#### CPSC 436C



#### Cloud Computing for Data Science

**Big Data** 

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#### Last Week's Review

- Virtualization
- Virtualization types
- VM categories
- Partitioning
- VM Live migration
- How to launch a VM?



# Today's Topics

- Big data definition
- Big data properties
- Big data sources
- Big data analytics stack



"THAT'S your Ark for the Big Data flood? Noah, you will need a lot more storage space!"

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small data

#### ience – Big Data.

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 Big Data refers to datasets and flows large enough that has outpaced our capability to store, process, analyze, and understand.

## Big Data





# Four Attributes of Big Data



- Volume: data size
- Velocity: data generation rate
- Variety: data heterogeneity
- Veracity: data quality





#### Volume

- More data than fits in a computer's RAM
- More data than fits on a single hard drive
- Facebook's 2+billion users



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# Velocity

- Rapid generation of new data
- Nearly 200 million emails are sent each minute of each day
- Nearly 5 billion videos are watched on YouTube every day





# Variety

- Data in many formats
- Videos, photos, audio
- GPS coordinates
- Social network connections



# Veracity



- Data reliability and trustworthiness
- Important to make informed decisions or draw meaningful insights.
- Data quality processes, data validation procedures, and data governance practices are required to maintain and improve the accuracy and trustworthiness of their data.



#### Where does big data come from?

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# **Big Data Market Driving Factors**

- Social Media
  - Social media begets more social media
  - Posts get liked
  - Images get tags
  - Followers share content





# **Big Data Market Driving Factors**

- Internet of Things (IoT)
  - IoT sensors
  - Smart homes
  - Smart grids
  - Self-driving cars



 More than 65 billion devices were connected to the Internet by 2010, and this number exceeded 230 billion by 2020.

 $\ast$  "The Internet of Things Is Coming" [John Mahoney et al., 2013]



# How to store and process big data?

#### Scale Up vs. Scale Out



► Scale up or scale vertically: adding resources to a single node in a system.

► Scale **out** or scale **horizontally**: adding **more nodes** to a system.







#### Scale Up vs. Scale Out

- Scale up: more expensive than scaling out.
- Scale out: more challenging for fault tolerance and software development.

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#### Taxonomy of Parallel Architectures



DeWitt, D. and Gray, J. "Parallel database systems: the future of high performance database systems". ACM Communications, 35(6), 85-98, 1992.



# **Big Data Tools and Frameworks**

• Two main types of tools: HBASE Data Store Graph **StratoSphere**  Data Processing Above the Clouds Storm Grap distributed stream computing platform HIVE cassan



#### **Big Data Analytics Stack**





# Big Data – Storage (File systems)

- Traditional file-systems are not well-designed for large-scale data processing systems.
- Efficiency has a higher priority than other features, e.g., directory service.
- Massive size of data tends to store it across multiple machines in a distributed way.
- ► HDFS/GFS, Amazon S3, ...



#### Big Data - Database



- Relational Databases Management Systems (RDMS) were not designed to be distributed.
- NoSQL databases relax one or more of the ACID properties:
  - BASE (Basically Available, Soft state, Eventually consistent).
- Different data models: key/value, column-family, graph, document.
- NoSQL database exeamples: Hbase/BigTable, Dynamo, Scalaris, Cassandra, MongoDB, Voldemort, Riak, Neo4J, ...



# Big Data – Resource Management



- Different frameworks require different computing resources.
- Large organizations need the ability to share data and resources between multiple frameworks.
- Resource management share resources in a cluster between multiple frameworks while providing resource isolation.
- Mesos, YARN, Borg, ...



# **Big Data- Execution Engines**



- Scalable and fault tolerance parallel data processing on clusters of unreliable machines.
- Data-parallel programming model for clusters of commodity machines.
- MapReduce, Spark, Stratosphere, Dryad, Hyracks, ...





# Big Data – Query/Scripting Languages

- Low-level programming of execution engines, e.g., MapReduce, is not easy for end users.
- Need high-level language to improve the query capabilities of execution engines.
- It translates user-defined functions to low-level API of the execution engines.
- Pig, Hive, Shark, Meteor, DryadLINQ, SCOPE, ...



# **Big Data: Graph Processing**



- Many problems are expressed using graphs: sparse computational dependencies, and multiple iterations to converge.
- Data-parallel frameworks, such as MapReduce, are not ideal for these problems: slow
- Graph processing frameworks are optimized for graph-based problems.
- Pregel, Giraph, GraphX, GraphLab, PowerGraph, GraphChi, ...



# Big Data – Machine Learning



- Implementing and consuming machine learning techniques at scale are difficult tasks for developers and end users.
- There exist platforms that address it by providing scalable machinelearning and data mining libraries.
- Mahout, MLBase, Tensorflow, ...





## Big Data – Stream Processing

- Providing users with fresh and low latency results.
- Database Management Systems (DBMS) vs. Data Stream Management Systems (DSMS)
- Storm, S4, SEEP, D-Stream, Naiad, ...



# Big Data – Configuration and Synchronization



- A means to synchronize distributed applications accesses to shared resources.
- Allows distributed processes to coordinate with each other.
- Zookeeper, Chubby, …





#### Recap

- Big data definition
- Big data properties
- Big data sources
- Big data analytics stack



#### Next Topic: Data Store

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#### Module 1 Quiz

<u>https://join.iclicker.com/XXOJ</u>

#### **Resource Utilization**



- Compare the resource utilization efficiency of virtual machines (VMs), containers, and serverless functions. In which scenario do you expect the most efficient resource usage, and why?
  - a) VM
  - b) Container
  - c) Serverless
  - d) All of them
  - e) None of them

#### **Isolation and Security**



2. How does containerization technology, such as Docker, achieve process isolation compared to virtual machines and serverless? What one is more secure?

- a) VM
- b) Container
- c) Serverless
- d) All of them
- e) None of them

#### Portability



3. Compare the concept of portability of virtual machines in contrast to containers and serverless functions. What service does bring portability to cloud deployments?

- a) VM
- b) Container
- c) Serverless
- d) All of them
- e) None of them

# Scaling and Cost



4. Compare how VMs, containers, and serverless computing handle auto-scaling based on workload demand. Which approach is likely to be more cost-effective for a highly variable workload?

- a) VM
- b) Container
- c) Serverless
- d) All of them
- e) None of them

## **Cold Starts and Latency**



5. Which computing service has the minimum "cold starts" and latency when responding to requests?

- a) VM
- b) Container
- c) Serverless
- d) All of them
- e) None of them