

Using Big Data to Explore New Opportunities

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Introduction to Big Data



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The Myth About Big Data

- **Big Data Is New**
- **Big Data Is Only About *Massive Data Volume***
- **Big Data Means *Hadoop***
- **Big Data Means *Unstructured Data***
- **Big Data Is for *Social Media & Sentiment Analysis***

Source: *Big Data: New Era of Analytic*, Omer Sever, IBM SWG TR, Enterprise Content Manager

Big Data Is..

It is all about **better Analytic** on a **broader** spectrum of **data**, and therefore represents an **opportunity** to **create** even more **differentiation** among **industry peers**.

Source: Big Data: New Era of Analytic, Omer Sever, IBM SWG TR, Enterprise Content Manager



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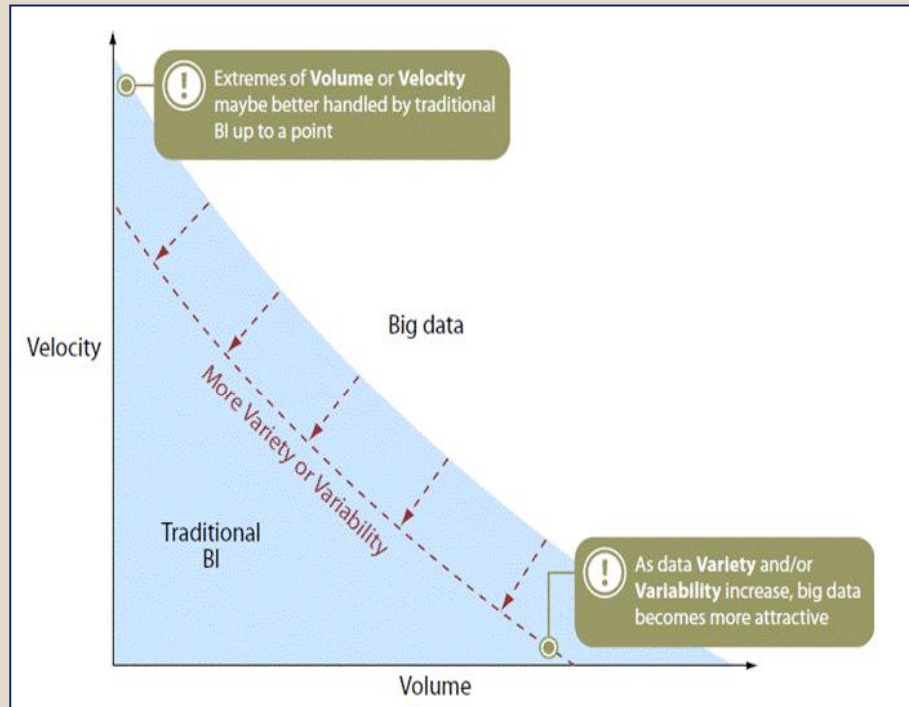


What's Big Data?

- **Big data** is the term for a collection of data sets so large and complex that it becomes difficult to process using on-hand database management tools or traditional data processing applications.
- The challenges include capture, curation, storage, search, sharing, transfer, analysis, and visualization.
- The trend to larger data sets is due to the additional information derivable from analysis of a single large set of related data, as compared to separate smaller sets with the same total amount of data, allowing correlations to be found to "spot business trends, determine quality of research, prevent diseases, link legal citations, combat crime, and determine real-time roadway traffic conditions."

Source: Big Data: New Era of Analytic, Omer Sever, IBM SWG TR, Enterprise Content Manager

What is Big Data?



Volume

Exceeds physical limits of vertical scalability

Velocity

Decision window small compared to data change rate

Variety

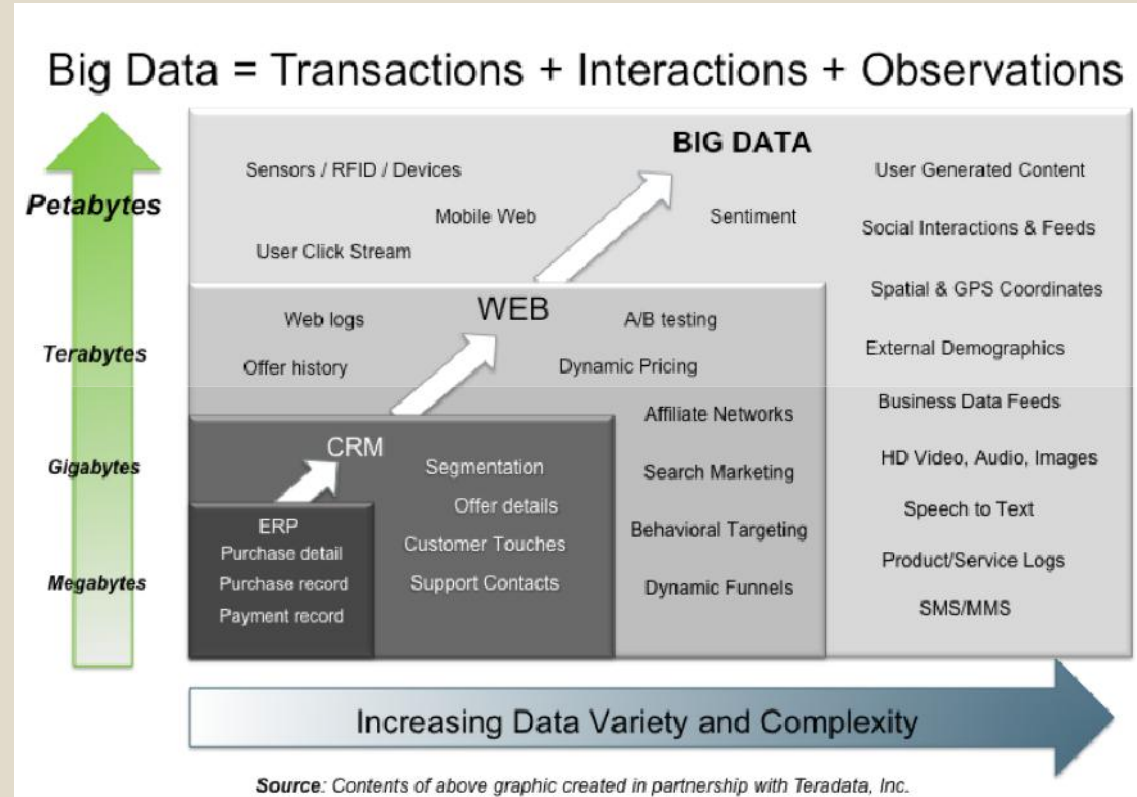
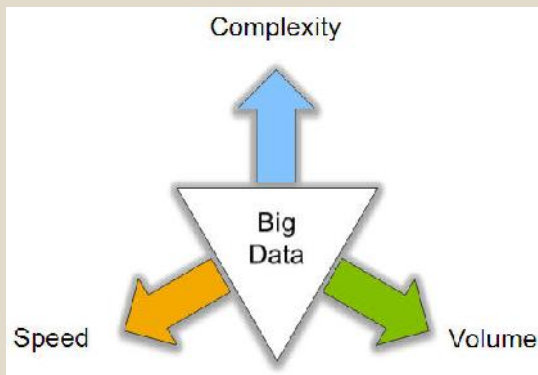
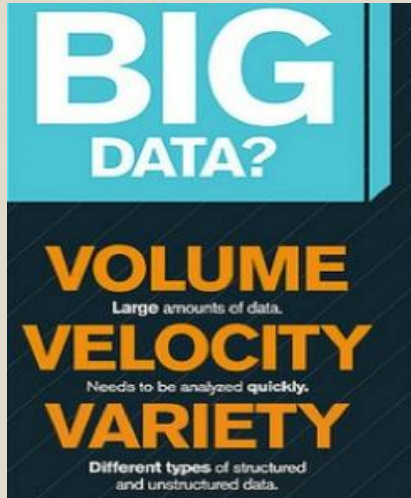
Many different formats makes integration expensive

Variability

Many options or variable interpretations confound analysis

Source: Ensuring Compliance of Patient Data with Big Data and BI, Ayad Shammout & Denny Lee, PASS Business Analytics Conference

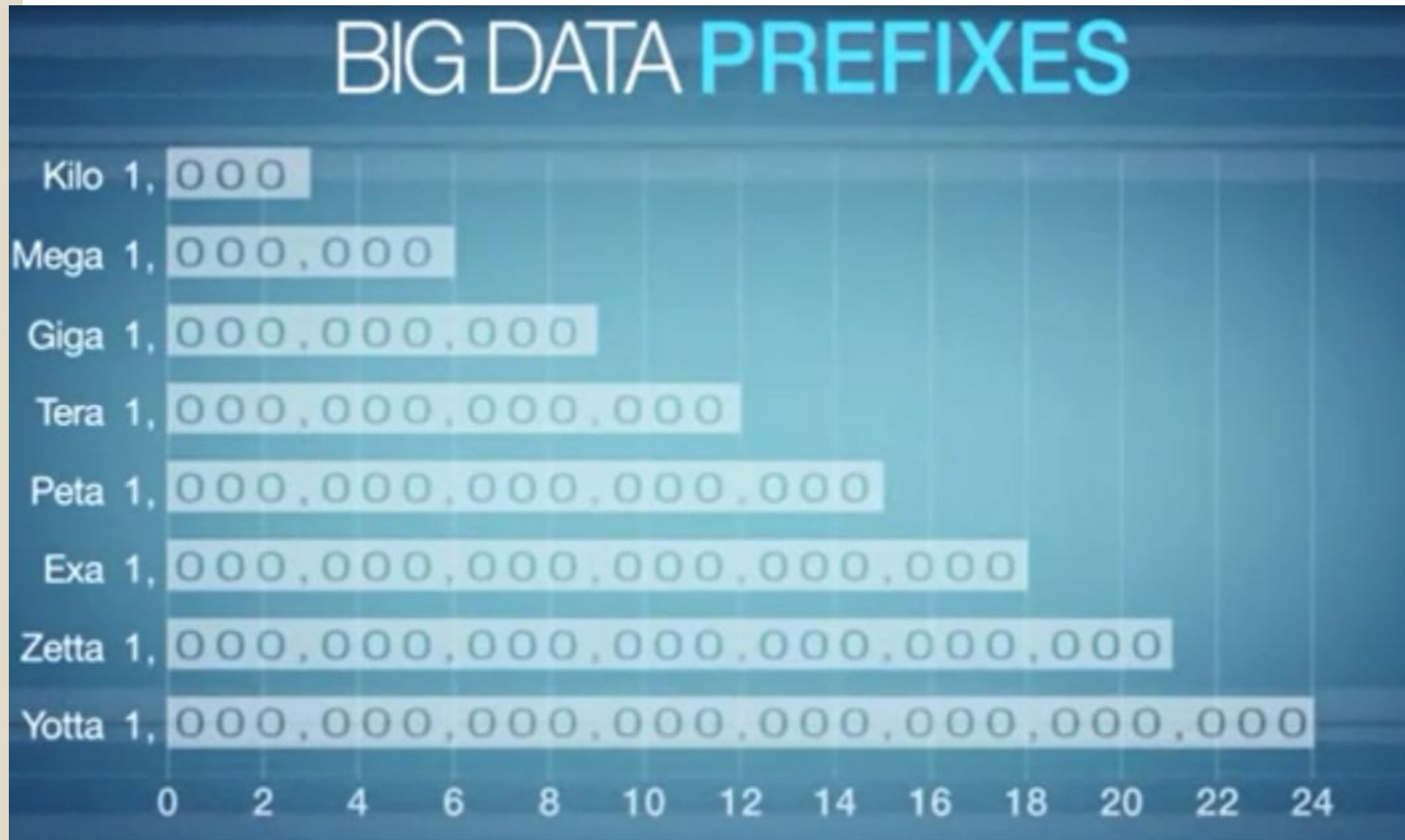
Big Data: From 3V's to 4V's



Source: Introduction to Big Data and Basic Data Analysis, Ruoming Jin, Kent University

Volume (Scale)

Source: Introduction to Big Data and Basic Data, Analysis, Ruoming Jin, Kent University



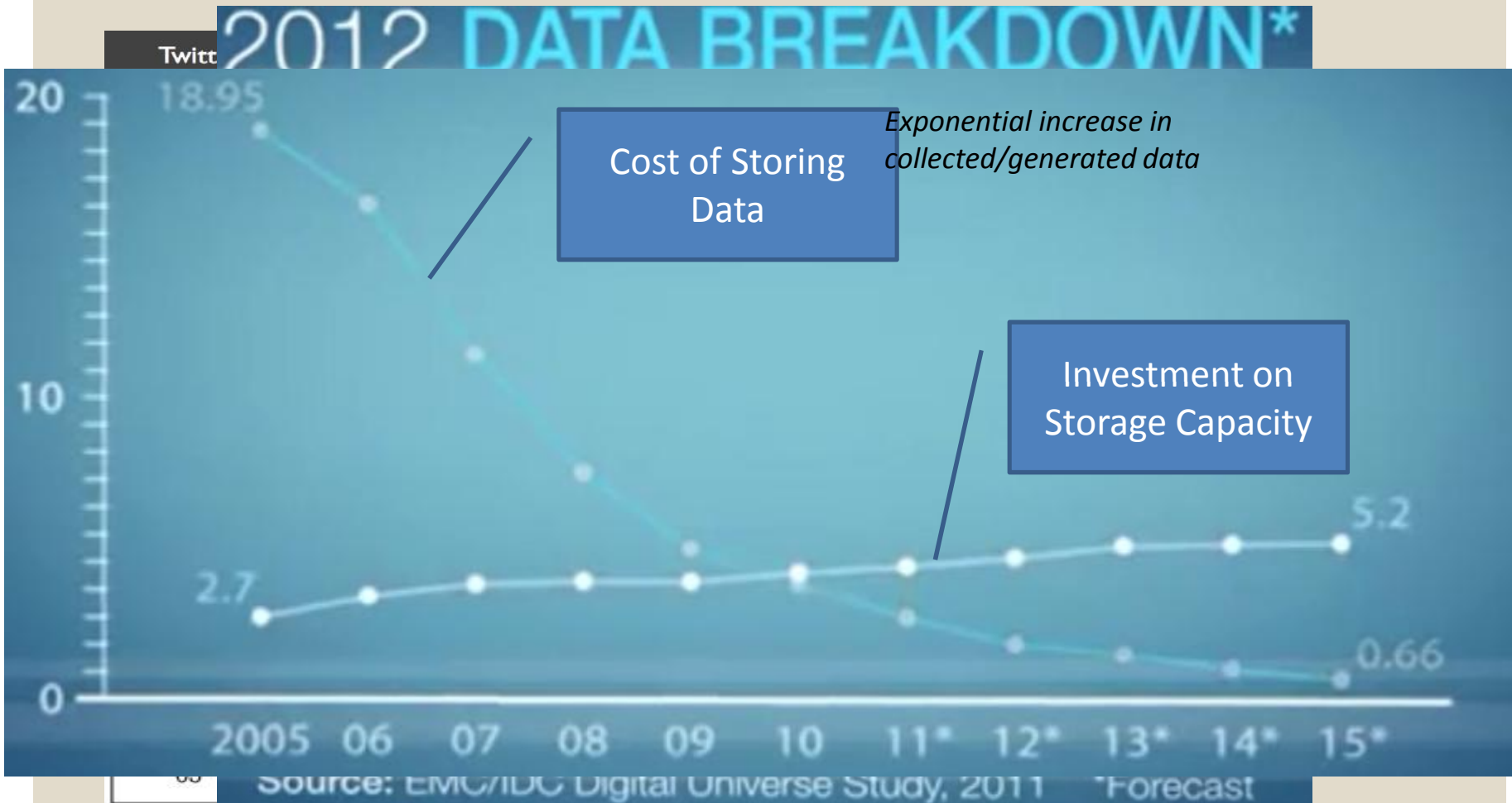
Source: Contents of above graphic created in partnership with Teradata, Inc.



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Volume (Scale)



Source: Introduction to Big Data and Basic Data, Analysis, Ruoming Jin, Kent University

Where Is This “Big Data” Coming From ?

? TBs of data every day

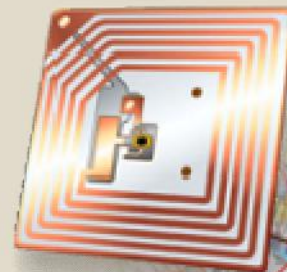


12+ TBs of tweet data every day



25+ TBs of log data every day

30 billion RFID tags today (1.3B in 2005)



76 million smart meters in 2009... 200M by 2014



4.6 billion camera phones world wide



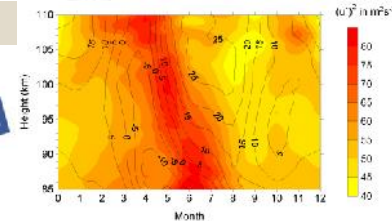
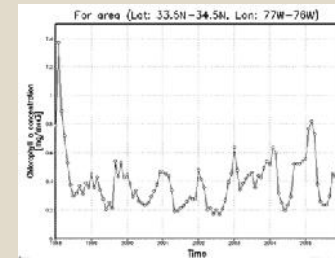
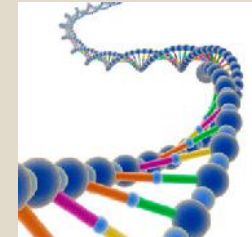
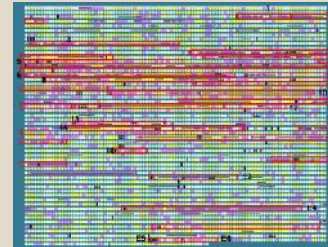
100s of millions of GPS enabled devices sold annually

2+ billion people on the Web by end 2011



Variety (Complexity)

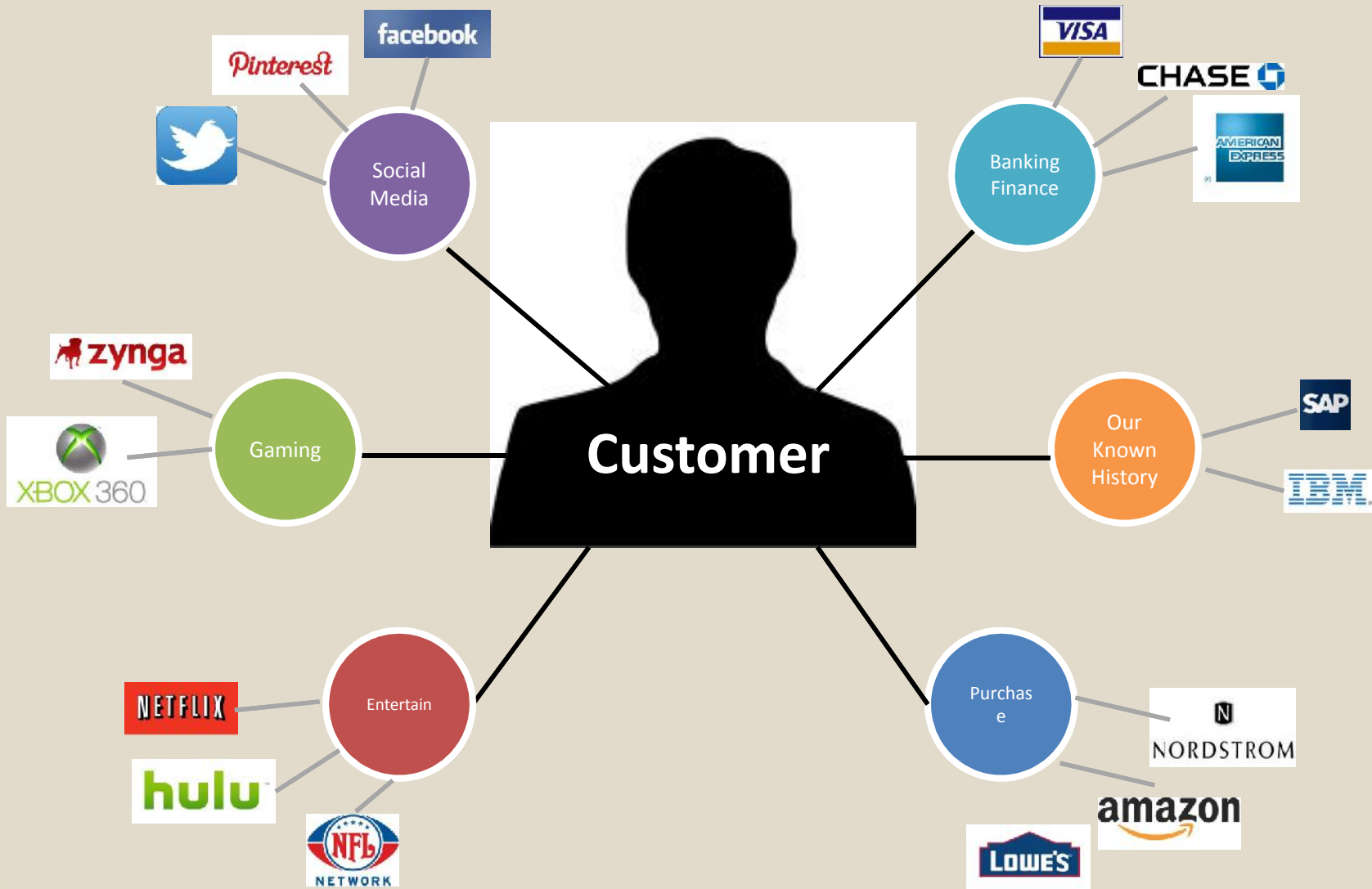
- Relational Data (Tables/Transaction/Legacy Data)
- Text Data (Web)
- Semi-structured Data (XML)
- Graph Data
 - Social Network, Semantic Web (RDF), ...
- Streaming Data
 - You can only scan the data once
- A single application can be generating/collecting many types of data
- Big Public Data (online, weather, finance, etc)



To extract knowledge → all these types of data need to be linked together

Source: Introduction to Big Data and Basic Data, Analysis, Ruoming Jin, Kent University

A Single View to the Customer



Velocity (Speed)

- Data is begin generated fast and need to be processed fast

- Online Data Analytics

- Late decisions → missing opportunities

- **Examples**

- **E-Promotions:** Based on your current location, your purchase history, what you like → send promotions right now for store next to you
- **Healthcare monitoring:** sensors monitoring your activities and body → any abnormal measurements require immediate reaction



Source: Introduction to Big Data and Basic Data, Analysis, Ruoming Jin, Kent University

Real-time/Fast Data



Social media and networks
(all of us are generating data)



Scientific instruments
(collecting all sorts of data)



Mobile devices
(tracking all objects all the time)

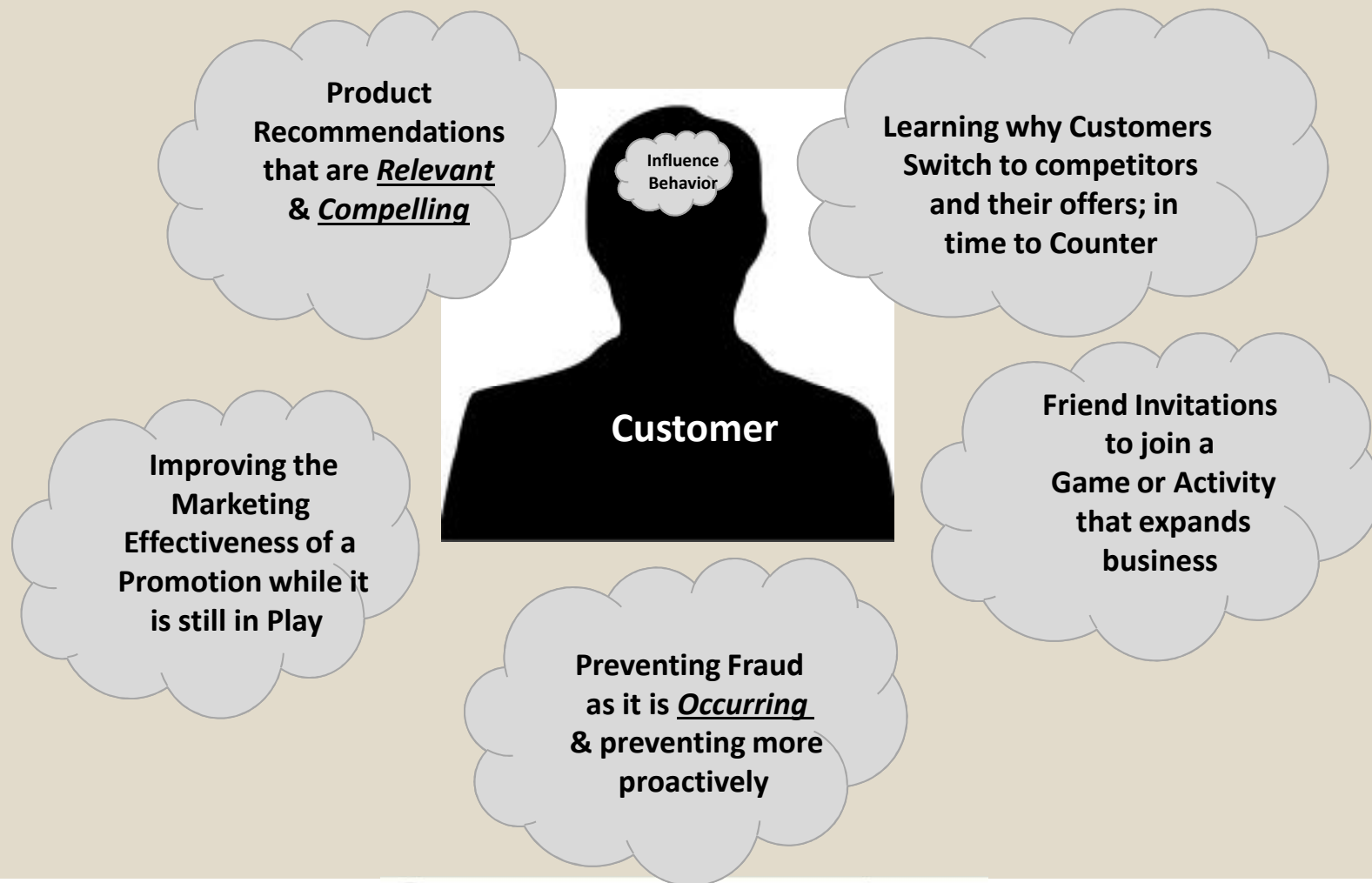


Sensor technology and networks
(measuring all kinds of data)

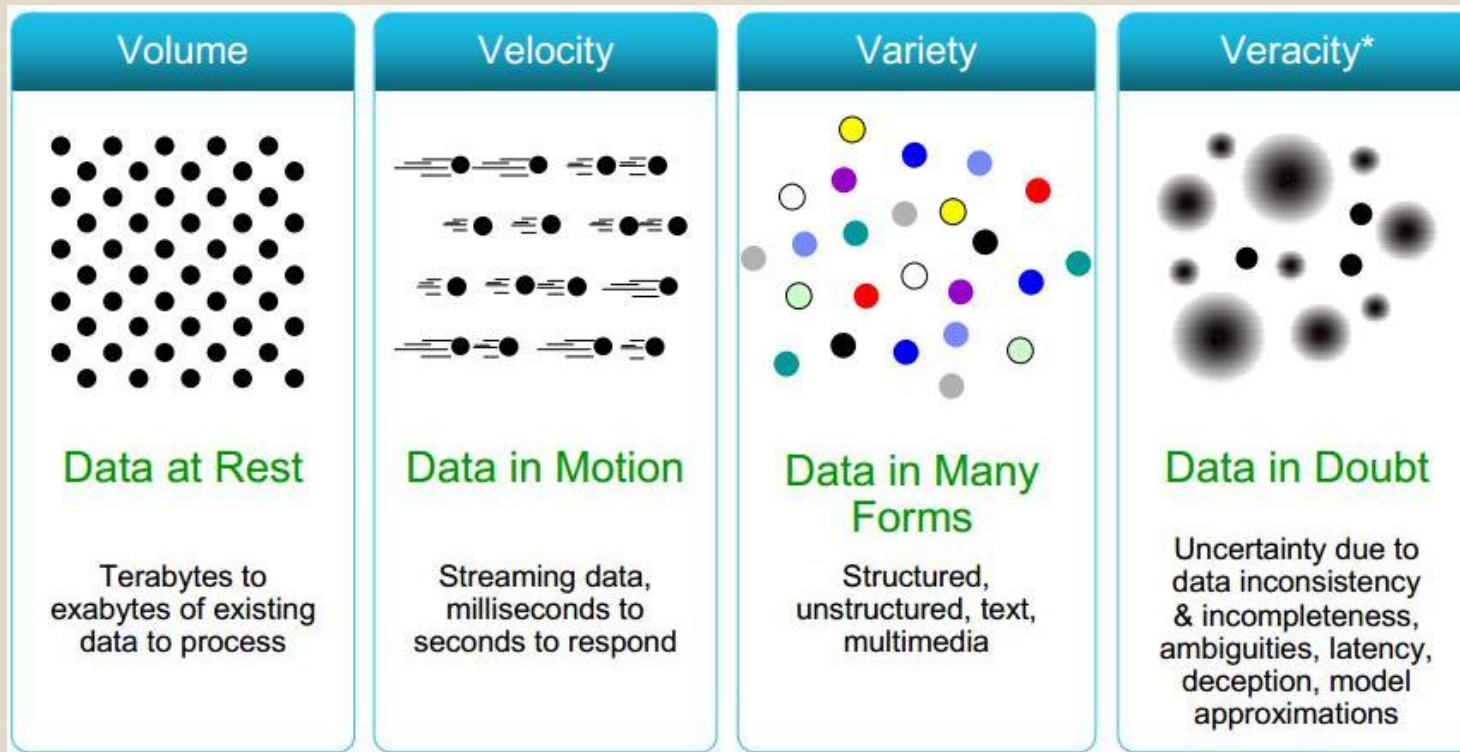
- The progress and innovation is no longer hindered by the ability to collect data
- But, by the ability to manage, analyze, summarize, visualize, and discover knowledge from the collected data in a timely manner and in a scalable fashion

Source: Introduction to Big Data and Basic Data, Analysis, Ruoming Jin, Kent University

Real-Time Analytics/Decision Requirement



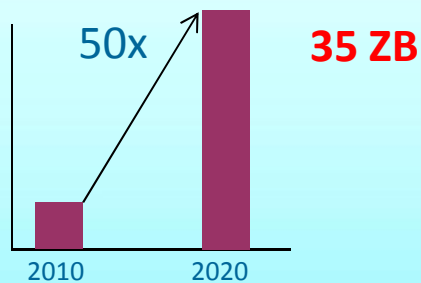
Some Make it 4V's



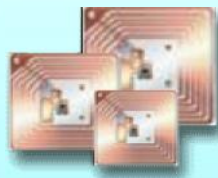
Source: *Big Data in Engineering Applications*, Jasti Aswini

Summary: Four Characteristics of Big Data

Cost efficiently processing the growing **Volume**



Responding to the increasing **Velocity**



30 Billion
RFID sensors
and counting

Collectively Analyzing the broadening **Variety**



80% of the worlds
data is unstructured



Establishing the **Veracity** of big data sources

1 in 3 business leaders don't trust the information they use to make decisions

Source: Big Data: New Era of Analytic, Omer Sever, IBM SWG TR, Enterprise Content Manager

Big Data:
Big Opportunity, Big Challenge



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Data explosion

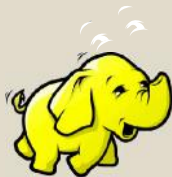


10x increase every five years

85% from new data types

Volume
Velocity
Variety

Hadoop



Cheap, Distributed Storage & Processing



Cloud

Easy Accessibility of External Data

By 2015, organizations that build a modern information management system will outperform their peers financially by 20 percent.

- - Gartner, Mark Beyer
"Information Management in the 21st Century"

Source: Ensuring Compliance of Patient Data with Big Data and BI, Ayad Shammout & Denny Lee, PASS Business Analytics Conference

The Big Potential Opportunities of Big Data

- **Bollen, Mao, and Zeng (2011)** - use Twitter data to predict daily fluctuations of the Dow Jones Industrial Average (DJIA). Google's Profile of Mood States and OpinionFinder (Wilson et al. 2005) measure public mood based on 10 million public tweets.
- **Chan (2003) and Mittermayer (2004)** - use news articles to predict stock price movements. Social and Mass Media data can be utilized to help measure financial health of a firm, and better evaluate the audit engagement
- **Mofitt and Vasarhelyi (2013)** - propose to use news, audio and video streams, cell phone recordings, social media comments to obtain new forms of audit evidence, confirm existence of events, and validate reporting elements.

Source: Big Data Analytics in Financial Statement Audits, Min Cao, Roman Chychyła, Trevor Stewart, 2015

Big Data Business Value

15 out of 17

sectors in the US have more data stored per company than the US Library of Congress

140,000-190,000

more deep analytical talent positions

1.5 million

more data savvy managers in the US alone

€250 billion

Potential annual value to Europe's public sector

50-60%

increase in the number of Hadoop developers within organizations already using Hadoop within a year

\$300 billion

Potential annual value to US healthcare

Source: *Ensuring Compliance of Patient Data with Big Data and BI*, Ayad Shammout & Denny Lee, PASS Business Analytics Conference



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The number of organizations who ~~see~~ analytics as a competitive advantage ^{70%} is growing.

57%

37%

2010

58%

2011

63%

2012

analytics IQ
business initiative

Studies show that organizations competing on analytics outperform their peers

substantially outperform

220%

1.6x

Revenue
Growth

2.5x

Stock Price
Appreciation

2.0x

EBITDA
Growth

The Big Opportunities for Financial Industry



The Big Challenge of Big Data

McKinsey & Co. recently reported that two-thirds of C-suite executives surveyed consider big data to be a top strategic priority.

McKinsey survey, however, noted a big gap between what organizations want to do with big data and their capabilities to do so, given their existing IT infrastructure and expertise.

Source: Big Data, Collect It, Respect It, IIA Tone at the Top, Aug 2013



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The Big Challenge for Internal Auditor

- The Los Angeles Police Department analyzes data from crime scenes, including time, location, nature, and actors in order to predict the most likely timing and location of crimes on that day.
- Similar analytical tools are used in other industries, and direct audit efforts aimed at identifying risks and opportunities.
- Population vs. sample size: The IIA's *Internal Audit* journal has published articles on the implications of big data in its February 2015 issue. Article, author Russell Jacks, notes that internal auditors should address the implications of big data, including integration with other data sources, data security, and data destruction policies.
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“The audit tool kit looks the same as it did 50 or 60 years ago,” he says. “If we were doctors, that would be pretty frightening. This has tremendous potential, but it’s still early. We’re still experimenting.”

Dorsey Baskin, Grant Thornton

Source: *Big Data Analytics in Financial Statement Audits*, Min Cao, Roman Chychyła, Trevor Stewart, 2015

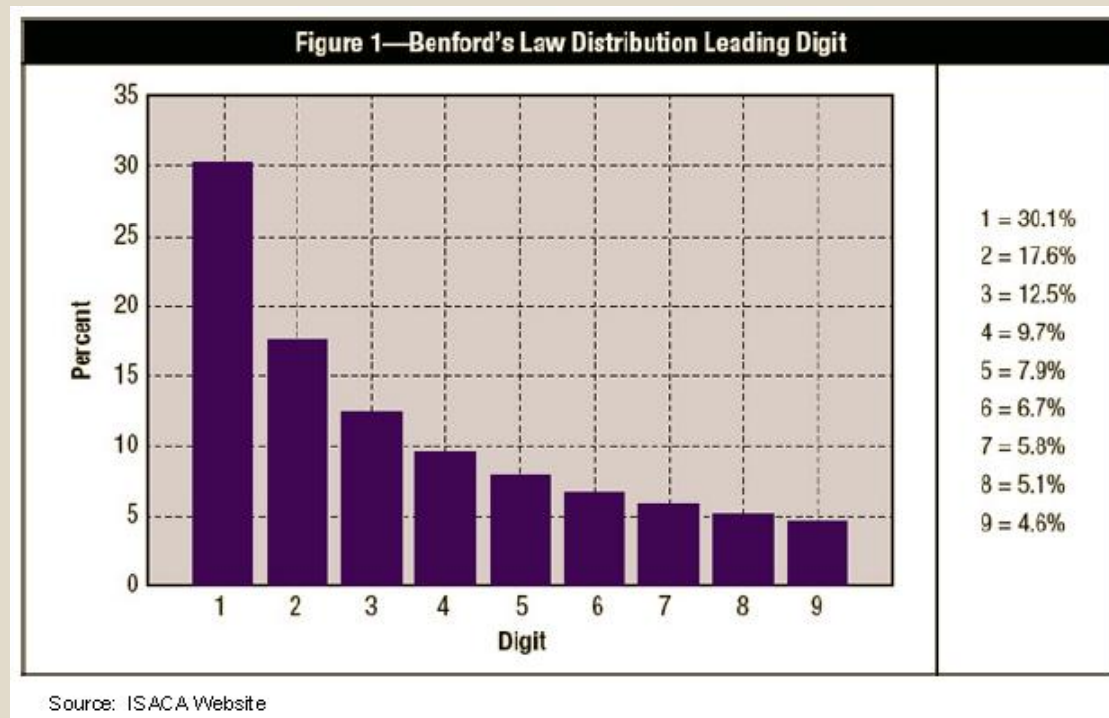


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Benford Law

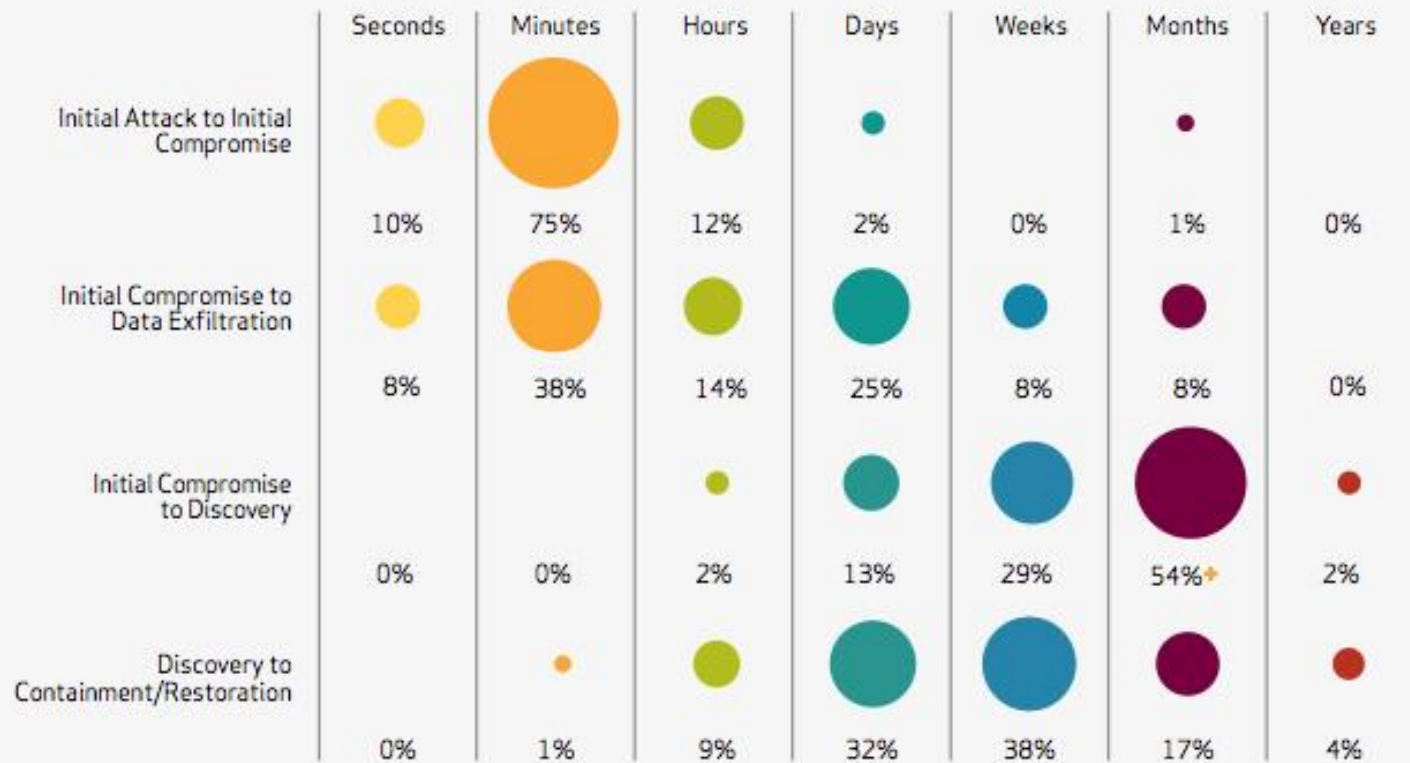
- Mathematical theory of leading digits. Leading digits are distributed in a specific, non-uniform way.
- Simon Newcomb, 1881: Described theoretical frequency that is Benford's Law
- Frank Benford, 1938: Numbers starting with 1, 2, or 3 are more common in nature than those with initial digits 4 – 9.



Source: *An Auditors Guide to Data Analysis*, Natasha DeKroon, Duke University

Velocity of Data Generation vs Fraud/Breach Detection

Figure 40. Timespan of events by percent of breaches



Source: Verizon Data Breach Investigation Report 2013

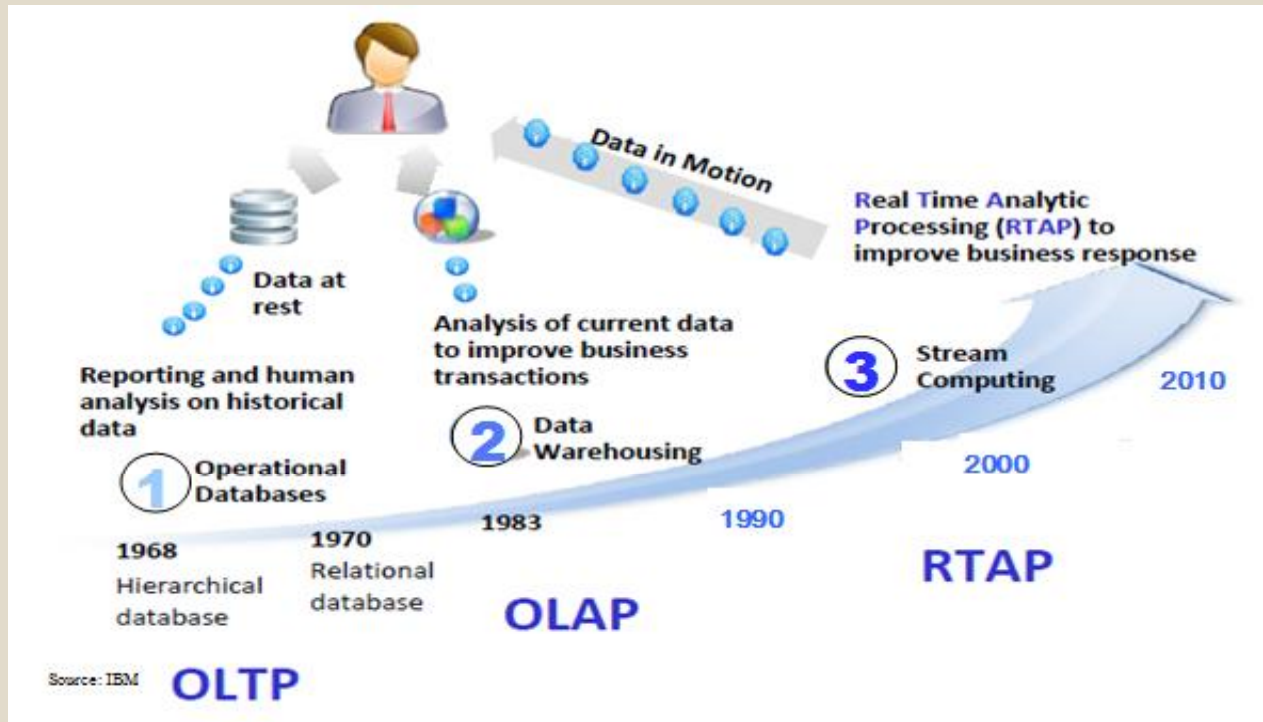
Development of Big Data



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Harnessing Big Data



Source: Big Data in Engineering Applications, Jasti Aswini

- **OLTP:** Online Transaction Processing (DBMSs)
- **OLAP:** Online Analytical Processing (Data Warehousing)
- **RTAP:** Real-Time Analytics Processing (Big Data Architecture & technology)

The Model Has Changed...

The Model of Generating/Consuming Data has Changed

Old Model: Few companies are generating data, all others are consuming data

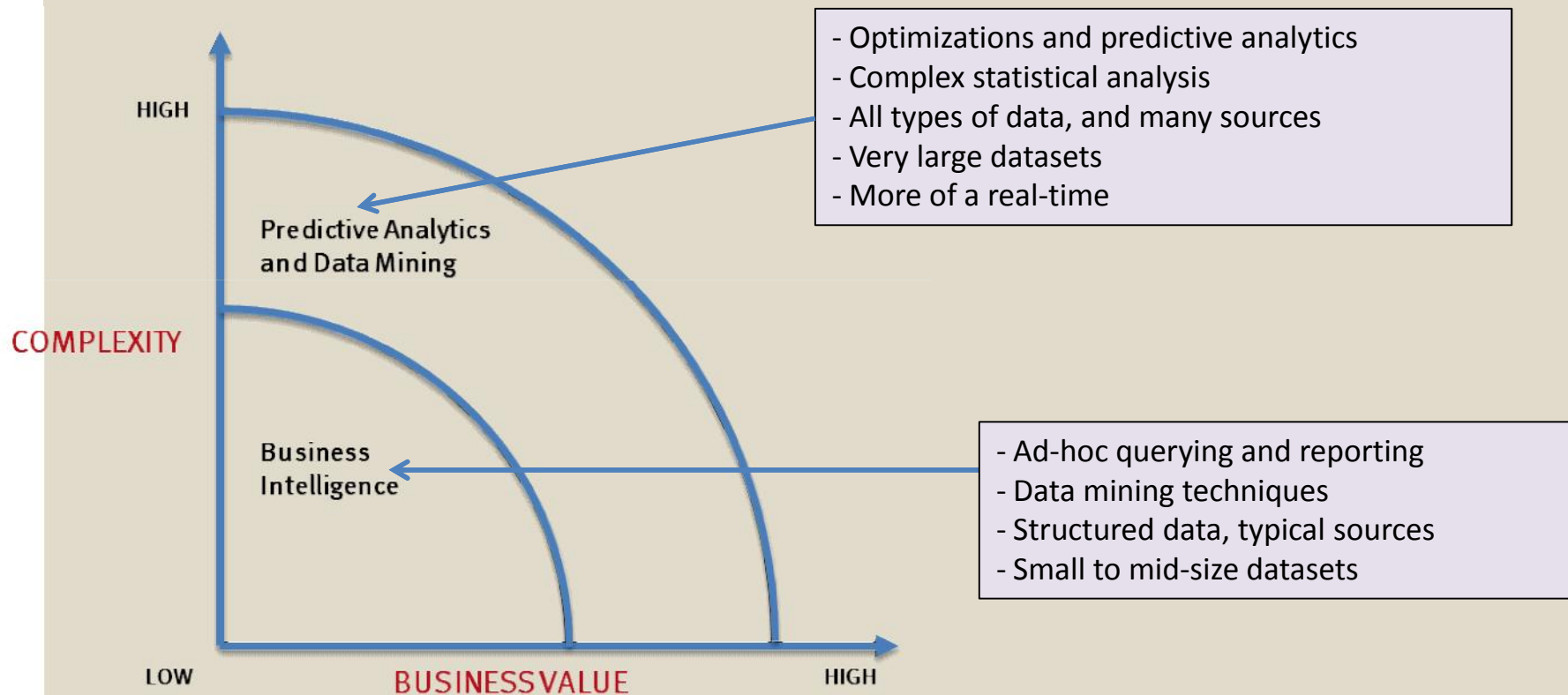


New Model: all of us are generating data, and all of us are consuming data



Source: Introduction to Big Data and Basic Data, Analysis, Ruoming Jin, Kent University

What's driving Big Data



Source: Introduction to Big Data and Basic Data, Analysis, Ruoming Jin, Kent University

Big Data Implementation

*From CAATTs, Open Source Project to Full
Package Commercial*

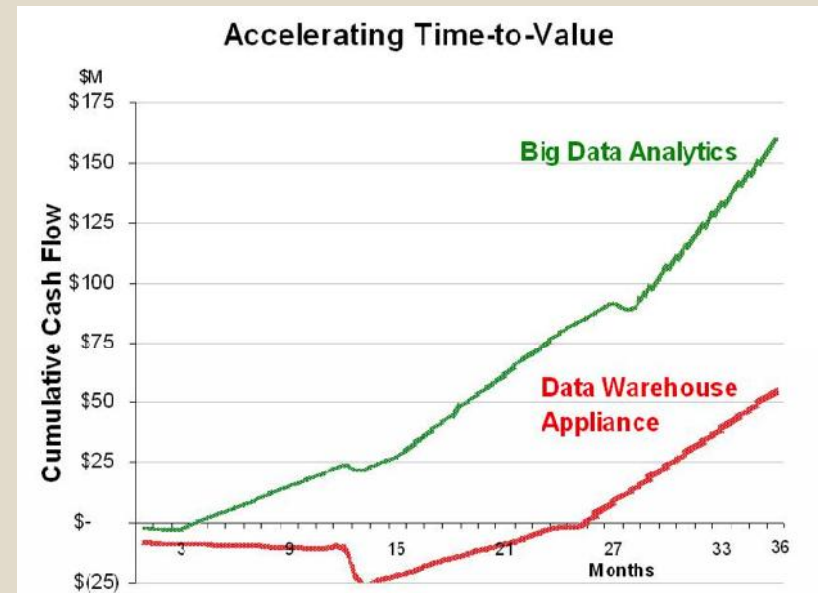


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Big Data Analytics

- Big data is more real-time in nature than traditional DW applications
- Traditional DW architectures (e.g. Exadata, Teradata) are not well-suited for big data apps
- Shared nothing, massively parallel processing, scale out architectures are well-suited for big data apps



Source: Introduction to Big Data and Basic Data, Analysis, Ruoming Jin, Kent University

Big Data & CAATTs

Types of CAATTs	Description
Test Data	Fictitious, auditor-prepared data, which will be processed by the audited systems. The evaluation bases on a comparison between the results of the test data and the auditor's expectations. The processing within the audited systems is a "black box".
Integrated Test Facility	Processing of Test Data in separated areas or modules within the audited system. The results of the internal system controls are visible for the auditor.
Parallel Simulation	Auditor-developed application, which is completely separated from the client's systems. The results of processing real data are compared with the results of the client's systems.
Embedded Audit Module, System Control and Audit Review Files (EAM/ SCARF)	Auditor-developed module which is implemented within a client's system. EAM evaluates real data by predefined criteria while it is processed. Results of EAM evaluations can be written into a SCARF, which is send to the auditors for further examination
Generalized Audit Software	Auditor-developed and self-contained applications, which evaluate extracted real data and analyze them, regarding predefined criteria.
Snapshot Method (tagging and tracing)	Selection and marking of accounting transactions and monitoring their processing within the AIS. After every step, a snapshot is created and analyzed.

Source: Continuous Auditing in Big Data Computing Environments, Andreas Kiezow, University of Osnabruck

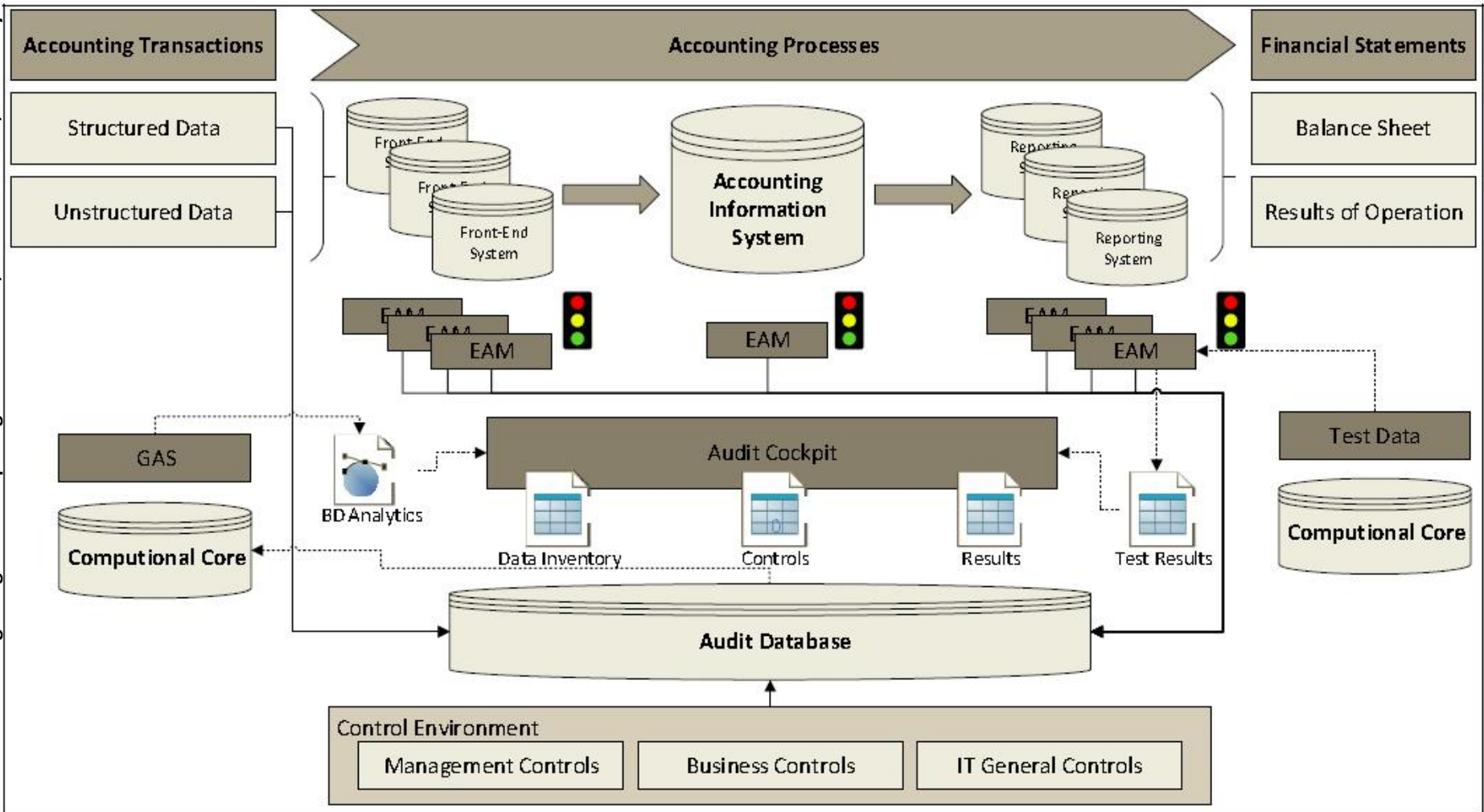
Big Data & CAATTs

Types of CAATTs	Dimensions of Big Data					Overall Applicability
	Volume	Velocity	Variety	Veracity	BDA	
Test Data	●	○	●	●	○	●
ITF	○	○	○	●	○	○
PS	○	○	○	○	○	○
EAM, SCARF	●	●	●	●	●	●
GAS	○	○	○	●	●	●
Snapshot Method	○	○	○	●	○	○

Legend: ○ = low, ● = medium, ● = high

Source: Continuous Auditing in Big Data Computing Environments, Andreas Kiezow, University of Osnabruck

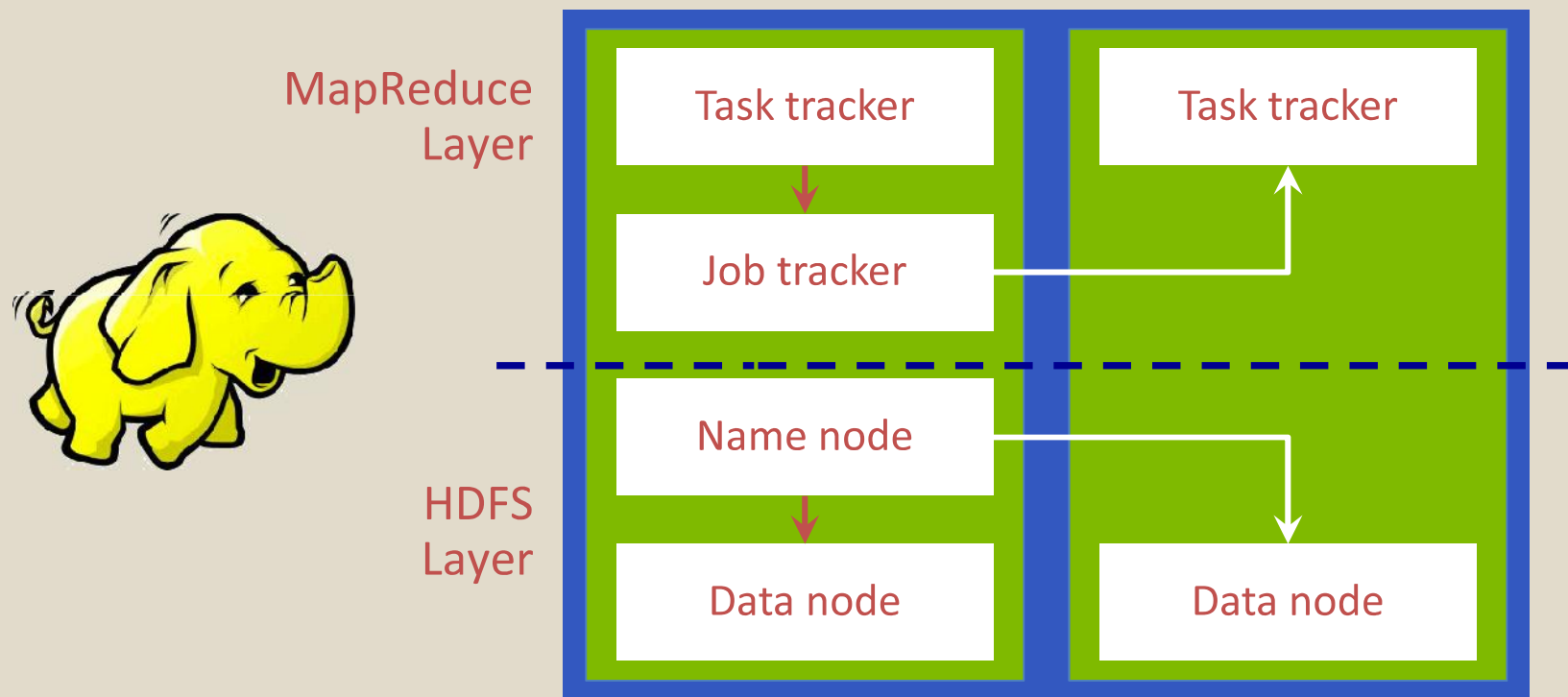
Embedded Audit Module - On-going Monitoring



Cloud based implementation

- **Infrastructure as a Service (IaaS)**
 - Elastic Compute Cloud – EC2 (IaaS)
 - Simple Storage Service – S3 (IaaS)
 - Elastic Block Storage – EBS (IaaS)
- **Platform as a Service**
 - SimpleDB (SDB) (PaaS)
 - Simple Queue Service – SQS (PaaS)
 - CloudFront (S3 based Content Delivery Network – PaaS)

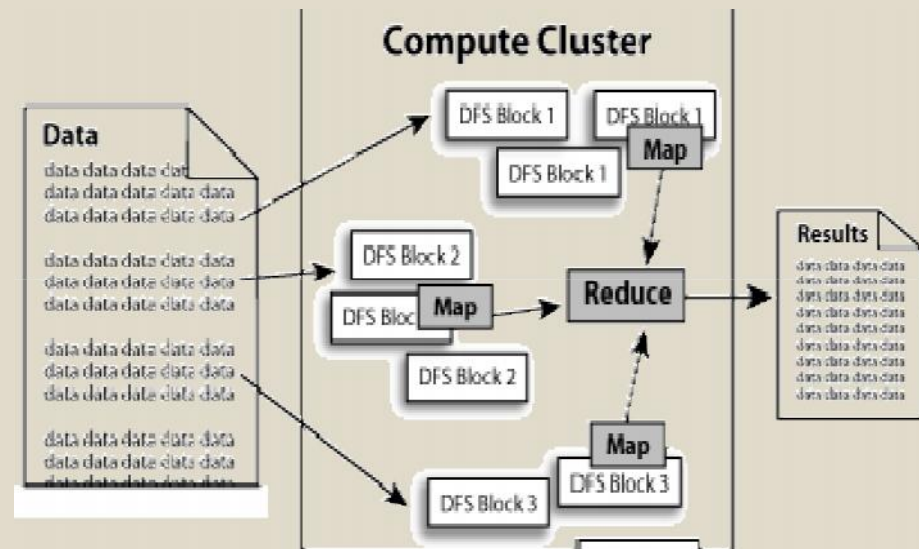
Hadoop: The most visible face of Big Data



Source: Ensuring Compliance of Patient Data with Big Data and BI, Ayad Shammout & Denny Lee, PASS Business Analytics Conference

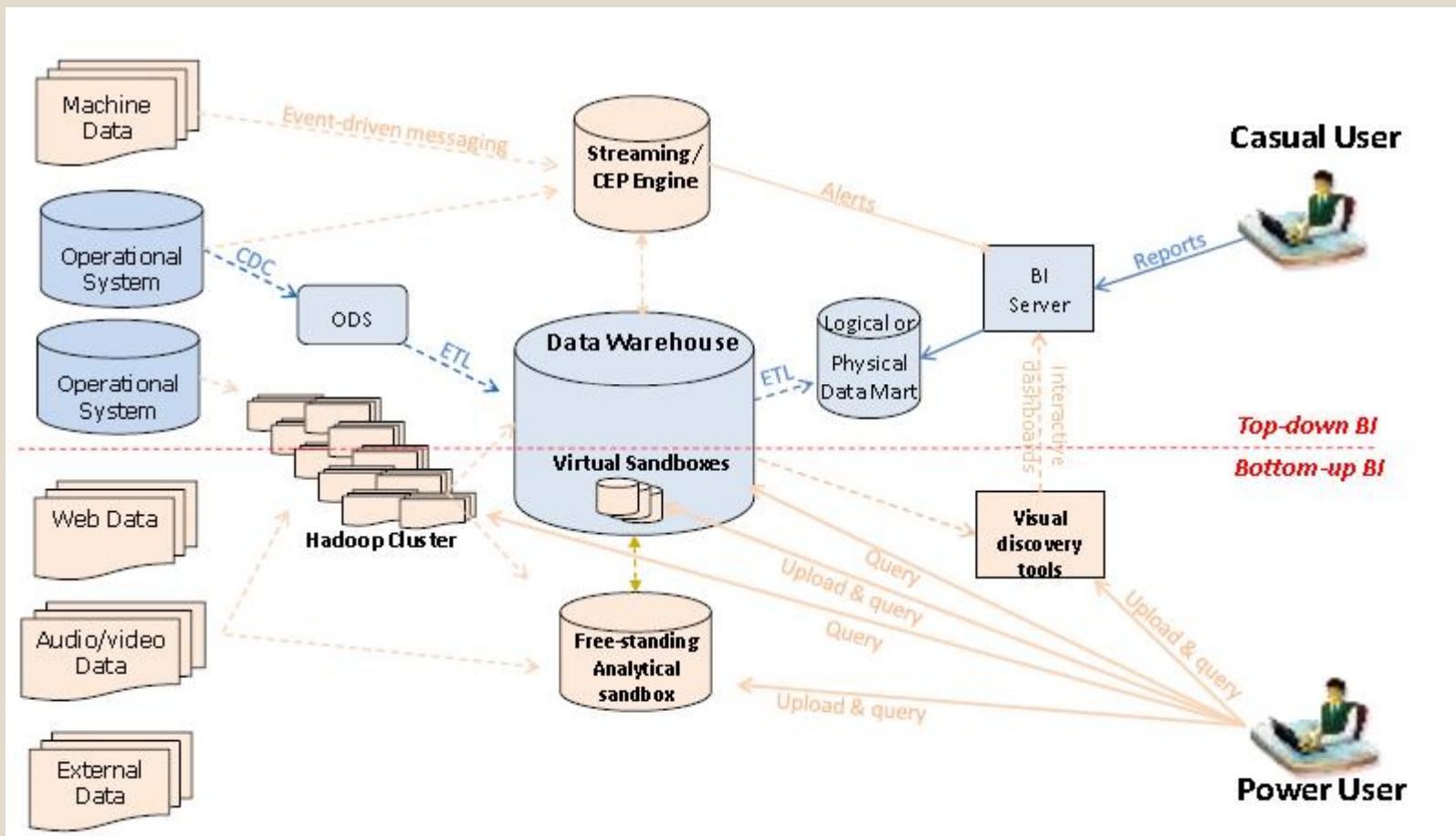
Hadoop: How does it do?

- Hadoop implements Google's MapReduce, using HDFS
- MapReduce divides applications into many small blocks of work/job.
- HDFS creates multiple replicas of data blocks for reliability, placing them on compute nodes around the cluster.
- MapReduce can then process the data where it is located.
- Hadoop 's target is to run on clusters of the order of 10,000-nodes.



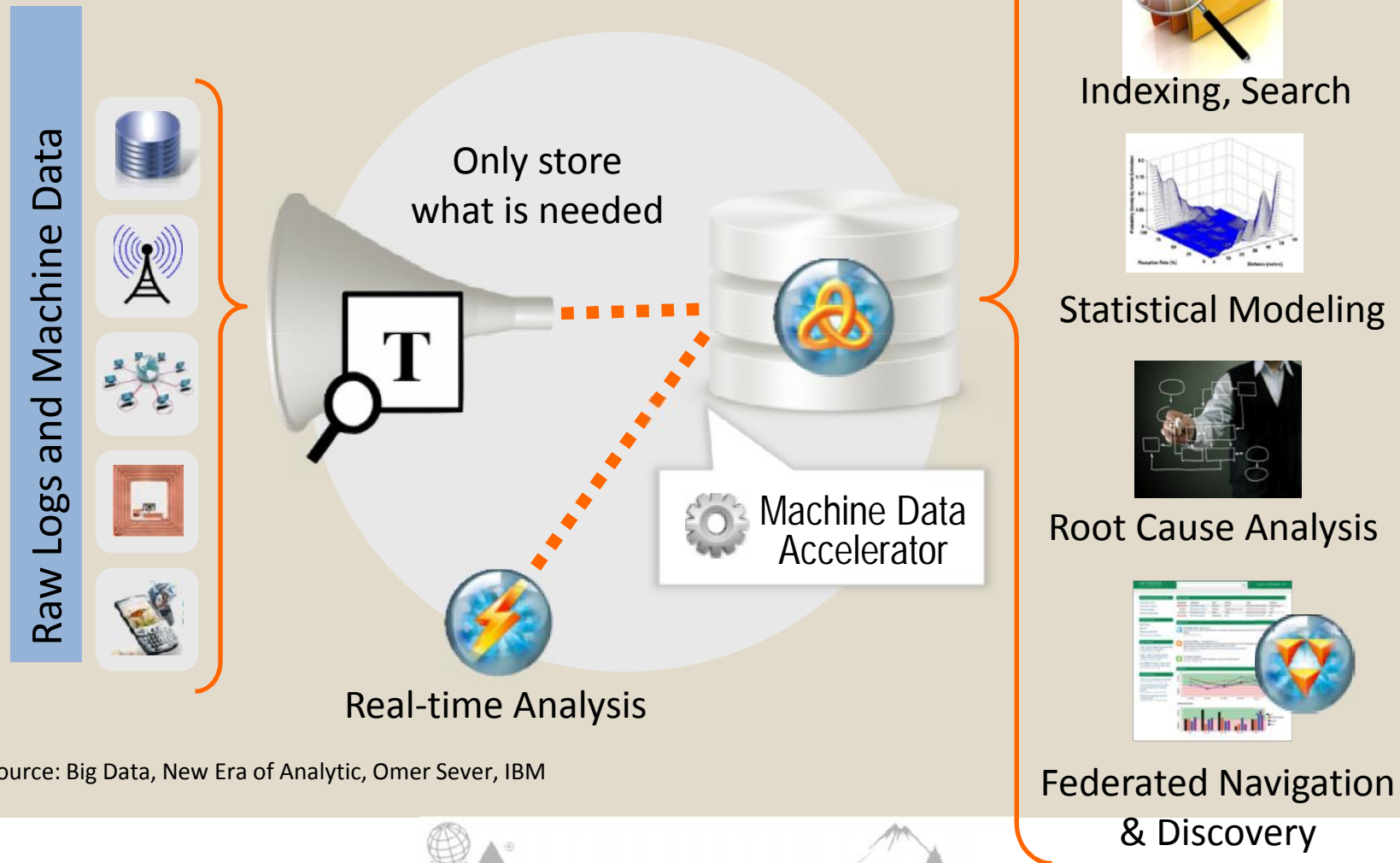
Source: Hadoop: A Software Framework for Data Intensive Computing Applications, Ravi Mukkamala, Old Dominion University

Integrated Analytics Architecture with Hadoop



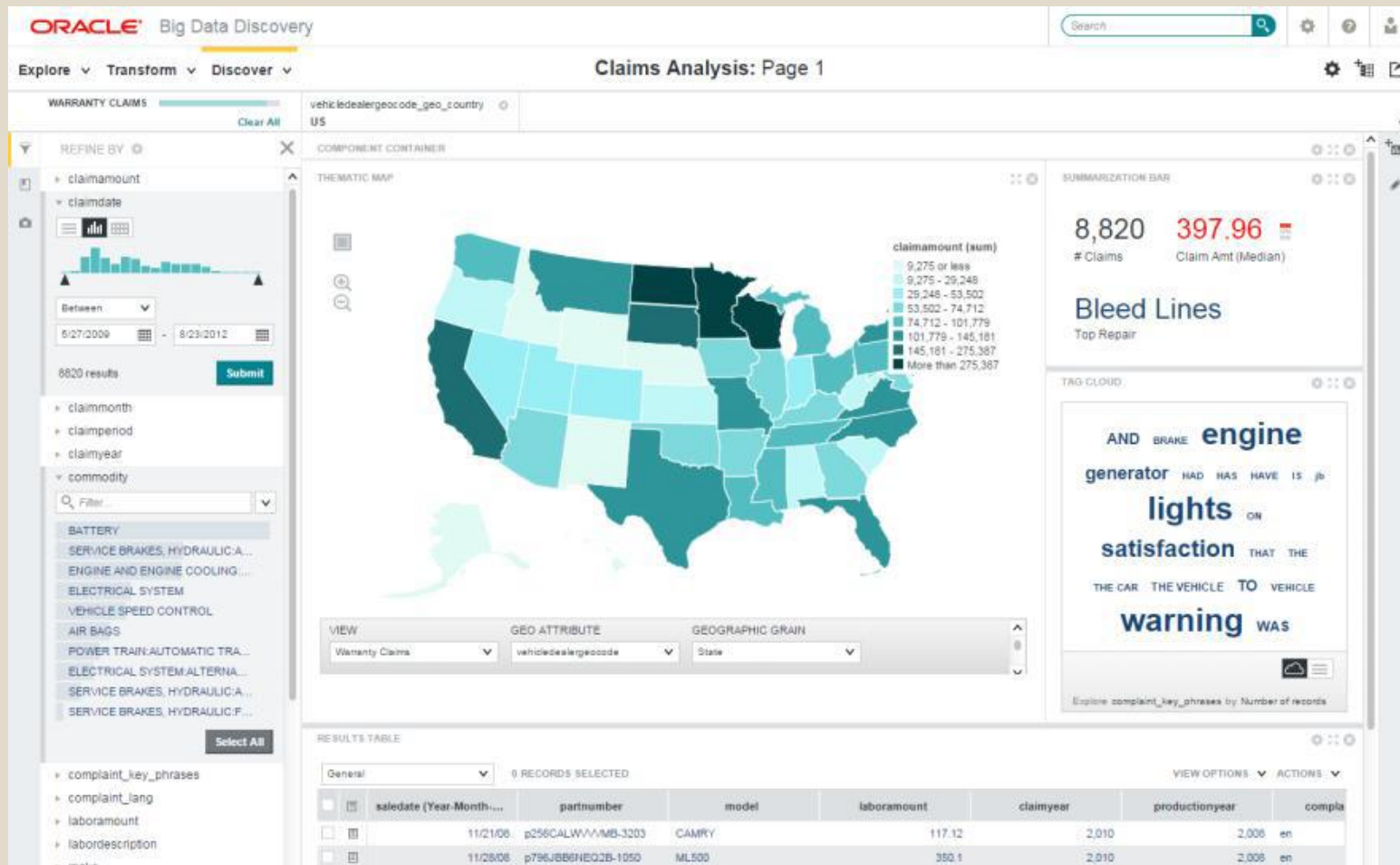
Source: Watson, Tutorial Big Data Business Analytics

Sample Machine Data Refinery with IBM BigInsights®



Source: Big Data, New Era of Analytic, Omer Sever, IBM

Visual Discovery with Oracle Big Data Discovery®



Questions & Discussion



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When do we need to be aware of? *Some of them already in the game...*

- Google processes 20 PB a day (2008)
- Wayback Machine has 3 PB + 100 TB/month (3/2009)
- Facebook has 2.5 PB of user data + 15 TB/day (4/2009)
- eBay has 6.5 PB of user data + 50 TB/day (5/2009)
- CERN's Large Hydron Collider (LHC) generates 15 PB a year



640K ought to be
enough for anybody.

Thank You

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