



# Industry 4.0: **Big Data and Advanced Analysis**

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## What is Big Data?

- Big data means
  - > data which is "too big" to be handled by
    - > processing tools, and
    - > conventional databases.
- Big data consists of
  - > structured and
  - > non-structured data

such as web blogs, FB chats, images, news, tweets, comments, etc.

Source: cs.kent.edu: Big data





## **Big Data: Definition**

> "Big data will represent the data of which acquisition speed, data volume or data characterization restricts the capacity of using conventional associated methods to manage successful analysis or the data which can be successfully operated with important horizontal zoom technologies."

[NIST(National Institute of Standards and Technology)]

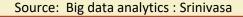
Source: cs.kent.edu: Big data





## **Data Types**

- > Structured data
  - > Data that can be easily organized.
  - > It is stored in relational databases.
  - > It is managed by Structured Query Language (SQL) in databases.
  - > It accounts for only 20% of the total available data today in the world.

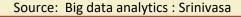






## Data Types(Contd.)

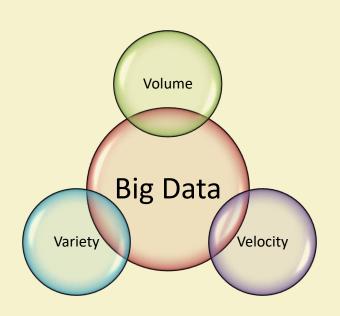
- Unstructured data
  - > Data that do not possess any pre-defined model.
  - > Traditional RDBMSs are unable to process unstructured data.
  - > Enhances the ability to provide better insight to huge datasets.
  - > It accounts for 80% of the total data available today in the world.



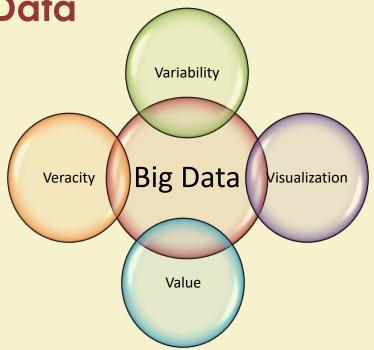




Characteristics of Big Data



➤ There are mainly 3 Vs in Big Data



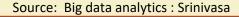
Some authors also include another 4 Vs

Source: Big data analytics: Srinivasa





- > Volume
  - Quantity of created data.
  - Sources of data are added continuously.
  - > Example of volume -
    - ➤ More than 32TB of pictures will be created each night from the Large Synoptic Survey Telescope (LSST).
    - > In every minute, 70 hours of video is uploaded to Youtube.







## > Velocity

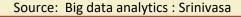
- > Speed of generation of data.
- > Data processing time is decreasing day by day to provide real-time services.
- Older processing technologies can not help to handle high velocity of data.
- > Example of *velocity*
  - > 140 million tweets per day on average (according to a survey conducted in 2011)
  - > NYSE(New York Stock Exchange) measures 1TB of exchange data during every exchanging session.

Source: Big data analytics: Srinivasa





- > Variety
  - > Category of the data.
  - > No restriction over the input data formats.
  - Mostly data are not structured.
  - > Example of variety -
    - ➤ Pure text, images, audio, video, web, GPS data, sensor data, SMS, documents, PDFs, flash etc.

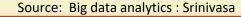






## Variability

- Variability is different from variety.
- > Data whose meaning is constantly changing.
- > Such data appear as an indecipherable mass without structure.
- > Example:
  - Language processing, Hashtags, Geo-spatial data, Multimedia, Sensor events.







- Veracity
  - > Veracity indicates biasness in the data, unusualness and noise in data.
  - > It is important in programs which involve automated decision-making.
  - ➤ It is also important for feeding the data into an unsupervised machine learning algorithm.
- Veracity deals about the data understandability, not just the data quality.

Source: Big data analytics: Srinivasa





#### Visualization

- > Data can be in form of pictures or in form of a graphical format.
- > Visualization provides the power to decision makers to see visually.
- > It is helpful to identify new patterns.

#### > Value

- > It means extracting useful business information from scattered data.
- > Simple to access and provides quality investigation that empowers informed decisions.

Source: Big data analytics: Srinivasa





## **Data Sources**

**Enterprise data** 

Online trading & data analysis

Production and inventory data

Sales and other financial data

IoT data

Industrial data

Healthcare data

Agricultural data





### **Data Sources**

Biomedical data



Data from medical clinics

Others

- Computational biology
- Astronomy
- Nuclear research





## **Data Acquisition**

#### Data collection

- ➤ Data sources automatically generate log files or record files to record activities for further analysis.
- ➤ Complex and variety of data collection through mobile devices. E.g. geographical location, 2D barcodes, pictures, videos etc.

#### Data transmission

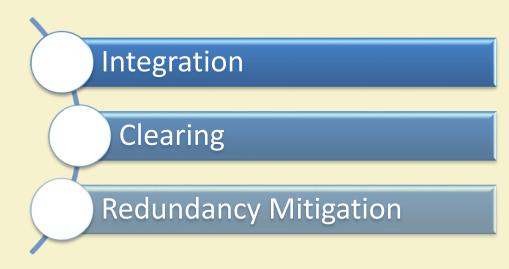
- > Categorized as Inter-DCN transmission and Intra-DCN transmission.
- Collect data and transfer to storage system for further processing and analysis of the data.





# Data Acquisition (Contd.)

- Data pre-processing
  - Pre-processing of data is necessary as collected datasets suffer from noise, redundancy etc.
  - Pre-processing of relational data mainly follows-







## Data Acquisition (Contd.)

- ➤ Integration:
  - > combine data from various sources and
  - > delivers the users a constant data view.
- > Clearing:
  - > spot incorrect, insufficient, or uncooperative data, and
  - > correct or remove such data.
- > Redundancy mitigation:
  - ➤ eliminate data repetition through detection, filter and compression of data to avoid unnecessary transmission.





## Data Storage

## > File system

- ➤ Distributed file systems that store massive data and ensure consistency, accessibility, and fault tolerance of data.
  - > GFS is a distributed file system that supports large-scale file system.
  - ➤ HDFS(Hadoop Distributed File System) is a notable file systems, derived from the open source codes of GFS.

#### Databases

Emergence of non-traditional relational databases (NoSQL) in order to deal with the characteristics that big data possess.





## Why Data Analytics?

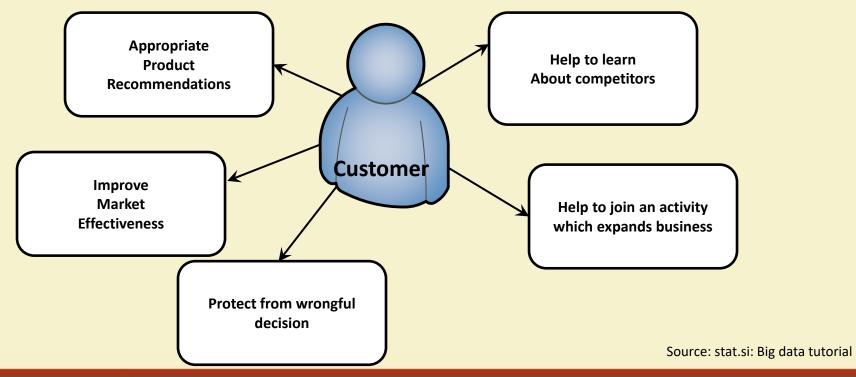
Sensors are very small in sizes. They can be placed anywhere and transfer the data over wireless technology, because of this explosion of data moving to systems from sensors. Some data are irrelevant for systems. How can one know which data are relevant, this requires analysis of the data.

Source: Industry 4.0:The Industrial Internet of Things: Gilchrist





## Why Data Analytics?(Contd.)







# **Big Data Analytics**

- ➤ Big data is different from conventional Data Warehouse (DW) approaches.
- ➤ Big data apps cannot be fit in traditional DW architectures (e.g. Exadata, Teradata).
- ➤ Distributed nothing, mighty parallel performing, scale out frameworks are convenient for big data apps.

Source: Industry 4.0:The Industrial Internet of Things: Gilchrist





# Big Data Analytics for Industry 4.0

- ➤ Industrial Internet require an approach to manage and process data coming from thousand of sensors for precious perceptions .
- To manage and handle the huge data in health services and manufacturing etc. is not new. For example-
  - An event is detected by a sensor and sent to the operational recorder. An operational recorder is a database which stores data. After that this data is optimized by querying such as, what about this hour's production from the norm.

Source: Industry 4.0:The Industrial Internet of Things: Gilchrist





# Big Data Analytics for Industry 4.0 (Contd.)

- > IIoT can be recognized as a big benefactor of Big Data.
- > It needs new technologies to manage vast data.
- > Cloud services are accessible to handle Big Data with no-limit of storage on demand.
- ➤ In IIoT, Hadoop (open source cloud based distributed data storage) is also available for managing the data.





## Cloud-Based Method for Analytics

- Essential features (according to NIST)
  - > On-demand self service
  - ➤ Wide network access
  - ➤ Method grouping
  - > Fast flexibility
  - ➤ Measured service





## Types of Analytics

#### **Prescriptive Analytics**

- -> Best action?
- -> Should we try this?

#### **Predictive Analytics**

- ->What next?
- ->Pattern?

#### **Descriptive Analytics**

- ->When, where?
- ->What happened?

Source: Industry 4.0:The Industrial Internet of Things: Gilchrist





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# Thank You!!



