



Real Time Processing with Hadoop

Hortonworks. We do Hadoop.

Topics – Hortonworks and Apache Storm

- Hortonworks and Enterprise Hadoop
- Stream Processing with Apache Storm – Introduction
- Apache Storm – Key Constructs
- Apache Storm – Data Flows and Architecture
- Apache Storm Usage Cases and Real World Walkthrough
- Apache Storm Demo and How To



Hortonworks and Enterprise Hadoop

Hortonworks. We do Hadoop.



Our Mission:

Power your Modern Data Architecture
with HDP and Enterprise Apache Hadoop

Who we are

June 2011: Original 24 architects, developers, operators of Hadoop from Yahoo!

June 2014: An enterprise software company with 420+ Employees

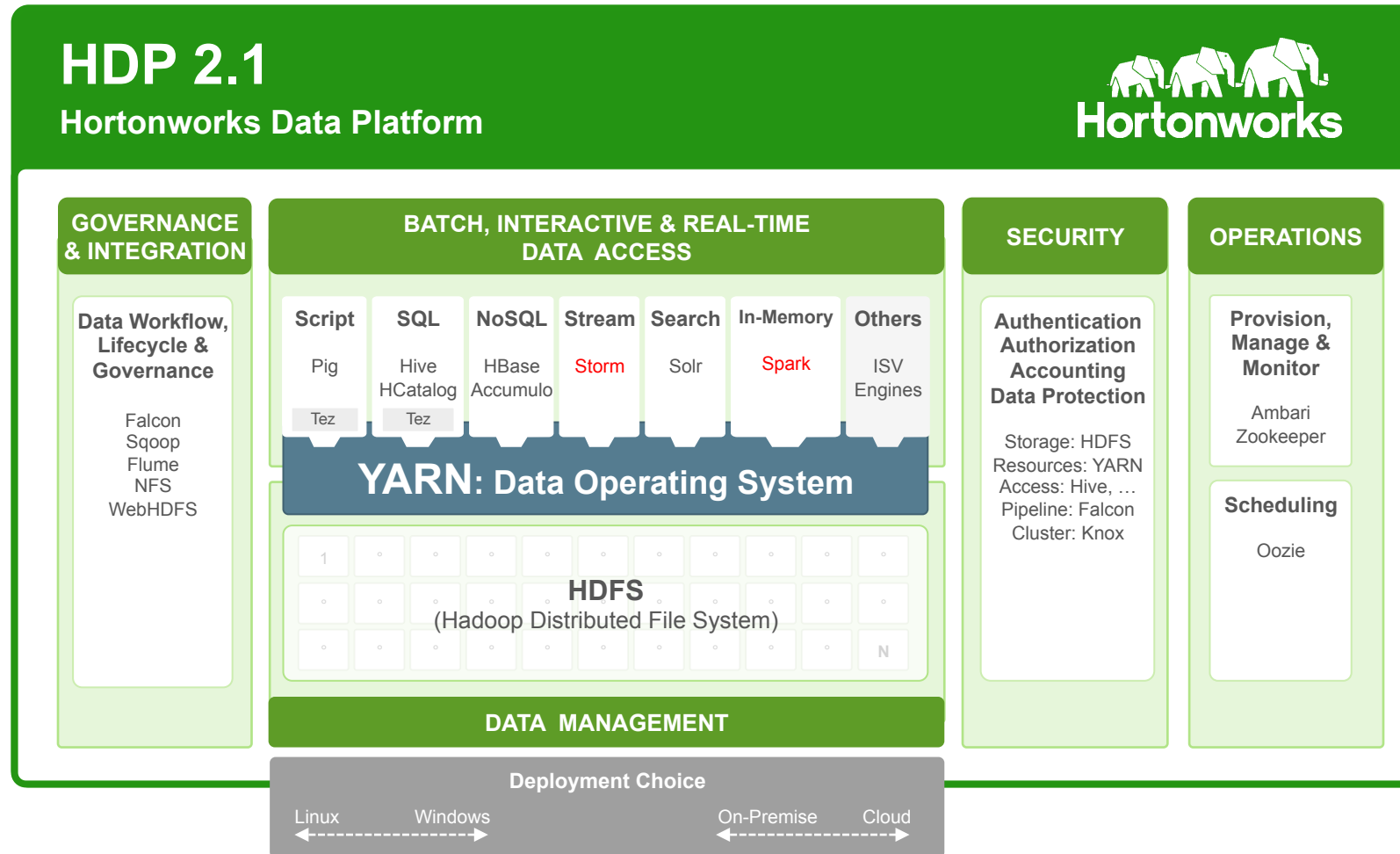
Our model

Innovate and deliver Apache Hadoop as a complete enterprise data platform
completely in the open, backed by a world class support organization

Key Partners

 Microsoft  TERADATA  SAP  hp  sas  CSC  accenture
High performance. Delivered. ...and hundreds more

HDP delivers a comprehensive data management platform



YARN is the architectural center of HDP

- Enables batch, interactive and real-time workloads
- Single SQL engine for both batch and interactive
- Enable existing ISV apps to plug directly into Hadoop via YARN

Provides comprehensive enterprise capabilities

- Governance
- Security
- Operations

The widest range of deployment options

- Linux & Windows
- On premise & cloud

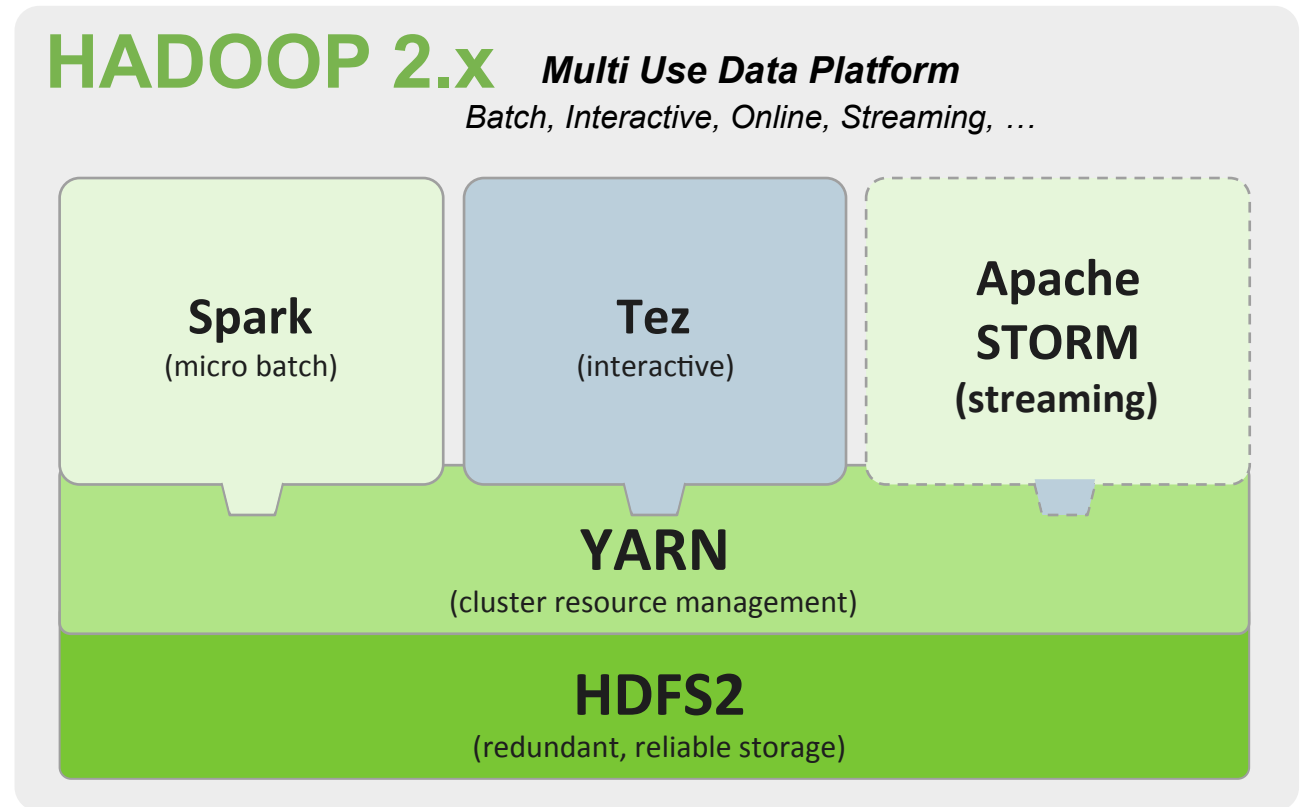
Choice Of Tool For Stream Processing

Storm

- Event by Event processing
- Suited for sub second real time processing
 - Fraud detection

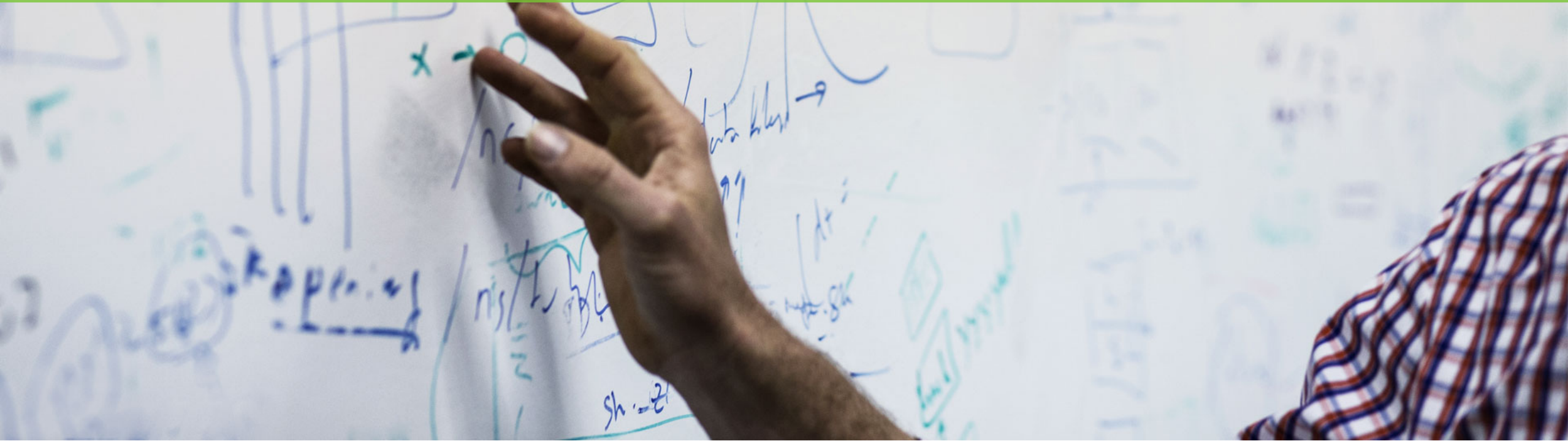
Spark Streaming

- Micro-Batch processing
- Suited for near real time processing
 - Trending Facebook likes





Stream Processing with Apache Storm



Why stream processing IN Hadoop?

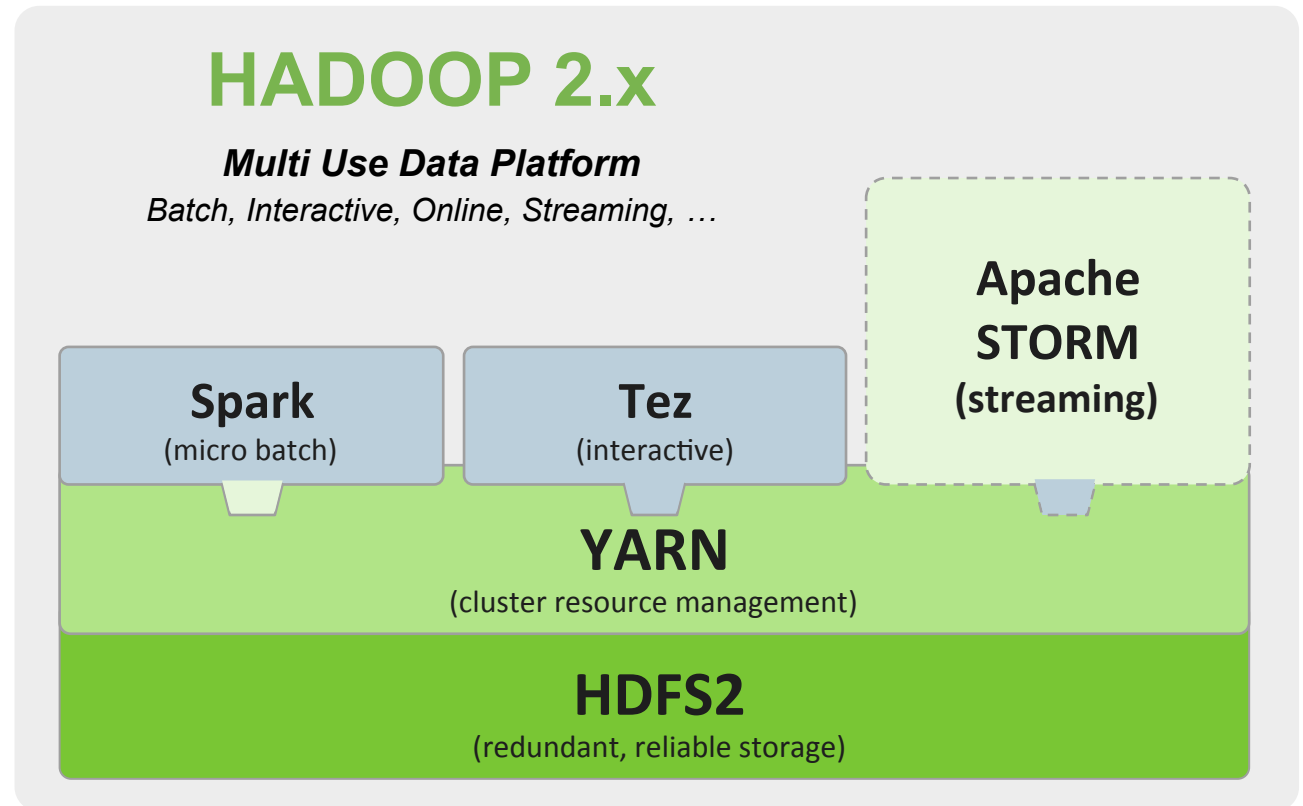
Stream processing has emerged as a key use case

What is the need?

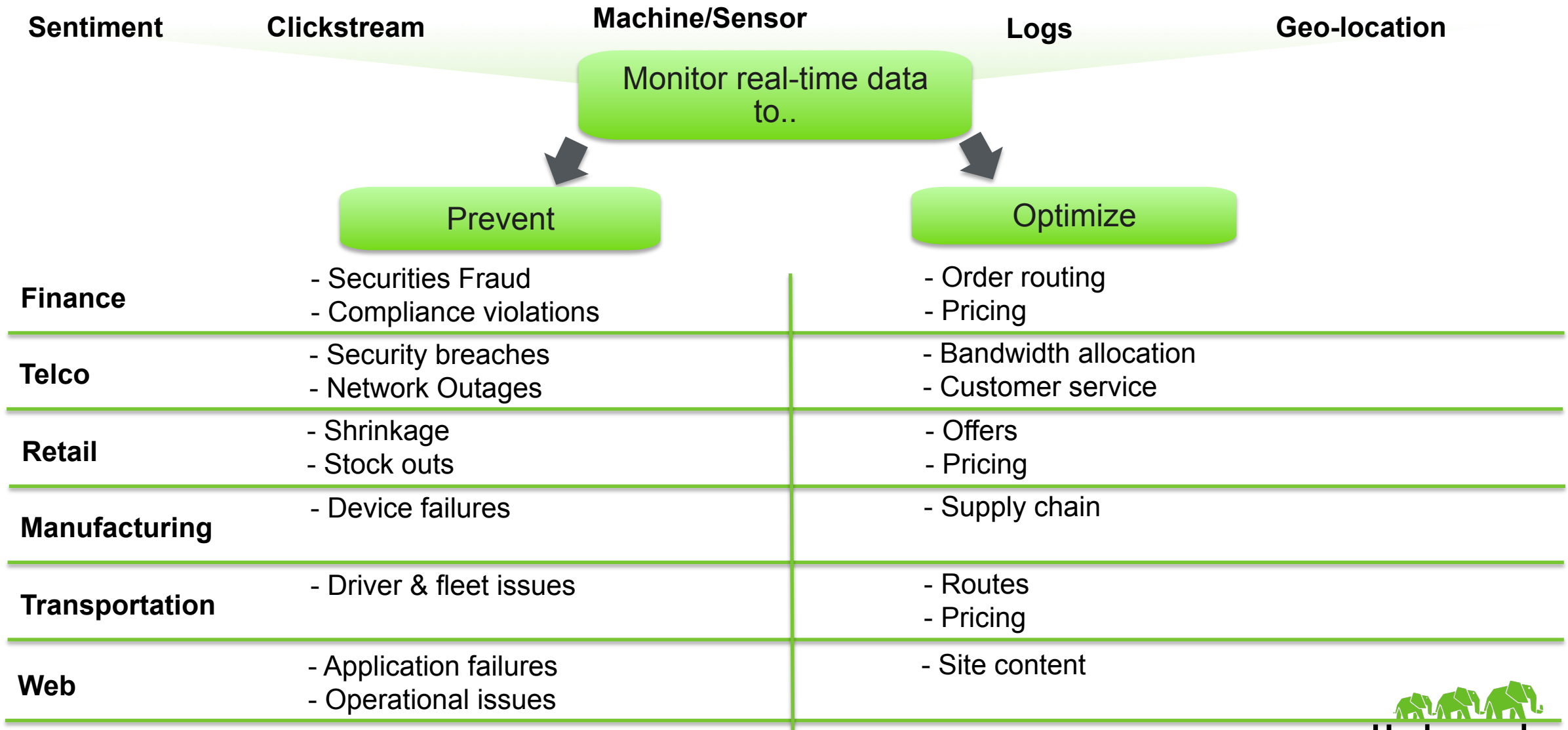
- Exponential rise in real-time data
- Ability to process real-time data opens new business opportunities

Why Now?

- Economics of Open source software & commodity hardware
- YARN allows multiple computing paradigms to co-exist in the data lake



Storm Use Cases – Business View



Stream processing very different from batch

Factors		Streaming	Batch
Data	Freshness	Real-time (usually < 15 min)	Historical – usually more than 15 min old
	Location	Primarily memory (moved to disk after processing)	Primarily in disk moved to memory for processing
Processing	Speed	Sub second to few seconds	Few minutes to hours
	Frequency	Always running	Sporadic to periodic
Clients	Who?	Automated systems only	Human & automated systems
	Type	Primarily operational systems	Primarily analytical applications

Key Requirements of a Streaming Solution

Data Ingest

- Extremely high ingest rates – millions of events/second

Processing

- Ability to easily plug different processing frameworks
- Guaranteed processing – at least once processing semantics

Persistence

- Ability to persist data to multiple relational and non- relational data stores

Operations

- Security, HA, fault tolerance & management support

Apache Storm Leading for Stream Processing

Open source, real-time event stream processing platform that provides fixed, continuous, & low latency processing for very high frequency streaming data

Highly scalable

- Horizontally scalable like Hadoop
- Eg: 10 node cluster can process 1M tuples per second

Fault-tolerant

- Automatically reassigns tasks on failed nodes

Guarantees processing

- Supports at least once & exactly once processing semantics

Language agnostic

- Processing logic can be defined in any language

Apache project

- Brand, governance & a large active community

Storm Use Cases – IT operations view

Continuous processing

- Continuously ingest high rate messages, process them and update data stores

High speed data aggregation

- Aggregate multiple data streams that emit data at extremely high rates into one central data store

Data filtering

- Filter out unwanted data on the fly before it is persisted to a data store

Distributed RPC response time reduction

- Extremely resource(CPU, mem or I/O) intensive processing that would take long time to process on a single machine can be parallelized with Storm to reduce response times to seconds



Apache Storm Key Constructs

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Key Constructs in Apache Storm

Basic Concepts of a Topology

Tuples and Streams, Sprouts, Bolts

Field Grouping

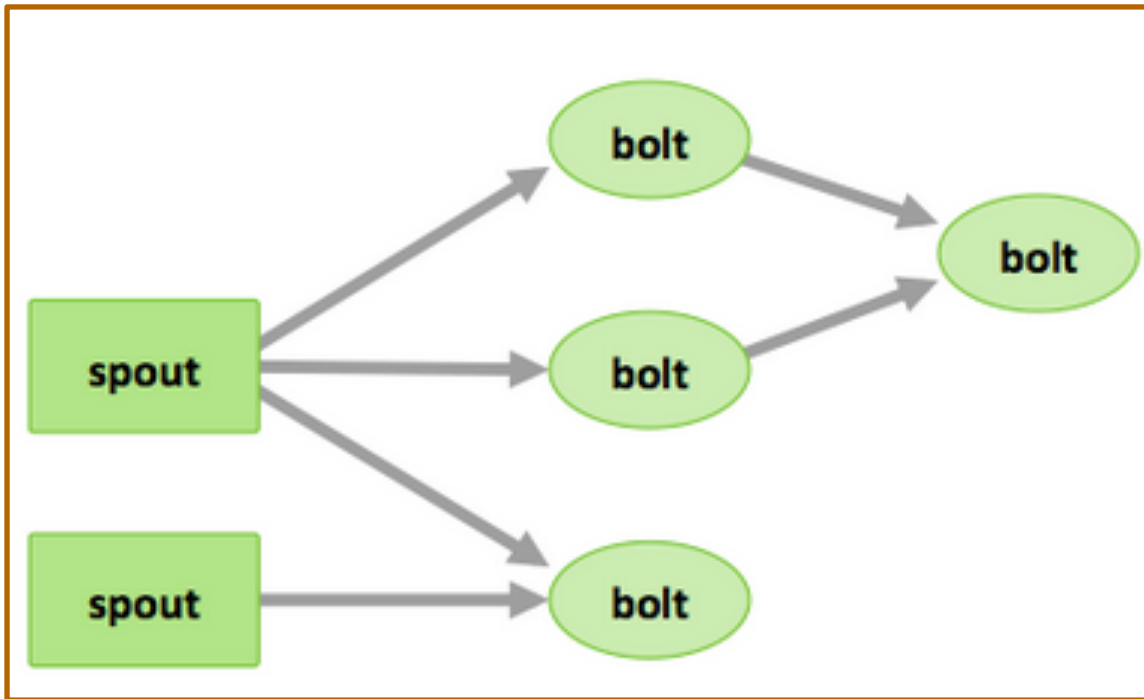
Architecture and Topology Submission

Parallelism

Processing Guarantee

Basic Concepts

Topology



Spouts: Generate streams.

Tuple: Most fundamental data structure and is a named list of values that can be of any datatype

Streams: Groups of tuples

Bolts: Contain data processing, persistence and alerting logic. Can also emit tuples for downstream bolts

Tuple Tree: First spout tuple and all the tuples that were emitted by the bolts that processed it

Topology: Group of spouts and bolts wired together into a workflow

Tuples and Streams

What is a Tuple?

- Fundamental data structure in Storm. Is a named list of values that can be of any data type.

```
new Values(driverId, truckId, eventTime, eventType, longitude, latitude, eventKey, correlationId);
```

- **What is a Stream?**

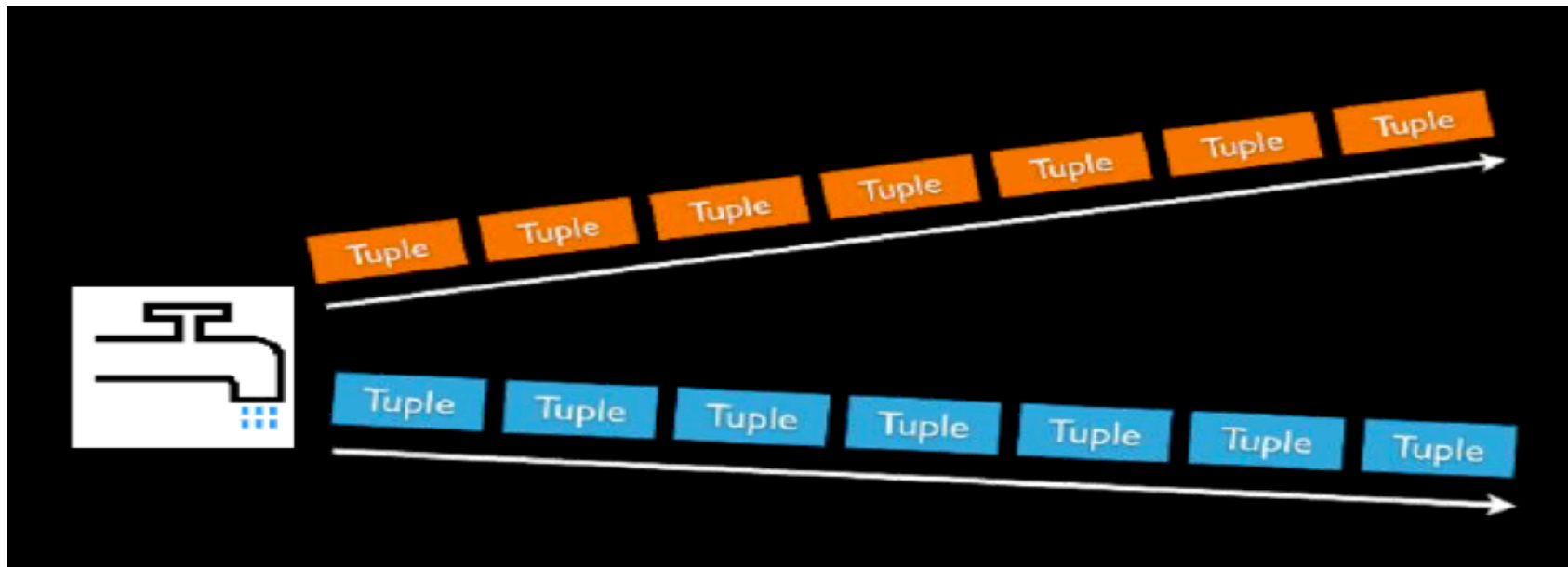
- An unbounded sequences of tuples.
- Core abstraction in Storm and are what you “process” in Storm



Spouts

What is a Spout?

- Generates or a source of Streams
 - E.g.: JMS, Twitter, Log, Kafka Spout
- Can spin up multiple instances of a Spout and dynamically adjust as needed



Bolts

What is a Bolt?

- Processes any number of input streams and produces output streams
- Common processing in bolts are functions, aggregations, joins, read/write to data stores, alerting logic
- Can spin up multiple instances of a Bolt and dynamically adjust as needed

Example of Bolts:

1. **HBaseBolt**: persist stream in Hbase
2. **HDFS Bolt**: persist stream into HFDS as Avro Files using Flume
3. **MonitoringBolt**: Read from Hbase and create alerts via email and messaging queues if given thresholds are exceeded.

Fields Grouping

Question: If multiple instances of single bolt can be created, when a tuple is emitted, how do you control what instance the tuple goes to?

Answer: Field Grouping

What is a Field grouping?

- Provides various ways to control tuple routing to bolts.
- Many field grouping exist include shuffle, fields, global..

Example of Field grouping in Use Case

- Events for a given driver/truck must be sent to the same monitoring bolt instance.

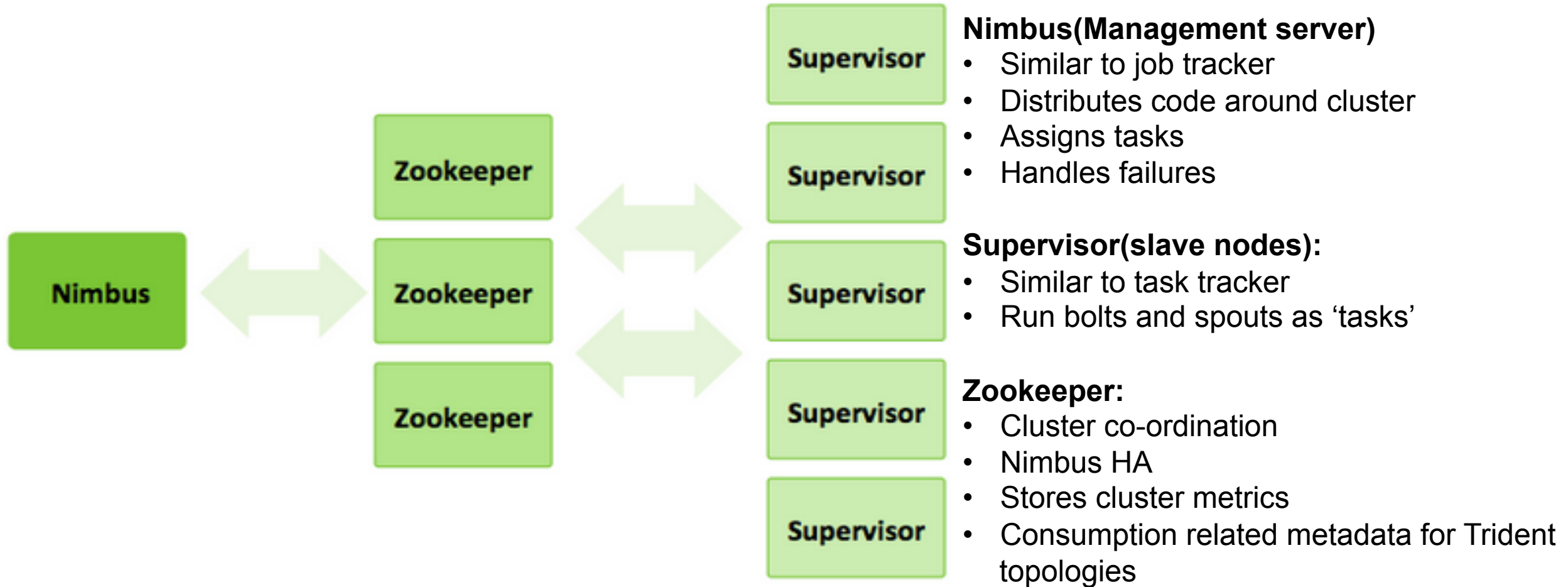
Different Fields Grouping

Grouping type	What it does	When to use
Shuffle Grouping	Sends tuple to a bolt in random round robin sequence	- Doing atomic operations eg.math operations
Fields Grouping	Sends tuples to a bolt based on or more field's in the tuple	<ul style="list-style-type: none">- Segmentation of the incoming stream- Counting tuples of a certain type
All grouping	Sends a single copy of each bolt to all instances of a receiving bolt	<ul style="list-style-type: none">- Send some signal to all bolts like clear cache or refresh state etc.- Send ticker tuple to signal bolts to save state etc.
Custom grouping	Implement your own field grouping so tuples are routed based on custom logic	- Used to get max flexibility to change processing sequence, logic etc. based on different factors like data types, load, seasonality etc.
Direct grouping	Source decides which bolt will receive tuple	- Depends
Global grouping	Global Grouping sends tuples generated by all instances of the source to a single target instance (specifically, the task with lowest ID)	- Global counts..

Data Processing Guarantees in Storm

Processing guarantee	How is it achieved?	Applicable use cases
At least once	Replay tuples on failure	<ul style="list-style-type: none">- Unordered idempotent operations- Sub second latency processing
Exactly once	Trident	<ul style="list-style-type: none">- Need ordered processing- Global counts- Context aware processing- Causality based- Latency not important

Architecture



Parallelism in a Storm Topology

Supervisor

- The supervisor listens for work assigned to its machine and starts and stops worker processes as necessary based on what Nimbus has assigned to it.

Worker (JVM Process)

- Is a JVM process that executes a subset of a topology
- Belongs to a specific topology and may run one or more executors (threads) for one or more components (spouts or bolts) of this topology.

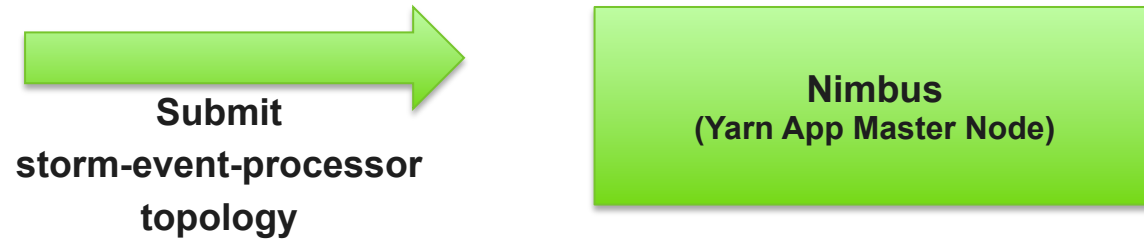
Executors (Thread in JVM Process)

- Is a thread that is spawned by a worker process and runs within the worker's JVM
- Runs one or more tasks for the same component (spout or bolt)

Tasks

- Performs the actual data processing and is run within its parent executor's thread of execution

Storm Components and Topology Submission



Nimbus (Management server)

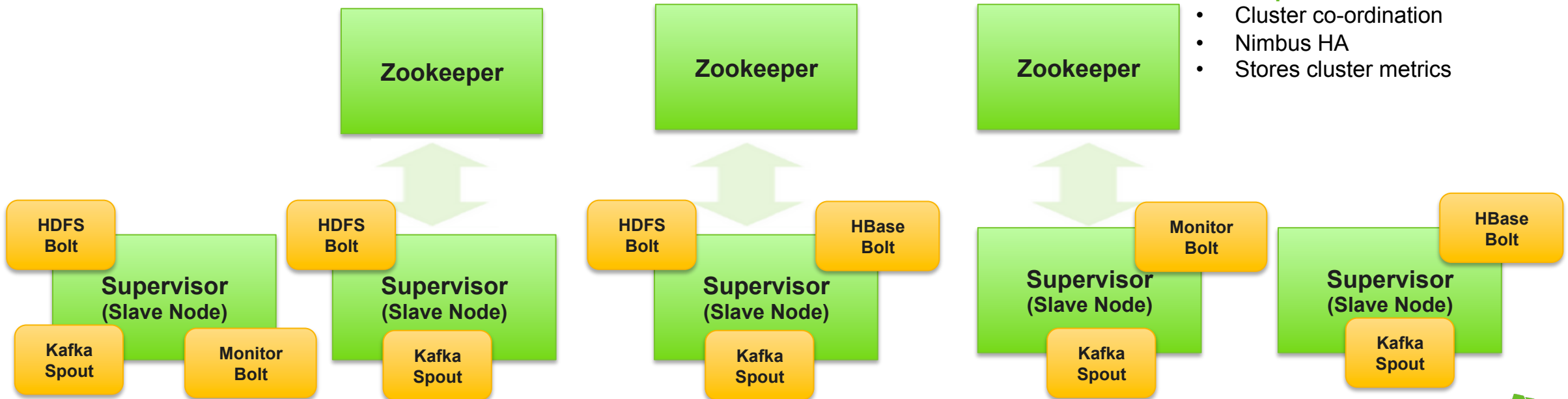
- Similar to job tracker
- Distributes code around cluster
- Assigns tasks
- Handles failures

Supervisor (Worker nodes)

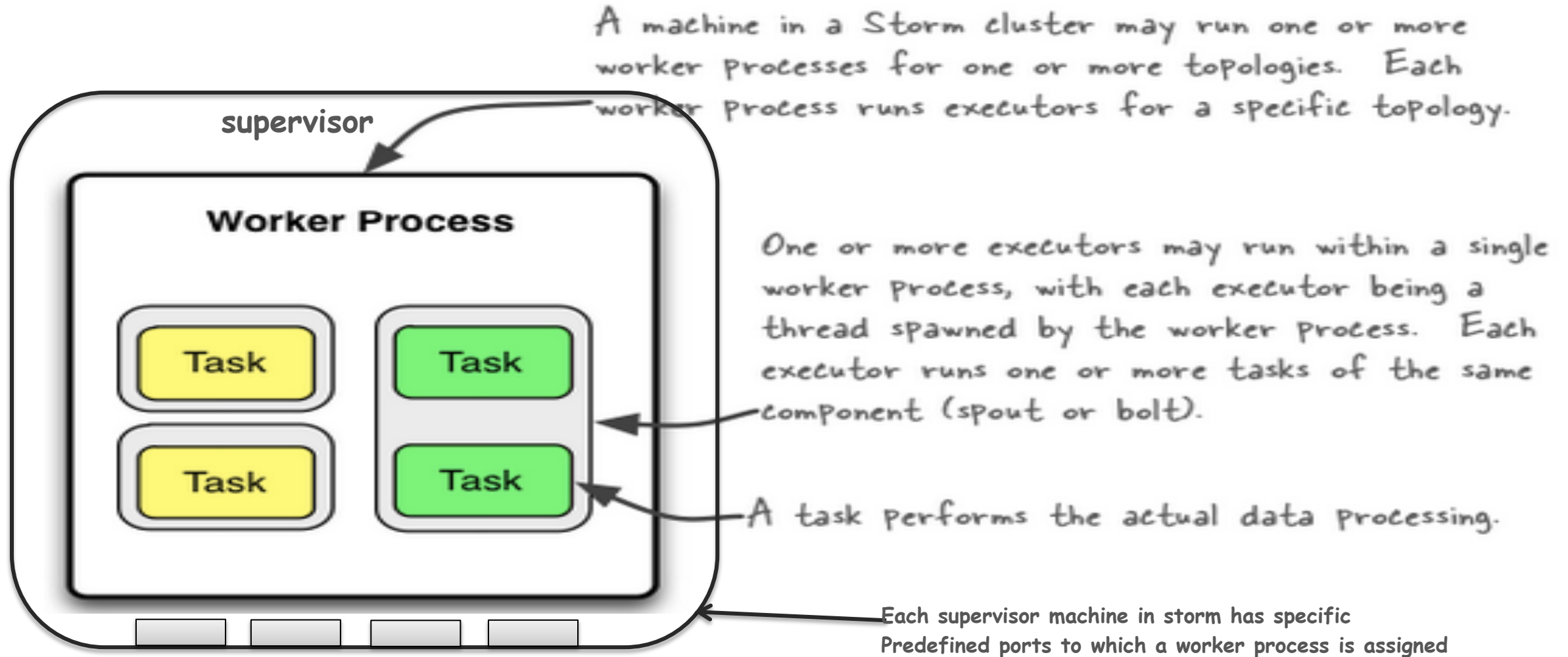
- Similar to task tracker
- Run bolts and spouts as 'tasks'

Zookeeper:

- Cluster co-ordination
- Nimbus HA
- Stores cluster metrics



Relationship Between Supervisors, Workers, Executors & Tasks



Parallelism Example

Example Configuration in Use Case:

- Supervisors = 4
- Workers = 8
- `spout.count` = 5
- `bolt.count` = 16

How many worker threads per supervisor?

- $8/4 = 2$

How many total executors(threads)?

- $16 + 5 = 21$

How many executors(threads) per worker?

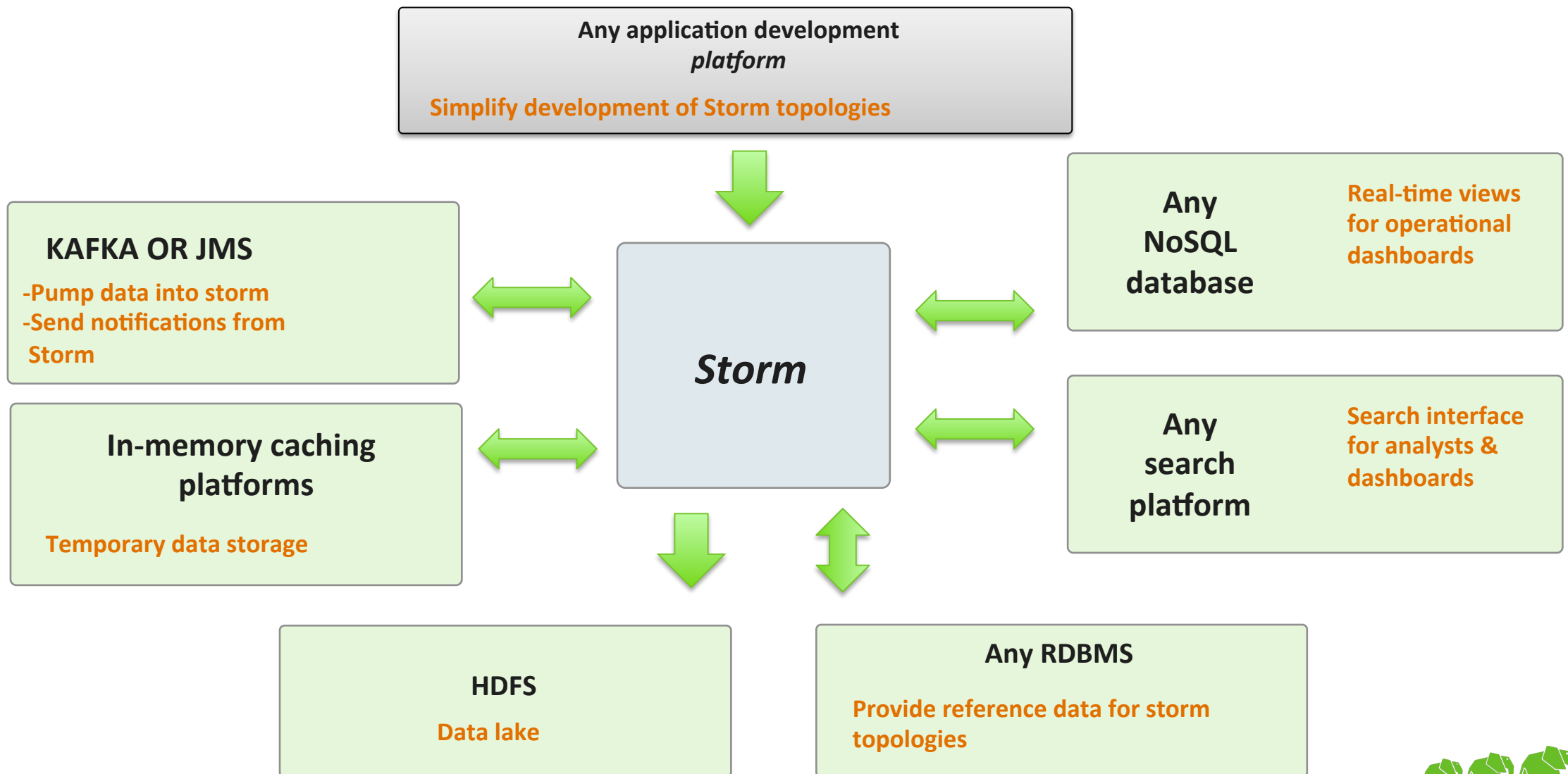
- $21/8 = 2 \text{ to } 3$



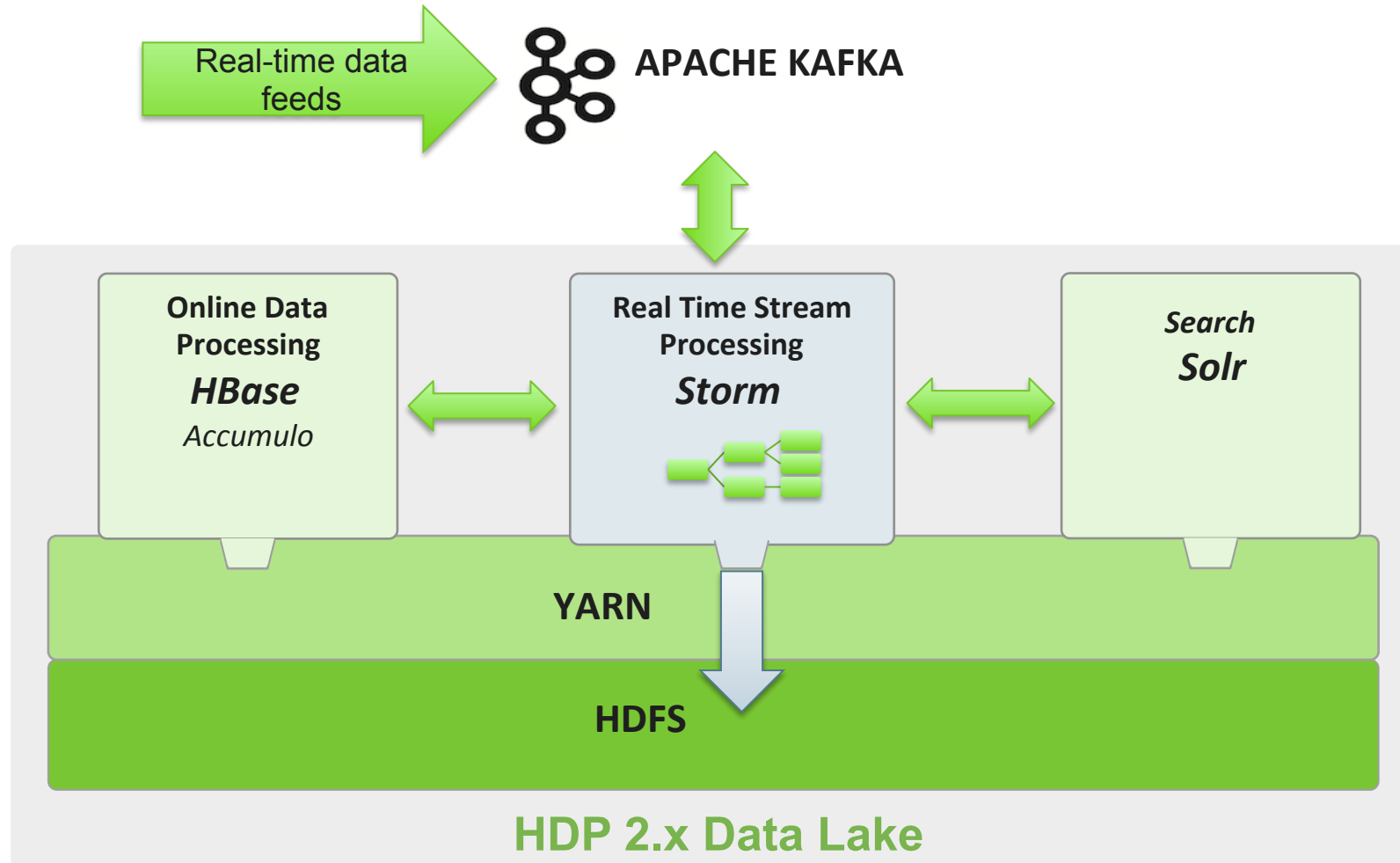
Apache Storm Data Flows and Architecture

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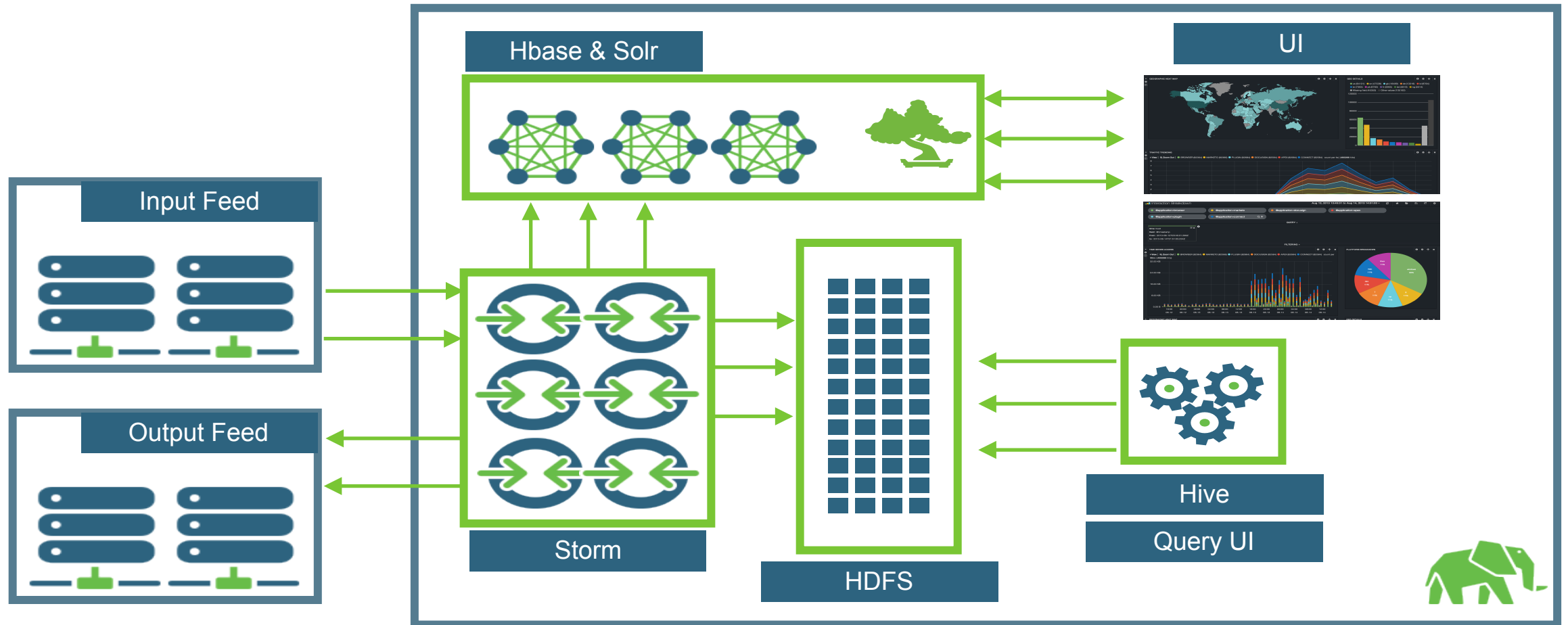
Flexible Architecture with Apache Storm



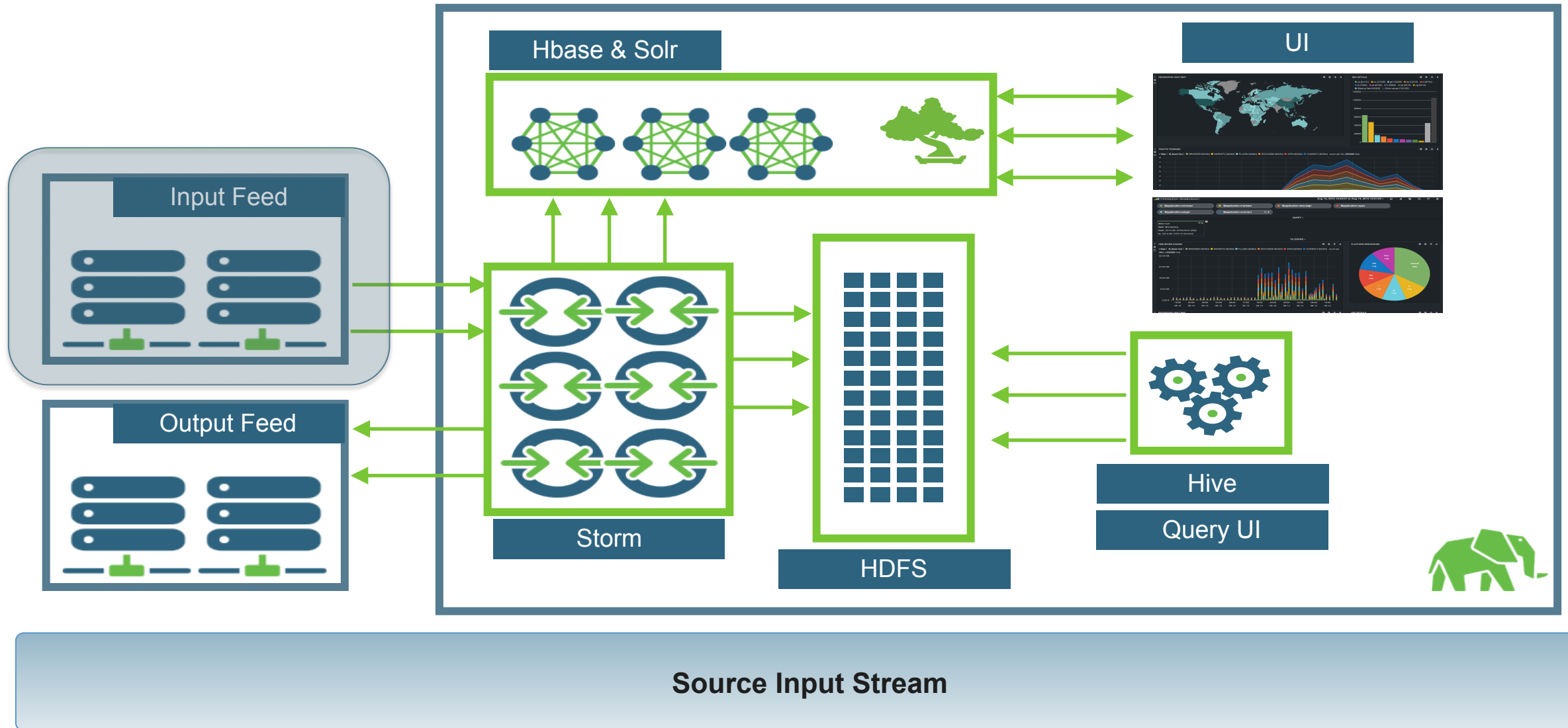
An Apache Storm Solution Architecture Example



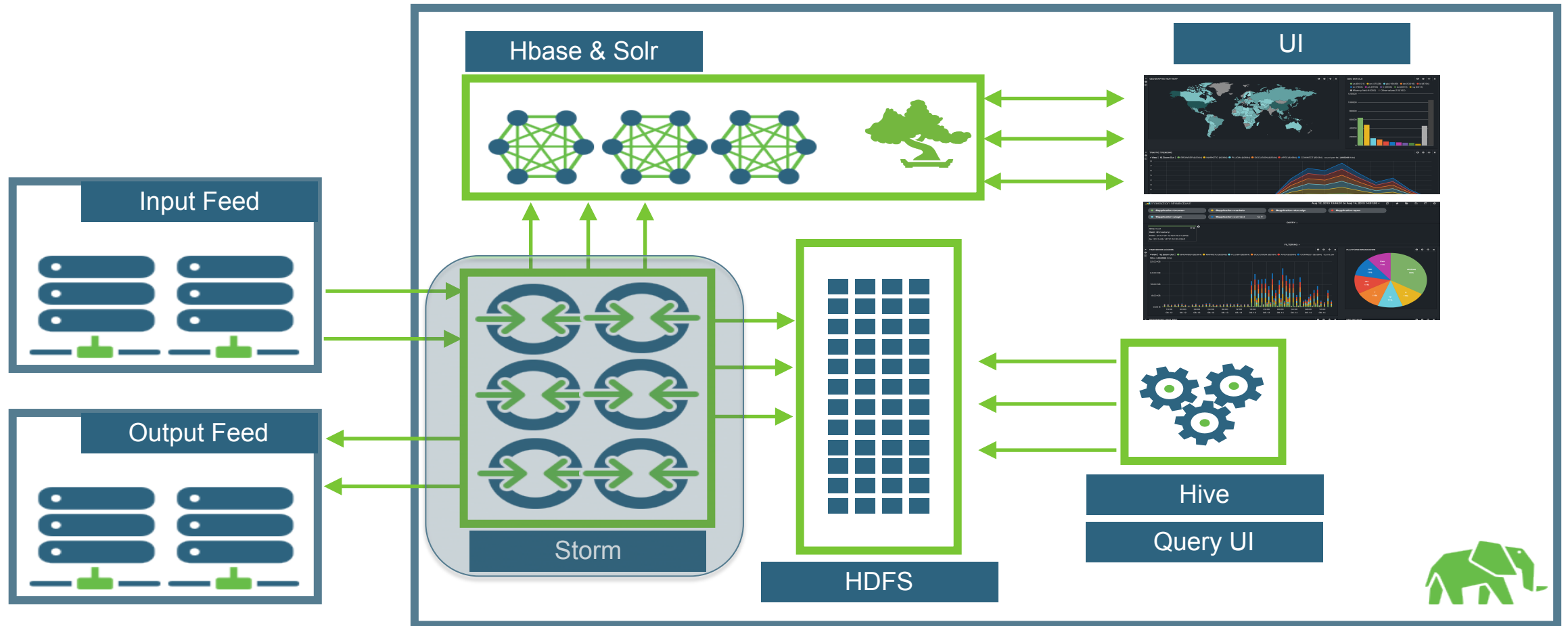
HPD Data Flow and Architecture



HDP Data Flow and Architecture

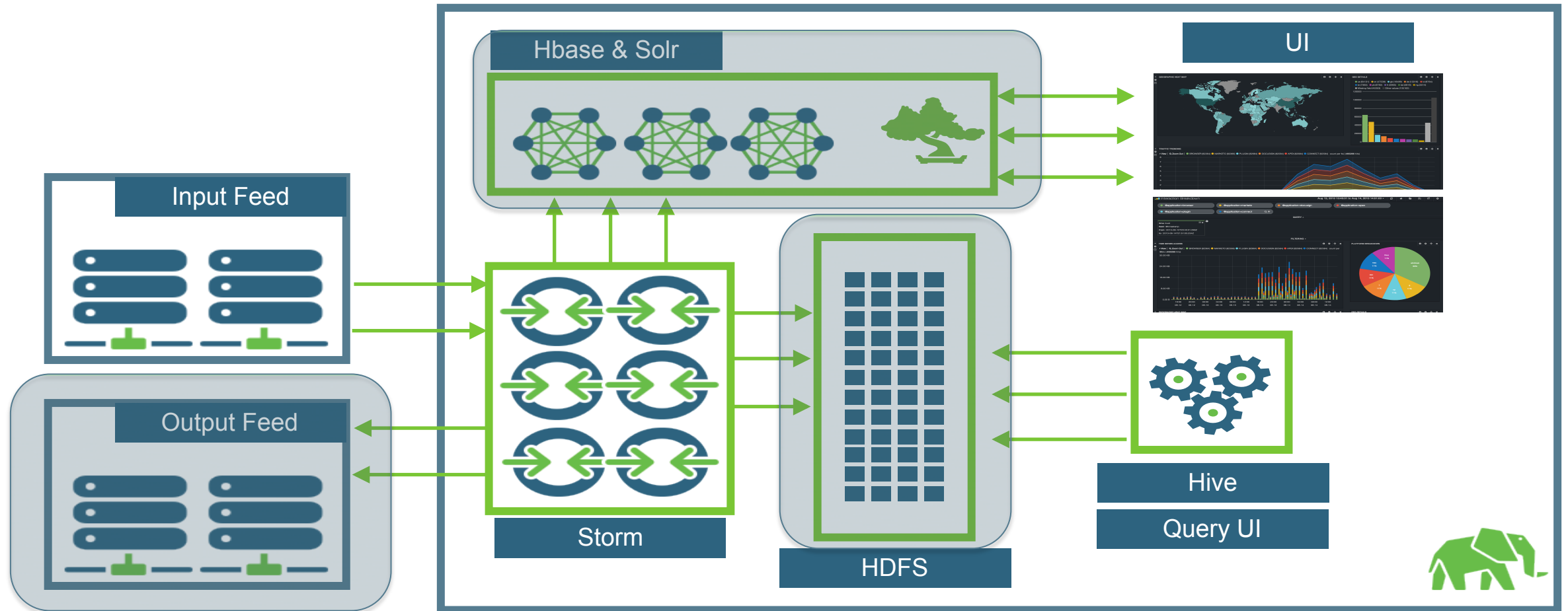


HDP Data Flow and Architecture



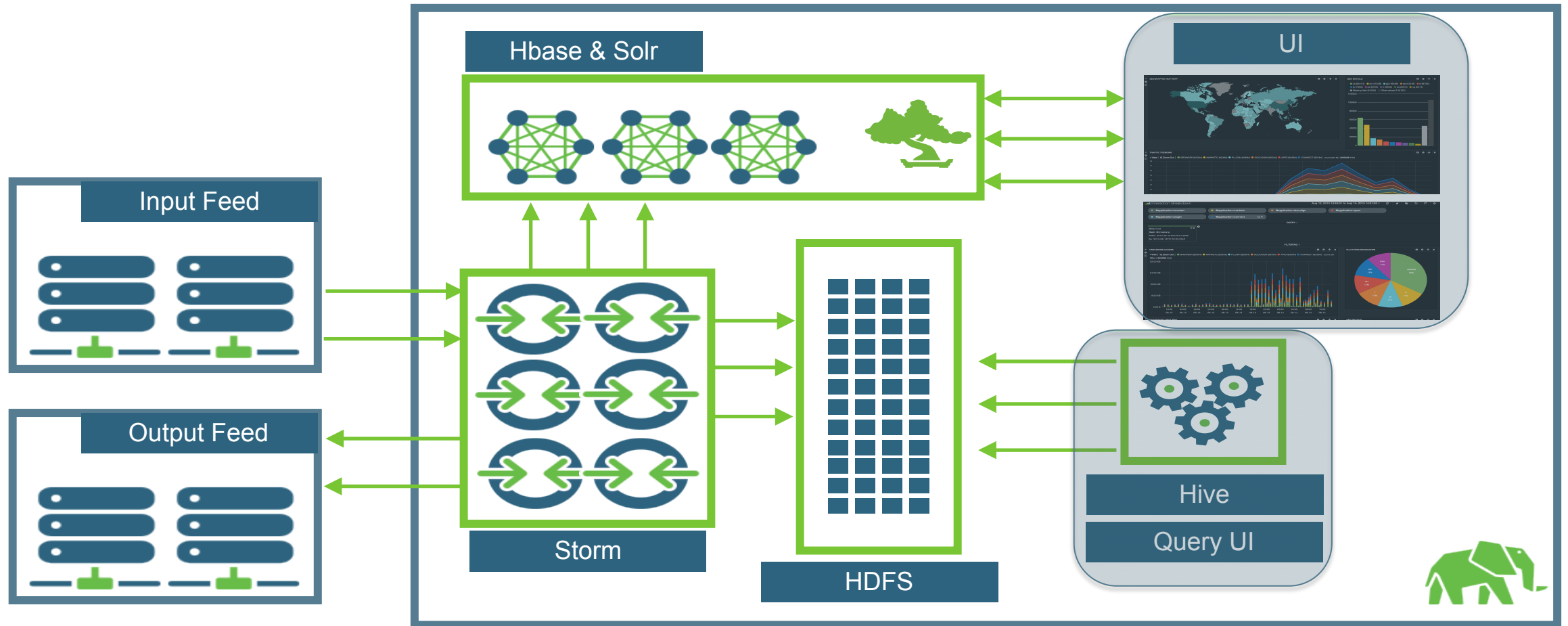
Ingest and Process Stream (Normalization/Micro ETL) in Real Time Using Storm

HDP Data Flow and Architecture



Publish to Output Feed (Available via RESTful API)
Push to Hbase & Solr and Sink to HDFS/HCatalog (Meta Model) for Ad-Hoc SQL

HDP Data Flow and Architecture



Data Available for Search and Query (RESTful API Calls an Option as Well)



Apache Storm – Code Walkthrough

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Automated Stock Insight

Monitor Tweet Volume For Early Detection of Stock Movement

- Monitor Twitter stream for S&P 500 companies to identify & act on unexpected increase in tweet volume

Solution components

- **Ingest:**

- Listen for Twitter streams related to S&P 500 companies

- **Processing**

- Monitor tweets for unexpected volume
- Volume thresholds managed in HBASE

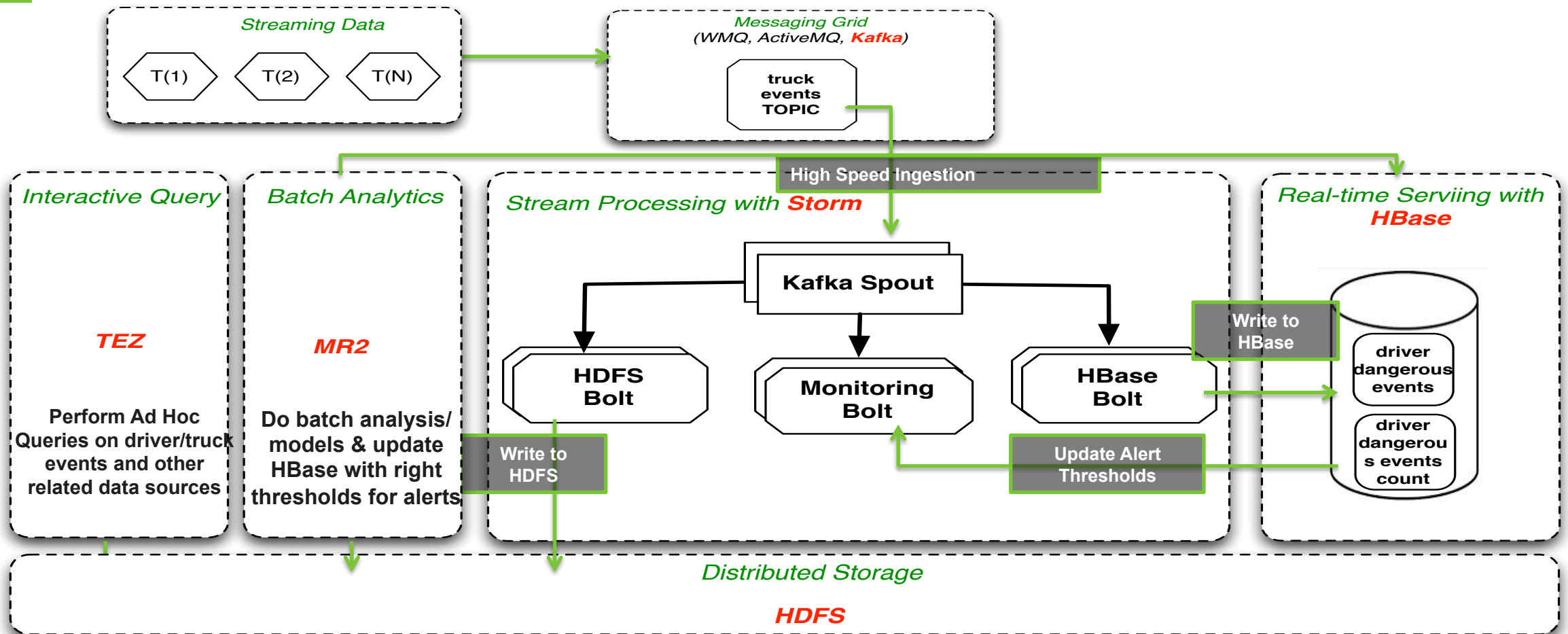
- **Persistence**

- HBase & HDFS

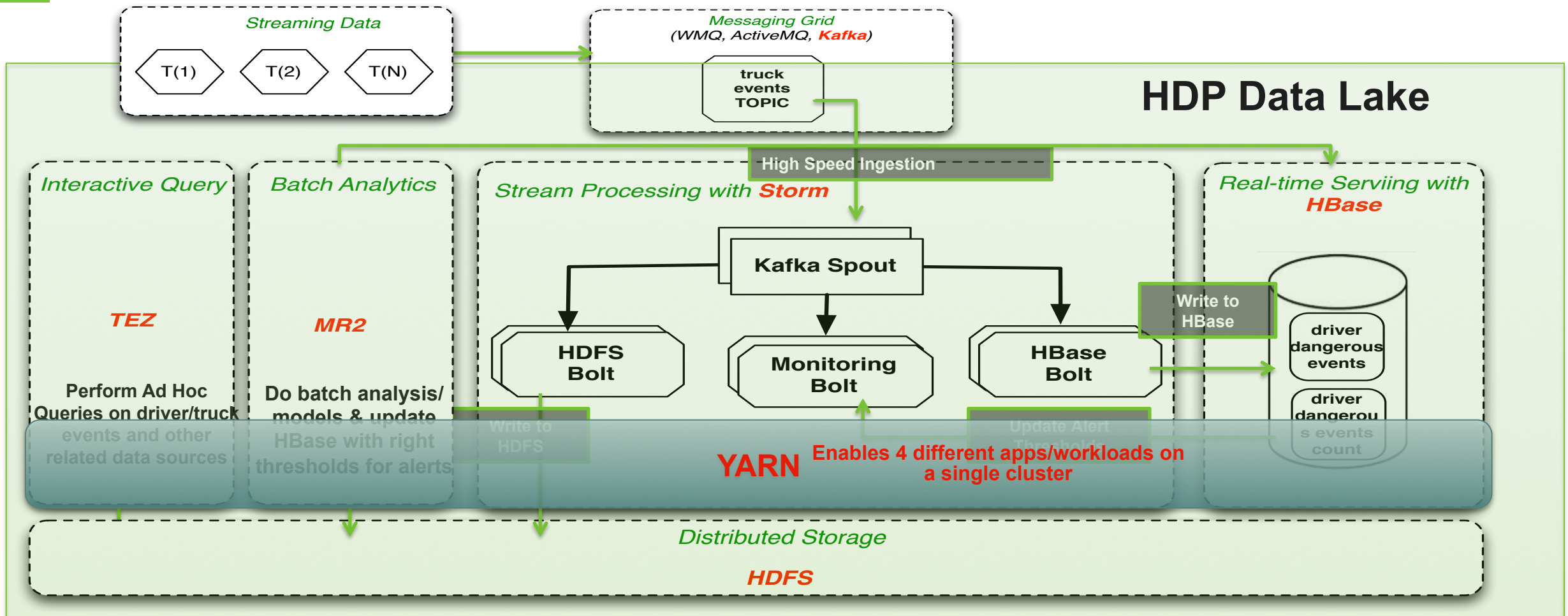
- **Refine**

- Update threshold values based on historical analysis of tweet volumes

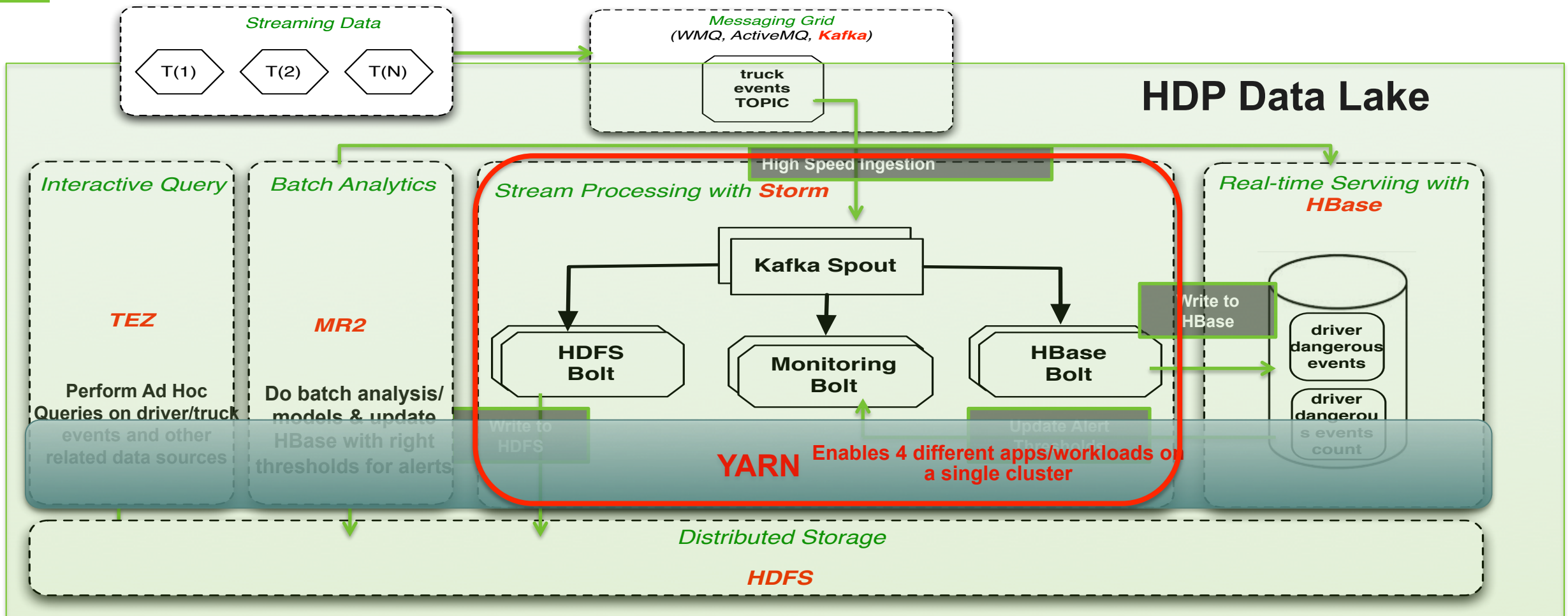
High Level Architecture



HDP Provides a Single Data Platform



HDP Provides a Single Data Platform





Q&A

