



Data Integration and Analysis 08 Cloud Computing Fundamentals

Matthias Boehm

Last update: Dec 06, 2019

Graz University of Technology, Austria
Computer Science and Biomedical Engineering
Institute of Interactive Systems and Data Science
BMVIT endowed chair for Data Management





Announcements/Org

#1 Video Recording

Link in TeachCenter & TUbe (lectures will be public)



#2 DIA Projects

- 13 Projects selected (various topics)
- 4 Exercises selected (distributed data deduplication)
- All project discussions last Friday (Nov 29)

Reminder:

Dec 15 for CIDR'20





Course Outline Part B:

Large-Scale Data Management and Analysis

12 Distributed Stream Processing [Jan 24]

13 Distributed Machine Learning Systems [Jan 31]

Compute/ Storage 11 Distributed Data-Parallel Computation [Jan 17]

10 Distributed Data Storage [Jan 10]

Infra

09 Cloud Resource Management and Scheduling [Dec 13]

08 Cloud Computing Fundamentals [Dec 06]





Agenda

- Motivation and Terminology
- Cloud Computing Service Models
- Cloud, Fog, and Edge Computing





Motivation and Terminology





Motivation Cloud Computing

Definition Cloud Computing

- On-demand, remote storage and compute resources, or services
- User: computing as a utility (similar to energy, water, internet services)
- Cloud provider: computation in data centers / multi-tenancy

Service Models

- laaS: Infrastructure as a service (e.g., storage/compute nodes)
- PaaS: Platform as a service (e.g., distributed systems/frameworks)
- SaaS: Software as a Service (e.g., email, databases, office, github)

→ Transforming IT Industry/Landscape

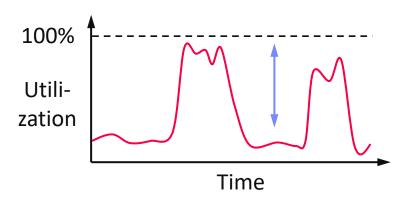
- Since ~2010 increasing move from on-prem to cloud resources
- System software licenses become increasingly irrelevant
- Few cloud providers dominate laaS/PaaS/SaaS markets (w/ 2018 revenue):
 Microsoft Azure Cloud (\$ 32.2B), Amazon AWS (\$ 25.7B), Google Cloud (N/A),
 IBM Cloud (\$ 19.2B), Oracle Cloud (\$ 5.3B), Alibaba Cloud (\$ 2.1B)





Motivation Cloud Computing, cont.

- Argument #1: Pay as you go
 - No upfront cost for infrastructure
 - Variable utilization → over-provisioning
 - Pay per use or acquired resources



Argument #2: Economies of Scale

- Purchasing and managing IT infrastructure at scale → lower cost (applies to both HW resources and IT infrastructure/system experts)
- Focus on scale-out on commodity HW over scale-up → lower cost
- Argument #3: Elasticity
 - Assuming perfect scalability, work done in constant time * resources
 - Given virtually unlimited resources allows to reduce time as necessary

100 days @ 1 node

≈

1 day @ 100 nodes

(but beware Amdahl's law: max speedup sp = 1/s)





Characteristics and Deployment Models

Extended Definition

 ANSI recommended definitions for service types, characteristics, deployment models [Peter Mell and Timothy Grance: The NIST Definition of Cloud Computing, **NIST 2011**]



Characteristics

- On-demand self service: unilateral resource provision
- Broad network access: network accessibility
- Resource pooling: resource virtualization / multi-tenancy
- Rapid elasticity: scale out/in on demand
- Measured service: utilization monitoring/reporting

Deployment Models

- Public cloud: general public, on premise of cloud provider
- Hybrid cloud: combination of two or more of the above
- Community cloud: single community (one or more orgs)
- Private cloud: single org, on/off premises

MS Azure Private Cloud

IBM Cloud Private





Cloud Computing Service Models

(computing as a utility)





Anatomy of a Data Center





Xeon E5-2440: 6/12 cores Xeon Gold 6148: 20/40 cores



Server:

Multiple sockets, RAM, disks



Rack:

16-64 servers + top-of-rack switch



Cluster:

Multiple racks + cluster switch



Data Center:

>100,000 servers



[Google Data Center, Eemshaven, Netherlands]





Fault Tolerance

[Christos Kozyrakis and Matei Zaharia: CS349D: Cloud Computing Technology, lecture, **Stanford 2018**]



Yearly Data Center Failures

- ~0.5 overheating (power down most machines in <5 mins, ~1-2 days)
- ~1 PDU failure (~500-1000 machines suddenly disappear, ~6 hrs)
- ~1 rack-move (plenty of warning, ~500-1000 machines powered down, ~6 hrs)
- ~1 network rewiring (rolling ~5% of machines down over 2-day span)
- ~20 rack failures (40-80 machines instantly disappear, 1-6 hrs)
- ~5 racks go wonky (40-80 machines see 50% packet loss)
- ~8 network maintenances (~30-minute random connectivity losses)
- ~12 router reloads (takes out DNS and external vIPs for a couple minutes)
- ~3 router failures (immediately pull traffic for an hour)
- ~dozens of minor 30-second blips for dns
- ~1000 individual machine failures (2-4% failure rate, at least twice)
- "thousands of hard drive failures (1-5% of all disks will die)





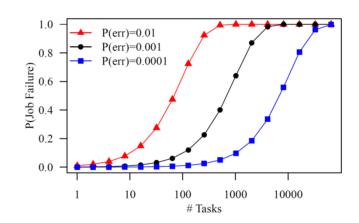
Fault Tolerance, cont.

Other Common Issues

- Configuration issues, partial SW updates, SW bugs
- Transient errors: no space left on device, memory corruption, stragglers

Recap: Error Rates at Scale

- Cost-effective commodity hardware
- Error rate increases with increasing scale
- Fault Tolerance for distributed/cloud storage and data analysis



→ Cost-effective Fault Tolerance

- BASE (basically available, soft state, eventual consistency)
- Effective techniques
 - ECC (error correction codes), CRC (cyclic redundancy check) for detection
 - Resilient storage: replication/erasure coding, checkpointing, and lineage
 - Resilient compute: task re-execution / speculative execution





Virtualization

#1 Native Virtualization

- Simulates most of the HW interface
- Unmodified guest OS to run in isolation
- Examples: VMWare, Parallels, AMI (HVM)

Applications Libraries Operating System Hardware

#2 Para Virtualization

- No HW interface simulation, but special API (hypercalls)
- Requires modified quest OS to use hyper calls, trapped by hypervisor
- Examples: Xen, KVM, Hyper-V, AMI (PV)

#3 OS-level Virtualization

- OS allows multiple secure virtual servers
- Guest OS appears isolated but same as host OS
- Examples: Solaris/Linux containers, Docker

[Prashant Shenoy: Distributed and Operating Systems - Module 1: Virtualization, **UMass Amherst**, **2019**]

#4 Application-level Virtualization

Examples: Java VM (JVM), Ethereum VM (EVM), Python virtualenv







Containerization

Docker Containers

- Shipping container analogy
 - Arbitrary, self-contained goods, standardized units



- Containers reduced loading times → efficient international trade
- #1 Self-contained package of necessary SW and data (read-only image)
- #2 Lightweight virtualization w/ shared OS and resource isolation via cgroups
- Cluster Schedulers (see Lecture 09)
 - Container orchestration: scheduling, deployment, and management
 - Resource negotiation with clients
 - Typical resource bundles (CPU, memory, device)
 - Examples: Kubernetes, Mesos, (YARN),Amazon ECS, Microsoft ACS, Docker Swarm

[Brendan Burns, Brian Grant, David Oppenheimer, Eric Brewer, John Wilkes: Borg, Omega, and Kubernetes. **CACM 2016**]



→ from machine- to applicationoriented scheduling







Excursus: AWS Snowmobile (since 2016)

■ Snowmobile Service: Data transfer on-premise → cloud via 100PB trucks



Real-World
"Containerization"

100PB
~26 years
(1Gb Link)
→ weeks

[https://aws.amazon
.com/snowmobile/?
nc1=h ls]





Infrastructure as a Service (IaaS)

Overview

- Resources for compute, storage, networking as a service
 - → Virtualization as key enabler (simplicity and auto-scaling)
- Target user: sys admin / developer

Storage

- Amazon AWS Simple Storage Service (S3)
- OpenStack Object Storage (Swift)
- IBM Cloud Object Storage
- Microsoft Azure Blob Storage

Compute

- Amazon AWS Elastic Compute Cloud (EC2)
- Microsoft Azure Virtual Machines (VM)
- IBM Cloud Compute













Infrastructure as a Service (IaaS), cont.

- Example AWS Setup
 - Create user and security credentials
- Example AWS S3File Upload
 - Setup and configure S3 bucket
 - WebUI or cmd for interactions
- Example AWS EC2Instance Lifecycle

> aws2 configure

```
AWS Access Key ID [None]: XXX
AWS Secret Access Key [None]: XXX
Default region name [None]: eu-central-1
Default output format [None]:
```

- > aws2 s3 cp data s3://mboehm7datab/air \
 --recursive
- > aws2 s3 ls s3://mboehm7datab/air \
 --recursive

```
2019-12-05 15:26:45 20097 air/Airlines.csv
2019-12-05 15:26:45 260784 air/Airports.csv
2019-12-05 15:26:45 6355 air/Planes.csv
2019-12-05 15:26:45 1001153 air/Routes.csv
```

> aws2 ec2 allocate-hosts \
 --instance-type m4.large \
 --availability-zone eu-central-1a \
 --quantity 2





Platform as a Service (PaaS)

Overview

- Provide environment setup (libraries, configuration), platforms, and services to specific applications → additional charges
- Target user: developer

Example AWS Elastic MapReduce (EMR)

 Environment for Apache Hadoop, MapReduce, and Spark over S3 data, incl entire eco system of tools and libraries

```
> clusterId=$(aws emr create-cluster --applications Name=Spark \
    --ec2-attributes ... --instance-type m4.large --instance-count 100 \
    --steps '[{"Args":["spark-submit","--master","yarn",'${sparkParams}'"--class", \
        "org.tugraz.sysds.api.DMLScript","./SystemDS.jar","-f","./test.dml"], ...]' \
        --scale-down-behavior TERMINATE_AT_INSTANCE_HOUR --region eu-central-1)
```

- > aws emr wait cluster-running --cluster-id \$clusterId
- > aws emr wait cluster-terminated --cluster-id \$clusterId





salesforce

Software as a Service (SaaS)

Overview

- Provide application as a service, often via simple web interfaces
- Challenges/opportunities: multi-tenant systems (privacy, scalability, learning)
- Target user: end users

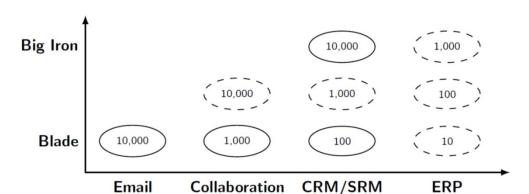
Examples

- Email/chat services: Google Mail (Gmail), Slack
- Writing and authoring services: Micrsoft Office 365, Overleaf
- Enterprise: Salesforces, ERP as a service (SAP HANA Cloud)
- Database as a Service (DaaS)



as a service: schema-mapping

techniques. SIGMOD 2008]



HCM







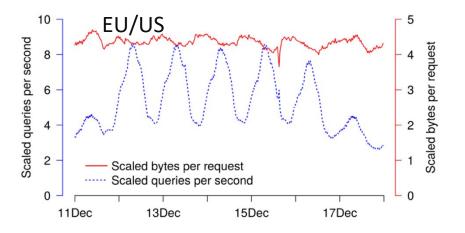
Software as a Service (SaaS)

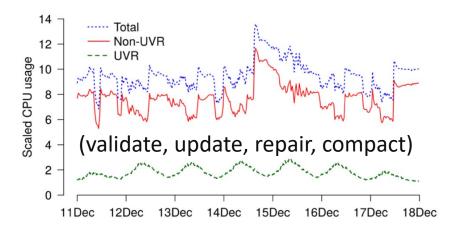
Performance Analysis on Gmail Data

- Coordinated bursty tracing via time
- Vertical context injection into kernel logs

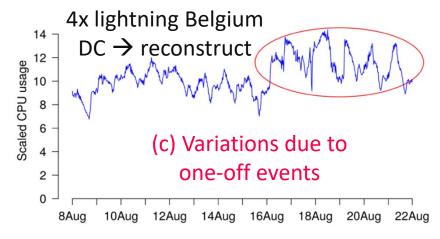
[Dan Ardelean, Amer Diwan, Chandra Erdman: Performance Analysis of Cloud Applications. NSDI 2018]







(a) Variations in rate and mix of user visible requests (UVR)



(b) Variations in rate and mix of essential non-UVR work

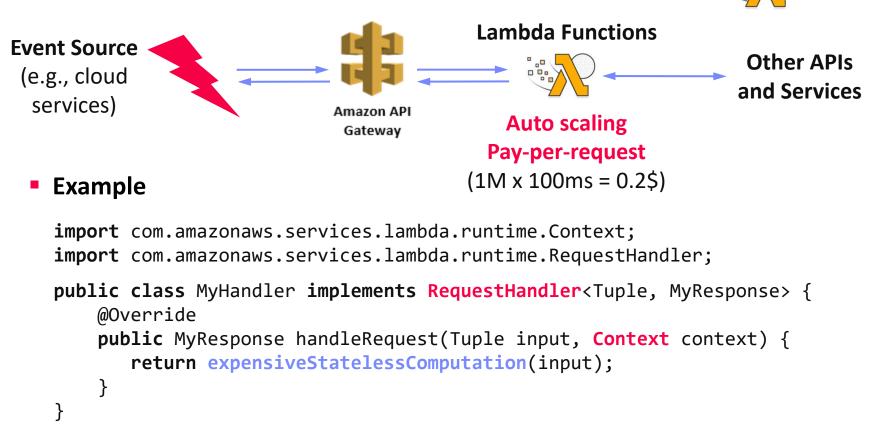


Serverless Computing (FaaS)

Definition Serverless

- FaaS: functions-as-a-service (event-driven, stateless input-output mapping)
- Infrastructure for deployment and auto-scaling of APIs/functions
- Examples: Amazon Lambda, Microsoft Azure Functions, etc







Serverless Computing (FaaS), cont.

- Advantages (One Step Forward)
 - Auto-scaling (the workload drives the allocation and deallocation of resources)

[Joseph M. Hellerstein et al: Serverless Computing: One Step Forward, Two Steps Back. CIDR 2019]



- Use cases: embarrassingly parallel functions, orchestration functions
 (of proprietary auto scaling services), function composition (workflows)
- Disadvantages (Two Steps Backward)
 - Lacks efficient data processing (limited lifetime of state/caches,
 I/O bottlenecks due to lack of co-location)
 - Hinders distributed systems development (communication through slow storage, no specialized hardware)

Func. Invoc.	Lambda I/O	Lambda I/O	EC2 I/O	EC2 I/O	EC2 NW
(1KB)	(S3)	(DynamoDB)	(S3)	(DynamoDB)	(0MQ)
303ms	108ms	11ms	106ms	11ms	290μs
1,045×	372×	37.9×	365×	37.9×	1×

→ "Taken together,
these challenges seem
both interesting and surmountable. [...] Whether
we call the new results
'serverless computing' or
something else, the
future is fluid."





Example AWS Pricing (current gen)

Amazon EC2 (Elastic Compute Cloud)

- laaS offering of different node types and generations
- On-demand, reserved, and spot instances

v(Cores		Mem		
m4.large	2	6.5	8 GiB	EBS Only	\$0.117 per Hour
m ^{4.large}	2	6.5	8 GiB	EBS Only	\$0.12 per Hour
n m4.xlarge	4	13	16 GiB	EBS Only	\$0.24 per Hour
m m4.2xlarge	8	26	32 GiB	EBS Only	\$0.48 per Hour
n m4.4xlarge	16	53.5	64 GiB	EBS Only	\$0.96 per Hour
m m4.10xlarge	40	124.5	160 GiB	EBS Only	\$2.40 per Hour
m4.16xlarge	64	188	256 GiB	EBS Only	\$3.84 per Hour

Amazon ECS (Elastic Container Service)

- PaaS offering for Docker containers
- Automatic setup of Docker environment

Amazon EMR (Elastic Map Reduce)

- PaaS offering for Hadoop workloads
- Automatic setup of YARN, HDFS, and specialized frameworks like Spark
- Prices in addition to EC2 prices

Pricing	accoi	ding	to	EC	2

(in EC2 launch mode)

m4.large	\$0.117 per Hour	\$0.03 per Hour
m4.xlarge	\$0.234 per Hour	\$0.06 per Hour
m4.2xlarge	\$0.468 per Hour	\$0.12 per Hour
m4.4xlarge	\$0.936 per Hour	\$0.24 per Hour
m4.10xlarge	\$2.34 per Hour	\$0.27 per Hour
m4.16xlarge	\$3.744 per Hour	\$0.27 per Hour





Example AWS Pricing (current gen), cont.

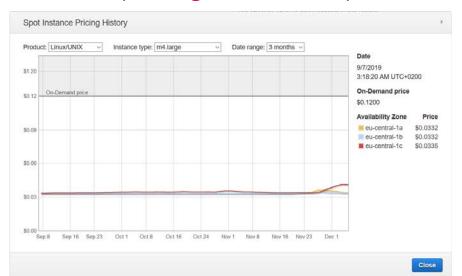
Spot Instances

Self-regulating effect

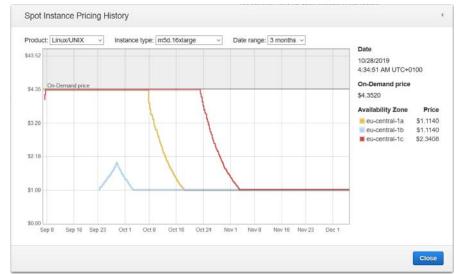
- Unused cloud recourses for much lower prices → bidding market
- Interruption behavior: hibernate, stop, terminate

Example Instance Types

(**m4.large**, 2 vCPU, 8GB)



(m5d.24xlarge, 96 vCPU, 384GB)



[AWS EC2 Management Console, Spot Requests, Dec 05 2019]





Cloud, Fog, and Edge Computing





Cloud vs Fog vs Edge Overview

ECE 590/COMPSI 590 Special Topics: Edge Computing IoT Meets the Cloud: The Origins of Edge Computing Wednesday August 29th, 2018

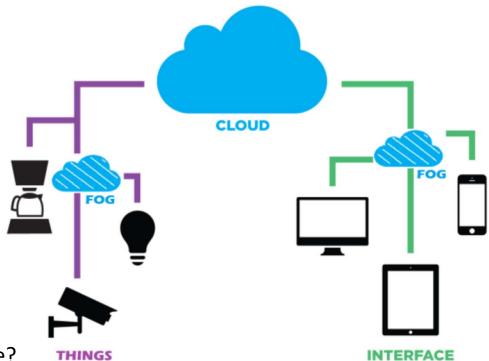
Overview Edge Computing

- Huge number of mobile / IoT devices
- Edge computing for latency, bandwidth, privacy

[Maria Gorlatova: Special Topics: Edge Computing; IoT Meets the Cloud – The Origins of Edge Computing, **Duke University 2018**]

Fog & Edge Computing

- Different degrees of application decentralization
- Reasons: energy, performance, data
- Natural hierarchy, heterogeneity
- Cloud as enabler for vibrant web ecosystem
- → fog/edge for IoT the same?







Example: AWS Greengrass

[Credit: https://aws.amazon.com/greengrass/?nc1=h ls]

Overview AWS Greengrass

reach the cloud"

- Combine cloud computing and groups of IoT devices
- Cloud configuration, group cores, connected devices to groups
- Run lambda functions (FaaS) in cloud, fog, and edge partial autonomy

System Architecture Central configuration and deployment Decentralized operation Customer Use cases: "My data doesn't





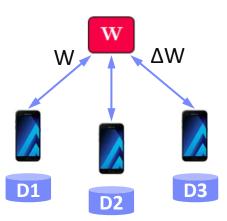
Federated ML

[Keith Bonawitz et al.: Towards Federated Learning at Scale: System Design. **SysML 2019**]



Overview Federated ML

- Learn model w/o central data consolidation
- Privacy vs personalization and sharing (example application: voice recognition)
- Adaptation of parameter server architecture,
 w/ random client sampling and distributed agg.
- Training when phone idle, charging, and on WiFi



Data Ownership Federated ML in the Enterprise

- **Example:** machine vendor middle-person customer equipment
- Who owns the data? Negotiated in bilateral contracts!

A Thought on a Spectrum of Rights and Responsibilities

 Federated ML creates new spectrum for data ownership that might create new markets and business models





Excursus: Federated ML in SystemDS

FFG ExDRa Project (Exploratory Data Science over Raw Data)

SIEMENS

Basic approach: Federated ML + ML over raw data

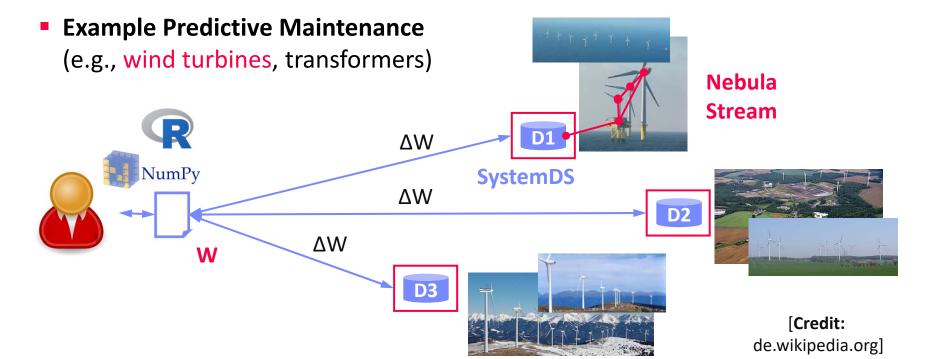
System infra, integration, data org & reuse, Exp DB, geo-dist.



Gefördert im Programm "IKT der Zukunft" vom Bundesministerium für Verkehr, Innovation, und Technologie (BMVIT)









Summary and Q&A

- Cloud Computing Motivation and Terminology
- Cloud Computing Service Models
- Cloud, Fog, and Edge Computing
- Projects and Exercises
 - 13 projects + 4 exercises
 - Few students w/o discussions → setup skype call if help needed

Next Lectures

- 09 Cloud Resource Management and Scheduling [Dec 13]
- 10 Distributed Data Storage [Jan 10]
- 11 Distributed, Data-Parallel Computation [Jan 17]
- 12 Distributed Stream Processing [Jan 24]
- 13 Distributed Machine Learning Systems [Jan 31]

