



# MSBA 6330 Big Data Analytics

Professor De Liu

## Agenda

- Instructor
- Introduction to Big Data
- Syllabus

Course introduction

## INSTRUCTOR

#### About the Instructor

- Dr. De Liu (刘德)
- Originally from Shandong Province, China
- Associate Professor & 3M Fellow in Business Analytics
- Research interests:





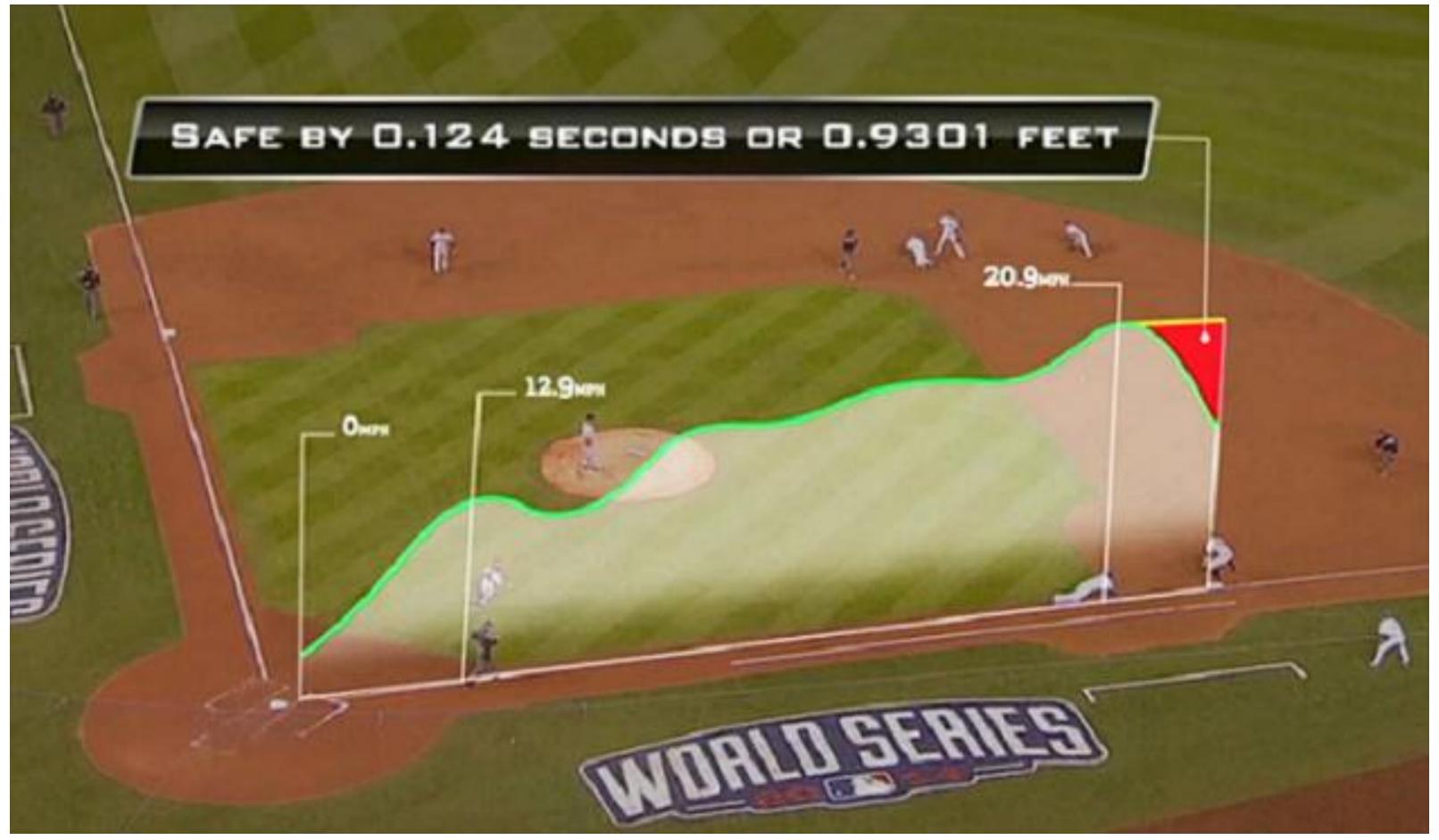
Course introduction

## START FROM A USE CASE

## How big data helps Major League Baseball (MLB)?



## Player Tracking Systems Powered by Big Data



Seconds after the play completed, the player tracker systems showed that if Hosmer had maintained his speed instead of diving to the bag, he would have been safe by about a foot

#### Behind the scene

#### Data capturing

- A Doppler radar system sits behind home plate, sampling ball position 2000 times a second.
- Two stereoscopic imaging devices, sampling positions of players on the field 30 times a second.
- Brief written notes of each play entered by personnel on the field after the action is over
- ~ 30 JSON docs per second per game, 7TB per game.

#### Data transmission

 Seconds after a player is completed, data is transmitted from stadium to cloud servers

#### Behind the scene (cont.)

- Data analytics
  - within milliseconds of data transmission, parallel processing of data began
  - e.g. measuring player speed, forecasting/what if analysis, visualization
- Delivering results
  - Results of analysis are delivered to the Internet destinations
  - e.g. customers' mobile phones and broadcaster's monitors

### More Use Cases of Big Data

- United Healthcare mines customer calls
  - Turn voice data into text, then analyze it with Natural Language Processing (NLP) software to detect consumer attitudes, using Hadoop and NoSQL.
- Medtronic: Using Hadoop + Spark + R to achieve 50+ speed up than SQL Server in analyzing billions of clinical observations to predict heart failure
- NeuroID: Loan fraud detection by analyzing real-time mousemovements data
  - Analyzing mouse trajectory when people fill out loan application forms online to flag fraudulent cases (Hibbeln et al 2014). Data is streamed and analyzed on Amazon cloud.



Course introduction

## BIG DATA: CONCEPT AND OPPORTUNITIES

#### What is big data?

**Big Data**: "large volumes of high velocity, complex, and variable data that require advanced techniques and technologies to enable the capture, storage, distribution, management and analysis of the information

-- from a US Congress Report in 2012

What are the characteristics of big bata?

Volume Velocity Variety

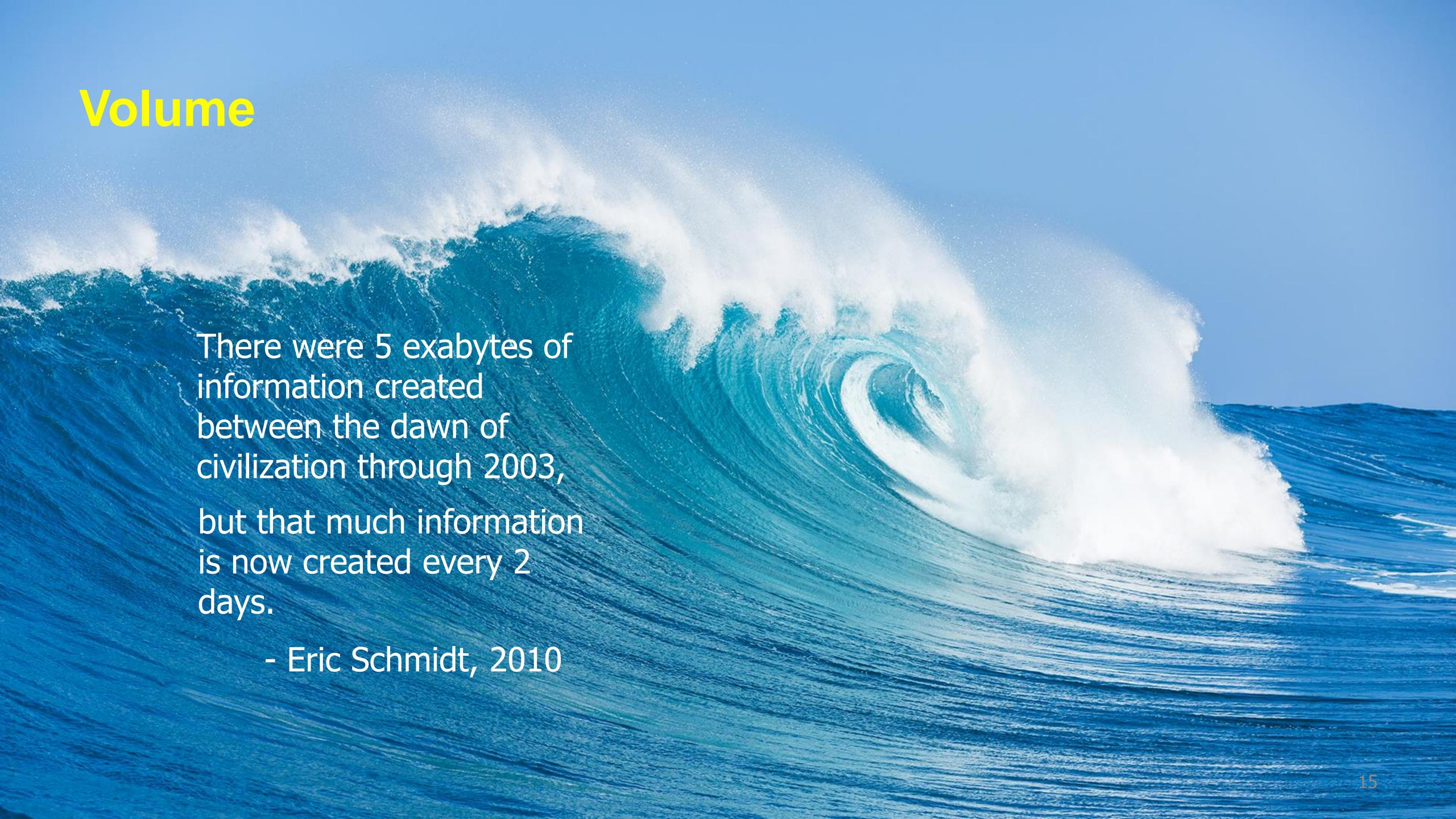
Doug Laney, Gartner (2001)

Veracity Value

Bernad Marr, "Big Data" (2015)

D. Laney, 3D data management: controlling data volume, velocity and variety, META Group Res. Note 6 (2001) 70.

B. Marr, Big Data: Using SMART Big Data, Analytics and Metrics to Make Better Decisions and Improve Performance, John Wiley & Sons, 2015.



#### Some Examples of Big Data

- Every day...
  - Over 2.25 billion shares are traded on the New York Stock Exchange
  - Facebook stores 4.5 billion "Likes"
  - Google processes about 24 petabytes of data
- Every minute...
  - Facebook users share nearly 2.5 million pieces of content
  - Email users send 204,000,000 messages





X 24,576

#### How big is "big"?

- 50% consider datasets between Terabyte and Petabye to be big.
- Whatever is considered "high volume" today will be even higher tomorrow.

Specific units of IEC 60027-2 A.2 and ISO/IEC 80000

IEC prefix		Representations				Customary prefix	
Name	Symbol	Base 2	Base 1024	Value	Base 10	Name	Symbol
kibi	Ki	2 <sup>10</sup>	1024 <sup>1</sup>	1024	≈ 1.02 × 10 <sup>3</sup>	kilo	k <sup>[13]</sup> or K
mebi	Mi	2 <sup>20</sup>	1024 <sup>2</sup>	1 048 576	≈ 1.05 × 10 <sup>6</sup>	mega	М
gibi	Gi	2 <sup>30</sup>	1024 <sup>3</sup>	1 073 741 824	≈ 1.07 × 10 <sup>9</sup>	giga	G
tebi	Ti	2 <sup>40</sup>	1024 <sup>4</sup>	1 099 511 627 776	≈ 1.10 × 10 <sup>12</sup>	tera	Т
pebi	Pi	2 <sup>50</sup>	1024 <sup>5</sup>	1 125 899 906 842 624	≈ 1.13 × 10 <sup>15</sup>	peta	Р
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#### How much data?

- There are about 5,000,000 articles in the English Wikipedia 2015.
   How much data is that
  - if the articles are stored in plain text (compressed)? 11.5 GB
  - If the articles and edit histories are stored in XML text (compressed)?
     100 GB
  - If the articles and edit histories are stored in XML text (uncompressed)?

10 TB



#### Opportunities of Big Volume

"It's not who has the best algorithm that wins, it's who has the most data" (Andrew Ng)

### The Challenge of Big Volume

 Question: How much time does it take to read one Terabyte of data from hard disk into memory?

```
1 TB = 1024 GB = 1024*1024 MB = 1,048,576 MB
1,048,576/100/3600 = 2.91 hour
```

#### Velocity

- Velocity: Data in motion
  - The speed at which data is created processed and analyzed continues to accelerate.
- Examples of high velocity data:
  - Twitter processes 340 million messages / per day
  - Trend Micro processes 6 TB of data/day to identify new security threats
  - Financial institutions process more than 10,000 credit card transactions/second
  - Amazon Web Services fields more than 650,000 requests / second
  - Large Hadron Collider produces 572 terabytes of data per second
  - MLB game generates 2.5 Terabytes / hour.

### Fantastic velocity and where to find them

 Can you think of an every-day example of high velocity data around you?





#### An example of high velocity data

Imagine you work for an e-commerce company that has 5 TB of web log

data per day

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                                                      "GET /blog/test-data/rowgen-development-update-2/feed/ HTTP/1.1" 404 93
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```

- How do you analyze such data?
  - Assume a Gigabit network

Why are we interested in processing log data?

#### Let's Do the Math

Assuming a gigabit network, 1024 Mbps = 1Gbps

Sending 1 GB requires 8 seconds

Sending 1 TB requires 1024\*8 /3600 = 2.27 Hours

Sending 5 TB requires 5\*1024\*8/3600 = 12 Hours!

#### Variety

- Variety: the complexity of multiple data types, including structured, semi-structured and unstructured data.
- They are also different forms:
  - Structured: transactional
  - Semi-structured: sensor data, logs, RFID
  - Unstructured: reviews, images, tweets, audio, video
- Inside or outside of enterprises
  - Internal: transactional systems, server logs, emails, chats, etc
  - External: social media, sensor networks, weather data, geographic data, census, macroeconomic data, third party data providers

## Fantastic variety and where to find them

 Can you think of an every-day example of high variety data around you?



#### Data Variety in Health Care

- 80% of information in healthcare industry is unstructured data, e.g.
  - Outputs from medical devices
  - Doctor's notes
  - Lab results
  - Medical imaging
  - Medical correspondence
  - Clinical data
  - Patient behavior and sentiment data
  - Genomic data



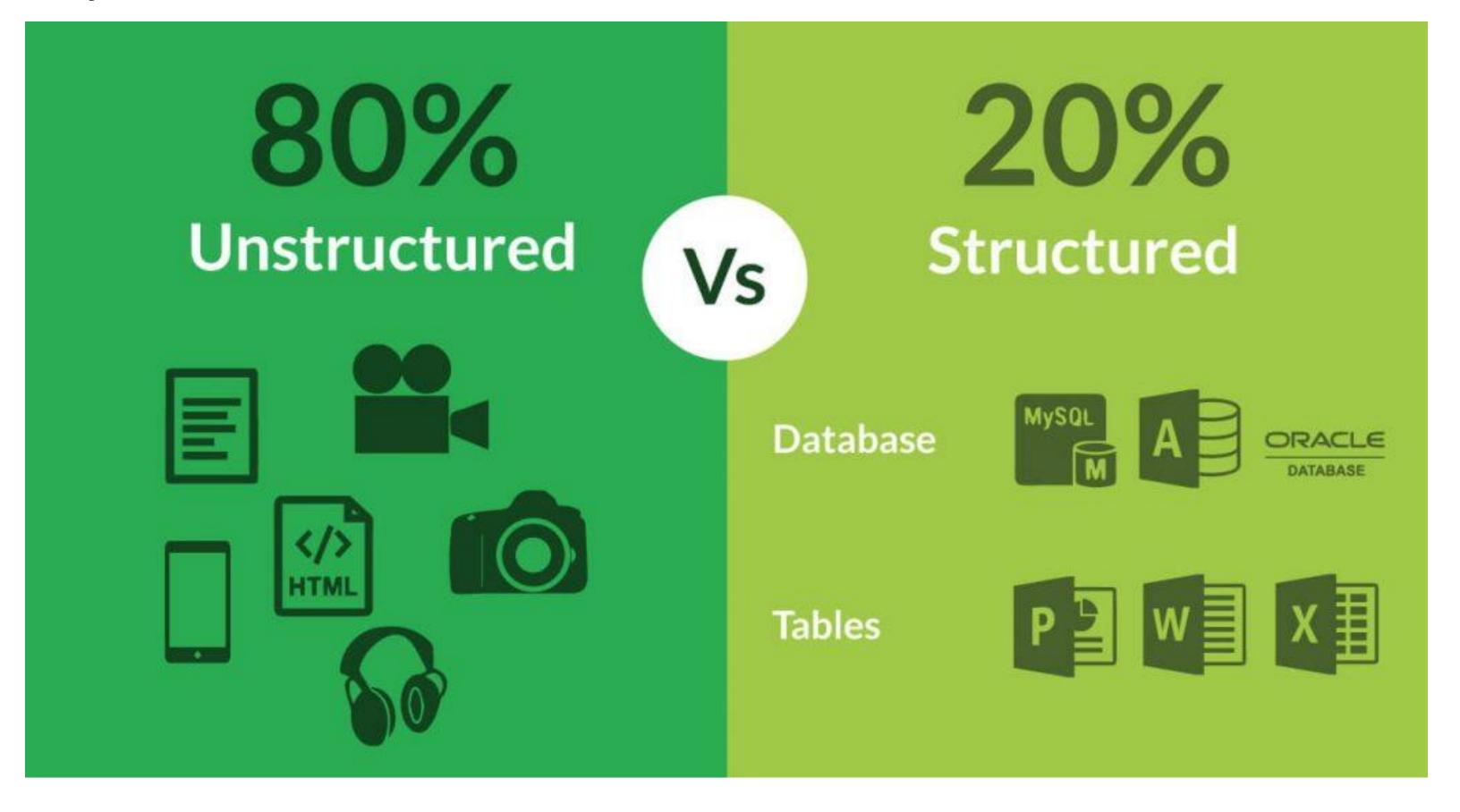
### Opportunity of high variety data

Half of the battle is to get quality signals

 Big data provides a way to capture and analyze novel data sources (e.g. social media, click stream, imagery, sensor data)

#### Challenges associated with variety

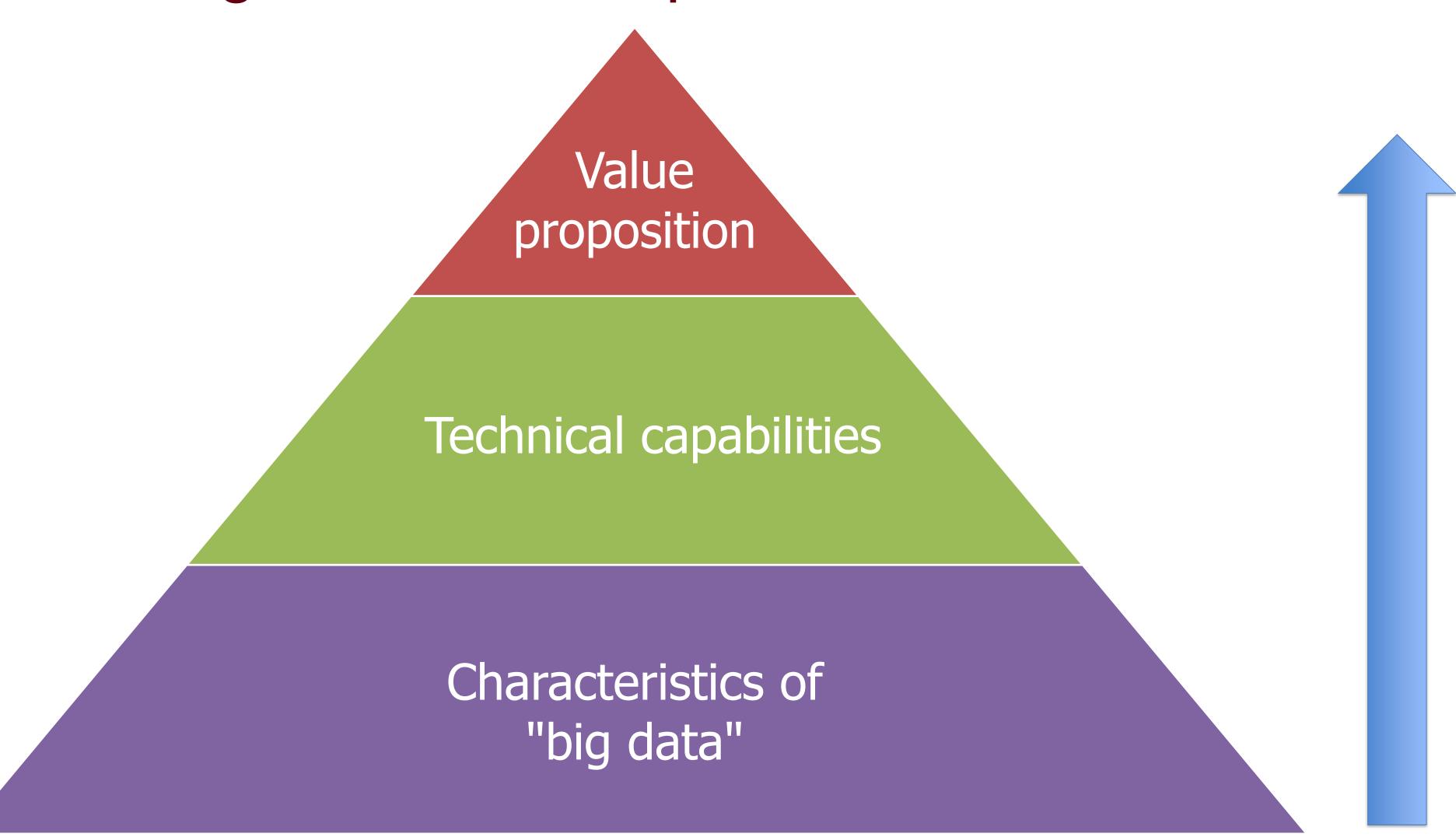
 Many earlier data technologies are not flexible enough to deal with large variety of semi- or un-structured data



#### What about 2 other V's

- Verasity refers to data uncertainty
  - Large volumes of disparate data being ingested at high speed are only useful if the information is correct. Incorrectly indexed data or spelling mistakes could make complete datasets useless and thus the veracity is important.
- Value: Big data has many valuable applications
  - Value is a multifaceted property of big data. As the volume of data grows the incremental value of each data point begins to decrease. As the variety of data available increases, not all the data may aid in product development, sales, or system management. Big data is not the retention of all data; some data needs to remain volatile.

### Organizational Implications of 5 V's



Course introduction

## BIG DATA: HOW ARE COMPANIES USING BIG DATA?



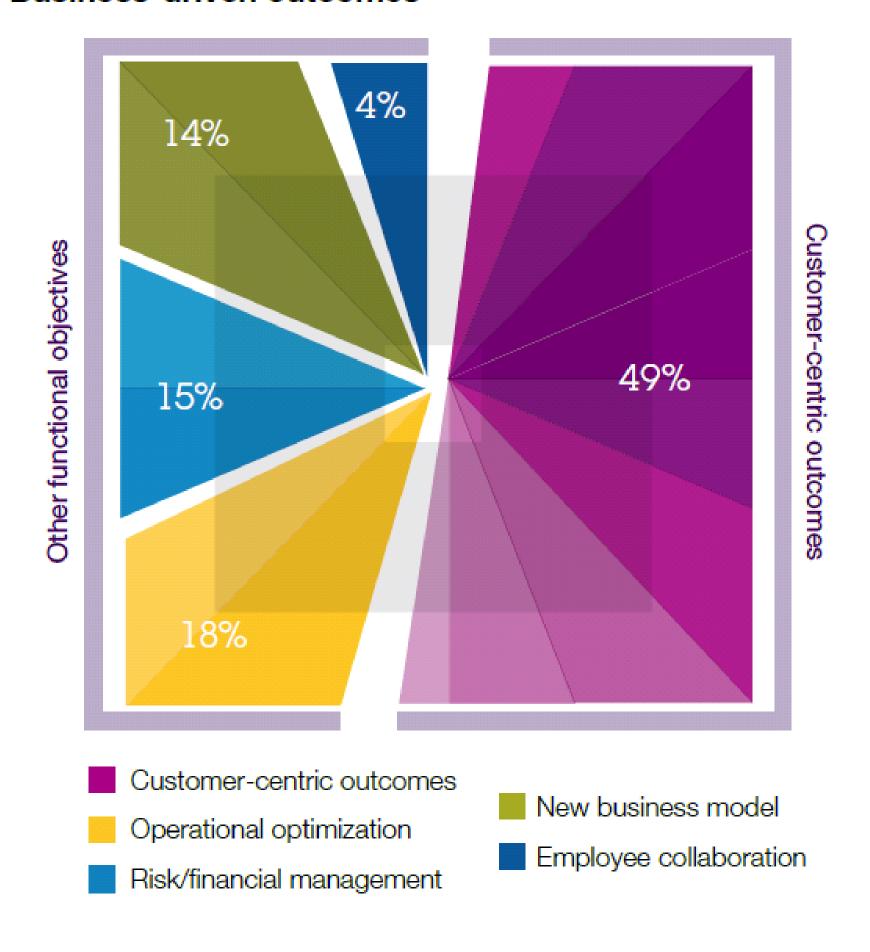
### How Are Companies Using Big Data?

- Big data has reached a point of mainstream adoption within Fortune 1000 firms
  - In 2016, 62.5% have at least one instance of big data in production. In 2018, 97.2% are investing in building or launching big data and Al initiatives
- · Chief Data Officer (CDO) is well established
  - 54% named a CDO in 2016, compared to 12% in 2012.
  - 62.5% named a CDO in 2018, compared to 12% in 2012
  - "Data is essentially the new oil, and the CDO is beginning to be recognized as the linchpin for tackling one of the most important problems in enterprises today: driving value from data"

## How Are Companies Using Big Data? (continue)

Customer-centric activities are the top priority

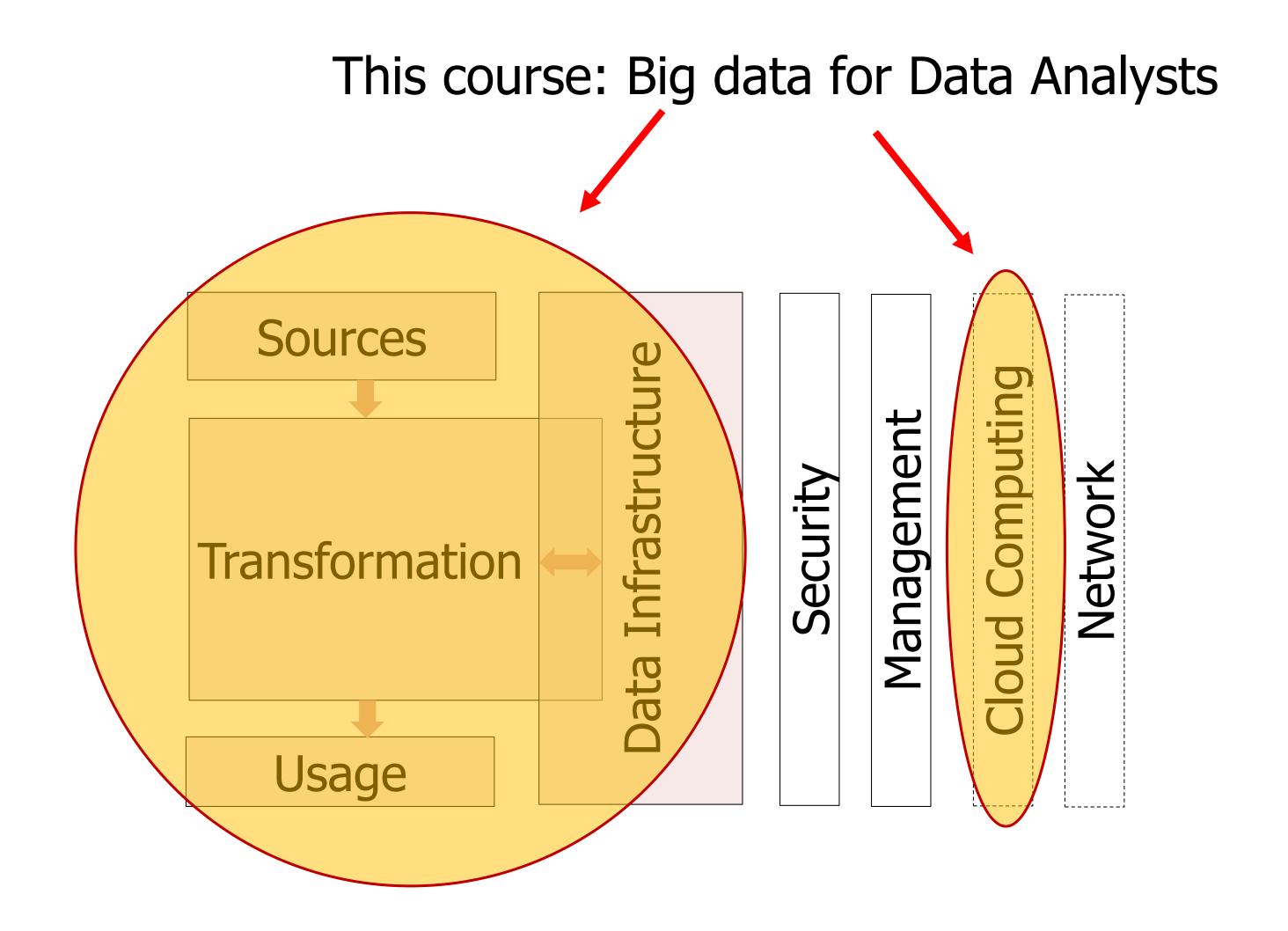
#### **Business-driven outcomes**



Course introduction

## THE FOCUS OF THIS COURSE

#### Where does this course fit?



#### Course Topics

- Learn through hands-on examples
  - Hadoop: MapReduce/HDFS/YARN
  - Data ingestion: Scoop
  - Data analysis / ETL: Hive
  - Spark: Core Spark, Spark SQL,
  - Machine Learning: Spark MLlib
  - Streaming: Spark Streaming
  - Cloud computing: Amazon AWS

#### Course Objectives



- Develop an understanding of the big data ecosystem, the kinds of problems it aims to solve, the characteristics of big data technologies, and their key advantages and disadvantages.
- Develop core competencies in using a variety of essentially big data tools (such as Scoop, Hive, Spark, and Cloud computing) and processes to solve data science problems at scale.