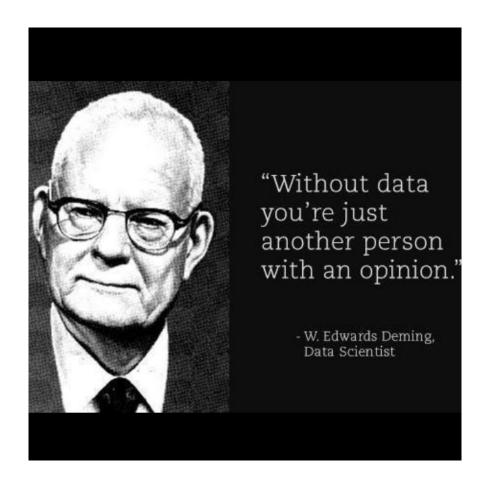


NIST Concepts and Definitions NIST.SP.1500-{1,2}





- What attributes define Big Data solutions?
- How is Big Data different from traditional data environments and related applications?
- What are the essential characteristics of Big Data environments?
- How do these environments integrate with currently deployed architectures?
- What are the central scientific, technological, and standardization challenges that need to be addressed to accelerate the deployment of robust Big Data solutions?



- Big Data is a term often used to describe the large amount of data in the networked, digitized, sensor-laden, information-driven world...
- ...data can overwhelm traditional technical approaches, and the growth of data is outpacing scientific and technological advances in data analytics



- Big Data also refers to the inability of traditional data architectures to efficiently handle the new datasets. Some characteristics of Big Data that force new architectures are:
- Volume (i.e., the size of the dataset);
- Variety (i.e., data from multiple repositories, domains, or types);
- Velocity (i.e., rate of flow); and
- Variability (i.e., the change in other characteristics).



- Big Data characteristics dictate the overall design of a Big Data system, resulting in different data system architectures which must exploit
 - extreme parallelism
 - ubiquitous caching & locality
 - state-of-the-art service/device speed



- The growth rates for data volumes, speeds, and complexity are outpacing scientific and technological advances in data analytics, management, transport, and data user spheres.
- The growth rates of data volumes are considered faster than Moore's Law, with data volumes more than doubling every eighteen months.
- One significant shift is in the amount of unstructured data.
 Historically, structured data has typically been the focus of most enterprise analytics, and has been handled through the use of the relational data model
- Non-relational models, frequently referred to as NoSQL, refer to logical data models that do not follow relational algebra for the storage and manipulation of data



- Data science combines various technologies, techniques, and theories from various fields, mostly related to computer science and statistics, to obtain actionable knowledge from data
- Data-intensive science, shortened to data science, refers to the conduct of data analysis as an <u>empirical science</u>, learning directly from data itself...open-ended analysis <u>often without</u> <u>preconceived hypotheses</u> (sometimes referred to as <u>discovery</u> or <u>data exploration</u>)



 A data scientist is a practitioner who has sufficient knowledge in the overlapping regimes of business needs, domain knowledge, analytical skills, and software and systems engineering to manage the end-toend data processes in the data life cycle



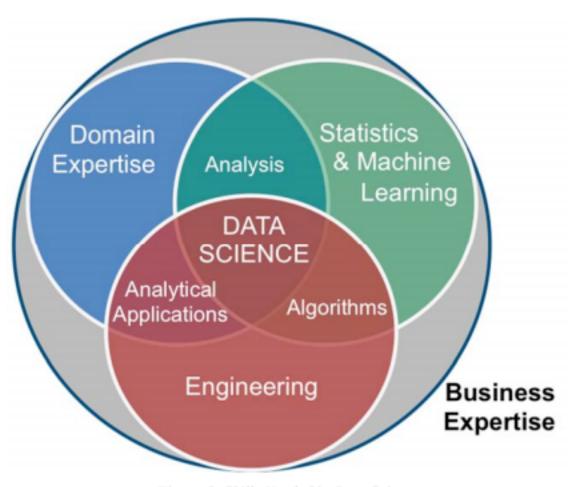


Figure 1: Skills Needed in Data Science



- The data life cycle consists of the following four stages:
- **Collection**: This stage gathers and stores data in its original form (i.e., raw data.).
 - [we've done a lot of collecting...the rest, not so much!]
- Preparation: This stage involves the collection of processes that convert raw data into cleansed, organized information.
- Analysis: This stage involves the techniques that produce synthesized knowledge from organized information.
- Action: This stage involves processes that use the synthesized knowledge to generate value for the enterprise.



- NIST.SP.1500-1.pdf:
 - Review Appendix A: Terms & Definitions
 - Review Appendix B: Abbreviations & Acronyms



How to talk about Big Data to different audiences

- For managers, the terms will distinguish the categorization of techniques needed to understand this changing field.
- For procurement officers, it will provide the framework for discussing organizational needs and distinguishing among offered approaches.
- For marketers, it will provide the means to promote Big Data solutions and innovations.
- For the technical community, it will provide a common language to better differentiate Big Data's specific HW/SW offerings.



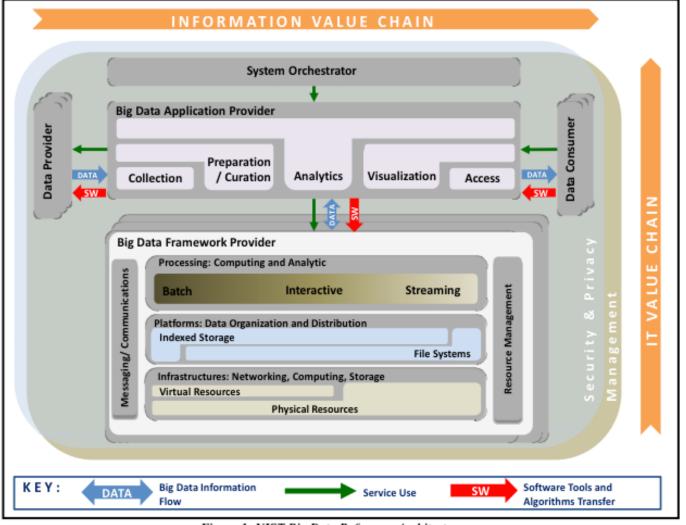


Figure 1: NIST Big Data Reference Architecture



System Orchestrator actors:

- Business Leadership
- Consultants
- Data Scientists
- > Information Architects
- Software Architects
- Security Architects
- Privacy Architects
- Network Architects

Data Provider actors:

- Enterprises
- Public Agencies
- Researchers and Scientists
- Search Engines
- > Web, FTP and Other Applications
- > Network Operators
- > End Users

Security and Privacy Fabric actors:

- Corporate Security Officer
- Security Specialist

Reference Architecture Taxonomy

Big Data Application Provider actors:

- Application Specialists
- Platform Specialists
- Consultants

Big Data Framework Provider actors:

- In-house Clusters
- > Data Centers
- Cloud Providers

Data Consumer actors:

- End Users
- Researchers
- Applications
- Systems

Management Fabric actors:

- > In-house Staff
- > Data Center Management
- Cloud Providers

Figure 2: Roles and a Sampling of Actors in the NBDRA Taxonomy



- NIST.SP.1500-2.pdf
 - Review Appendix B: Abbreviations & Acronyms