

## Cloud Infrastructures for Big Data Platforms

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# This is not a cloud computing lecture but how cloud computing is important for big data platforms

### Learning objectives

- Understand key cloud technologies
- Understand how cloud technologies empower big data platforms
- Understand cloud technologies enable us to acquire, utilize and manage resources for big data platforms

#### **Service Model**

#### Services offer well-defined interfaces for

- o access resources: data, things, machines, and people
- provide functions: ingestion, computation, sensing, analytics, inferences, etc.
- offer diverse service level agreements (SLAs) for different types of business models (e.g.,pay-per-use and subscription)

#### Services are

- characterized by scalability, reliability, elasticity, etc.
- provisioned in distributed systems of IoT, edge, cloud and HPC infrastructures, possibly a combination of different types of systems



#### Virtualization

#### Virtualization

- abstracts low-level compute, data and network resources to create *virtual version* of these resources
- virtualization software creates and manages "virtual resources" isolated from physical resources

#### Virtualization is a powerful concept

 we can apply virtualization techniques virtually for everything!



#### Virtualizing physical resources

File Ingestion Service

File Storage

Operating System (e.g. Ubuntu)

Physical resources (e.g. 4 dual-core CPUs +8GB RAM) File Ingestion Service

File Storage

Operating System (e.g. Windows)

Virtualization layer/resources (e.g.,1 dual-core CPU, 2 GB RAM)

Operating System (e.g. Ubuntu)

Physical resources (e.g. 4-core CPU +8GB RAM)



## Main types of virtualization of infrastructures

#### Compute resource virtualization

- o compute resources: CPU/GPU, memory, I/O, disks, etc.
- "virtual machines"/containers

#### Storage virtualization

- resources: storage devices, hard disks, etc.
- o for usage and management of data storage

#### Network Function virtualization

- o network resources: network equipment & functions
- dynamically provision and manage network functions



### **Cloud Computing**

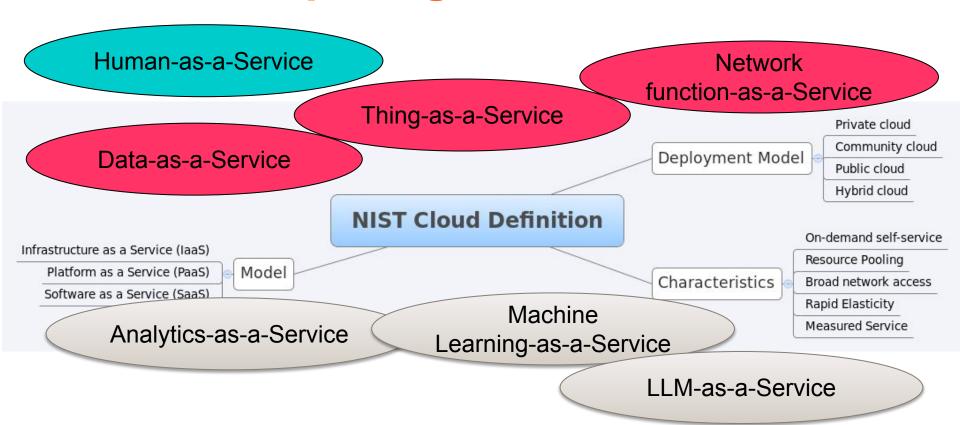
**Original definition from NIST** 

"This cloud model promotes availability and is composed of five essential characteristics, three service models, and four deployment models."

Source: NIST Definition of Cloud Computing v15, <a href="http://csrc.nist.gov/groups/SNS/cloud-computing/cloud-def-v15.doc">http://csrc.nist.gov/groups/SNS/cloud-computing/cloud-def-v15.doc</a>



### **Cloud Computing**





### Cloud computing principles

#### "Cloud"

- not just data centers or public cloud infrastructures
- For big data platforms: we need the "cloud mindset"
  - apply cloud principles for developing and operating big data platforms
  - big data platforms can be in on-premise infrastructures empowered with cloud technologies!
  - part of the big data platforms can be in the edge with similar technologies, such as microservices and Kubernetes
  - both multicloud and hybrid cloud systems can be used used



## Compute resource virtualization technologies

- Physical compute resources for big data platforms
  - o individual physical hosts/servers (CPU, memory, I/O)
  - clusters, supercomputer, and data centers
- At the low-level: two main streams
  - hypervisor/Virtual Machine monitor
    - Virtual machines (VirtualBox, VMWare, Zen, etc.)
  - containterlization
    - Containers (Linux Containers, Docker, Warden Container, OpenVZ, OCI based containers, etc.)



## Virtual infrastructural resources for big data platforms

- For big data platforms: we leverage clusters/infrastructures of VMs/containers
  - computing resources for core services and data
    - e.g., data storage, data ingestion, data processing, and messaging
- On-demand resources for large-scale deployments
  - compute nodes, storage, communication, etc.
  - virtual data centers work like a single distributed system
- On-demand resources for elastic workload
  - e.g., for data ingestion and analytics tasks



### **Example: OpenStack**

IBM

 A Big Data Platform can be built based on OpenStack (or similar)-based data center IBM Cloud OpenStack Services runs on OpenStack Icehouse to provide you with an environment built on the most current open standards.

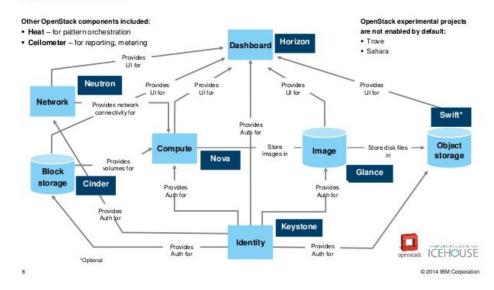


Figure source: http://www.slideshare.net/OpenStack\_Online/ibm-cloud-open-stack-services

#### **Example: Kubernetes**

- Support Docker, rkt, runc, etc.
- Act as a resource orchestration and management for big data platforms

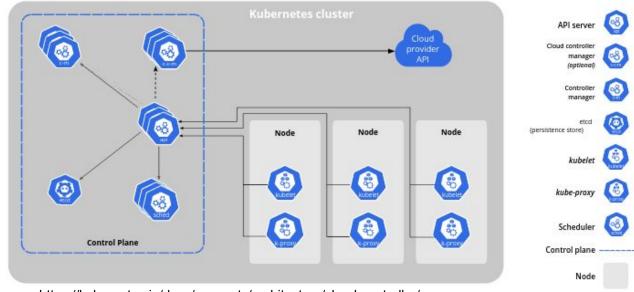


Figure source: https://kubernetes.io/docs/concepts/architecture/cloud-controller/



## **Example: storage virtualization**

#### Low-level storage

- e.g., VMware Virtual Machine File Systems
- High-level, e.g., database
  - MySQL Cluster + auto-sharding

Can be used as the Data Storage layer for Data Lake or Data Services

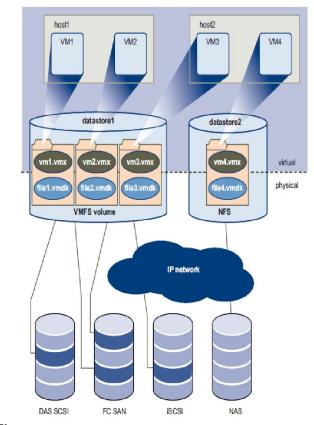


Figure source:

https://www.vmware.com/pdf/vi\_architecture\_wp.pdf



## #1: Enabling managed services

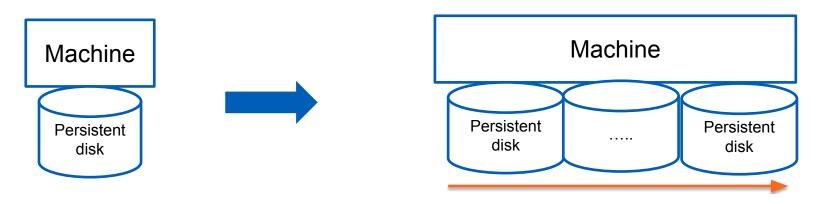
## Many options for resource provisioning in big data platforms

- Managing infrastructural resources for big data platforms must be easier with virtualization
  - different types of storage: block storage and file storage
  - o many CPU/memory configurations: single core to many cores
  - suitable for different workloads
- Different SLA offerings
  - reliability, security, performance, maintainability ...
- Elasticity
- Globalization support: important for many businesses
  - design with multiple clouds and hybrid clouds



### On-demand storage provisioning

Big data requires big storage which can be changed on-demand!



Dynamic numbers and sizes

- Examples of dynamic configurations:
  - Google persistent disks: different types of disks, can have 128 disks and max 257 TiB for a single VM



### Flexible computing capabilities

- Different workloads and programming models in big data platforms need flexible computing resource provisioning
  - Storage, Data Ingestion, and Analysis
- Examples
  - small data ingestion/analysis jobs
    - with pandas, DuckDB, scikit-learn (few cores)
  - large-scale MapReduce/Spark ⇒ clusters of VMs/containers
  - Machine Learning with TensorFlow ⇒ TPU (Tensor Processing Unit)
- Cloud technologies easily enable different computing capability configurations



## Security improvement

- Tenant's service isolation while platforms support multiple tenants
- Virtual private instances for security and performance of tenant's data!

File Storage (Tenant A) File Storage (Tenant B)

**Operating System** 

**Physical Machine** 

Virtual Private
Systems

File Storage (Tenant A)

Operating System

Virtual Private Systems

File Storage (Tenant B)

**Operating System** 

Physical Server



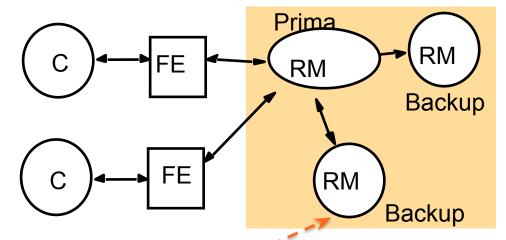
## #2: Achieving fault tolerance, performance and elasticity



### **Easing Replication Management**

## Passive (Primary backup) model:

- FE (Front-end) can interface to a Replication Manager (RM) to serve requests from clients.
- E.g., in MongoDB



Easy to deploy, globalize, manage and replace RM using cloud resources

**Figure source:** Coulouris, Dollimore, Kindberg and Blair, Distributed Systems: Concepts and Design Edn. 5



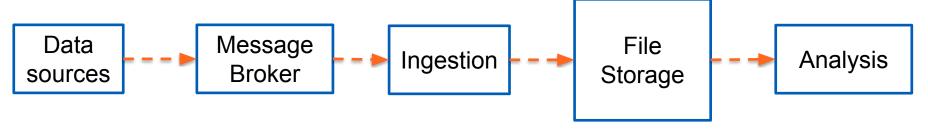
### High availability and performance

#### Cost/optimization

- elasticity, hot deployment, etc.
- cloud bursting with hybrid cloud (combining private/on-premise + public resources)
- edge-cloud continuum
- Improving service performance in incident management
  - e.g., spend time to fix a machine or just quickly relaunch a new one (and fix the old one later)?



## Scaling in every place of the data pipelines



#### Scaling

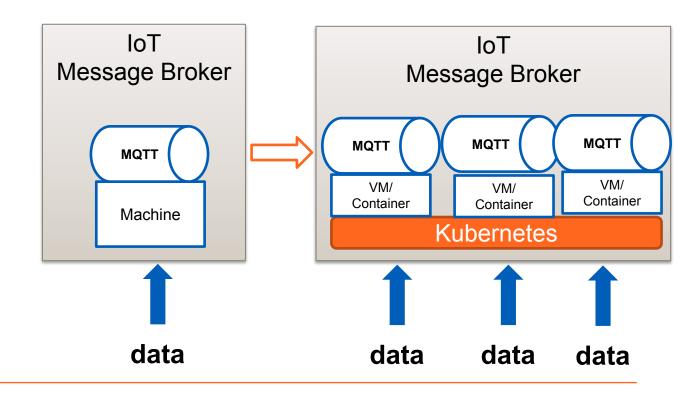
- disk spaces for file storage
- resources for data ingestion
- resources for data analysis





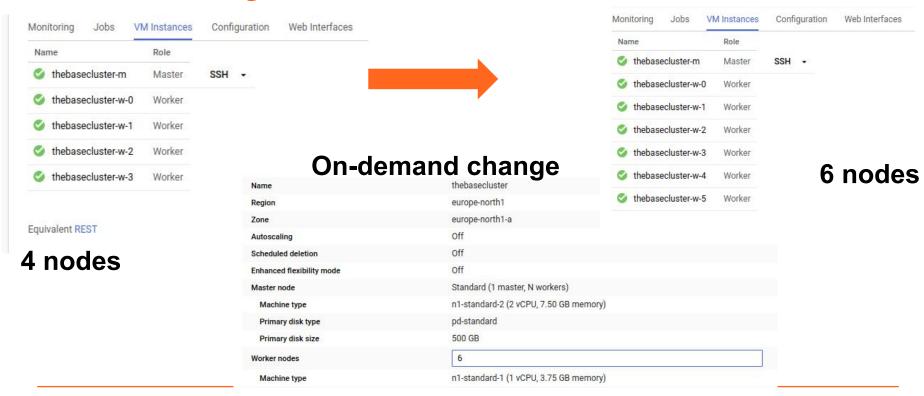
#### Scaling middleware nodes

- Increase the number of brokers when more data arrive
- Provide dedicated brokers on-demand





## Example: scaling compute nodes for data analysis



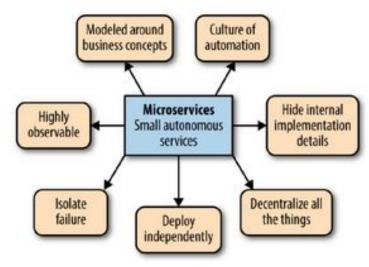


## #3: Living in the world of Microservices & DevOps



#### **Microservices**

- Many components for data storage, data processing and ingestion
  - microservices can be used to design components of big data platforms
  - in particular: services serving data requests and services for storing data in the platform
- Big data platforms provide features for other microservices



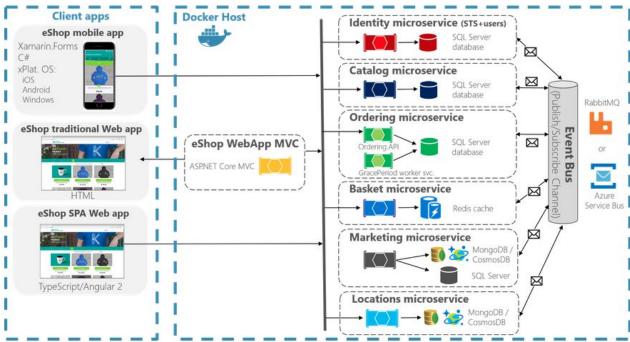
**Figure source:**Sam Newman, Building Microservices, 2015

#### **Examples**

#### eShopOnContainers reference application

(Development environment architecture)

Microservices for both services using big data platforms and components of data platforms



#### Figure source:

https://blogs.msdn.microsoft.com/dotnet/2017/08/02/microservices-and-docker-containers-architecture-patterns-and development-guidance/



### **DevOps**

- Close the gap between development/test environment and real/production environments
- Simplify testing, emulating real environments, etc.

DevOps for big data platforms are part of software systems under DevOps. However, many tasks have to deal with requirements from big data analytics and services!

For example: testing a data pipeline (data intensive test cases) != testing an integrated set of software components



Tools and frameworks and providers for infrastructural resources provisioning and service deployment:

Chef, Vagrant, Terraform, Amazon, Google, Microsoft, OpenStack, OpenShift,

. . .



Provisioning and management of platforms components Runtime Management code Registry build control Big data platforms VM/container images deploy Virtual machines/ configuration container instances/ functions artifacts backup /store Physical resources Google Container Registry Google Artefact Registry AWS EC2 Container Registry



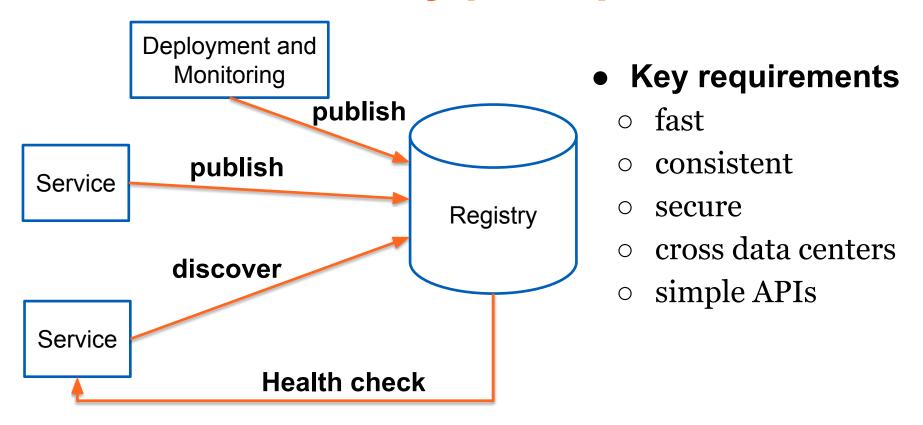
**Azure Container Registry** 

Etc.

### **#4: Service Discovery**



### Service Discovery principle



## Distributed coordination

- A lot of algorithms, etc.
  - Paxos family
- Well-known in the cloud

Notes from the paper: "server replication (SR), log replication (LR), synchronization service (SS), barrier orchestration (BO), service discovery (SD), group membership (GM), leader election (LE), metadata management (MM) and distributed queues (Q)"

Neo4i

ZooKeeper

Usage Patterns SD Project Consensus System SR LR BO GM LE MM Q **GFS** Chubby Borg Chubby/Paxos Kubernetes etcd Megastore Paxos Spanner Paxos Bigtable Chubby Hadoop/HDFS ZooKeeper 1 ZooKeeper **HBase** Hive ZooKeeper Zeus Configerator Cassandra ZooKeeper ZooKeeper Accumulo BookKeeper ZooKeeper Hedwig ZooKeeper Kafka ZooKeeper ZooKeeper Solr Giraph ZooKeeper Hama ZooKeeper Mesos ZooKeeper 1 CoreOS etcd 1 OpenStack ZooKeeper

TABLE 4. PATTERNS OF PAXOS USE IN PROJECTS

Source: Ailidani Ailijiang, Aleksey Charapkoy and Murat Demirbasz, Consensus in the Cloud: Paxos Systems Demystified, http://www.cse.buffalo.edu/tech-reports/2016-02.pdf



### Technology choices: ZooKeeper

- https://zookeeper.apache.org/
- Support service discovery, configuration information and distributed synchronization
- Centralized registry service
- Data is organized into a shared hierarchical name space
  - small data size
- Highly available and reliable



### **ZooKeeper Service**

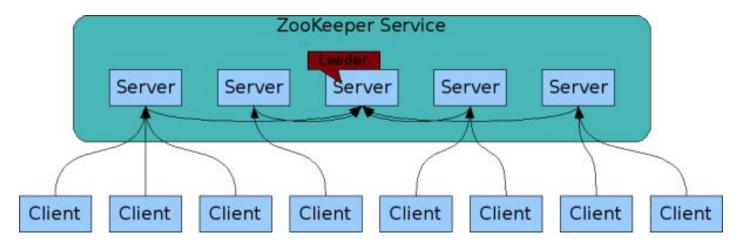
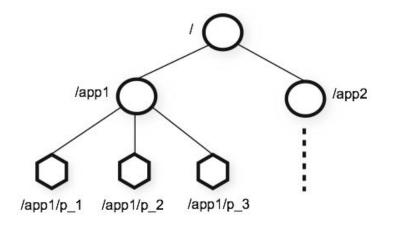


Figure source: https://zookeeper.apache.org/doc/r3.4.10/zookeeperOver.html

#### ZooKeeper data -- znodes

- Data nodes called znodes
- Missing data in a znode ⇒
   problems with the entity that the
   znode represents
- Persistent znode
  - /path deleted only through a delete call
- Ephemeral znode, deleted when
  - the client created it crashed
  - session expired

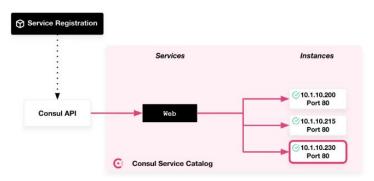


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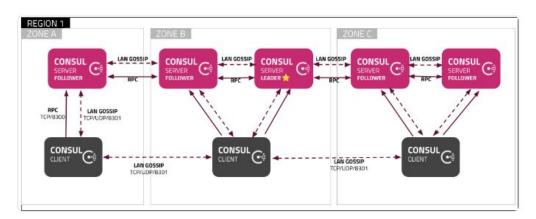
## **Technology choices: Consul**

- https://www.consul.io
- Cross data centers
- End-to-end service discovery
  - include health check



#### Figure source:

https://developer.hashicorp.com/consul/docs/concepts/service-discovery



#### Figure source:

https://developer.hashicorp.com/consul/tutorials/production-deploy/reference-ar chitecture



### Technology choices: etcd

- Consistent, distributed key-value store
- Allow monitor changes of keys/directories
  - enable reactive actions based on changes
- Widely used for
  - service discovery and state/configuration management
  - distributed key locking
  - e.g. in Kubernetes



#### Thanks!

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rdsea.github.io