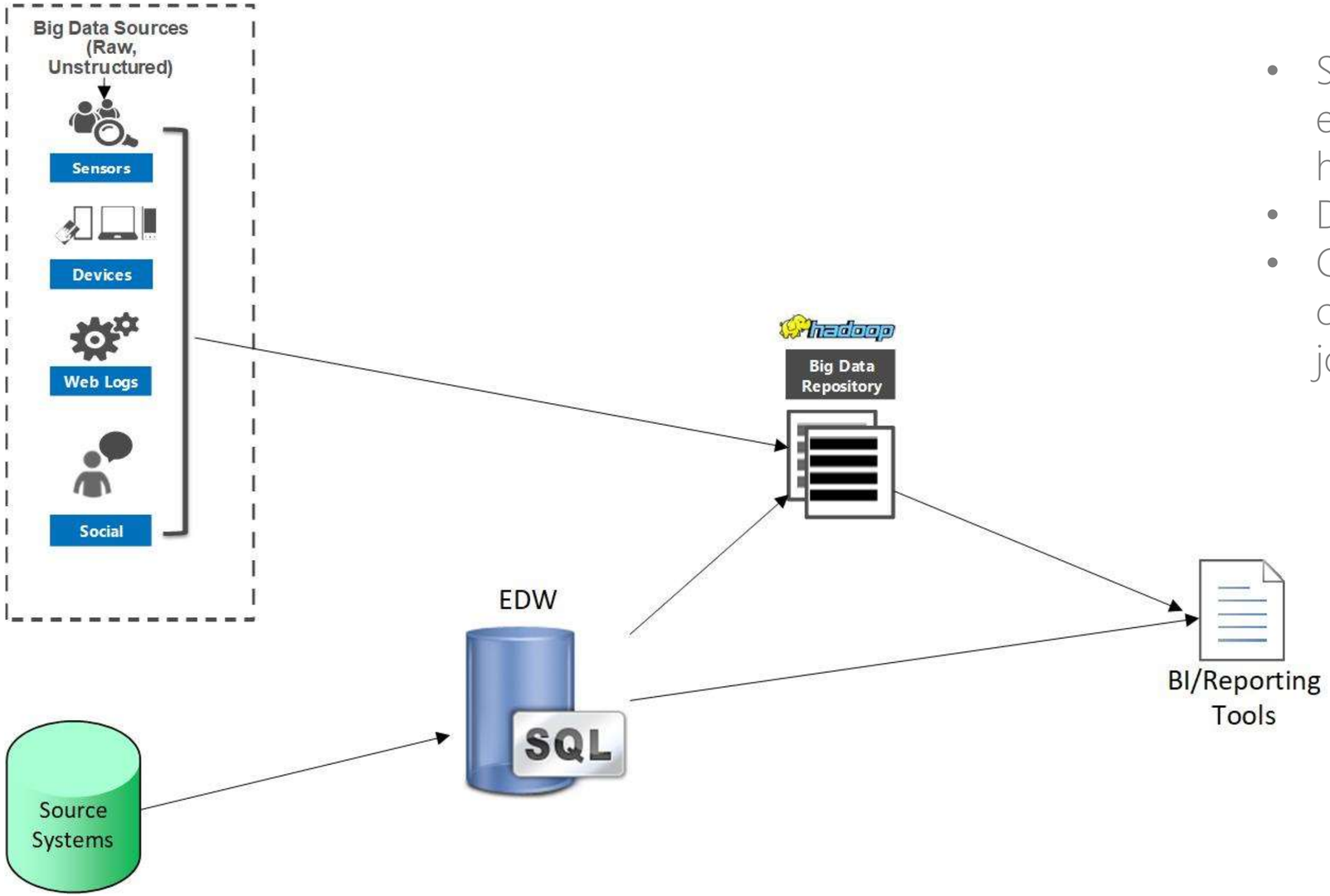
Agenda

- Big Data Architectures
- Why data lakes?
- Top-down vs Bottom-up
- Data lake defined
- Creating ADLS Gen2
- Data Lake Use Cases

Big Data Architectures

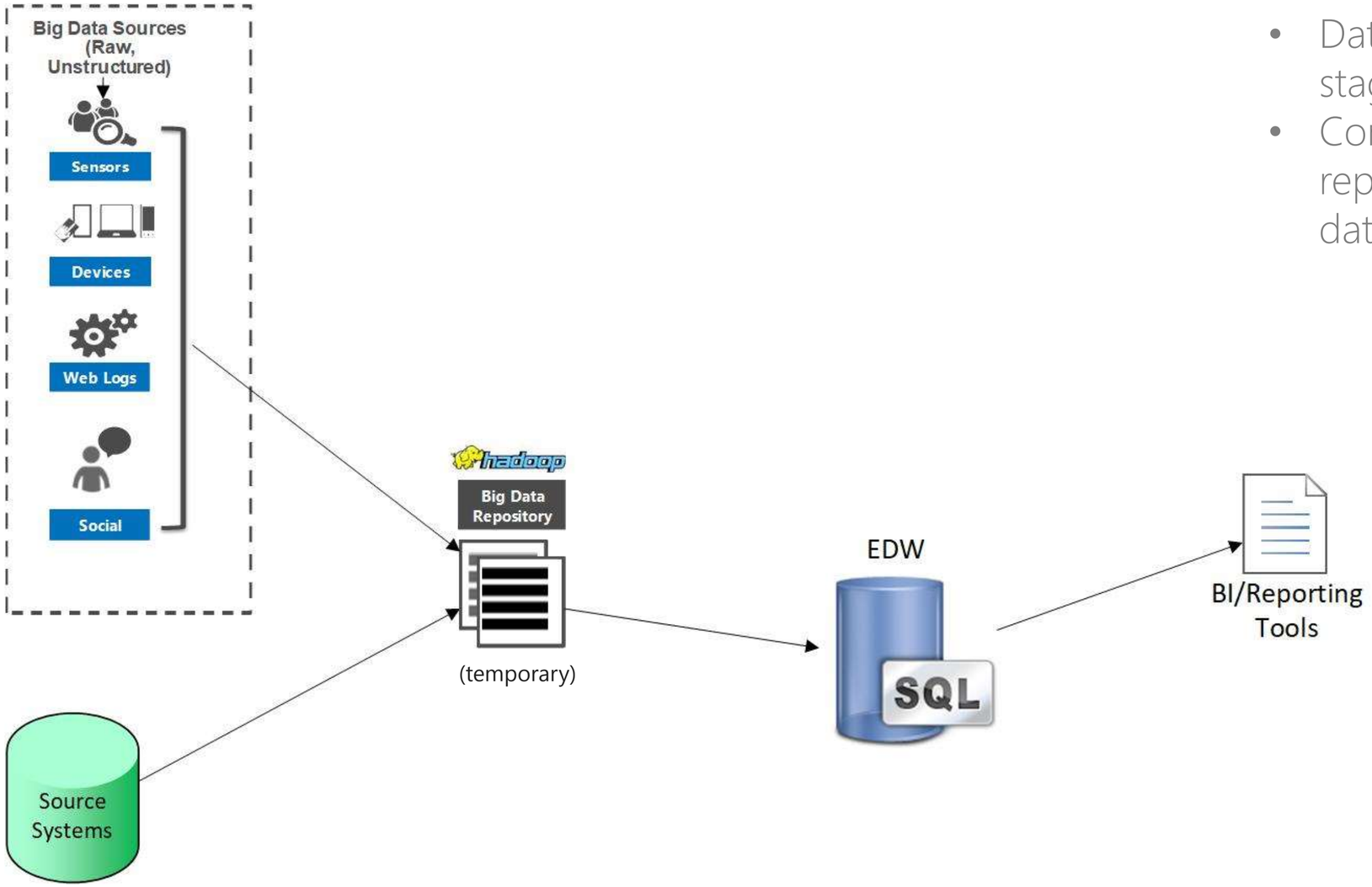


Enterprise data warehouse augmentation



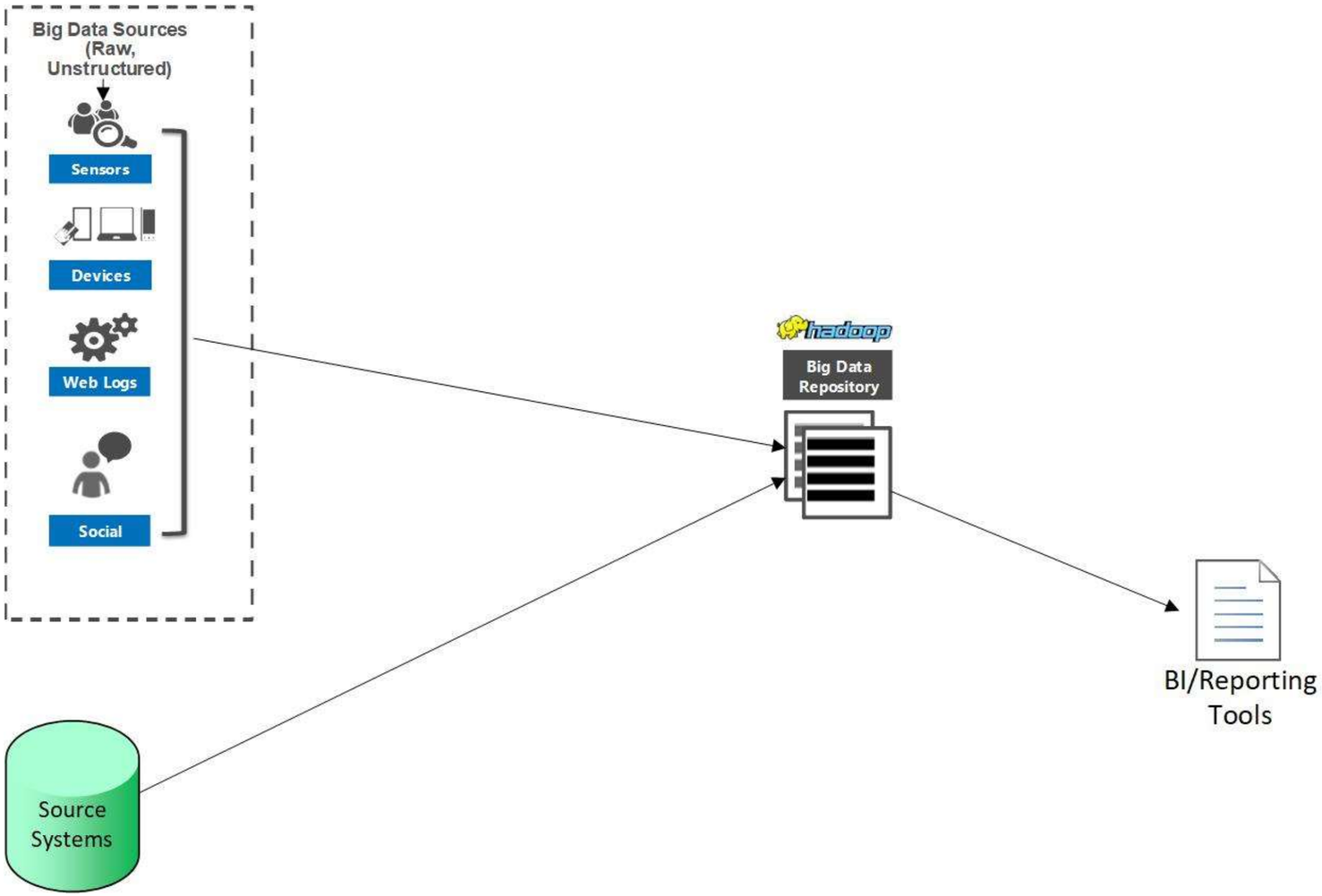
- Seen when EDW has been in existence a while and EDW can't handle new data
- Data hub, not data lake
- Cons: not offloading EDW work, can't use existing tools, difficulty joining data in data hub with EDW

Data hub plus EDW



- Data hub is used as temporary staging and refining, no reporting
- Cons: data hub is temporary, no reporting/analyzing done with the data hub

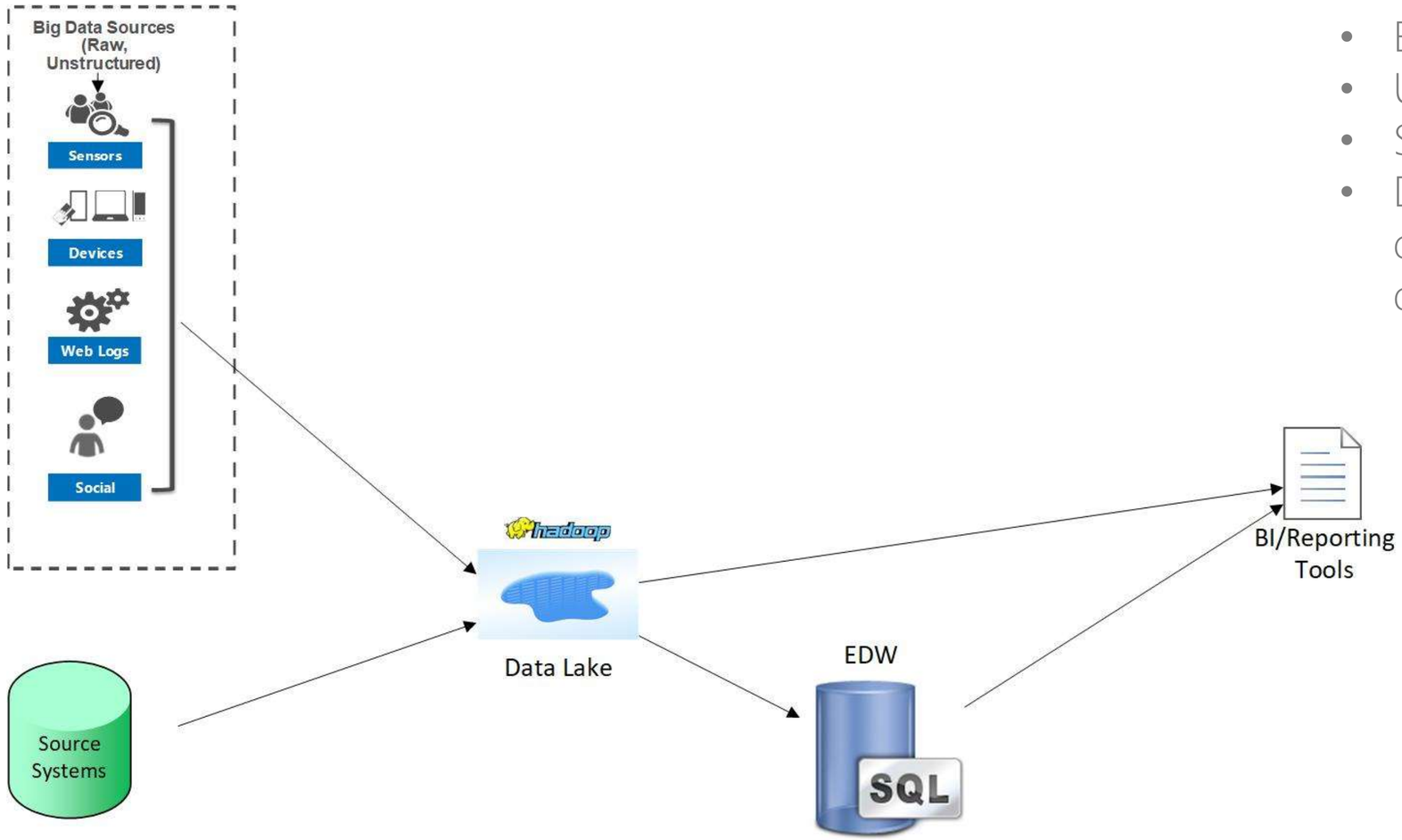
All-in-one



- Data hub is total solution, no EDW
- Cons: queries are slower, new training for reporting tools, difficulty understanding data, security limitations

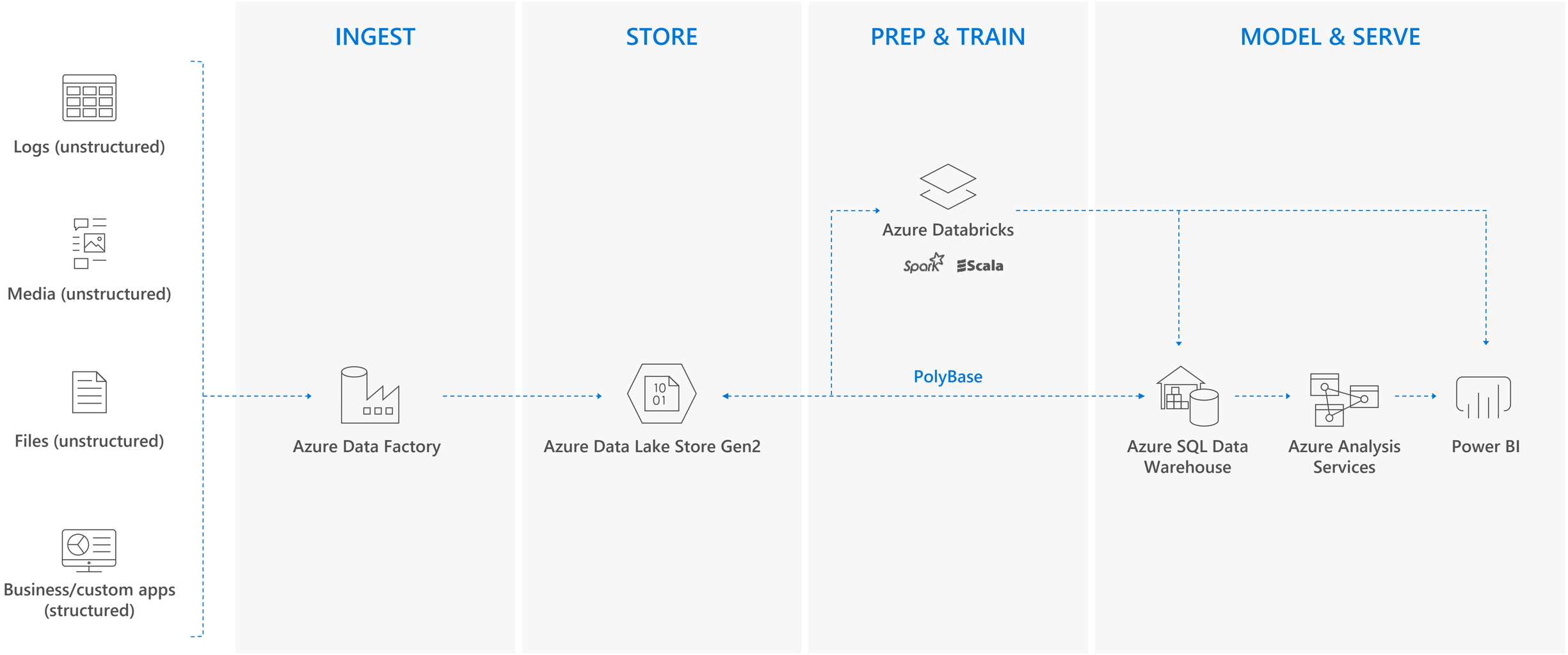
Is the traditional data warehouse dead? https://www.jamesserra.com/archive/2017/12/is-the-traditional-data-warehouse-dead/

Modern Data Warehouse



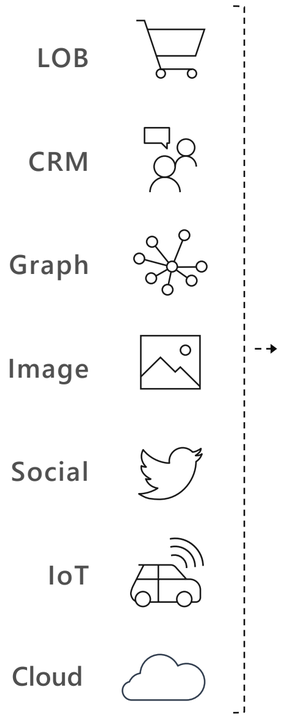
- Evolution of three previous scenarios
- Ultimate goal
- Supports future data needs
- Data harmonized and analyzed in the data lake or moved to EDW for more quality and performance

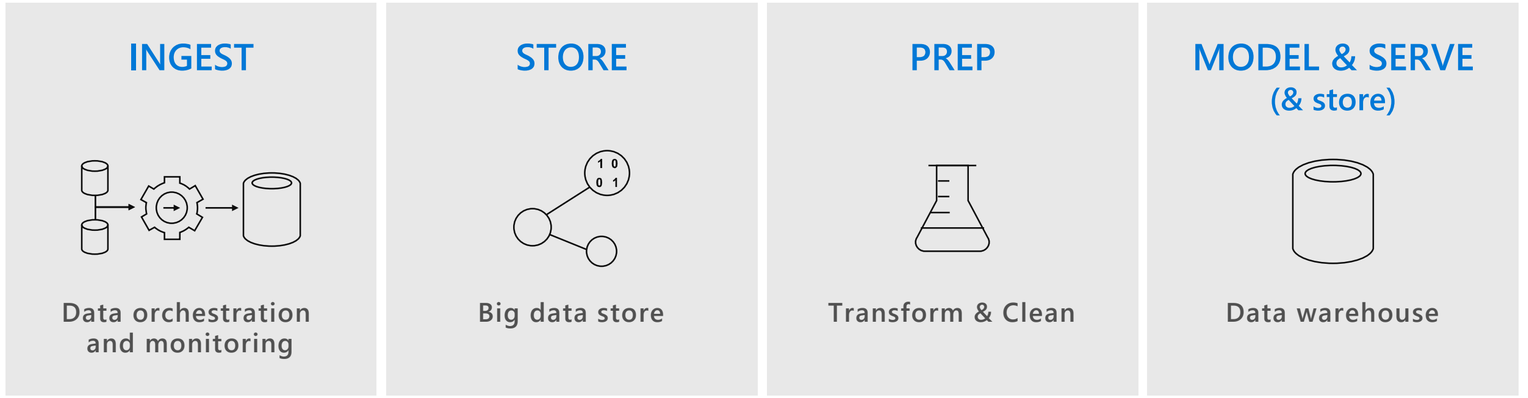
MODERN DATA WAREHOUSE



Microsoft Azure also supports other Big Data services like Azure HDInsight to allow customers to tailor the above architecture to meet their unique needs.

Modern Data Warehouse (possible products by four areas)





- Azure Data Factory
- SSIS

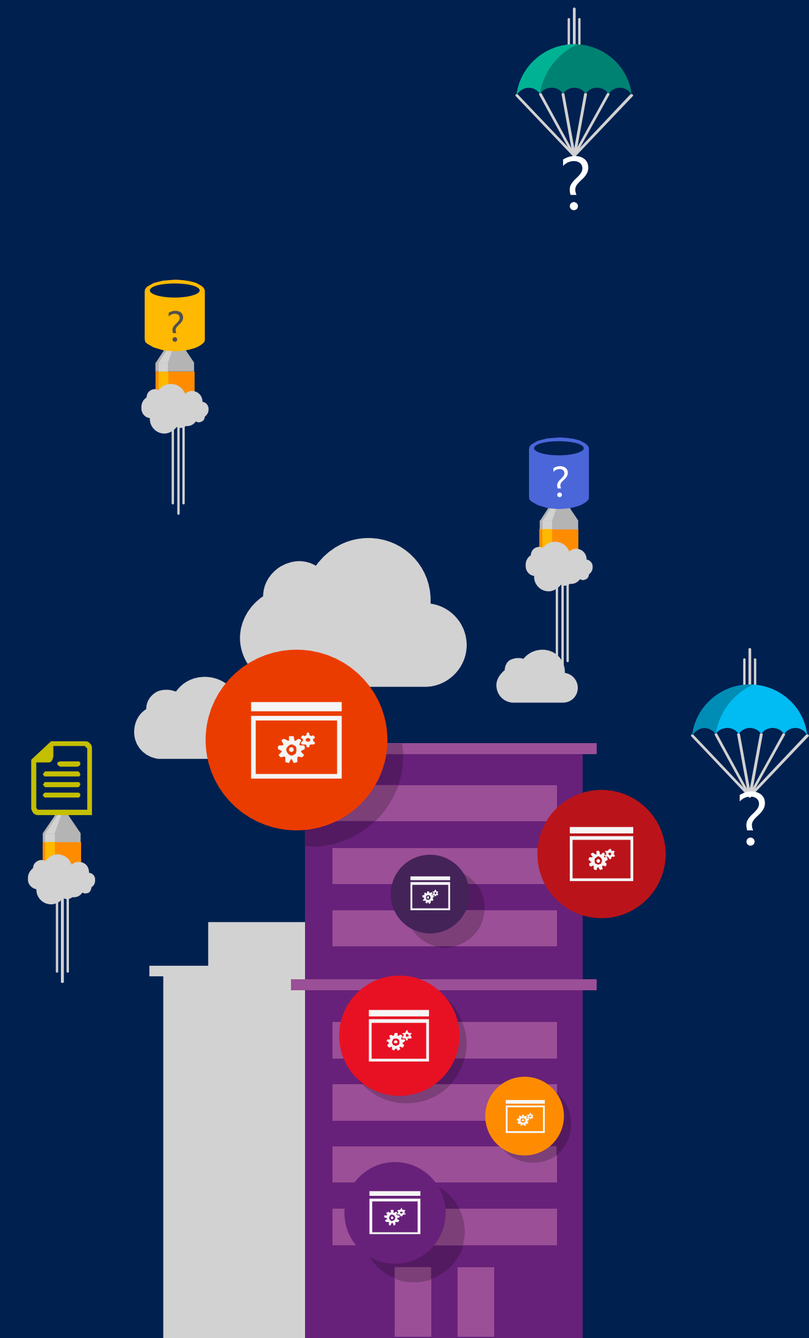
- Azure Data Lake Storage Gen2
- Blob Storage
- SQL Server 2019 Big Data Cluster

- Azure Databricks
- Azure HDInsight
- PolyBase & Stored Procedures
- Power BI Dataflows

- Azure SQL Data Warehouse
- Azure Analysis Services
- SQL Database (Single, MI, HyperScale, Serverless)
- SQL Server in a VM
- Cosmos DB
- Power BI Aggregations

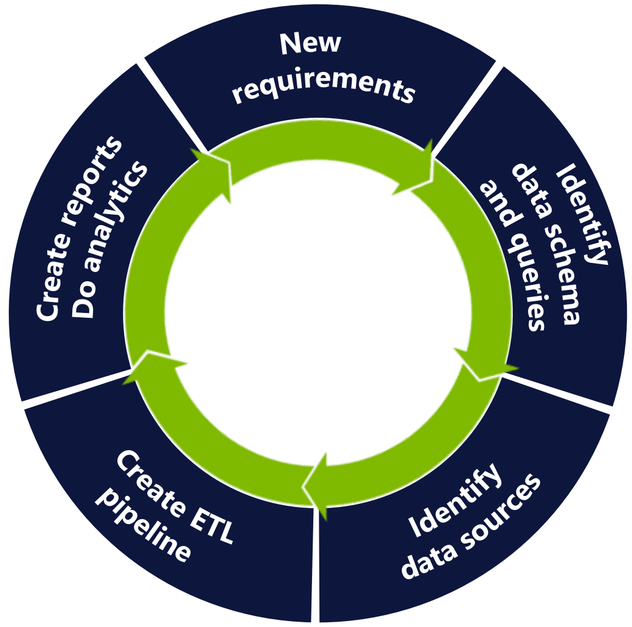
Note: Those products that span more than one area are listed in there primary area

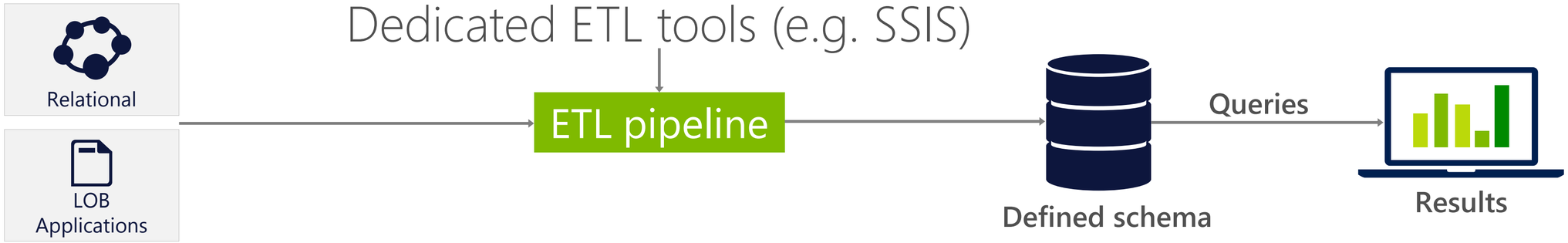
Why data lakes?



Traditional business analytics process

1. Start with end-user requirements to identify desired reports and analysis
2. Define corresponding database schema and queries
3. Identify the required data sources
4. Create a Extract-Transform-Load (ETL) pipeline to extract required data (curation) and transform it to target schema (‘*schema-on-write*’)
5. Create reports. Analyze data





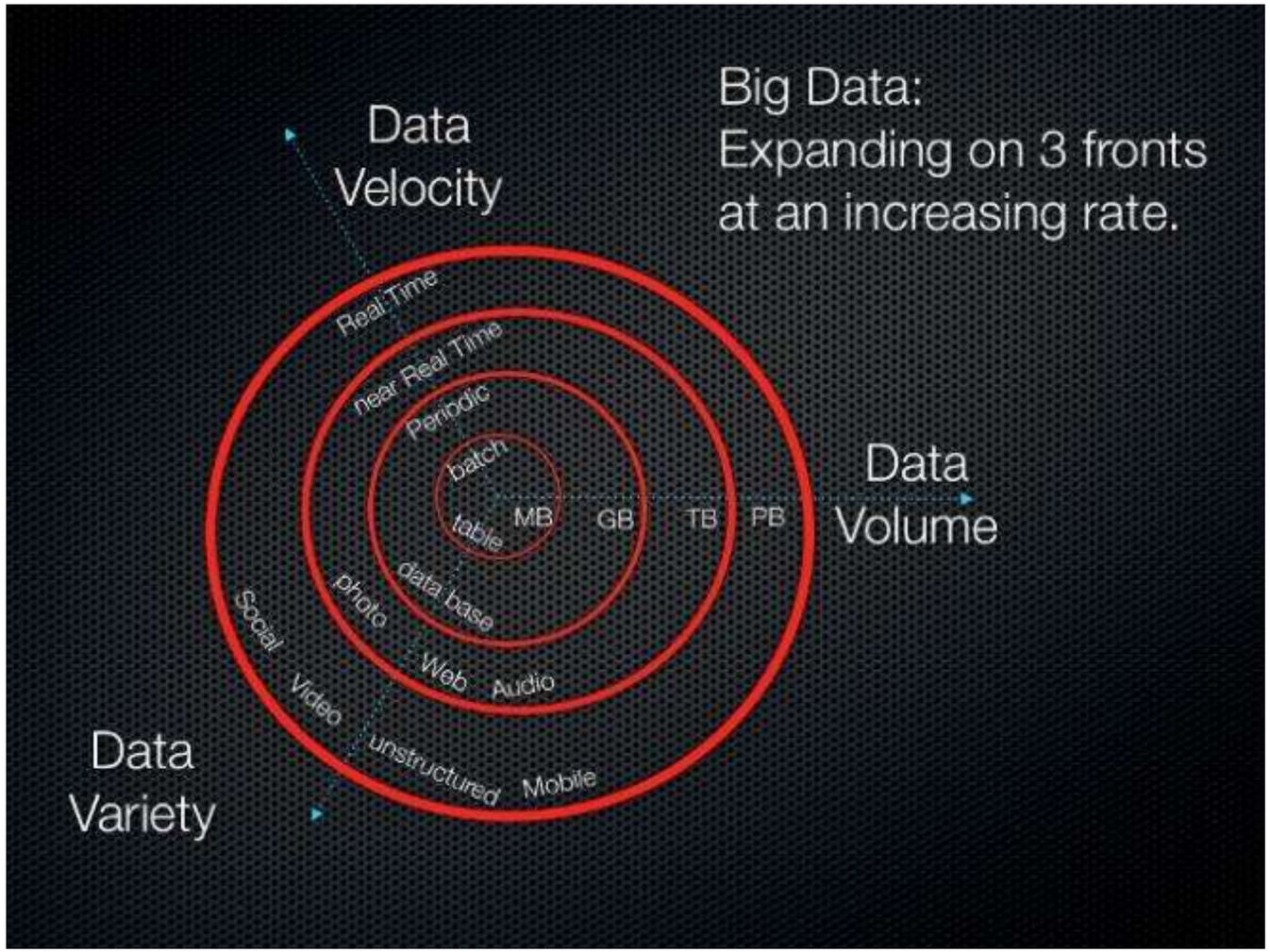
All data not immediately required is discarded or archived

Need to collect any data

Harness the growing and changing nature of data

- Challenge is combining transactional data stored in relational databases with less structured data
- *Big Data = All Data*
- Get the right information to the right people at the right time in the right format

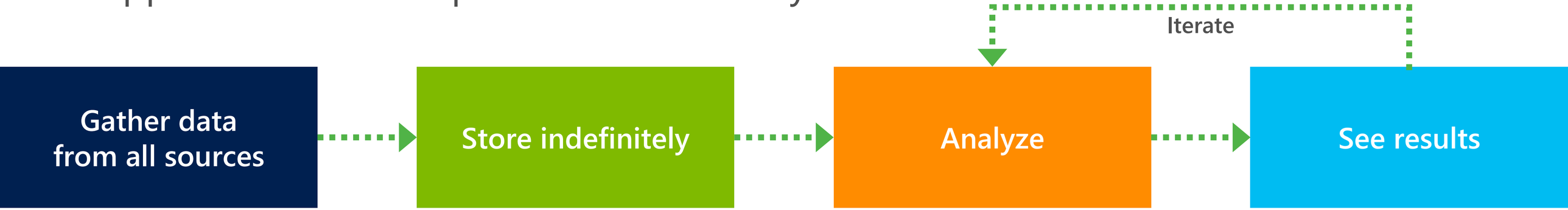
The three V's


Big Data:
Expanding on 3 fronts
at an increasing rate.
Data
Velocity
Real-Time
near Real Time
Periodic
batch
table
MB
GB
TB
PB
Data
Volume
data base
photo
Web
Audio
Social
Video
unstructured
Mobile
Data
Variety

New big data thinking: All data has value

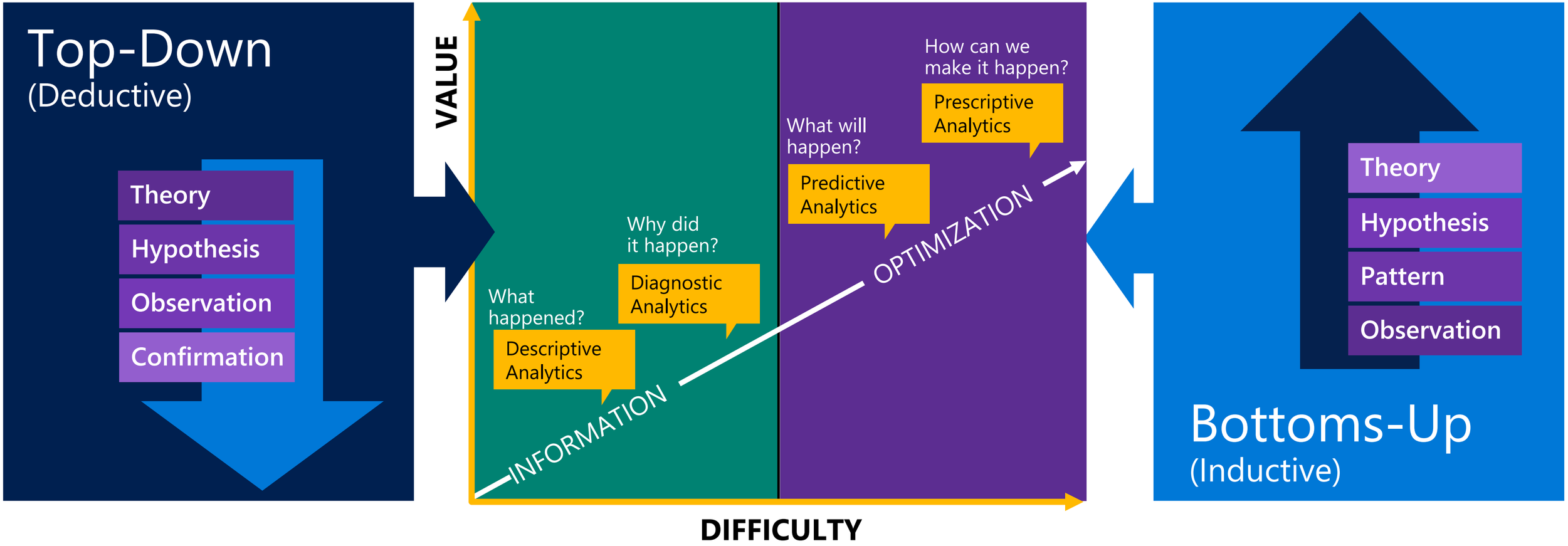
Use a data lake:

- All data has potential value
- Data hoarding
- No defined schema—stored in native format
- Schema is imposed and transformations are done at query time *(schema-on-read)*.
- Apps and users interpret the data as they see fit



Top-down vs Bottom-up

Two Approaches to getting value out of data: Top-Down + Bottoms-Up



- Know the questions to ask
- Lot's of upfront work to get the data to where you can use it
- Model first

- Don't know the questions to ask
- Little upfront work needs to be done to start using data
- Model later

Data Warehousing Uses A Top-Down Approach

Understand Corporate Strategy

Gather Requirements

Business Requirements

Technical Requirements

Implement Data Warehouse

- Reporting & Analytics Design → Reporting & Analytics Development
- Dimension Modelling → Physical Design
- ETL Design → ETL Development
- Setup Infrastructure → Install and Tune

BI and analytic

Dashboards Reporting

Data warehouse

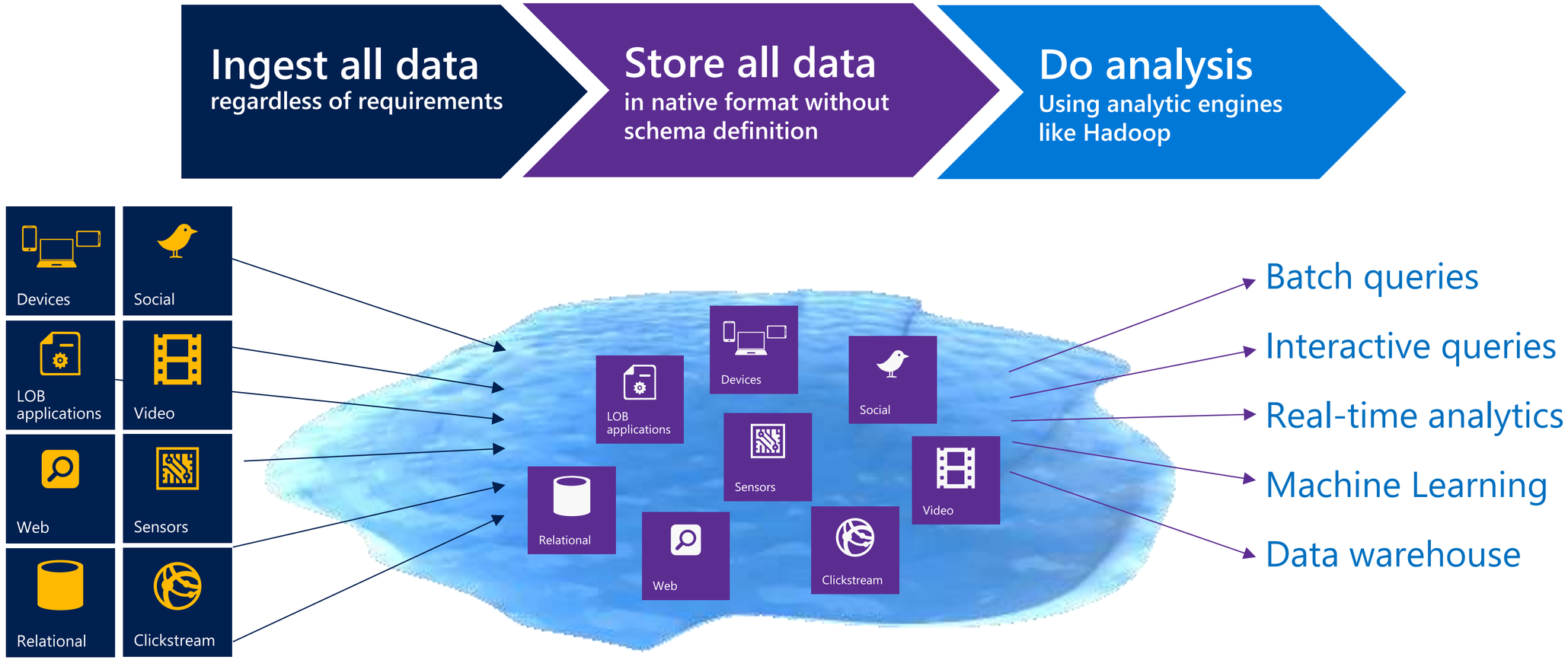
01100
10010
00101

ETL

Data sources

OLTP ERP CRM LOB

The “data lake” Uses A Bottoms-Up Approach



Data Lake + Data Warehouse Better Together

What happened?

Descriptive Analytics

Why did it happen?

Diagnostic Analytics

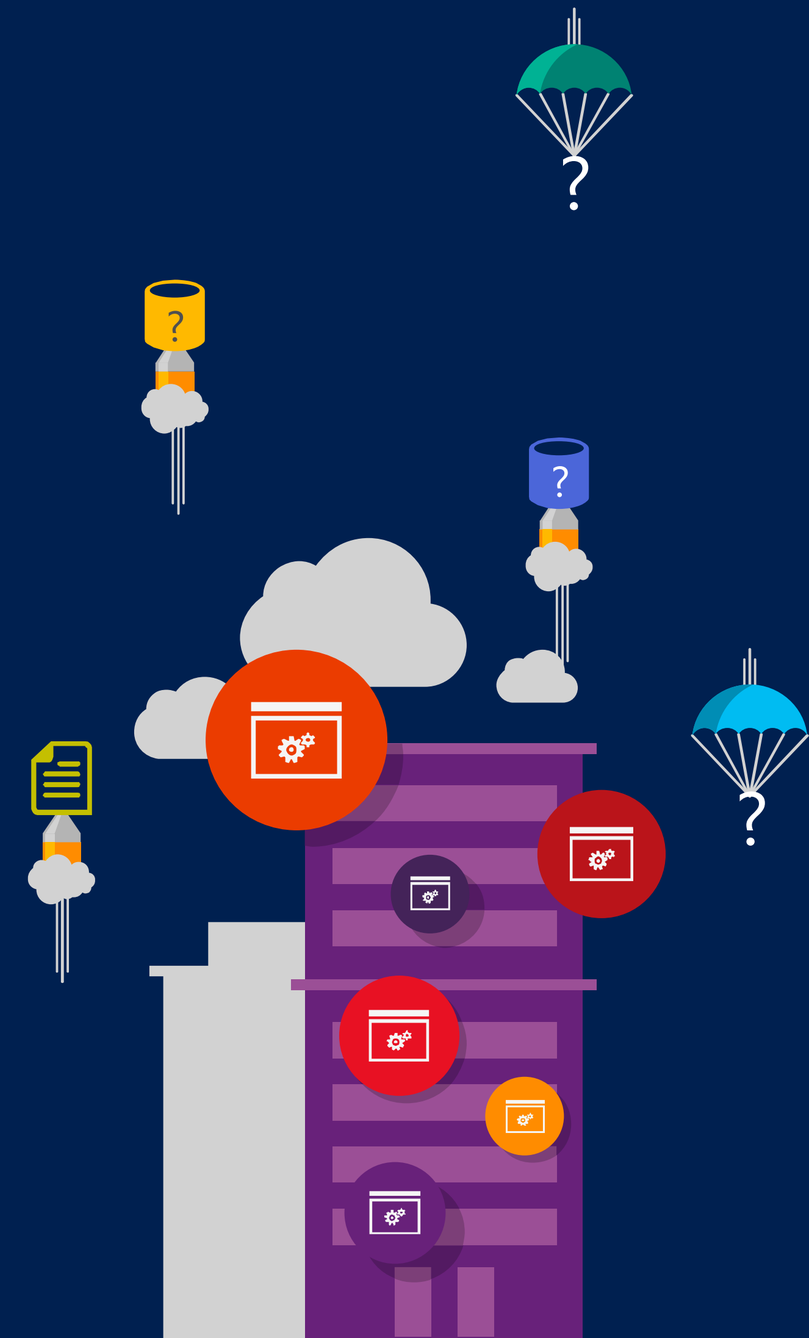
What will happen?

Predictive Analytics

How can we make it happen?

Prescriptive Analytics

Data lake defined



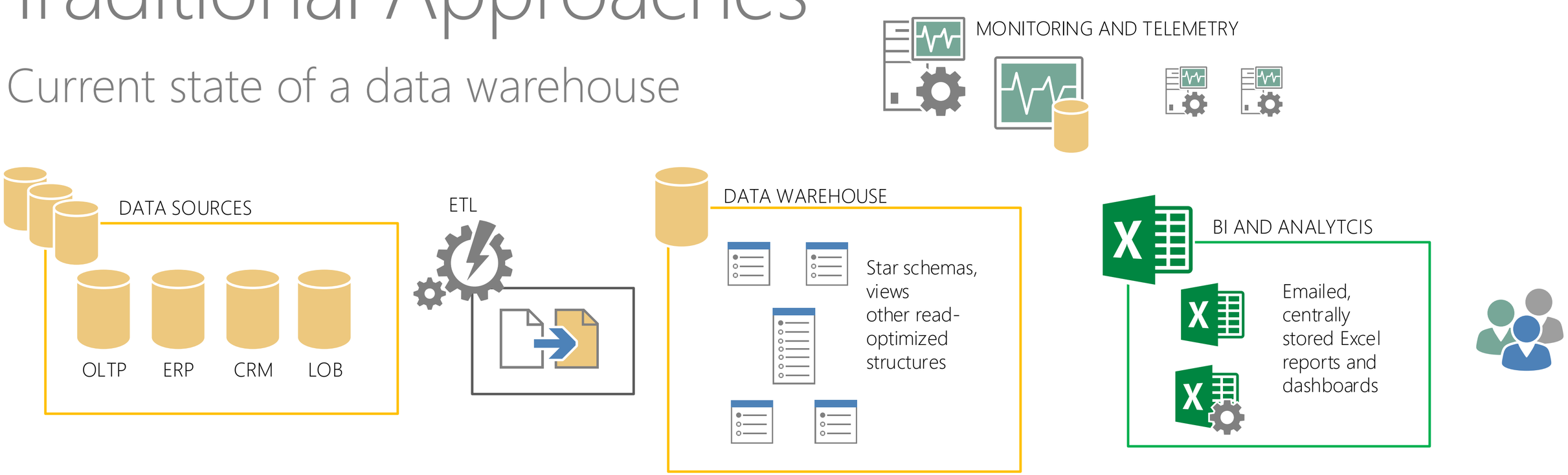
Exactly what is a data lake?

A storage repository, usually Hadoop, that holds a vast amount of raw data in its native format until it is needed.

- Inexpensively store unlimited data
- Centralized place for multiple subjects (single version of the truth)
- Collect all data "just in case" (data hoarding)
- Easy integration of differently-structured data
- Store data with no modeling – "Schema on read"
- Complements enterprise data warehouse (EDW)
- Frees up expensive EDW resources for queries instead of using EDW resources for transformations (avoiding user contention)
- Hadoop cluster offers faster ETL processing over SMP solutions
- Quick user access to data for power users/data scientists (allowing for faster ROI)
- Data exploration to see if data valuable before writing ETL and schema for relational database, or use for one-time report
- Allows use of Hadoop tools such as ETL and extreme analytics
- Place to land IoT streaming data
- On-line archive or backup for data warehouse data
- With Hadoop/ADLS, high availability and disaster recovery built in
- Keep raw data so don't have to go back to source if need to re-run
- Allows for data to be used many times for different analytic needs and use cases
- Cost savings and faster transformations: storage tiers with lifecycle management; separation of storage and compute resources allowing multiple instances of different sizes working with the same data simultaneously vs scaling data warehouse; low-cost storage for raw data saving space on the EDW
- Extreme performance for transformations by having multiple compute options each accessing different folders containing data
- The ability for an end-user or product to easily access the data from any location

Traditional Approaches

Current state of a data warehouse



Well manicured, often relational sources

Known and expected data volume and formats

Little to no change

Complex, rigid transformations

Required extensive monitoring

Transformed historical into read structures

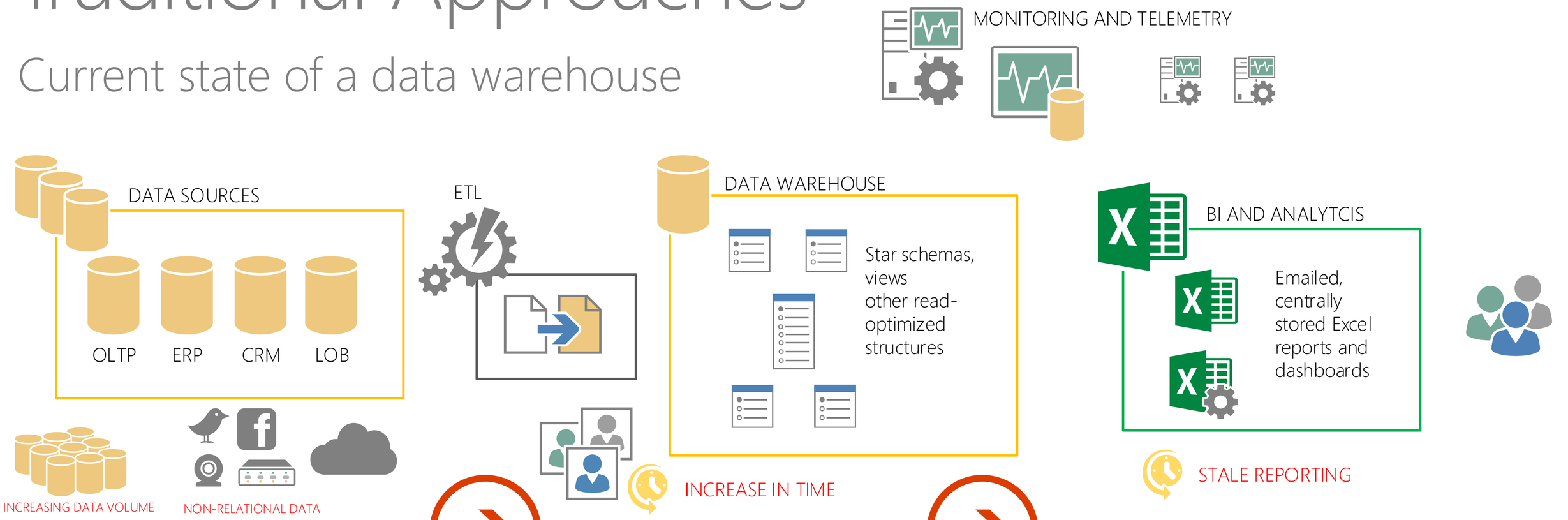
Flat, canned or multi-dimensional access to historical data

Many reports, multiple versions of the truth

24 to 48h delay

Traditional Approaches

Current state of a data warehouse



Increase in variety of data sources

Increase in data volume

Increase in types of data

Pressure on the ingestion engine

Complex, rigid transformations can't longer keep pace

Monitoring is abandoned

Delay in data, inability to transform volumes, or react to new sources

Repair, adjust and redesign ETL

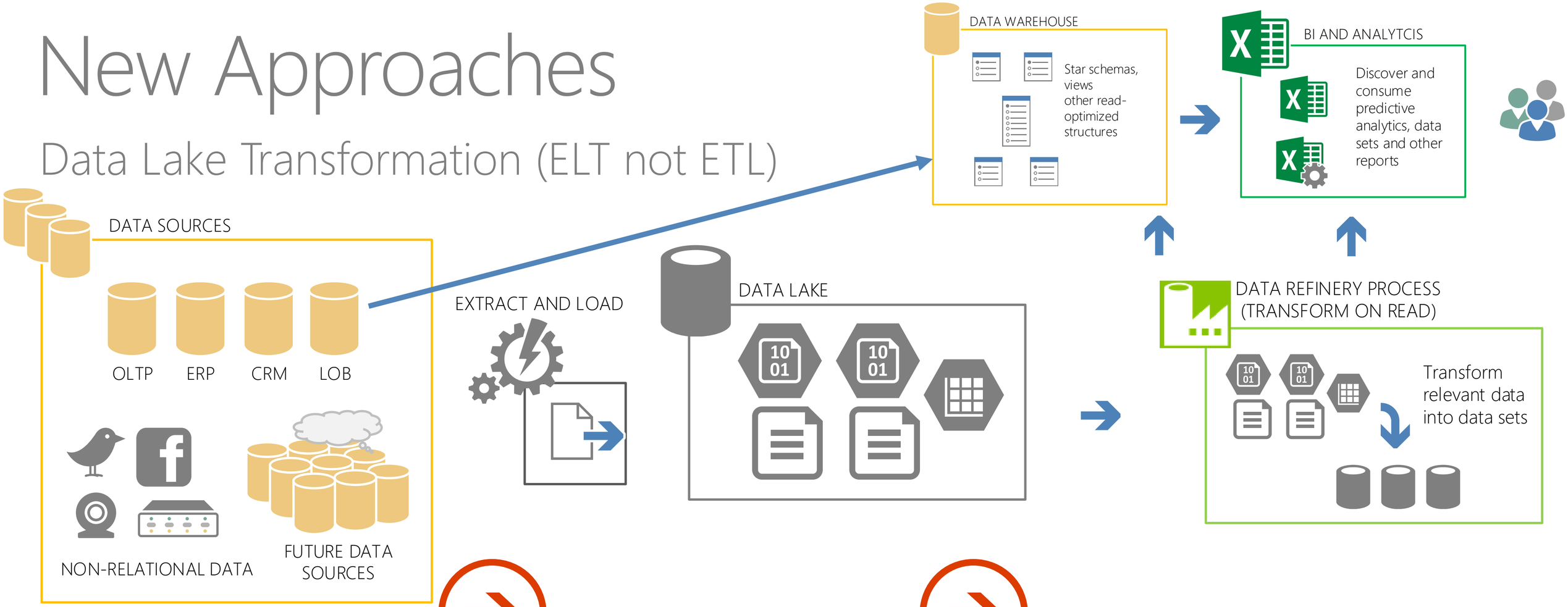
Reports become invalid or unusable

Delay in preserved reports increases

Users begin to "innovate" to relieve starvation

New Approaches

Data Lake Transformation (ELT not ETL)



All data sources are considered

Leverages the power of on-prem technologies and the cloud for storage and capture

Native formats, streaming data, big data

Extract and load, no/minimal transform

Storage of data in near-native format

Orchestration becomes possible

Streaming data accommodation becomes possible

Refineries transform data on read

Produce curated data sets to integrate with traditional warehouses

Users discover published data sets/services using familiar tools

Data Analysis Paradigm Shift

OLD WAY: Structure -> Ingest -> Analyze

NEW WAY: Ingest -> Analyze -> Structure

Data Lake Layers

Raw
Data Layer

Cleansed
Data Layer

Application
Data Layer

Sandbox
Data Layer

Needs data governance so your data lake does not turn into a data swamp!

Organizing a Data Lake – Folder structure

Objectives

- ✓ Plan the structure based on optimal data retrieval
- ✓ Avoid a chaotic, unorganized data swamp

Special thanks to:
Melissa Coates
CoatesDataStrategies.com

Common ways to organize 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…

Organizing a Data Lake

Raw Data Zone

Subject Area
  Data Source
    Object
      Date Loaded
        File(s)

Sales
  Salesforce
    CustomerContacts
      2016
        12
          01
            CustContact_2016_12_01.txt

Example 1

Pros: Subject area at top level, organization-wide
Partitioned by time

Cons: No obvious security or organizational boundaries

Curated Data Zone

Purpose
  Type
    Snapshot Date
      File(s)

Sales Trending Analysis
  Summarized
    2016_12_01
      SalesTrend_2016_12_01.txt

Thanks to Melissa Coates,
www.CoatesDataStrategies.com

Data Lake with DW use cases

Data Lake

Staging & preparation

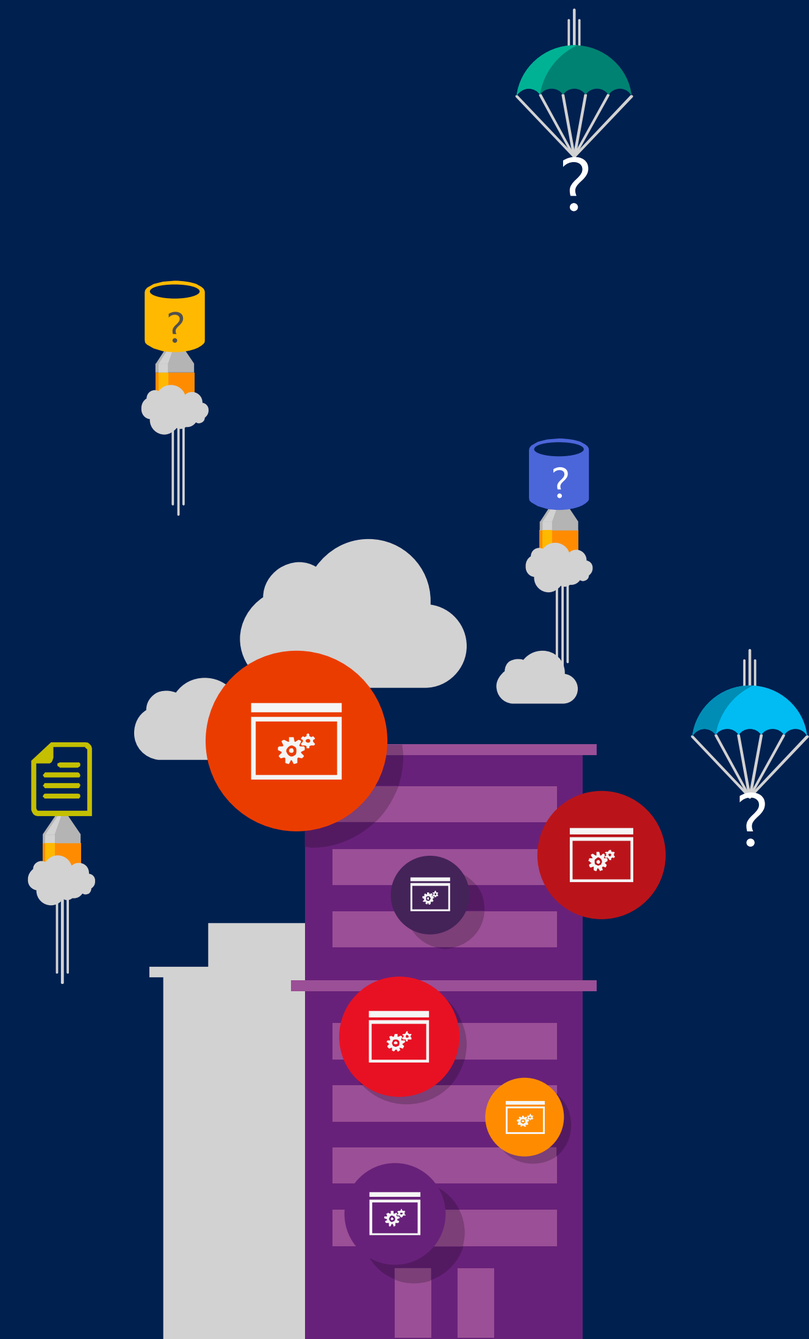
- Data scientists/Power users
- Batch processing
- Data refinement/cleaning
- ETL workloads
- Store older/backup data
- Sandbox for data exploration
- One-time reports
- Quick access to data
- Don’t know questions

Data Warehouse

Serving, Security & Compliance

- Business people
- Low latency
- Complex joins
- Interactive ad-hoc query
- High number of users
- Additional security
- Large support for tools
- Dashboards
- Easily create reports (Self-service BI)
- Know questions

Creating ADLS Gen2



Azure Data Lake Storage Gen2

A **"no-compromises"** Data Lake: secure, performant, massively-scalable Data Lake storage that brings the cost and scale profile of object storage together with the performance and analytics feature set of data lake storage

SECURE

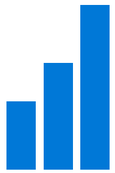
- ✓ Support for fine-grained ACLs, protecting data at the file and folder level
- ✓ Multi-layered protection via at-rest Storage Service encryption and Azure Active Directory integration

MANAGEABLE

- ✓ Automated Lifecycle Policy Management
- ✓ Object Level tiering

FAST

- ✓ Atomic file operations means jobs complete faster



SCALABLE

- ✓ No limits on data store size
- ✓ Global footprint (50 regions)

COST EFFECTIVE

- ✓ Object store pricing levels
- ✓ File system operations minimize transactions required for job completion

INTEGRATION READY

- ✓ Optimized for Spark and Hadoop Analytic Engines
- ✓ Tightly integrated with Azure end to end analytics solutions

Convergence of two Storage Services

Blob Storage

General Purpose Object Storage

Large partner ecosystem
Global scale – All 50 regions
Durability options
Tiered - Hot/Cool/Archive
Cost Efficient

Data Lake Store

Optimized for Big Data analytics

Built for Hadoop
Hierarchical namespace
ACLs, AAD and RBAC
Performance tuned for big data
Very high scale capacity and throughput

Azure Data Lake Storage Gen2

The best of Blobs and ADLS

Large partner ecosystem
Global scale – All 50 regions
Durability options
Tiered - Hot/Cool/Archive
Cost Efficient

Built for Hadoop
Hierarchical namespace
ACLs, AAD and RBAC
Performance tuned for big data
Very high scale capacity and throughput

Remaining known limitations with ADLS Gen2

Missing blob storage features:
- Archive and Premium tier
- Soft Delete
- Snapshots
- Some features in preview

https://docs.microsoft.com/en-us/azure/storage/blobs/data-lake-storage-known-issues

Missing ADLS Gen1 features:
- Microsoft product support: ADC, Excel, AAS
- 3rd-party products: Informatica, Attunity, Alteryx
- Some features in preview

https://docs.microsoft.com/en-us/azure/storage/blobs/data-lake-storage-upgrade

Azure Data Lake Store – Distributed File System

Files of any size can be stored because ADLS is a distributed system which file contents are divided up across backend storage nodes.

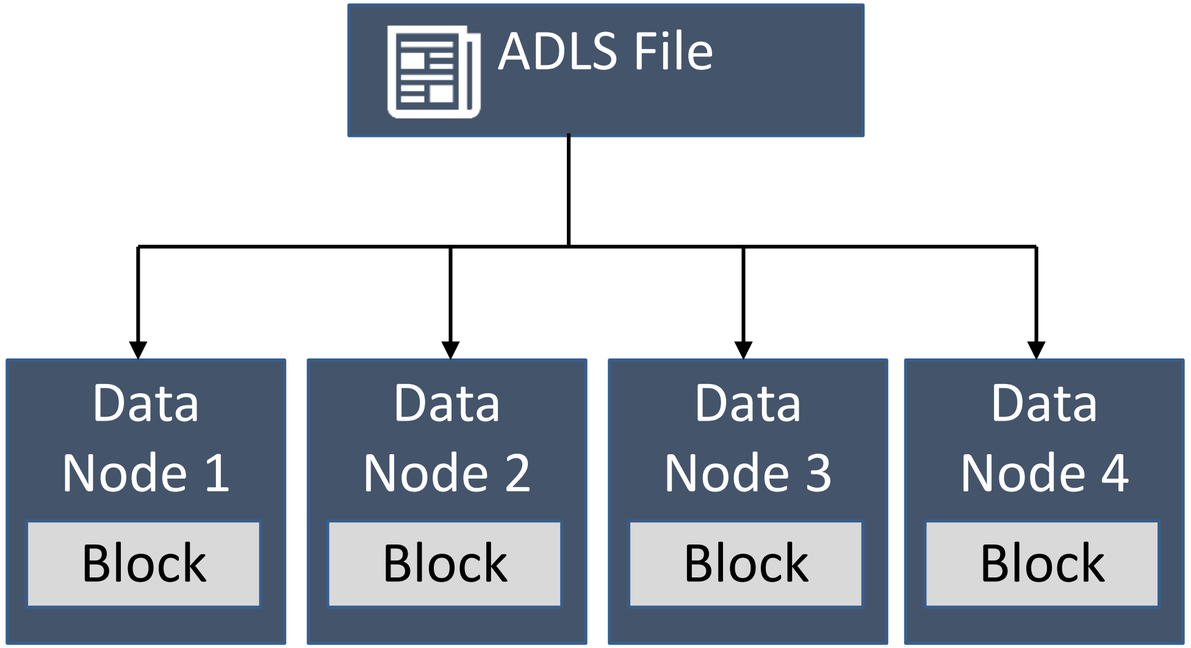
A read operation on the file is also parallelized across the nodes.

Blocks are also replicated for fault tolerance.

The ideal file size in ADLS is 256MB – 2GB in size.

Many very tiny files introduces significant overhead which reduces performance. This is a well-known issue with storing data in HDFS. Techniques:

- Append-only data streams



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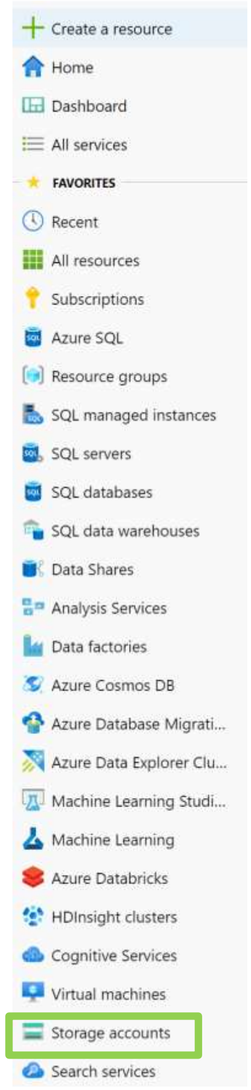
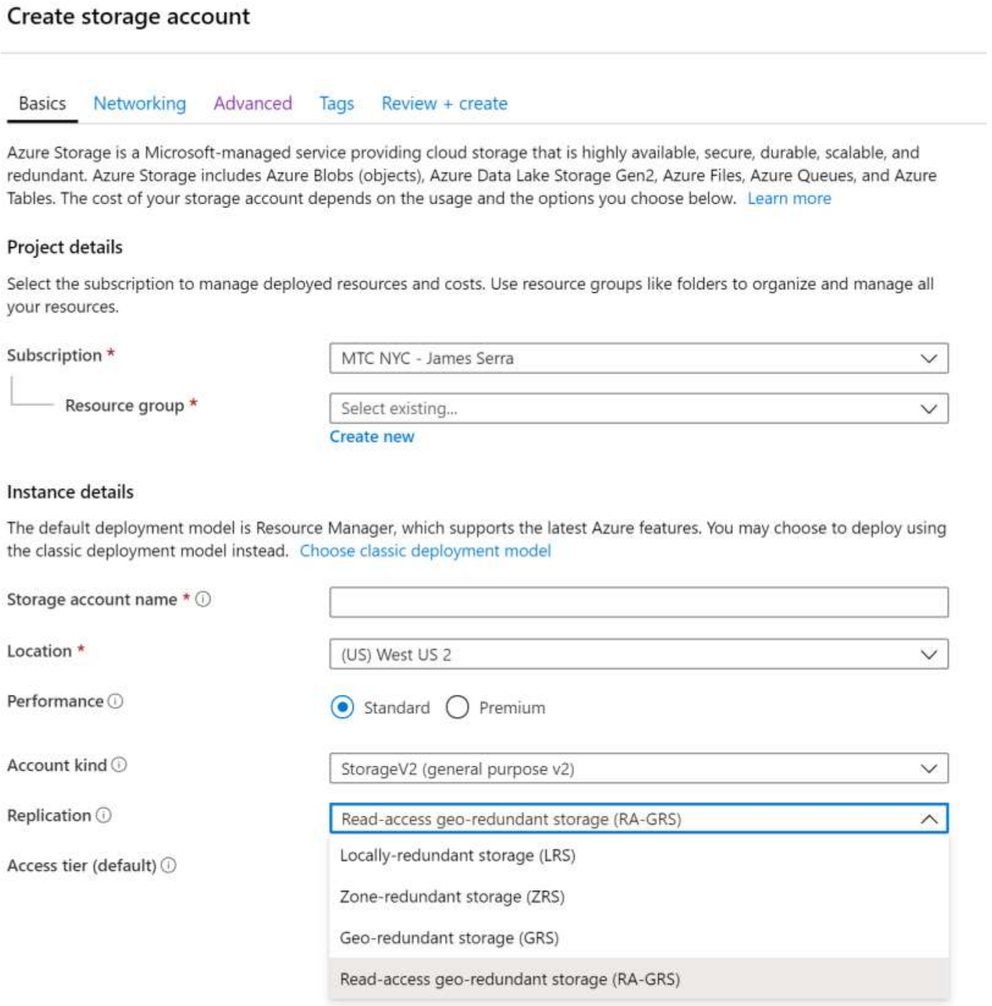
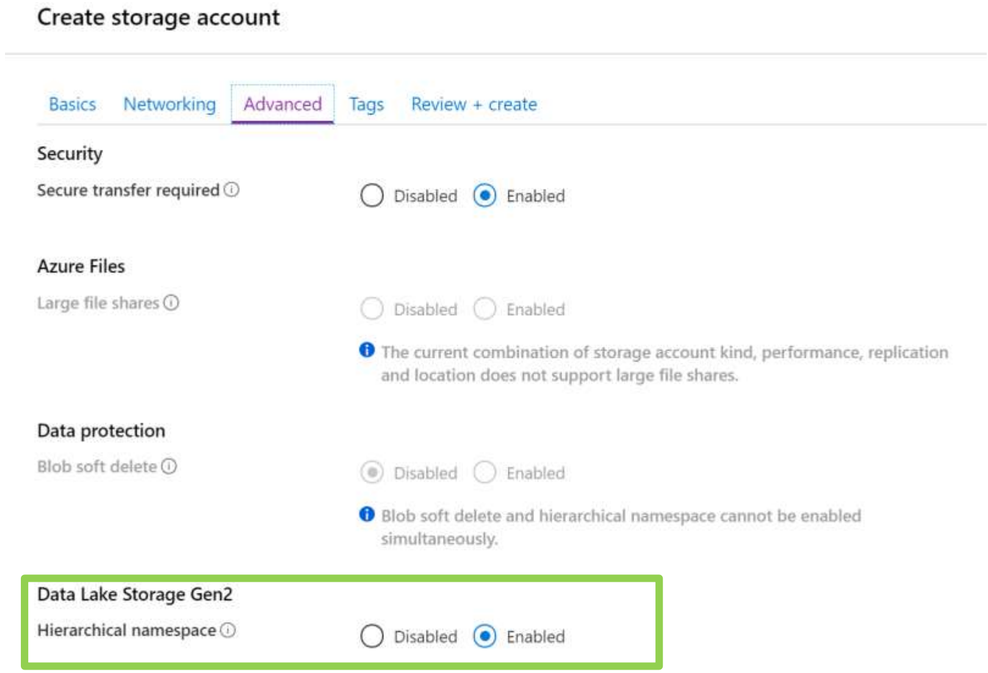
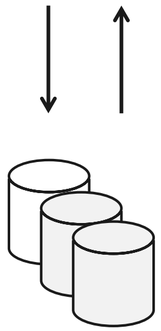
Create ADLS Gen2

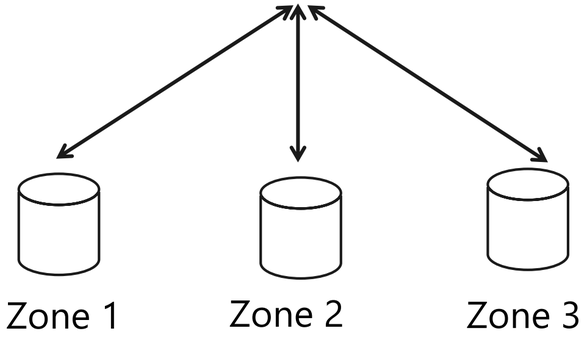


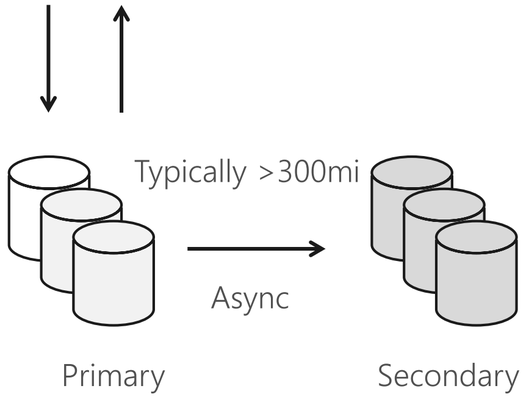


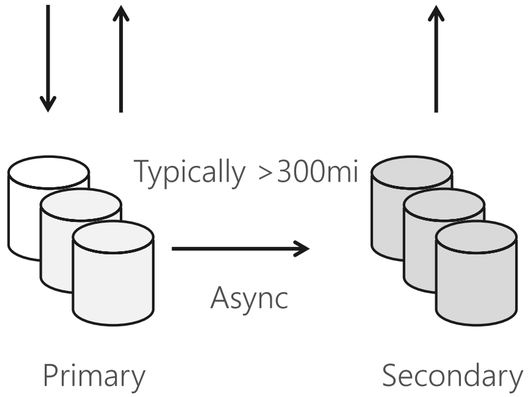


ADLS Gen2 Replication Options









LRS

Multiple replicas across a datacenter

Protect against disk, node, rack failures

Write is ack'd when all replicas are committed

Superior to dual-parity RAID

11 9s of durability

SLA: 99.9%

ZRS

Replicas across 3 Zones

Protect against disk, node, rack and zone failures

Synchronous writes to all 3 zones

12 9s of durability

Available in 8 regions

SLA: 99.9%

GRS

Multiple replicas across each of 2 regions

Protects against major regional disasters

Asynchronous to secondary

16 9s of durability

SLA: 99.9%

RA-GRS

GRS + Read access to secondary

Separate secondary endpoint

RPO delay to secondary can be queried

SLA: 99.99% (read), 99.9% (write)

Preview: Customer controlled failover to GRS location
Geo-zone-redundant storage (GZRS, RA-GZRS)

Data Transport Methods

File Sync
- Windows Srv <-> Azure
- Local caching
- With offline (Databox) can 'sync' remainder

Fuse
- Mount blobs as local FS
- Commit on write
- Linux

Site Replication
- On premise & cloud
- Windows, Linux
- Physical, virtual
- Hyper-V, VMWare

Network Acceleration
- Aspera
- Signiant

AZCopy
- Throughput +30%
- S3 to Azure Blobs
- Sync to cloud
- Hi Latency 10-100%

NetApp
- CloudSync
- SnapMirror
- SnapVault

Data Factory
- On premise & cloud sources
- Structured & unstructured
- Over 60 connectors
- UI design data flow

Partners
- Peer Global File Service
- Talon FAST
- Zerto
- …

Offline
- Data Box
- Data Box Heavy
- Data Box Disk
- Disk Import / Export

Fast Data Transfer
microsoft.com/en-us/garage/profiles/fast-data-transfer/

Azure Data Box Family

Offline Data Transfer

Data Box

- Capacity: 100 TB
- Weight: ~50 lbs
- Secure, ruggedized appliance

- Data Box enables bulk migration to Azure when network isn't an option.

Data Box Disk

- Capacity: 8TB ea.; 40TB/order
- Secure, ruggedized USB drives orderable in packs of 5 (up to 40TB).

- Perfect for projects that require a smaller form factor, e.g., autonomous vehicles.

Data Box Heavy PREVIEW

- Capacity: 1 PB
- Weight 500+ lbs
- Secure, ruggedized appliance

- Same service as Data Box, but targeted to petabyte-sized datasets.

Online Data Transfer

Data Box Gateway

- Virtual device provisioned in your hypervisor
- Supports storage gateway, SMB, NFS, Azure blob, files

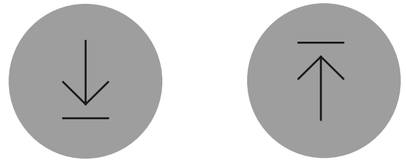
- Virtual network transfer appliance (VM), runs on your choice of hardware.

Data Box Edge

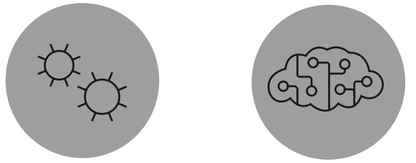
- Local Cache Capacity: ~12 TB
- Includes **Data Box Gateway** and **Azure IoT Edge**.

- Data Box Edge manages uploads to Azure and can pre-process data prior to upload.

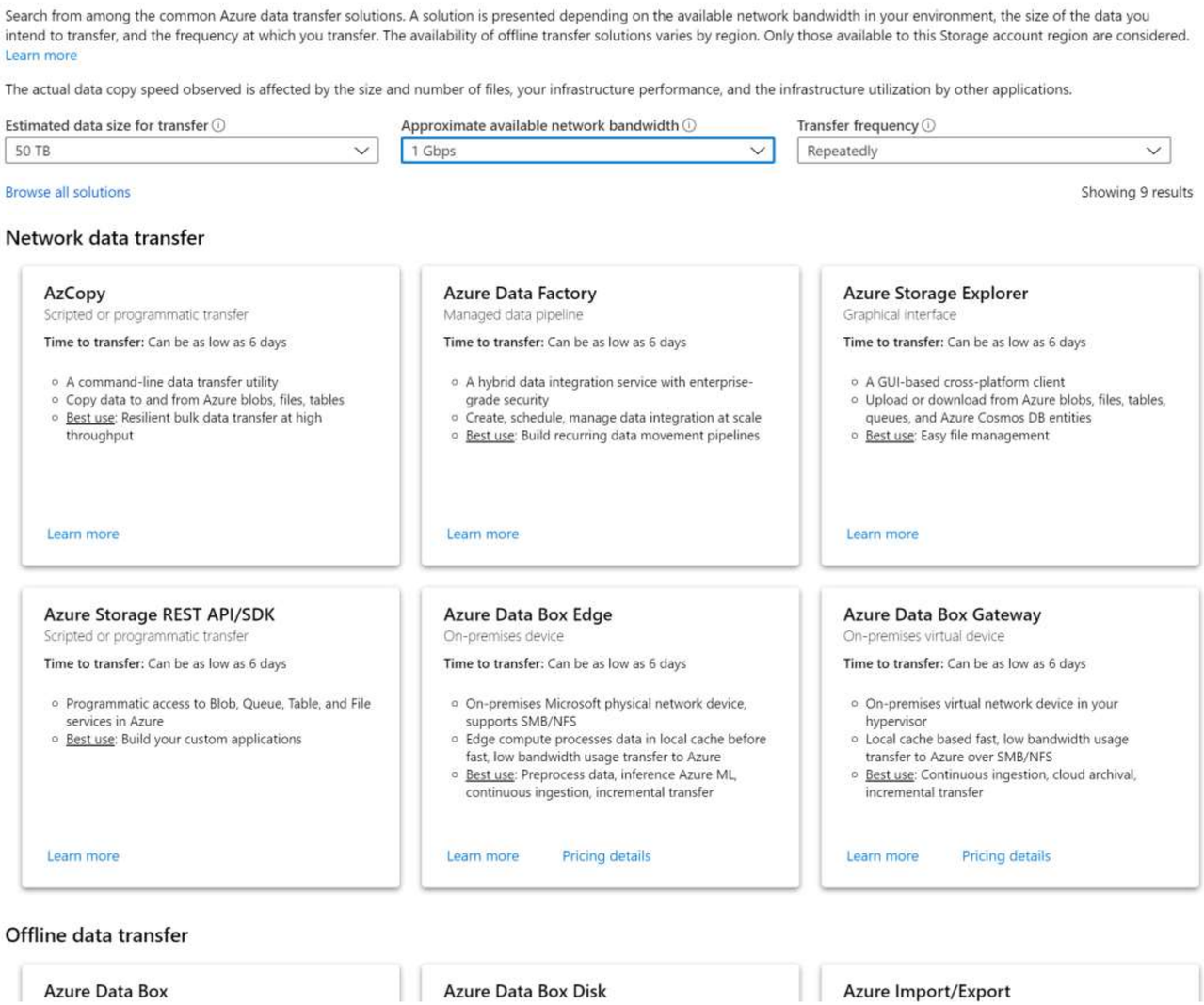
Network Data Transfer



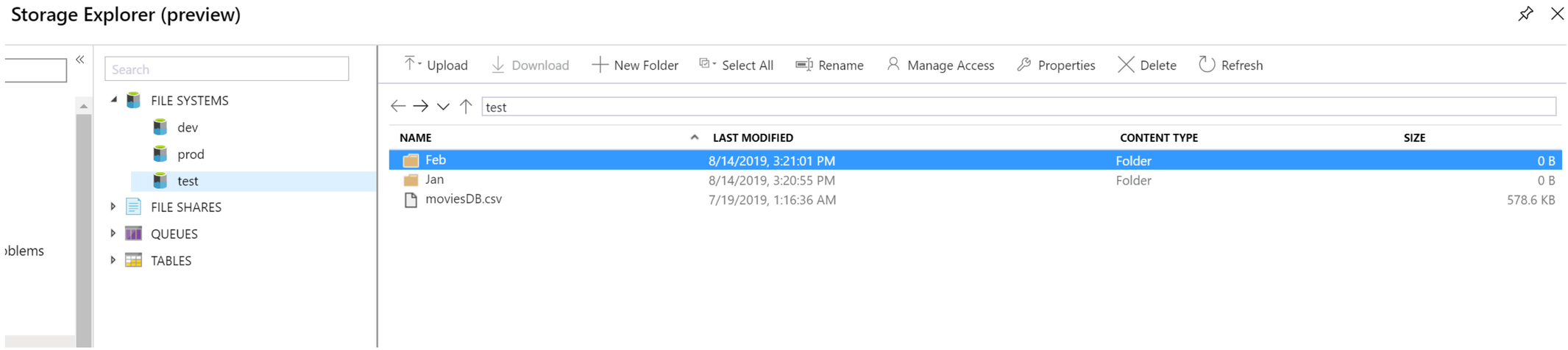
Edge Compute



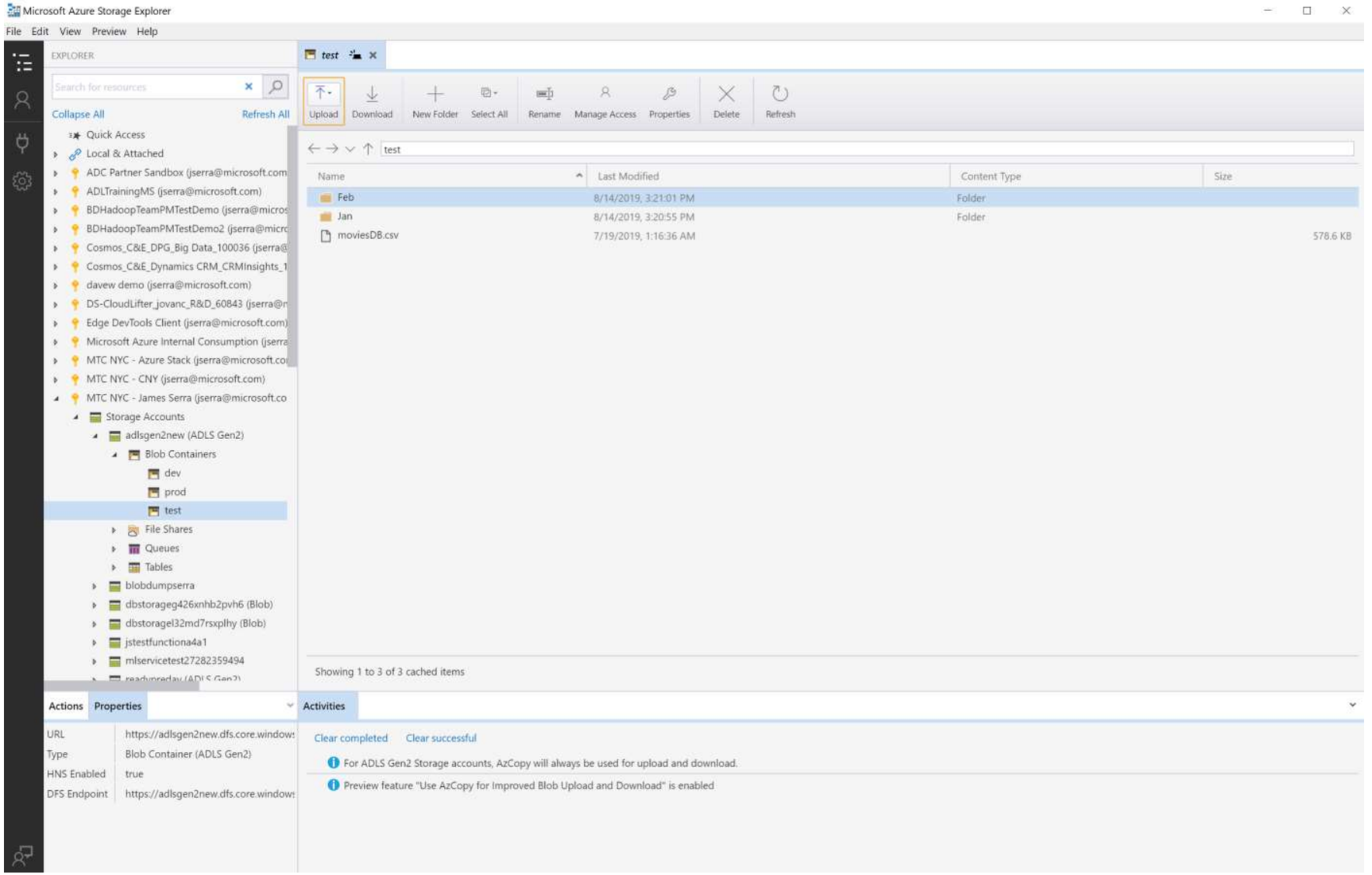
Data transfer



Storage Explorer (preview)



Microsoft Azure Storage Explorer



https://azure.microsoft.com/en-us/features/storage-explorer/

Geo-replication

Refresh

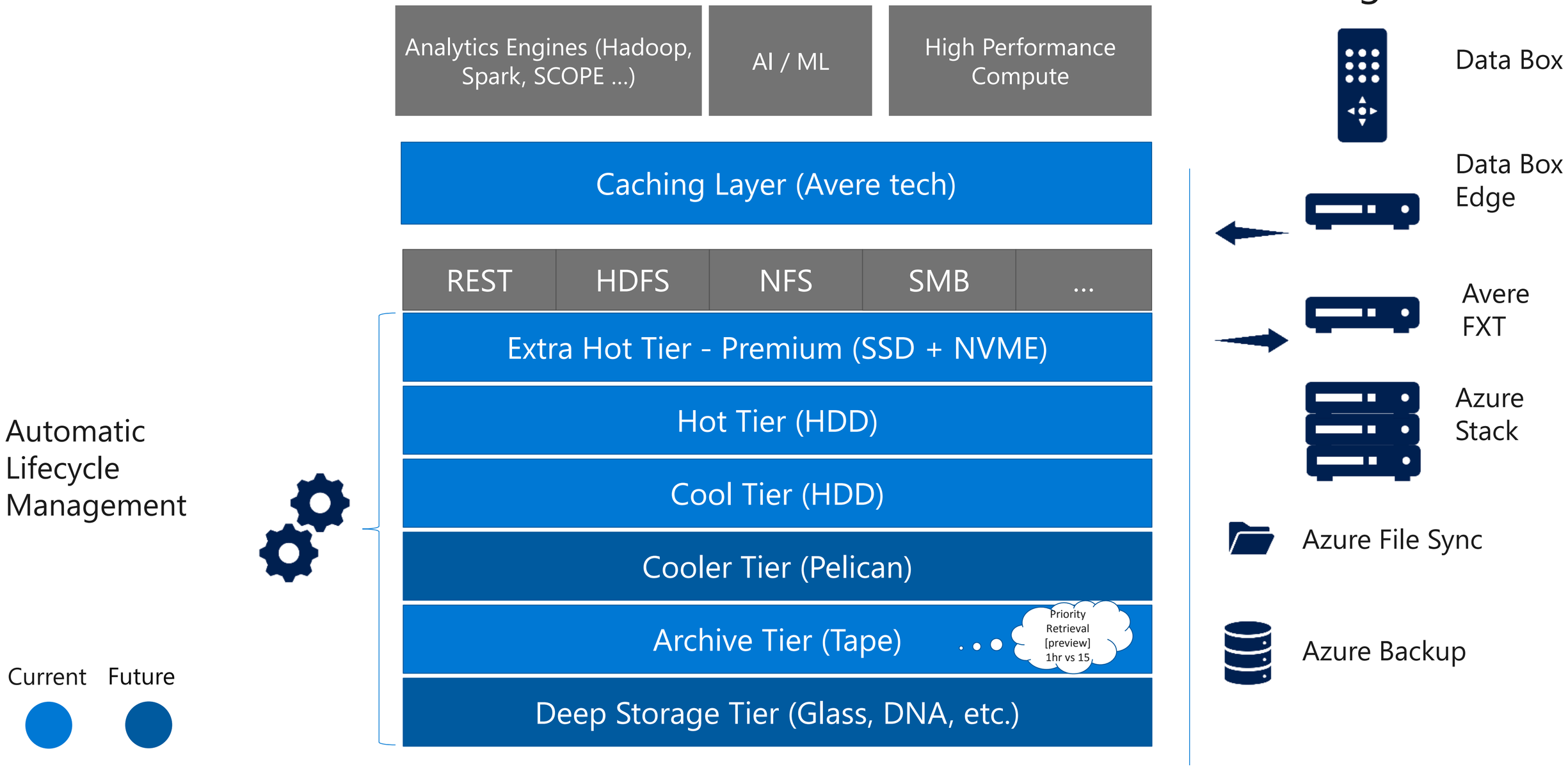
Azure Storage replication copies your data so that it is protected from transient hardware failures, network or power outages, and natural disasters. If an outage renders the primary endpoint unavailable, then you can initiate a failover to the secondary endpoint to rapidly restore write access to your data. To enroll in the failover preview, you will need to submit a request to register this feature to your subscription. Learn more

Replication
Read-access geo-redundant storage (RA-GRS)
Last failover time
-
Storage endpoints
View all

Primary location Secondary location

Location	Data center type	Status	Failover
West US 2	Primary	Available	-
West Central US	Secondary	Available	-

Where we are headed in Cloud Storage



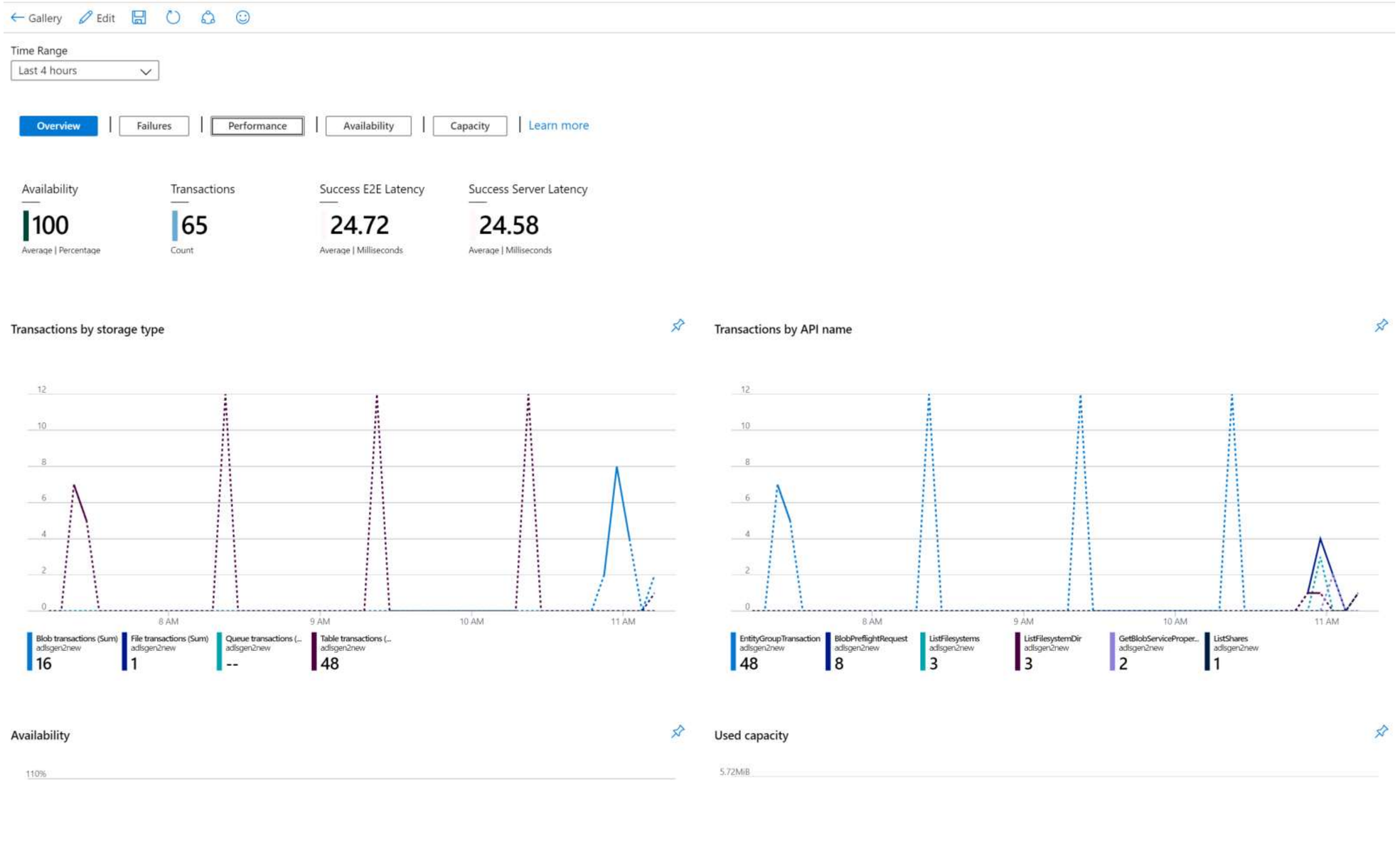
Comparing storage options

	Premium performance	Hot tier	Cool tier	Archive tier
Availability	99.9%	99.9%	99%	Offline
Availability (RA-GRS reads)	N/A	99.99%	99.9%	Offline
Usage charges	Higher storage costs, lower access and transaction cost	Higher storage costs, lower access, and transaction costs	Lower storage costs, higher access, and transaction costs	Lowest storage costs, highest access, and transaction costs
Minimum object size	N/A	N/A	N/A	N/A
Minimum storage duration	N/A	N/A	30 days[1]	180 days
Latency (Time to first byte)	Single-digit milliseconds	milliseconds	milliseconds	hours[2]

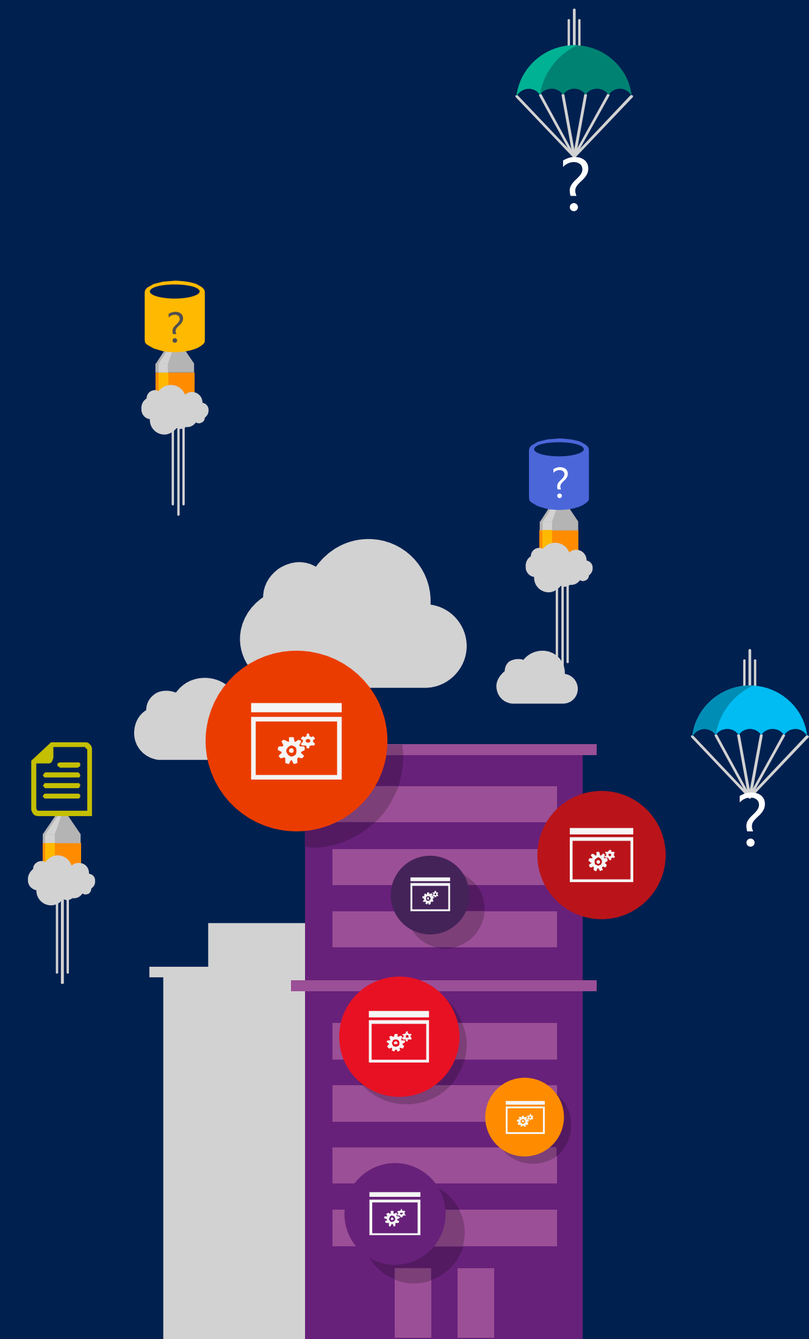
[1] Objects in the cool tier on GPv2 accounts have a minimum retention duration of 30 days. Blob storage accounts don't have a minimum retention duration for the cool tier.

[2] Archive Storage currently supports 2 rehydrate priorities, High and Standard, that offers different retrieval latencies. For more information, see Rehydrate blob data from the archive tier.

Insights (preview)

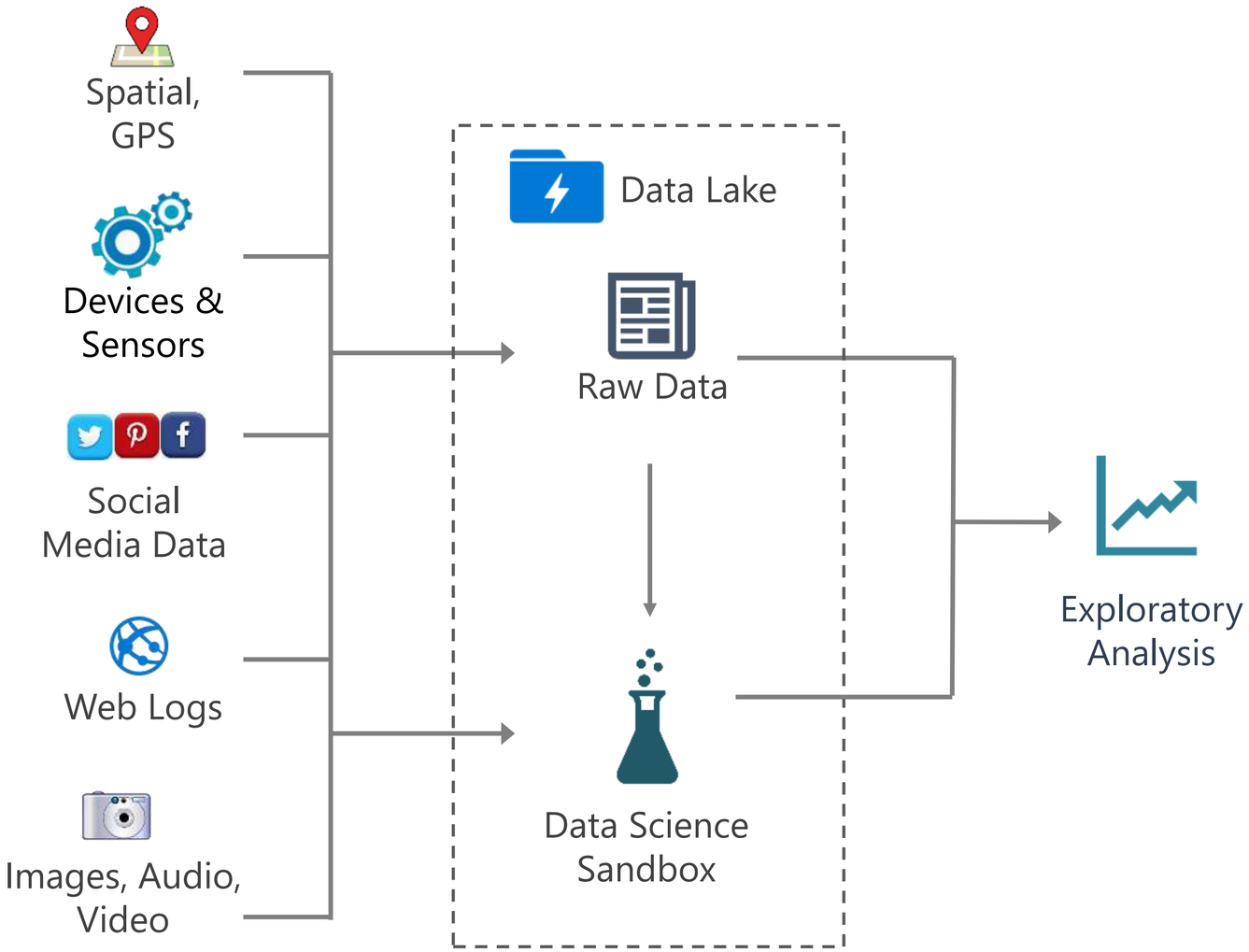


Data Lake Use Cases



Data Lake Use Cases

Ingestion of New File Types

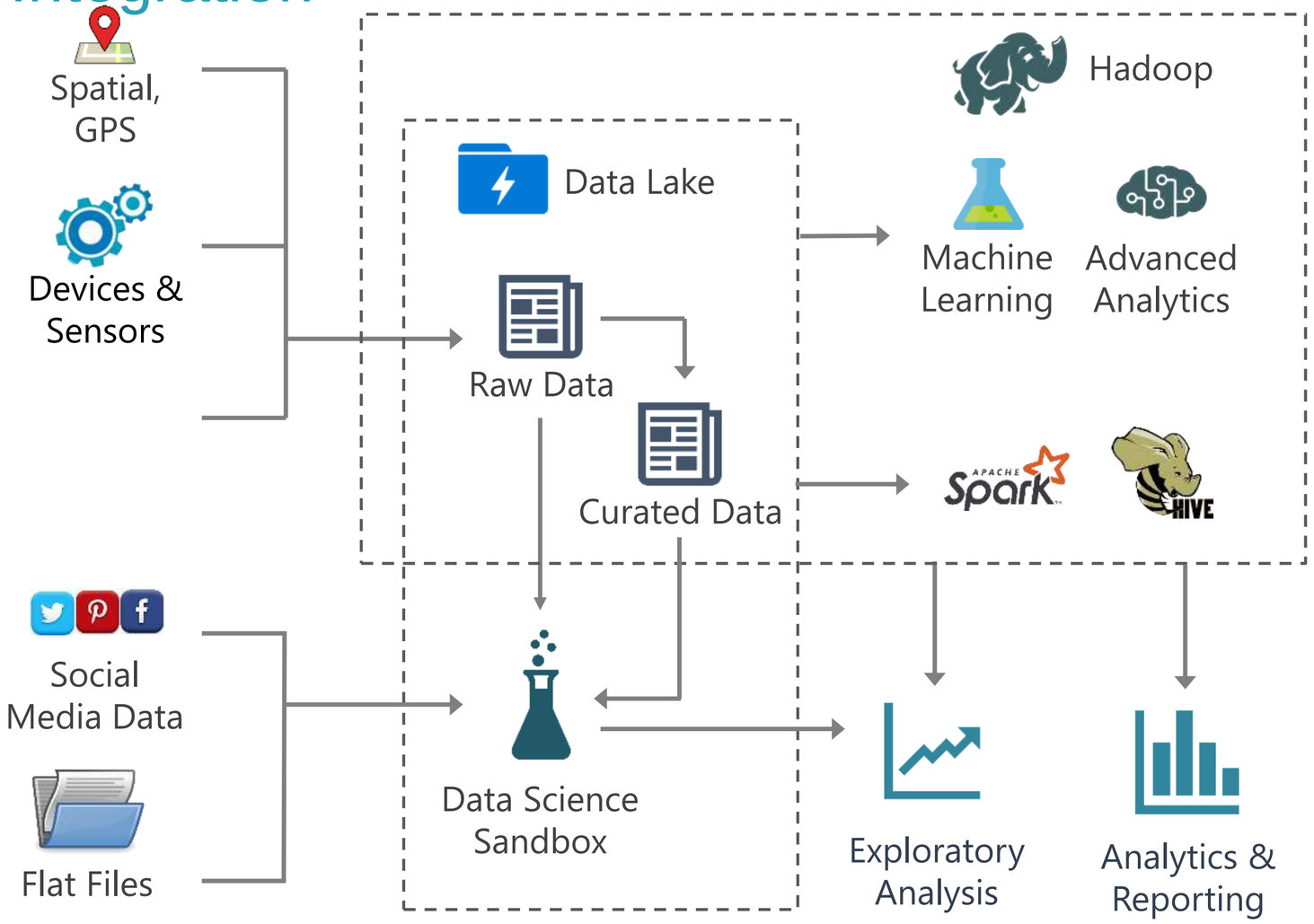


- ✓ Preparatory file storage for multi-structured data
- ✓ Exploratory analysis + POCs to determine value of new data types & sources
- ✓ Affords additional time for longer-term planning while accumulating data or handling an influx of data

Thanks to Melissa Coates, www.CoatesDataStrategies.com

Data Lake Use Cases

Data Science Experimentation | Hadoop Integration

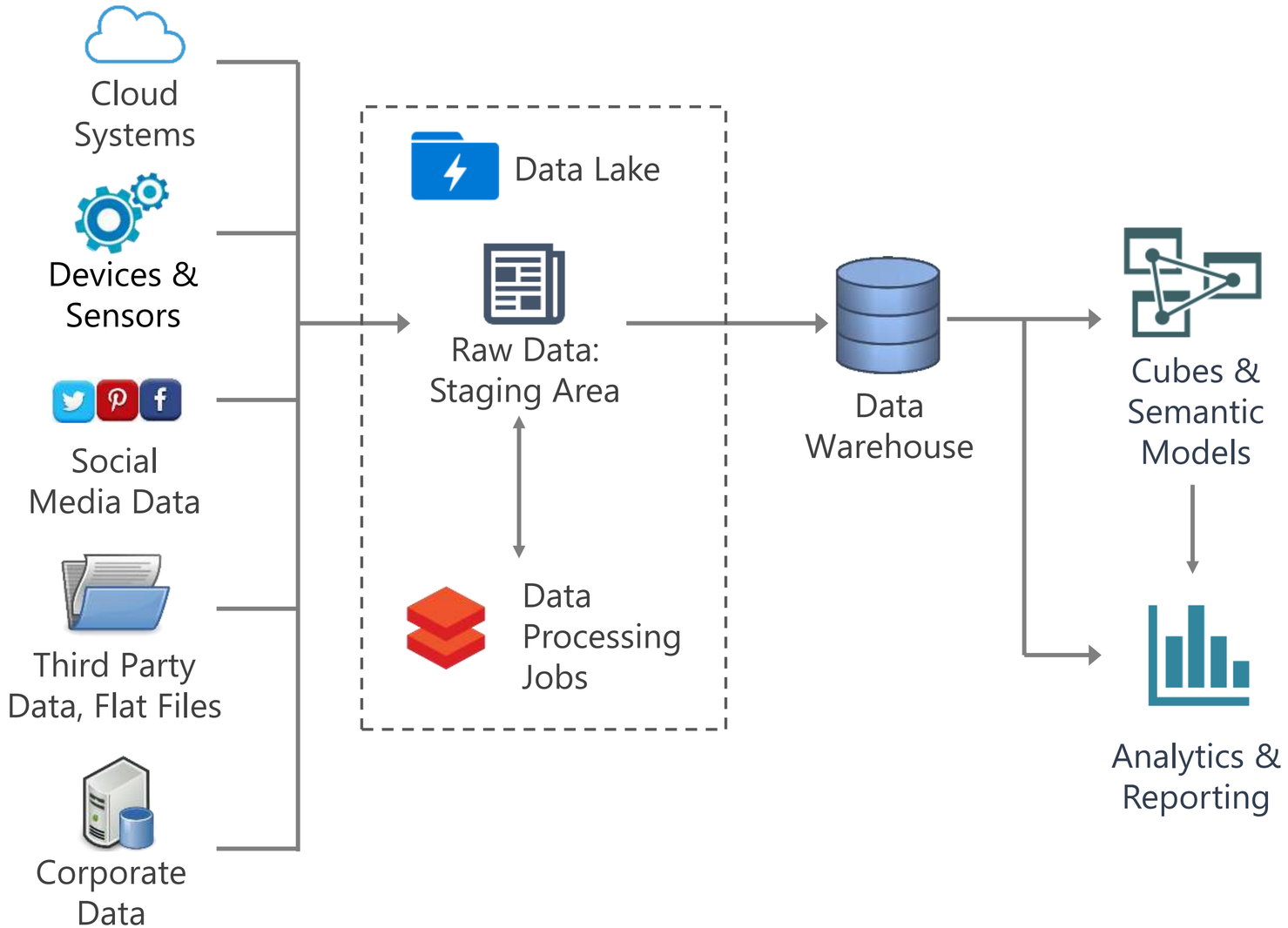


- ✓ Sandbox solutions for initial data prep, experimentation, and analysis
- ✓ Migrate from proof of concept to operationalized solution
- ✓ Integrate with open source projects such as Hive, Pig, Spark, Storm, etc.
- ✓ Big data clusters
- ✓ SQL-on-Hadoop solutions

Thanks to Melissa Coates, www.CoatesDataStrategies.com

Data Lake Use Cases

Data Warehouse Staging Area

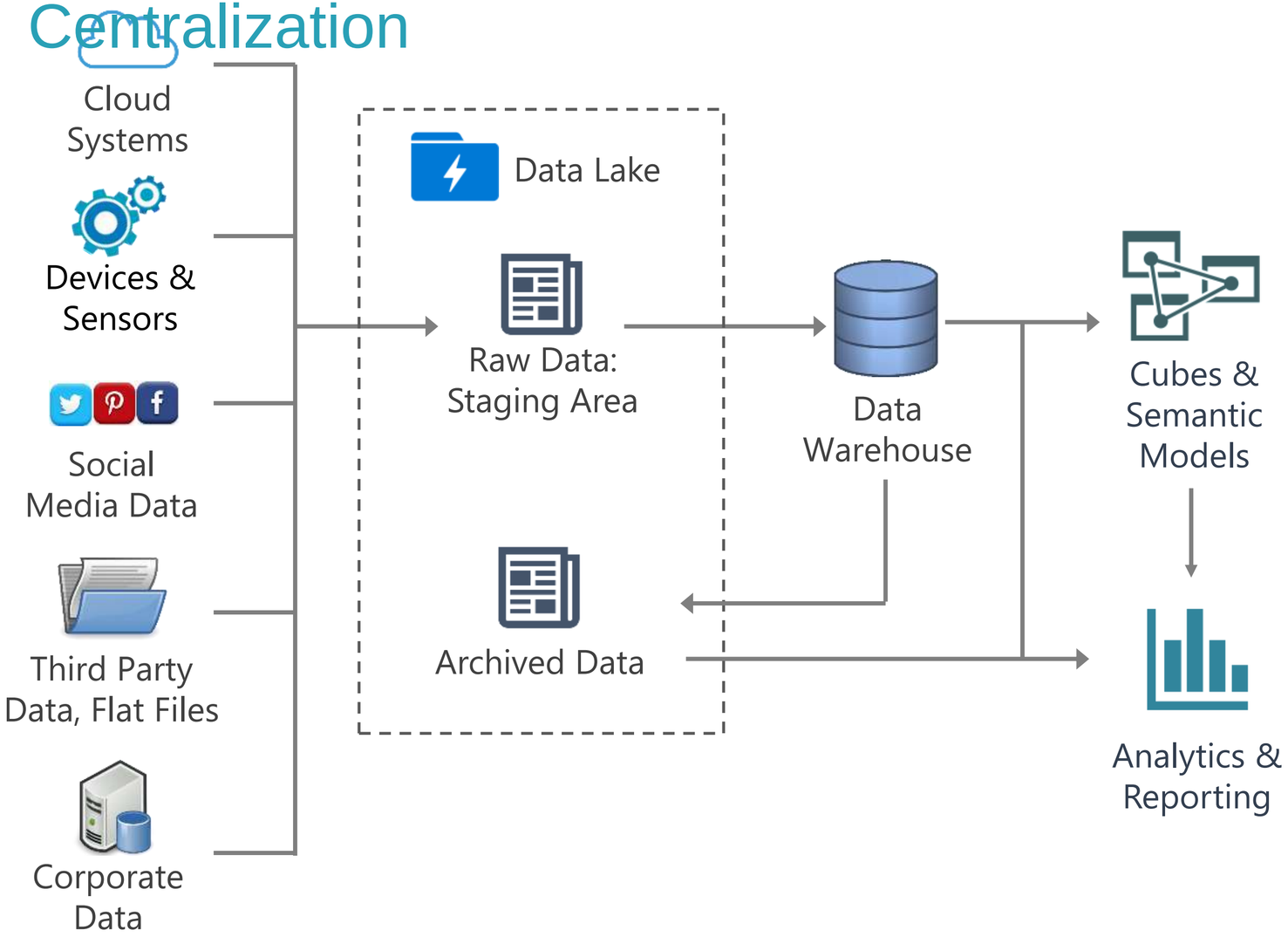


- ✓ ELT strategy
- ✓ Reduce storage needs in relational platform by using the data lake as landing area
- ✓ Practical use for data stored in the data lake
- ✓ Potentially also handle transformations in the data lake

Thanks to Melissa Coates, www.CoatesDataStrategies.com

Data Lake Use Cases

Integration with DW | Data Archival | Centralization

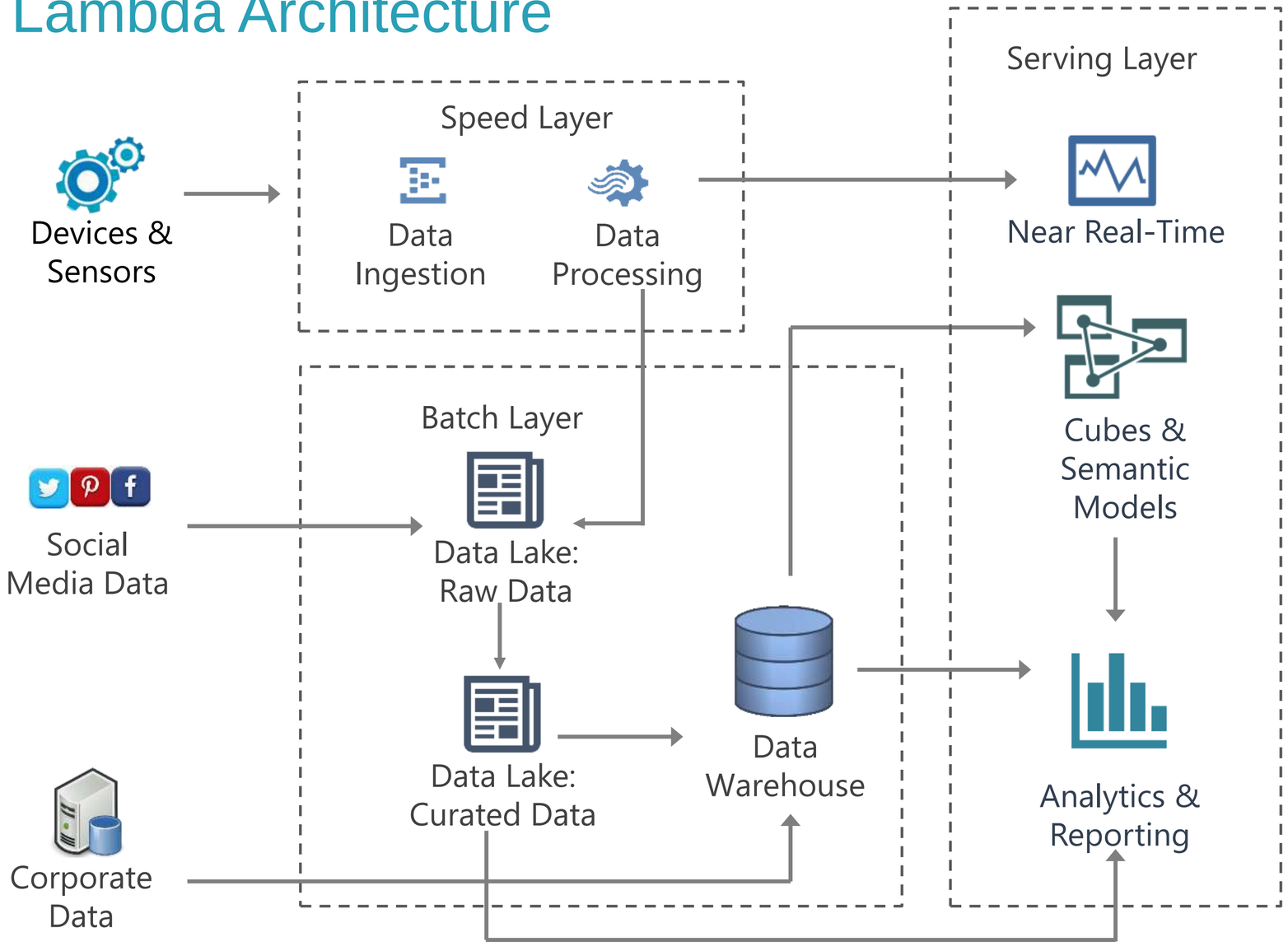


- ✓ Grow around existing DW
- ✓ Aged data available for querying when needed
- ✓ Complement to the DW via data virtualization
- ✓ Federated queries to access current data (relational DB) + archive (data lake)

Thanks to Melissa Coates, www.CoatesDataStrategies.com

Data Lake Use Cases

Lambda Architecture



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

Thanks to Melissa Coates, www.CoatesDataStrategies.com

Q & A

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