

A man in a dark suit is seen from the back, looking up towards a bright, futuristic sky. The sky is filled with blue and white light effects, including lens flares and a large, glowing text that reads "TIME FOR CHANGE". The overall scene has a high-tech, digital feel.

TIME FOR CHANGE

# Enterprise Data Warehouse Optimization

**Dr Barry Devlin**  
Founder & Principal  
9sight Consulting

**Piet Loubser**  
VP Product and Solutions Marketing  
Hortonworks



# **The EDW Lives On**

## **The Beating Heart of the Data Lake**

*10 August 2017*

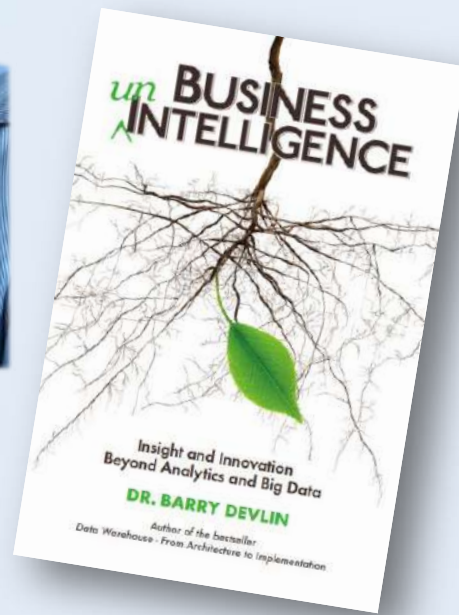
*Hortonworks Webinar*



**Dr Barry Devlin**

**Founder & Principal  
9sight Consulting**

# Dr. Barry Devlin



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**Founder and Principal**  
**9sight Consulting, [www.9sight.com](http://www.9sight.com)**

*Dr. Barry Devlin is a founder of the data warehousing industry, defining its first architecture in 1985. A foremost authority on business intelligence (BI), big data and beyond, he is respected worldwide as a visionary and thought-leader in the evolving industry. Barry has authored two ground-breaking books: the classic "Data Warehouse--from Architecture to Implementation" and "Business unIntelligence--Insight and Innovation Beyond Analytics and Big Data" ([http://bit.ly/Bunl\\_Book](http://bit.ly/Bunl_Book)) in 2013.*

*Barry has over 30 years of experience in the IT industry, previously with IBM, as a consultant, manager and distinguished engineer. As founder and principal of 9sight in 2008, Barry provides strategic consulting and thought-leadership to buyers and vendors of BI and Big Data solutions. He is an associate editor of TDWI's Journal of Business Intelligence, and a regular keynote speaker, teacher and writer on all aspects of information creation and use.*

*Barry operates worldwide from Cape Town, South Africa.*





# Agenda

1. Past – from a warehouse to a lake
2. Present – a warehouse *and* a lake
3. Emerging – a warehouse by a lake
4. Conclusions



# The data architecture since the mid-'80s

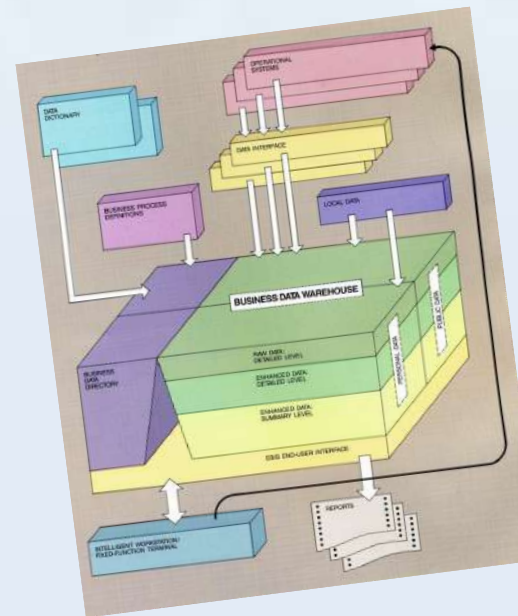
- Two layers within the Data Warehouse...

- Enterprise data warehouse
  - *Reconciled data*
- Data marts
  - *What the users need*

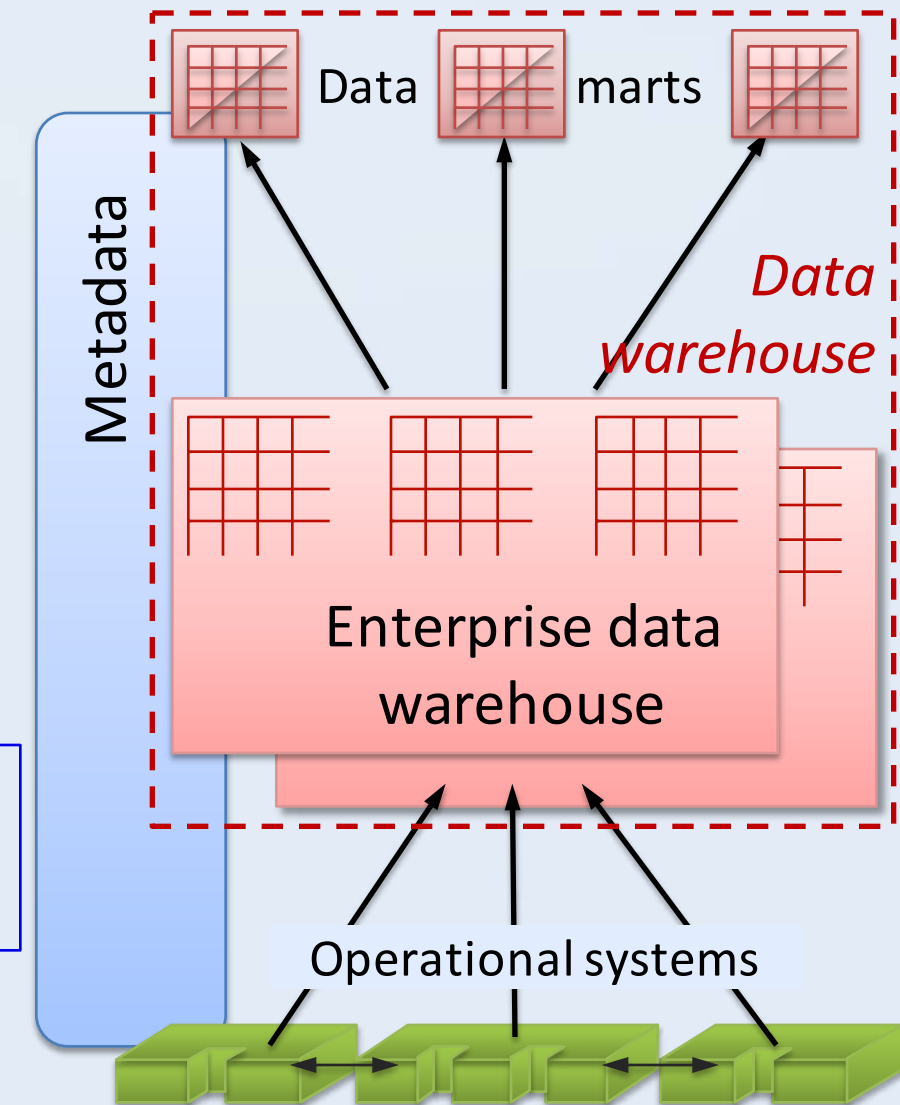
- ... fed from and separate to operational systems

- Data to run the business
- Created by the processes of the business

- All data created within the enterprise (or within partner ecosystem)



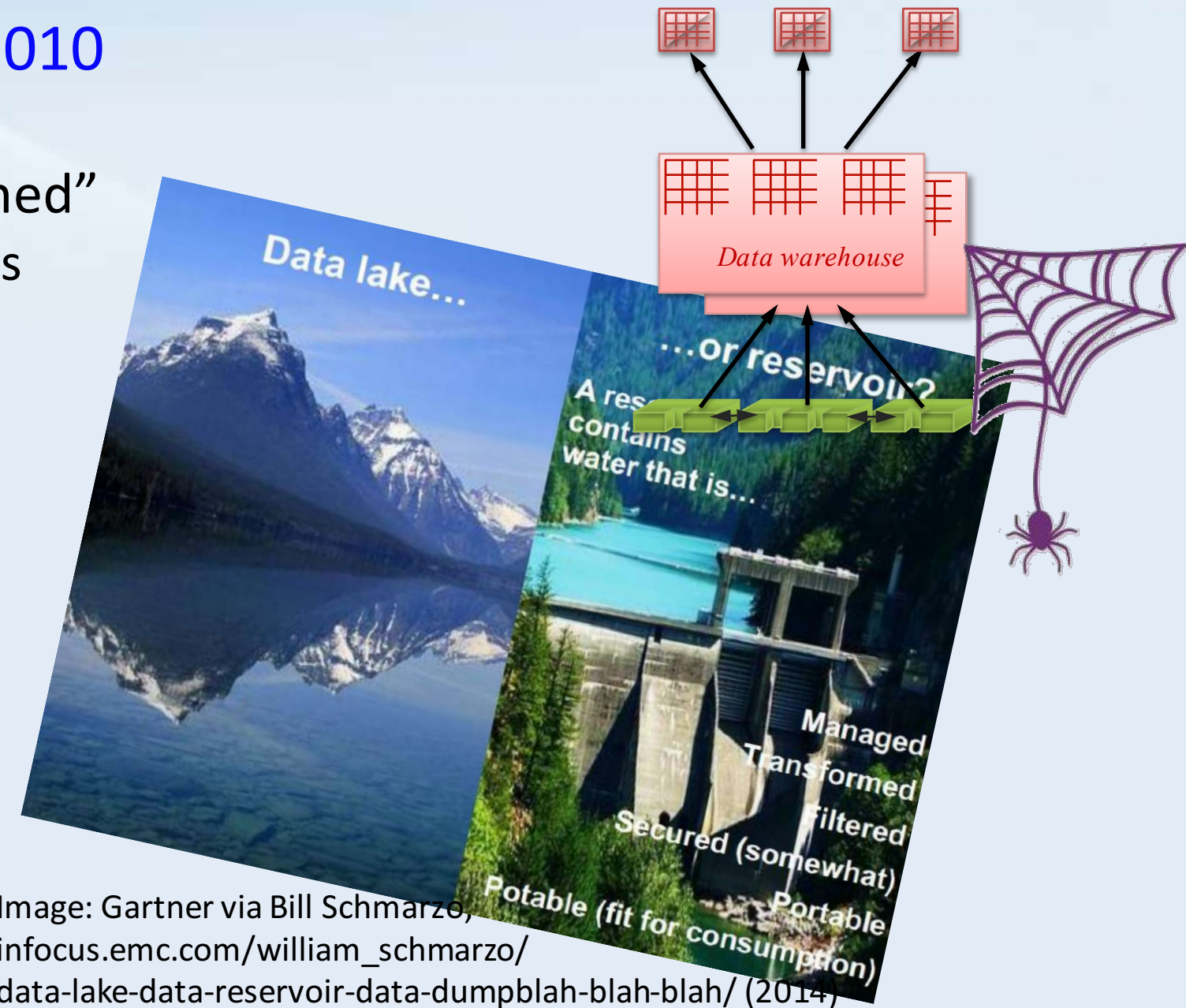
"An architecture for a business and information system",  
B. A. Devlin, P. T. Murphy,  
IBM Systems Journal, (1988)



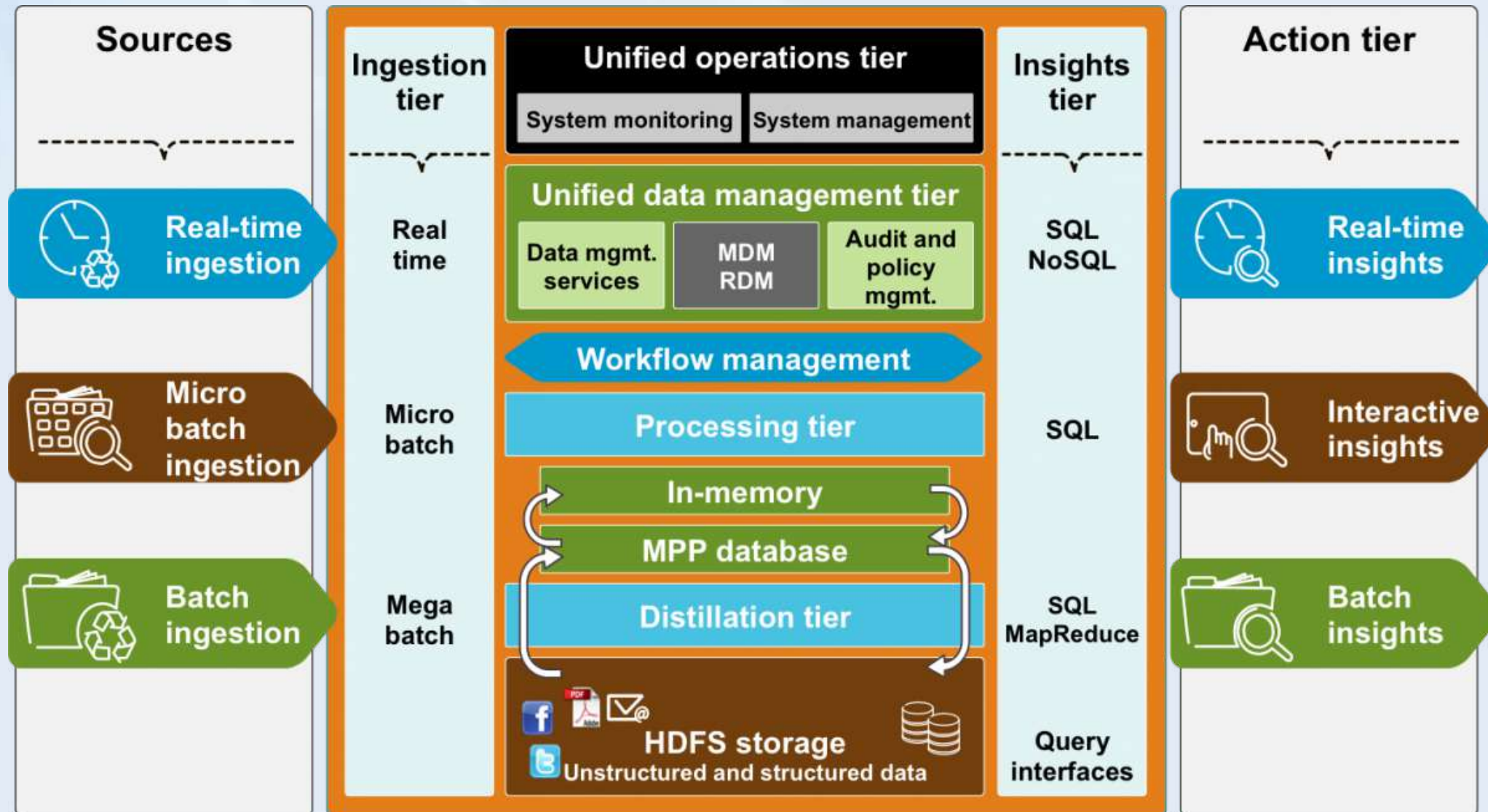


## The drive toward the data lake since 2010

- Data warehouse architecture “old-fashioned”
  - Linked to (traditional) relational databases
  - Too structured, schema-on-write
  - Too slow / complex to build
  - Lacking support for big data
  - No link to Hadoop
- Data lake proposed as alternative
  - Cheaper, bigger and more flexible
  - Structure-agnostic, schema-on-read (late binding)
  - Supports all data types
  - Agile, flexible, rapid implementation
  - Driven by Hadoop ecosystem
  - Data reservoir – a better(?) architected data lake



# Data lake architecture



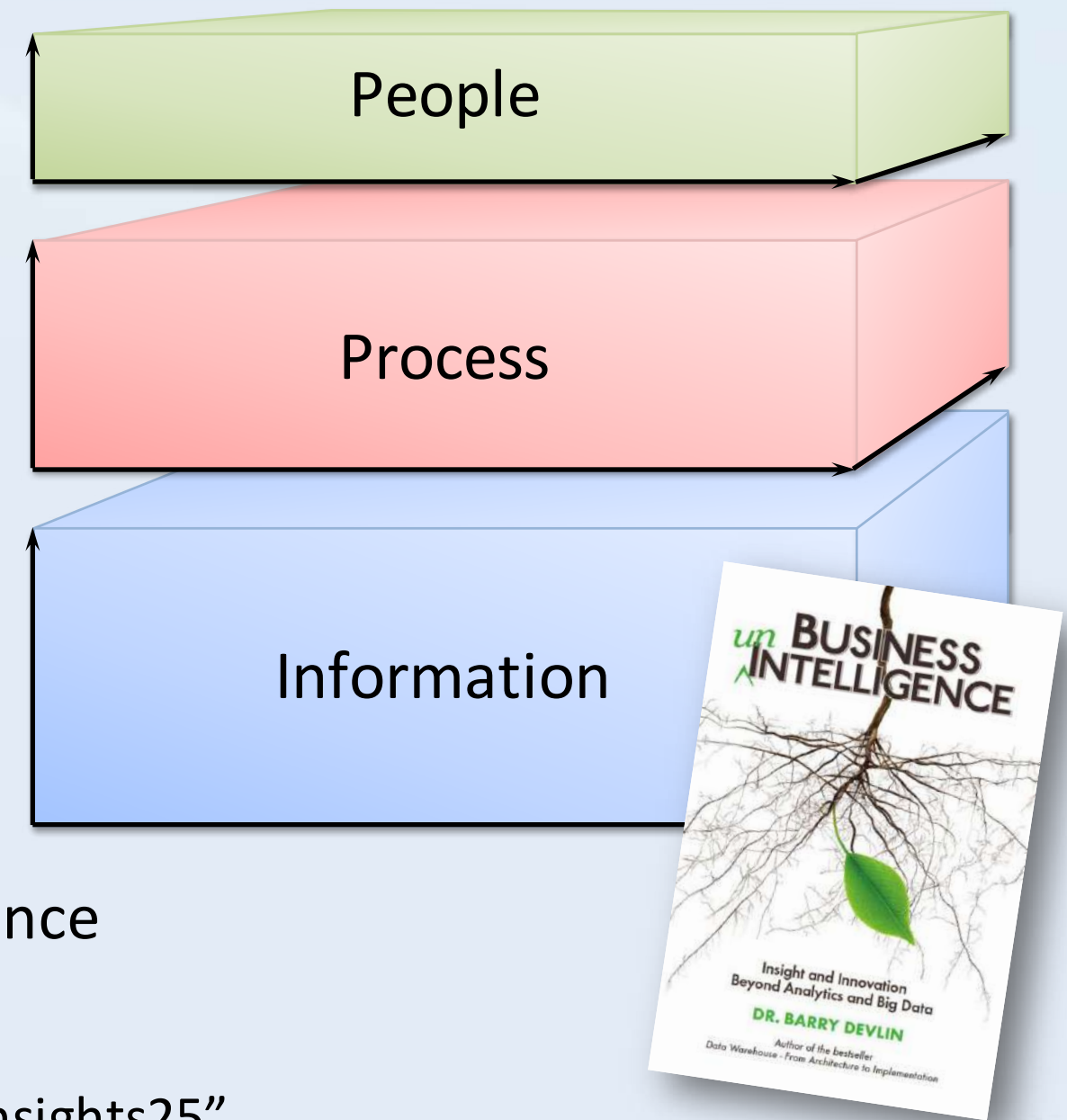
[www.capgemini.com/blog/capping-it-off/2014/08/you-have-to-manage-your-data-lake-the-fallacy-of-technology-being-magic](http://www.capgemini.com/blog/capping-it-off/2014/08/you-have-to-manage-your-data-lake-the-fallacy-of-technology-being-magic)





# From BI to Business unIntelligence

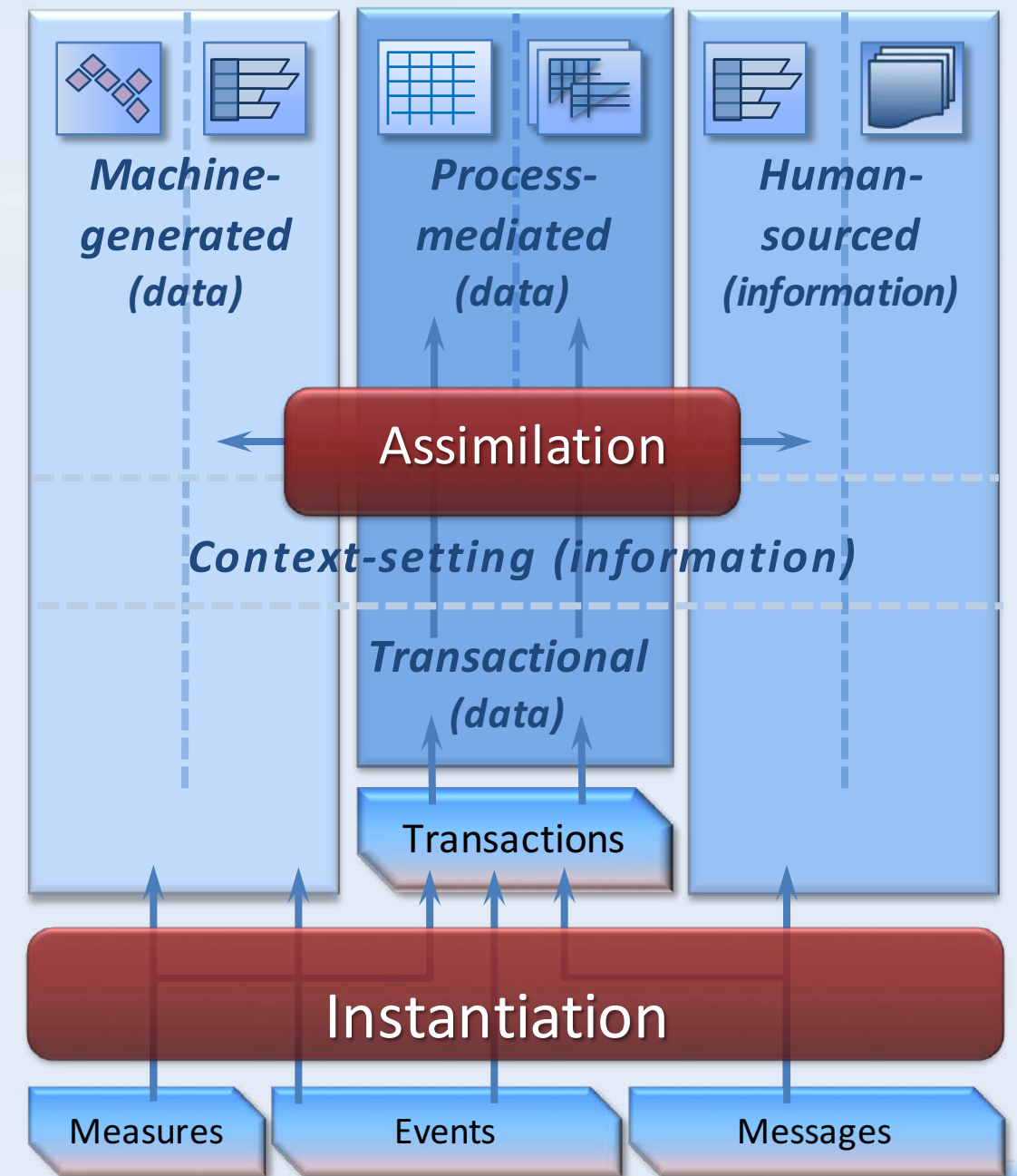
- People process information
- People: Rational thought and far beyond
  - People make all decisions!
- Process: Logic – predefined, emergent
  - Decision making is a process
- Information: Data, knowledge, meaning
  - Data/information is only the foundation
- Not business intelligence... Business unIntelligence
- Amazon: [http://bit.ly/Bunl\\_Book](http://bit.ly/Bunl_Book)
- Or <http://bit.ly/Bunl-TP2>: 25% discount with code “BIInsights25”





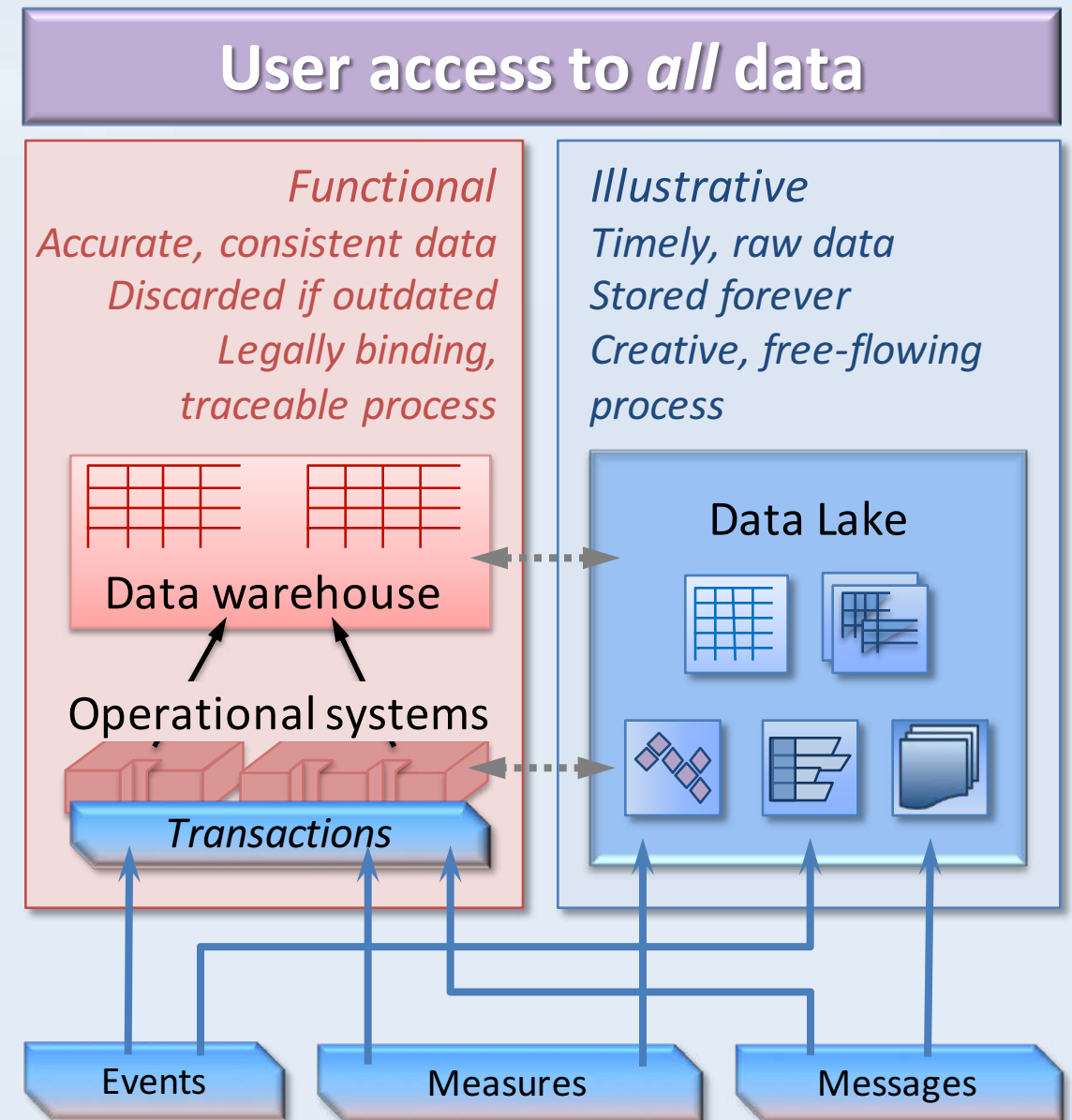
# Business unIntelligence – Information pillars

- One architecture for all types of information
  - Mix/match technology as needed
    - *Relational, NoSQL, Hadoop, etc.*
- Integration of sources and stores
  - Instantiation gathers inputs
  - Assimilation integrates stored info.
- Data flows as fast as needed and reconciled when necessary
  - No unnecessary storage or transformations
- Distinct data management / governance approaches as required



# Positioning of data lake and warehouse today

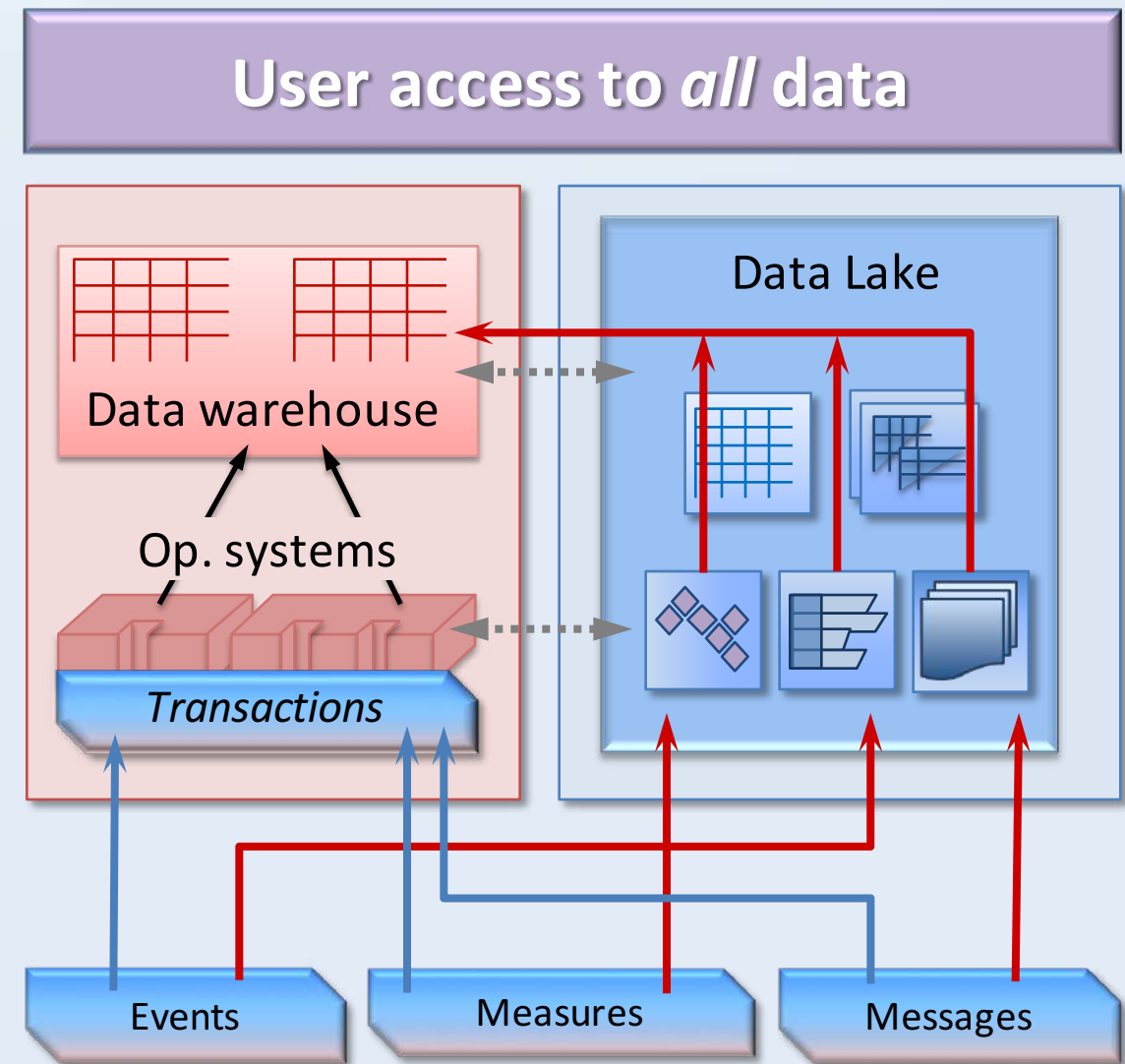
- Serve different purposes
  - Functional – run / manage the business
  - Illustrative – predict / influence the future
- Both required
  - Optimized for different strengths
  - Warehouse = accuracy and consistency
  - Lake = timeliness and rawness
- Links between environments
  - Better than copying everything into one (or both)
- Together – foundation for pervasive analytics





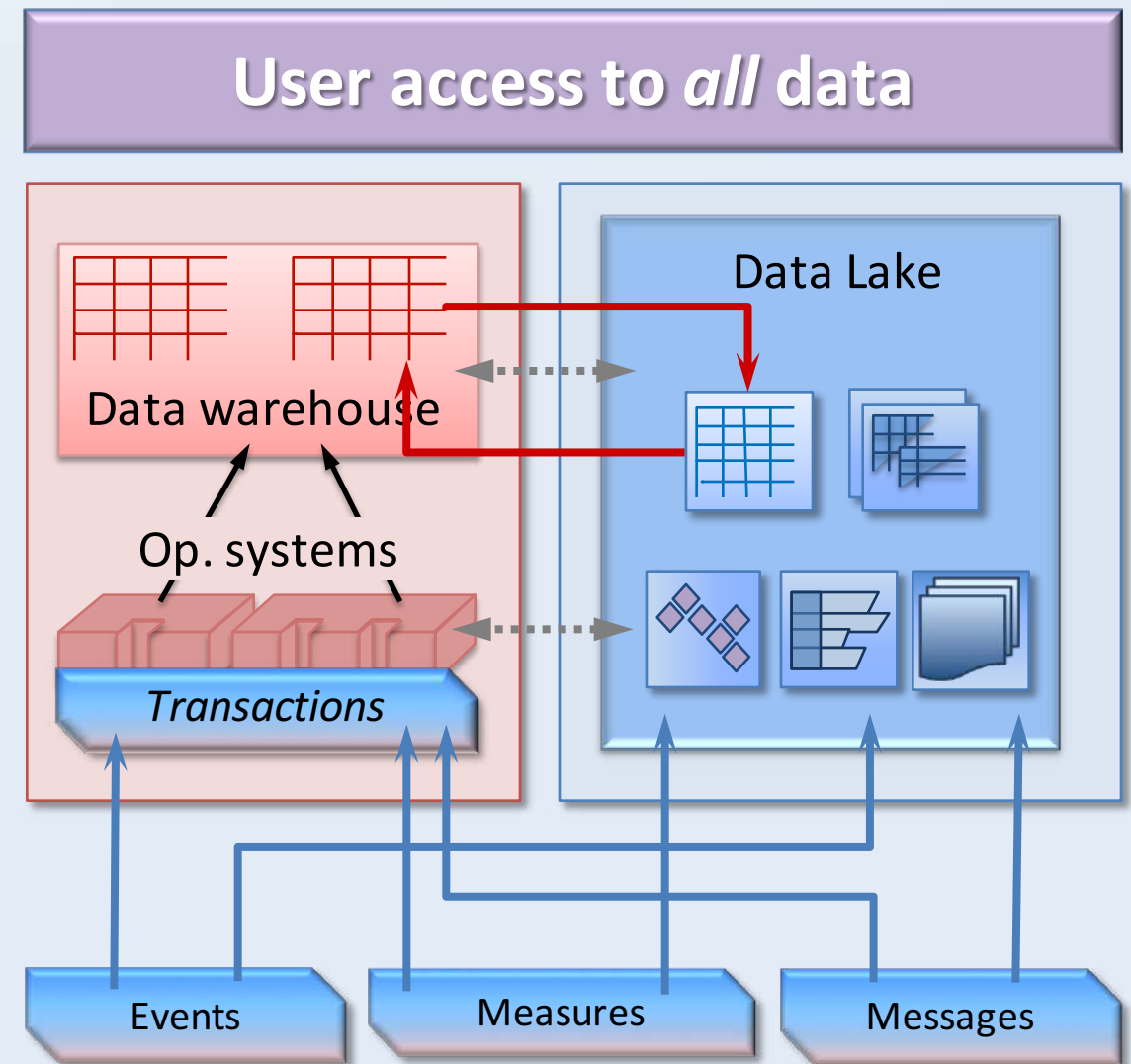
## A warehouse by a lake (1) Preparation and enrichment

- Challenge: ETL (extract, *transform* and load) to data warehouse complex and computationally expensive
- Transform in:
  - Proprietary ETL server – with high licensing cost
  - Data warehouse server – with impact on analytic tasks
- Solution: Pump some or all data through the data lake
  - Reduced processing cost and/or impact on DW work



## A warehouse by a lake (2) Archival

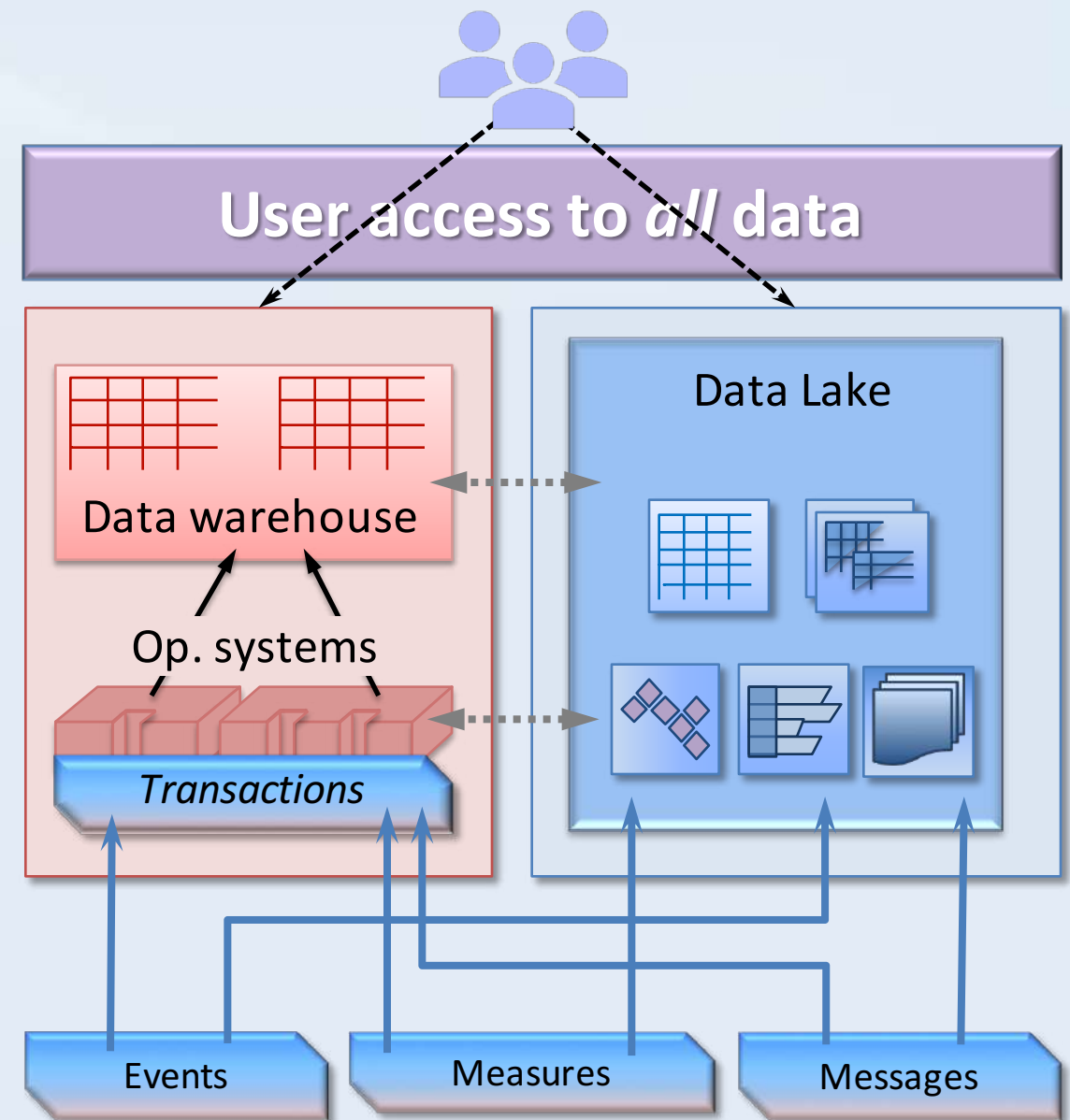
- Challenge: Storing seldom-used (cold) data in a data warehouse is an expensive waste of high-performance hardware
- Archiving to magnetic tape delays and complicates access to off-line data when needed
- Solution: archive to commodity servers and disks in data lake
  - Hadoop – no licensing costs
  - Faster access when needed – almost equal to DW
  - Same tools (SQL-based) for access as DW





## A warehouse by a lake (3) Access

- Challenge: Data increasingly resides on disparate platforms
  - Traditional business info in relational
  - Business people familiar with SQL
  - Social media, IoT on Hadoop / NoSQL / etc.
  - Copying back and forth is expensive
- Solution: Virtualize access to data on all platforms
  - SQL-based queries
  - Join data across platforms



## Conclusions

1. Enterprise data warehouse lives on
  - Focused on core business information
  - Traditional relational platforms still preferred
2. Data lake complements data warehouse
  - Focused on externally sourced data
  - Linked to data warehouse in multiple ways
3. Data lake can assist / offload data warehouse
  - Use commodity storage and processing power
  - Reduce costs and improve performance





# Thank You

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Founder & Principal  
9sight Consulting



# Hortonworks: Next-Gen Data Architectures To Power Modern Analytics

Piet Loubser

**VP Product and Solutions Marketing**  
Hortonworks





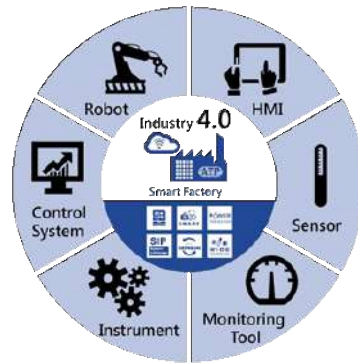
# The New Way of Business Is Fueled By Connected Data

## Development



- Connected Customers, Vehicles, Devices
- Socially crowd-sourced requirements
- Digital design and analysis
- Digital prototypes and tests (simulations)

## Manufacturing



- Connected Factories, Sensors, Devices
- Human-robotic interaction
- 3D-printing on demand

## Distribution



- Connected Trucks, Inventory
- Location, traffic, weather-aware distribution
- Real-time inventory visibility
- Dynamic rerouting

## Marketing/Sales



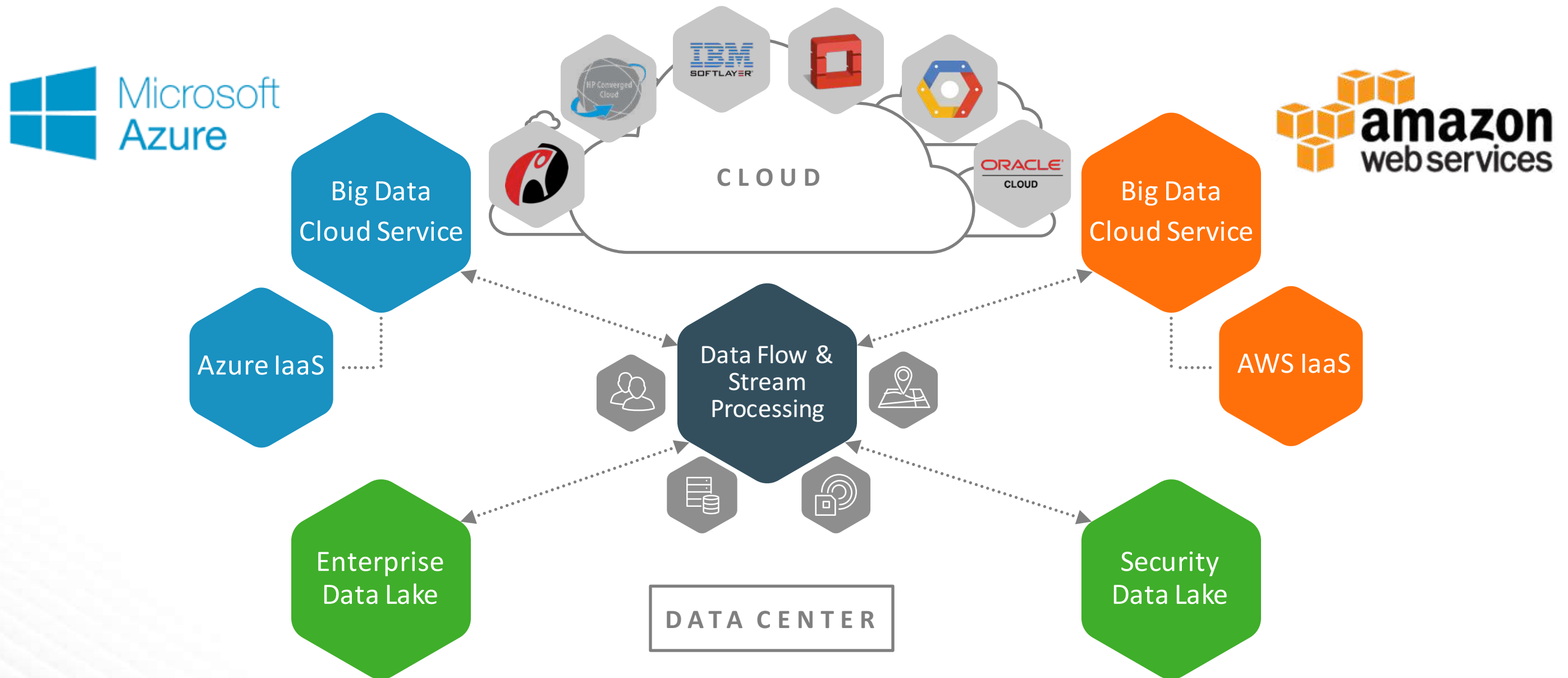
- Connected Customers, Devices
- Omni-channel demand sensing
- Real-Time Recommendations

## Service



- Connected Assets
- Remote service monitoring & delivery
- Predictive maintenance
- OTA Updates

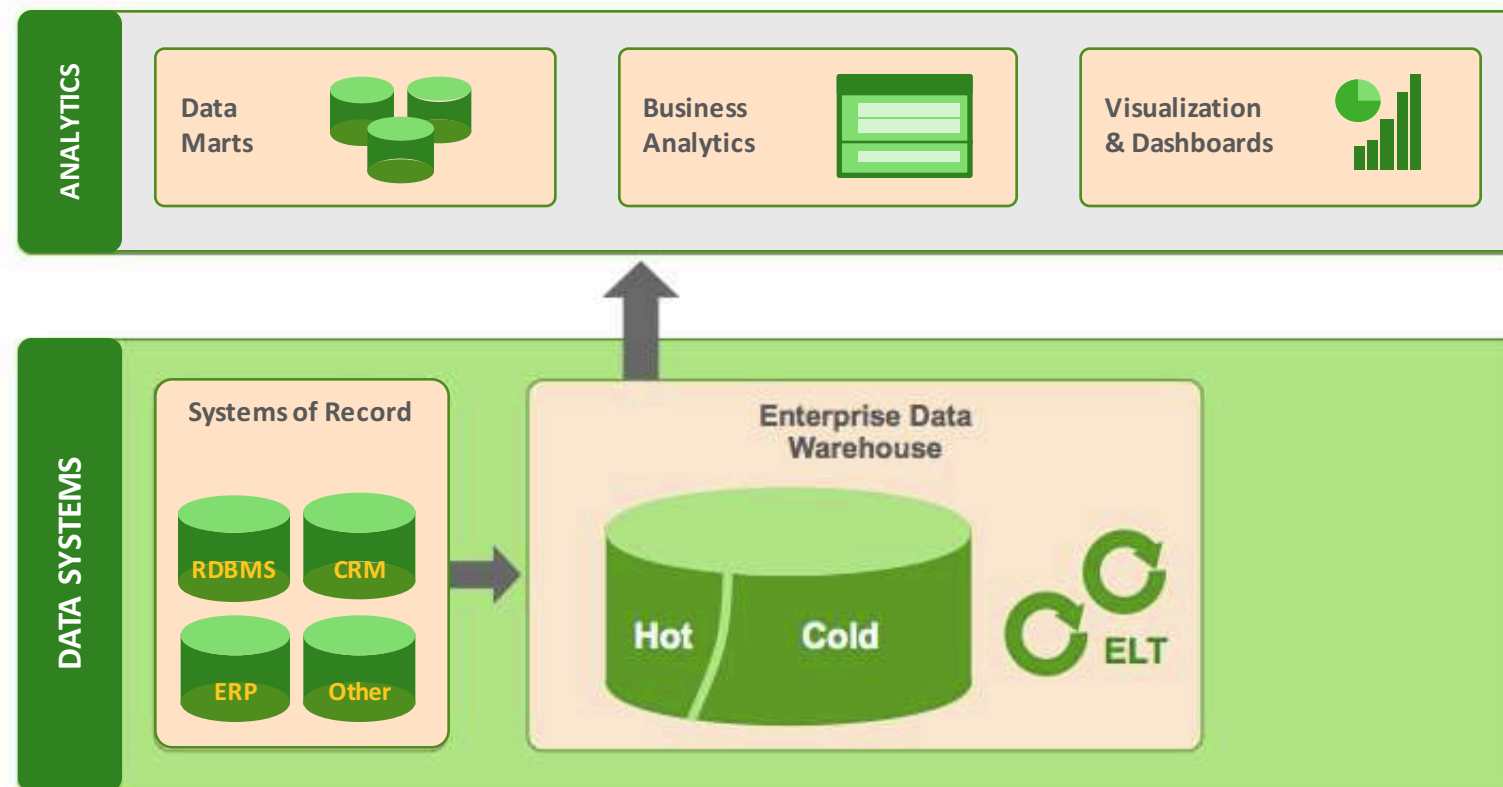
# A Connected Data Strategy Connects Data Center and Cloud





# Typical EDW Architecture

*Used inefficiently, from \$7,500 to \$35,000 per TB<sup>1</sup> of data stored and processed*



## In a typical EDW:

- **50-70%** of data is unused and/or cold
- **45-65%** of CPU capacity is ETL/ELT
- **25-35%** of CPU consumed by ETL is to load unused data
- **30-40%** of CPU is consumed by only **5%** of ETL workloads
- As little as **2.8%** of the data is **Hot**<sup>1</sup>

Source: Hortonworks Innovation and Strategy Team and Appfluent Analysis

1. EY Analysis shows typical range from \$10-15k / TB. Hortonworks experience shows a wide range observed in the field, from \$35k/TB for massive, in-memory EDW appliances to \$7.5k/TB for RDBMS based, home-grown EDW solutions

2. For example, for a client keeping a rolling 36-month window of data for reporting in an EDW, only 1 month of the 36 (2.8%) is new/hot.

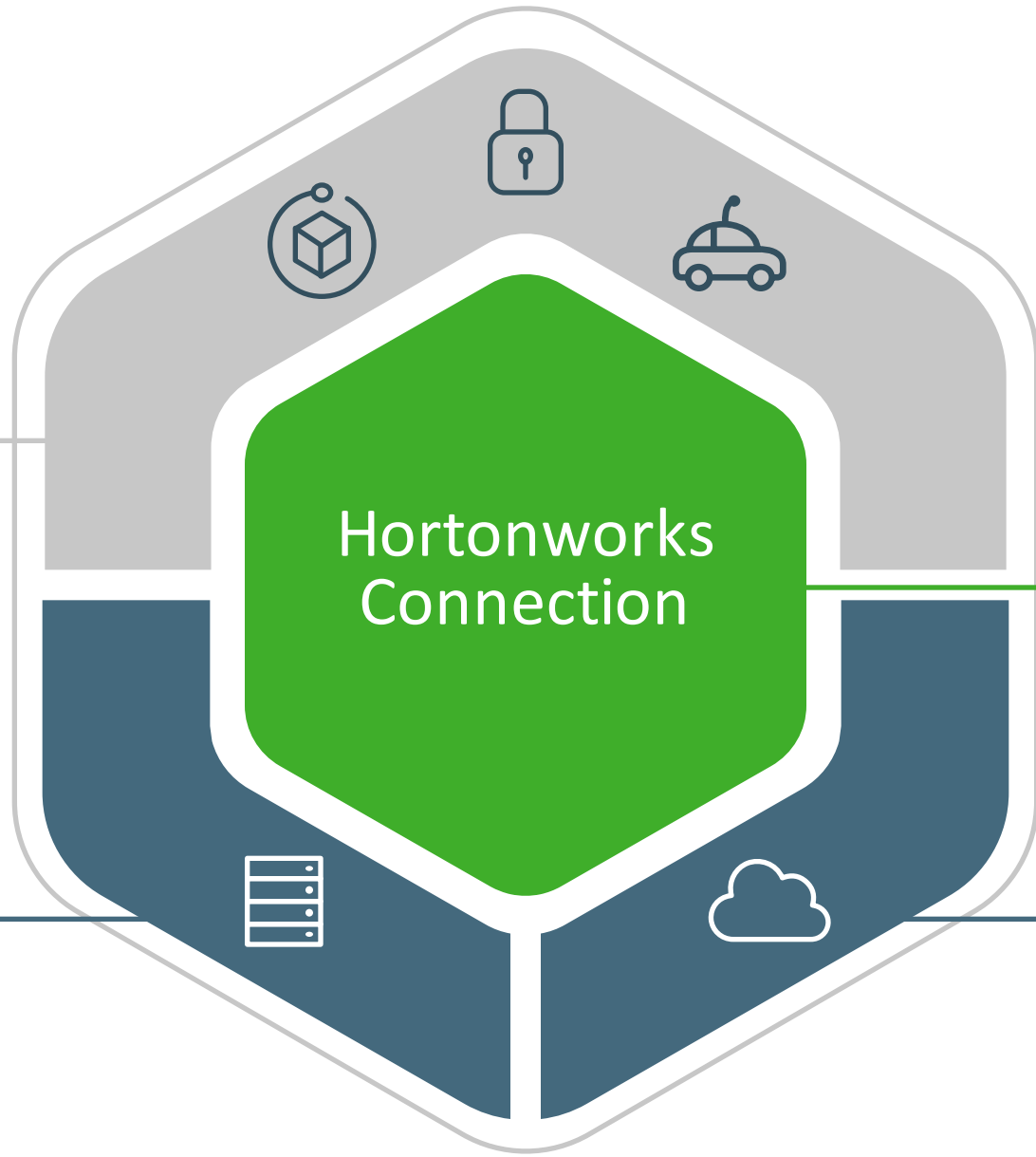
# Hortonworks Connection: Services and Solutions for Your Success

## Hortonworks Solutions

- Enterprise Data Warehouse Optimization
- Cyber Security and Threat Management
- Internet of Things and Streaming Analytics
- Data Science Experience
- Advanced SQL

## Data Center

- Hortonworks Data Suite
- HDP
- HDF



## Hortonworks Connection

- Enablement Subscription
- SmartSense™
- Premier Operational Support
- Educational Services
- Professional Services
- Community Connection

## Cloud

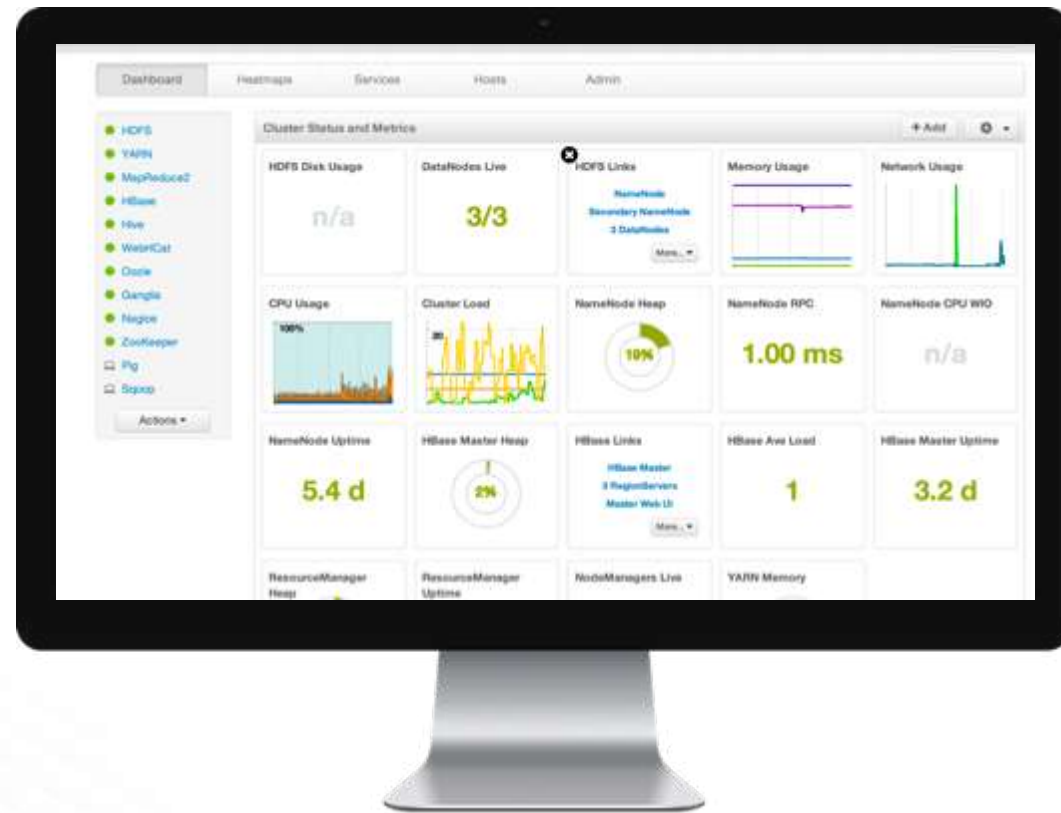
- Hortonworks Data Cloud
- AWS
- HDInsight

## Data Services





# Enterprise Data Warehouse Optimization



## Dramatic Cost Reductions

Reduce cost of your EDW Implementation by offloading ETL processes and archiving cold data

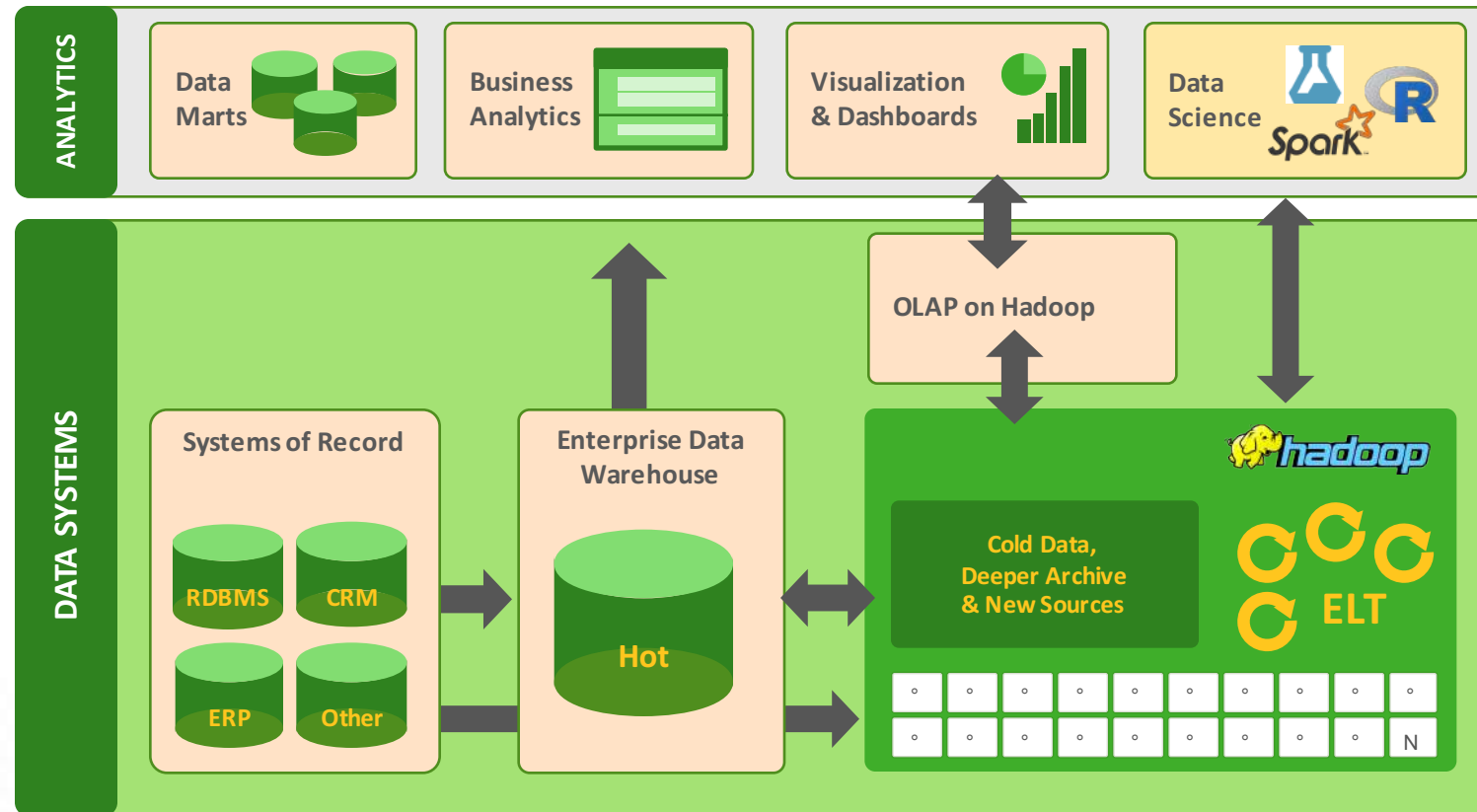
## Deploy Business Intelligence on Hadoop

Empower Business users with powerful reporting, new applications, visualization tools, and artificial intelligence

## Support More Types of Unstructured Data

Index and search images, videos, text & sound files

# EDW Plus Hadoop helps you optimize and reduce costs associated with your EDE



## Archive Cold Data away from EDW

- Move cold or rarely used data to Hadoop as active archive
- Store more of your data longer, cheaper

## Offload costly ETL process

- Free your EDW to perform high-value functions like analytics & reporting, not ETL
- Use Hadoop for advanced or massive-scale ETL/ELT



# EDW Optimization: ETL Offload

## ◆ The Problem:

- EDWs consume between 50% and 90% of CPU just on ETL/ELT tasks.
- These jobs interfere with more business-critical tasks like BI and advanced analytics.

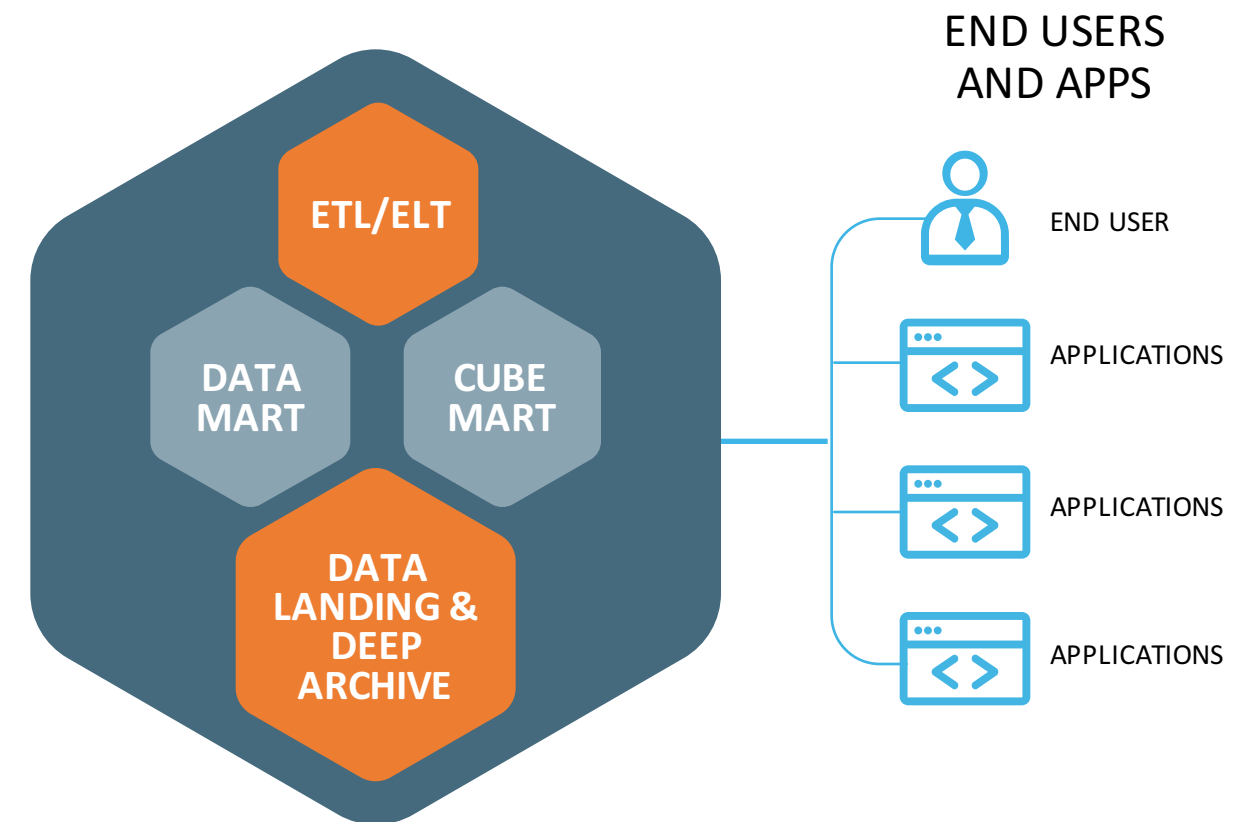
## ◆ The Solution:

- Hive and HDP deliver ETL that scales to petabytes.
- Syncsort DMX-h for simple drag-and-drop ETL workflows.
- Economical scale-out processing on commodity servers.

## ◆ The Result:

- Better SLAs for mission-critical analytics.
- Limit EDW expansion or retire old systems.

## EDW OPTIMIZATION SOLUTION



# EDW Optimization: Active Archive

## ◆ The Problem:

- Increasing data volumes and cost pressure force data to be archived to tape.
- Archived data not available for analytics, or must be retrieved at great expense.

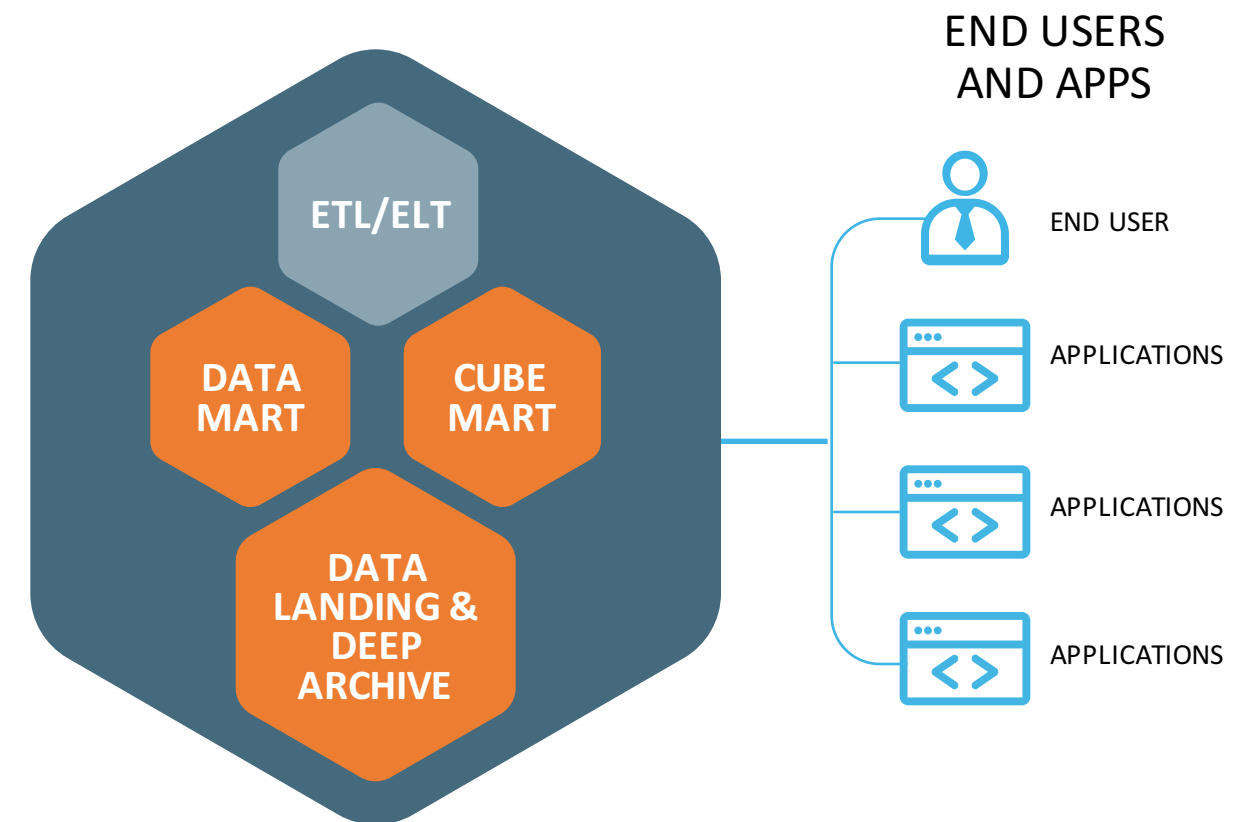
## ◆ The Solution:

- Adopting Hadoop delivers cost per terabyte on par with tape backup solutions.
- Data in Hadoop can be analyzed by all major BI tools, allowing analytics on archive data.

## ◆ The Result:

- Data always available for analytics.
- Store years of data rather than months.

## EDW OPTIMIZATION SOLUTION





# EDW Optimization: Fast BI on Hadoop

## ■ The Problem:

- Proprietary EDW systems were adopted for Fast BI and deep slice-and-dice analytics, but EDW prices are unsustainably high.

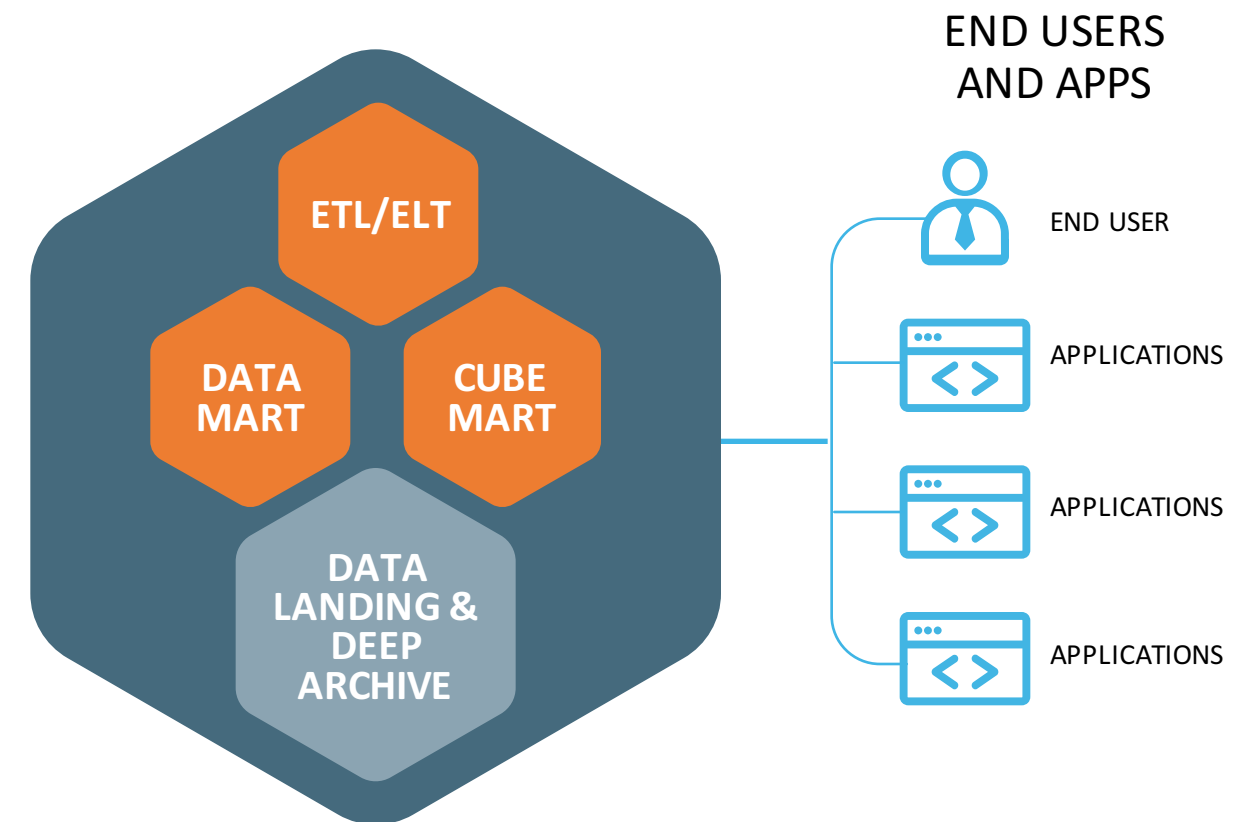
## ■ The Solution:

- Interactive SQL is a reality on Hadoop today.
- Partner Solutions (IBM BigSQL, Kyvos, Jethro) adds powerful SQL and OLAP capabilities for deep drilldown at scale.

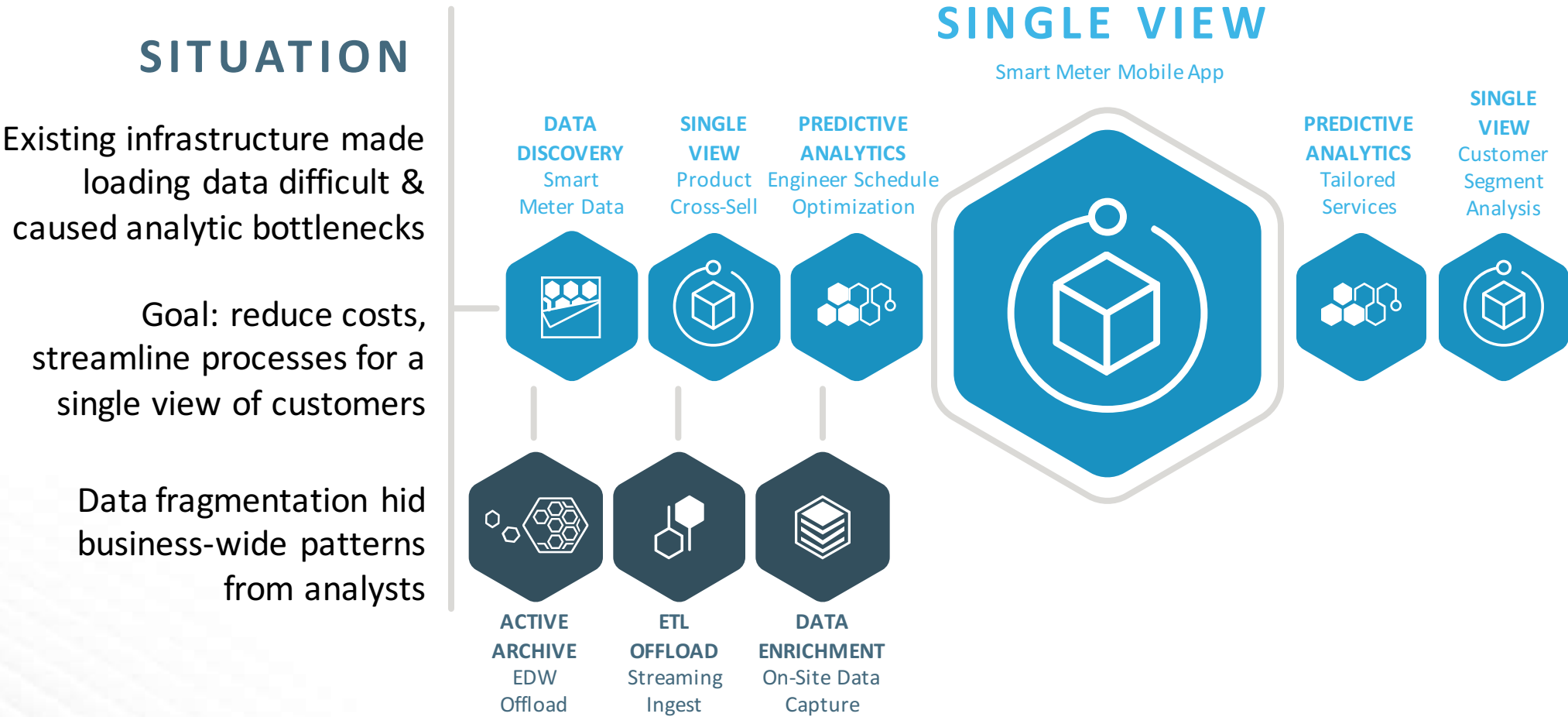
## ■ The Result:

- Query terabytes of data in seconds.
- Connect your favorite BI tools like Tableau and Excel through SQL and MDX interfaces.
- The EDW Optimization Solution is tailor-made to deliver Fast BI on Hadoop.

## EDW OPTIMIZATION SOLUTION



# Centrica Transforms Service For Utility Customers

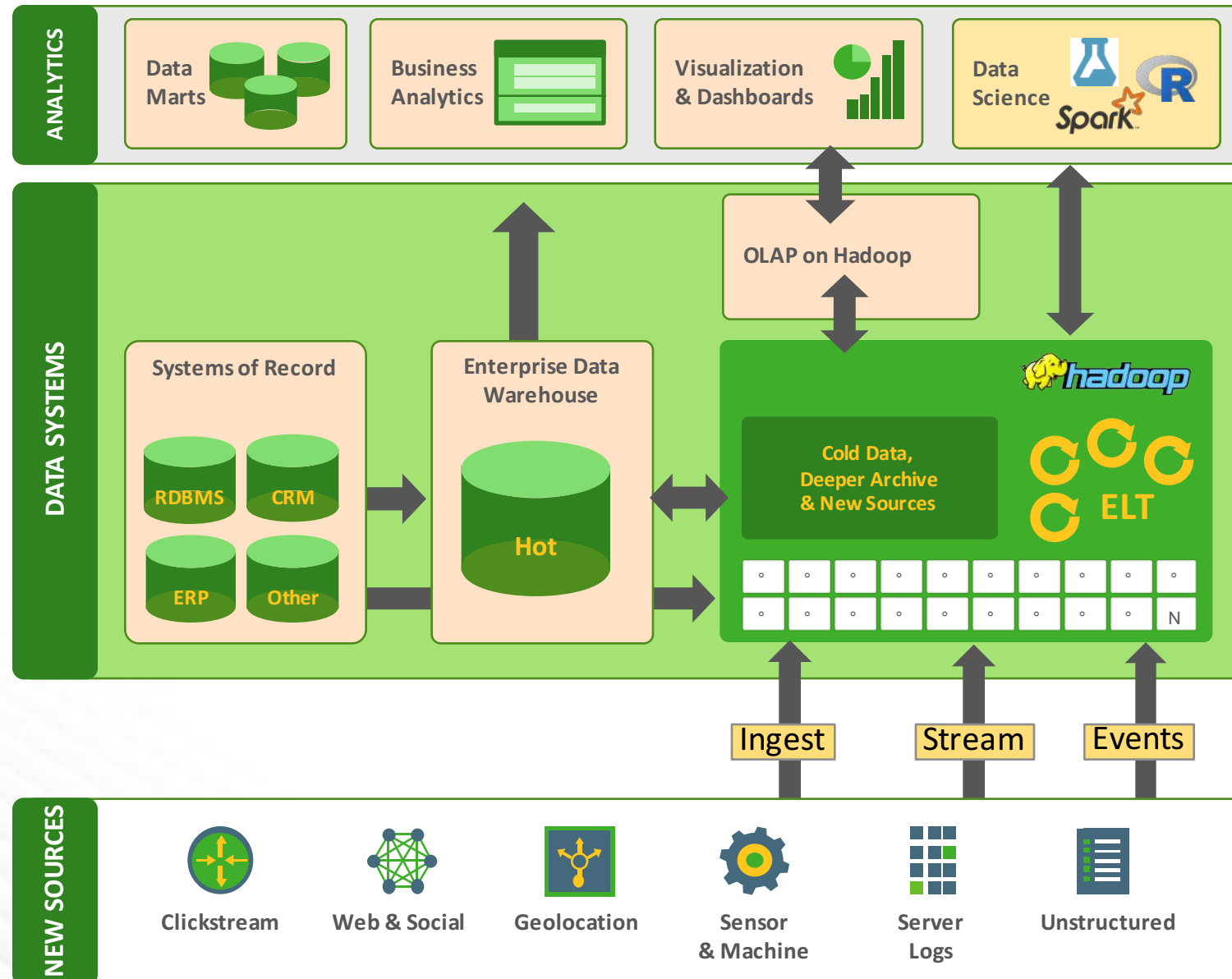


<b>3 Million Customers</b>	can access “smart energy reports”
<b>ETL efficiency gains</b>	from 11 hours to 45 minutes/job
<b>300 GB/Day Ingest</b>	rationalizes work of field engineers
<b>Decommissioned some EDWs</b>	saving millions annually

“Focusing on innovation, learning to forget traditional legacy ways of working and approaching it in new ways creates unexpected behavioural changes, because people feel freer and they also feel valued.”  
Dajit Rehal, Senior Systems Director



# EDW Plus Hadoop helps you land and enrich more data to respond faster to new business requests



## Archive Cold Data away from EDW

- Move cold or rarely used data to Hadoop as active archive
- Store more of your data longer, cheaper

## Offload costly ETL process

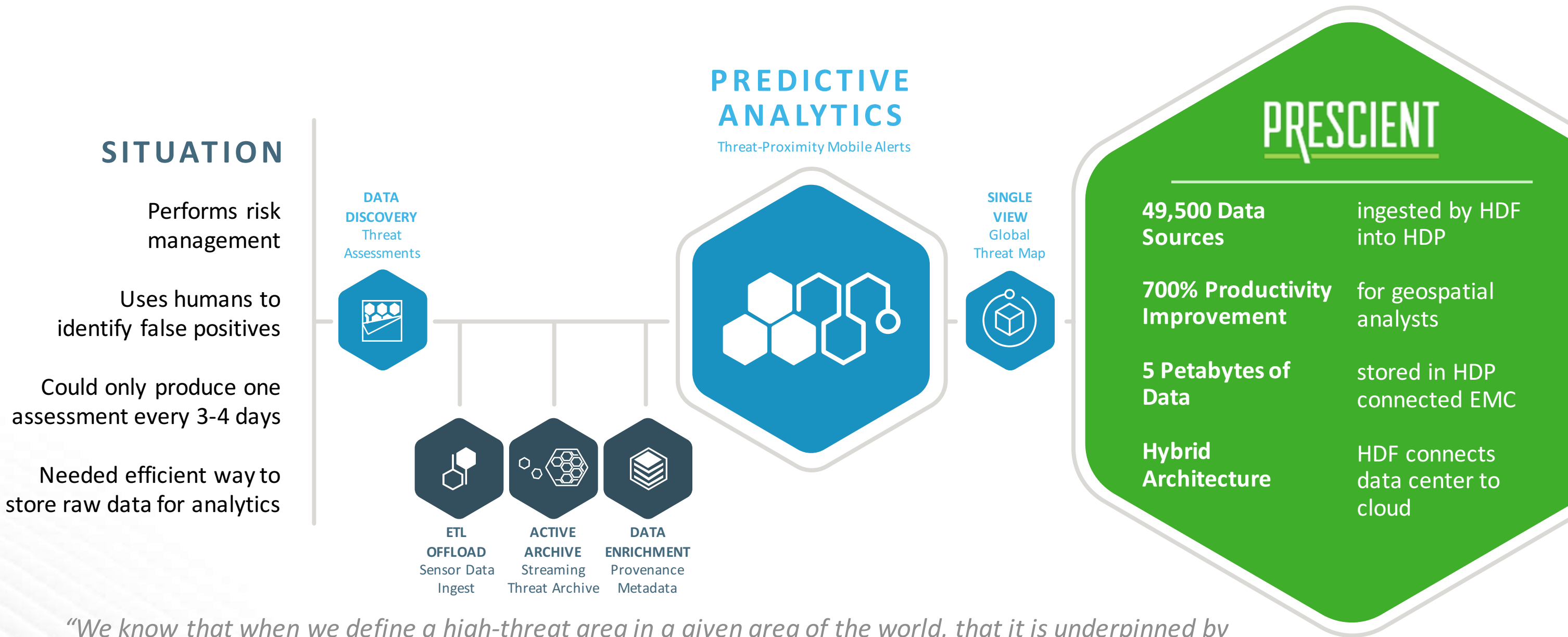
- Free your EDW to perform high-value functions like analytics & reporting, not ETL
- Use Hadoop for advanced or massive-scale ETL/ELT

## Land & Enrich more data to create more value-add analytics

- Use Hadoop to ingest new data sources, such as web and machine data for new analytical context from **unstructured and semi-structured sources**
- **Create an analytical sandbox** for advanced data science



# Prescient Harnesses Machine Learning for Traveler Safety Warnings



*“We know that when we define a high-threat area in a given area of the world, that it is underpinned by very specific data sources. It’s data-driven, and we can point to those sources—if ever asked—and say, ‘Here’s why.’”* Mike Bishop, Chief Systems Architect



# Why Hortonworks?

## Powered By 100% Open Source

Rapid innovation  
Dramatic cost reduction

## Enterprise Ready

Governance  
Fine grained security  
Lineage and data provenance

## Powering All Data

Data-at-Rest, Data-in-Motion  
Cloud, On-Premises  
Structured, unstructured



Forrester Wave: Big Data Warehouse, Q2 2017

[hortonworks.com/get-started/big-data-scorecard/](https://hortonworks.com/get-started/big-data-scorecard/)

# Thank You