

CLOSER 2018 Keynote: Lee Gillam, University of Surrey

Will Cloud Gain an Edge?

Or CLOSER, to the Edge



'Traditional' Cloud

- NIST SP800-145 (aka Mell and Grance): 3-4-5 & SPI
- Large (economically efficient, easily maintained but still expensive) datacentres in relatively few, geographical locations (regions) to support large user numbers, *centralized* corporate entity
- A 'Big Four' in Amazon, Microsoft, IBM, Google

'New' Cloud

- Containers (*Docker, kernel-locked*), and Functions (*multiplicity of approaches*) added "CaaS" and "FaaS"
- <u>Edge</u> (multiplicity of approaches and concerns)
- (Re-)*distributed* Computing, and 'new' problems (new 'traditional' problems)
- 'Big Four'?



<u>'Traditional' Cloud</u> (Big Four) Clouds are big Cost and performance (=cost) variation Performance variation and implications for energy efficiency

<u>'New' cloud</u> 'serverless' and performance Multiplicity of Edges 'serverless' Edges

An application Cloud Cars and exemplars

Summary and take home



'Traditional' Cloud



(Illusion of) Infinite capacity - consider one (big) provider:

AWS: 2014 based on 11 regions and 28 AZs: 2.8-5.6m servers (Morgan, 2014) based on datacenter of up to 80k servers

12 Jun 2012, 1 trillion objects in S3.13 April 2013, 2 trillion

2018, 18 regions, 54 AZs, 5 more regions coming: ~10m servers?

T. P. Morgan, A rare Peek Intro The Massive Scale of AWS, 2014, https://www.enterprisetech.com/2014/11/14/rare-peek-massive-scale-aws/



Regions: Contain big datacenters at distance but with big networking also

Edges: a shorter distance to something that does something useful for compute





Manchester United pitch ~80,000 sq ft

Figure sources: Datapath.io and Make IT Green: Cloud Computing and its Contribution to Climate Change, Greenpeace



Clouds are 'big' – with cost variation

27/4/16	vCPU	ECU	Mem (GiB)	US-E (NV)	EU-W (Ire)	EU-W (Fra)	SA (SP)
t2.nano	1	Variable	0.5	\$0.0065	\$0.007	\$0.0075	\$0.0135
t2.micro	1	Variable	1	\$0.013	\$0.014	\$0.015	\$0.027
m4.xlarge	4	13	16	\$0.239	\$0.264	\$0.285	N/A
m4.2xlarge	8	26	32	\$0.479	\$0.528	\$0.57	N/A
m4.4xlarge	16	53.5	64	\$0.958	\$1.056	\$1.14	N/A
m4.10xlarge	40	124.5	160	\$2.394	\$2.641	\$2.85	N/A
m3.medium	1	3	3.75	\$0.067	\$0.073	\$0.079	\$0.095
m3.large	2	6.5	7.5	\$0.133	\$0.146	\$0.158	\$0.19
m3.xlarge	4	13	15	\$0.266	\$0.293	\$0.315	\$0.381
m3.2xlarge	8	26	30	\$0.532	\$0.585	\$0.632	\$0.761
c4.large	2	8	3.75	\$0.105	\$0.119	\$0.134	N/A
c4.xlarge	4	16	7.5	\$0.209	\$0.238	\$0.267	N/A
c4.2xlarge	8	31	15	\$0.419	\$0.477	\$0.534	N/A
c4.4xlarge	16	62	30	\$0.838	\$0.953	\$1.069	N/A
c4.8xlarge	36	132	60	\$1.675	\$1.906	\$2.138	N/A
c3.large	2	7	3.75	\$0.105	\$0.12	\$0.129	\$0.163
c3.xlarge	4	14	7.5	\$0.21	\$0.239	\$0.258	\$0.325
c3.2xlarge	8	28	15	\$0.42	\$0.478	\$0.516	\$0.65
c3.4xlarge	16	55	30	\$0.84	\$0.956	\$1.032	\$1.3
`c3.8xlarge	32	108	60	\$1.68	\$1.912	\$2.064	\$2.6

Clouds are 'big' – with hardware variation



CPU model discovery - for ~700 EC2 FGS Instances for 1 user – <u>1 instance type</u>

Region	AZ	E5430	E5-2650	E5645	E5507
US East N.Virginia	us-east-1a	31%	0	25%	<u>44%</u>
2006 [year Region started]	us-east-1b	5%	<u>59%</u>	29%	7%
Cheapest – but	us-east-1c	0	47%	<u>52%</u>	1%
latencies	us-east-1d	18%	31%	<u>44%</u>	7%
EU West Dublin	eu-west-1a	4%	<u>75%</u>	19%	2%
2007	eu-west-1b	28%	0	<u>44%</u>	28%
	eu-west-1c	4%	0	<u>63%</u>	33%
US West N. California	us-west-1b	0	0	13%	<u>87%</u>
2009	us-west-1c	8%	0	18%	<u>74%</u>
SA San Paulo 2011	sa-east-1a	0	<u>81%</u>	19%	0
2011	sa-east-1b	0	<u>86%</u>	14%	0
US West Oregon 2011	us-west-2b	0	<u>73%</u>	27%	0
Asia Pacific Sydney	ap-southeast-2a	0	<u>64%</u>	36%	0
2012	ap-southeast-2b	0	<u>75%</u>	25%	0

Cost, latency, computational capability (moving up stack)

Performance varies by CPU







Heterogeneous hardware complicates costs

(at minimum) a user needs to:

- Identify suitable (cost-based?) instance offerings (price determination)
- Rank 'best' by workload (performance determination)
- Determine AZs (latency) offering those resources (location selection)
- Attempt to obtain them (instance lottery)
- For users with more than one account this may need to be done per account basis (account selection)

• <u>Costs</u> are incurred in (1) performance determination (2) location selection and (3) instance lottery, for every (4) account selection (AND infrastructure composition changes over time)

<u>Spend involved with getting cost-efficiency (performance gaming /</u> <u>deploy-and-ditch) - potentially expensive!</u>



Can't eliminate resource uncertainty

Performance can be 'stable' over a long period for a given benchmark – past a good indicator of future - but may be subject to abrupt changes and severe degradation

One instance, **<u>1379s</u>** for POV-Ray – ~13 standard deviations from the mean (639, 54)

Rarer: 'The requested Availability Zone is currently constrained and we are no longer accepting new customer requests for X/Y/Z instance types'

• Go elsewhere, but other AZs may not be cost-efficient – AZ lock-in

Unusually for a service: better can be cheaper

But much work around performance continues to assume homogeneity

John O.Loughlin and Lee Gillam (2014) "Should Infrastructure Clouds be Priced Entirely on Performance? An EC2 Case Study". International Journal of Big Data Intelligence

Cost-efficient use



Stationary Time Series, mean = 285s, sdev = 0.7s



Non-Stationary Time Series





Non-Stationary Time Series



An aside – 'brokers' seem popular



Cloud Service Brokers (e.g. aggregators) might address performance issues

Instances with known performance characteristics

• A match-making service between user application performance needs and available resources

Re-price based on desired performance

• Make more suitable instances more expensive, and less suitable less so.

Extensive simulations suggested:

- Assuming clouds are opaque makes it difficult to avoid instance gaming.
- Very difficult to make a profit, even with careful pool management! high vol.
- Opportunities in value of utility rather than price

Rare to find discussion of broker profit – Rogers & Cliff has been a notable exception

O'Loughin, J. (2018): A Workload-Specific Performance Brokerage for Infrastructure Clouds (unpub PhD thesis). Rogers, O. & Cliff, D., 2012. A Financial Brokerage Model for Cloud Computing. Journal of Cloud Computing: Advances, Systems and Applications, 1(2)



Performance and energy trade-off for different kinds of workload

- runtime variable with hardware (heterogeneity)
- how much power needed to deliver runtime on given hardware
- best performance might not equate to most energy efficient
- performance \rightarrow runtime; user cost higher with longer runtime

Put workloads on best machines for it: consolidation (implies migration)

- − risk of being own noisy neighbour → longer runtimes
- additional energy use for period of migration: at least costs of equvalent resources plus network
- question of recoverability depends on continued use
- for providers, opportunity to switch off / maintain (may not be an incentive to)



Workload-related CPU model ranking

- w1: A > B > C > D; w2: D > C > B > A
- VM allocation: B allocated as available; A preferred
- VM consolidation: migration beneficial if workload can recoup cost of migration – implies performance maintained (contention)

Consolidation with Migration Cost Recovery (CMCR)

- Migrate to more efficient hosts
- VM terminated before [t_{off}], the effort is wasted
- Recover migration overhead [Cost_{mig}], efficient gain after [t_{off}]





Broad characterisation: **9** scheduling approaches, several types of consolidation (+ none), CloudSim model using **12,583** heterogeneous hosts, **25m** VMs (tasks) in Google workload trace data, **5** minimum-runtime settings, migration rounds at 5 minute intervals (host utilization < 20%). **On-demand** VM allocation

CPU info not provided, so map Google priorities to benchmark results (range scaled) for preferences (Gratis (0) : POVRAY, Batch (2) : NAMD, Production (9) : STREAM). *Distributions typically skewed lognormal per CPU model.*

Then relate power ratings of these 'Cloud' CPUs – simulation results can be related to real VMs.

Workload	Bench	CPU	Real benchmarks					Google data				
	mark	Model	(µ)	(σ)	Min	Max	CoV	(µ)	(<i>o</i>)	Min	Max	CoV
Gratis	POVRAY	E5430	439	11	421	467	0.025	438.06	9.42	421	467	0.022
		E5-2650	468	12	451	500	0.026	473.87	11.93	451	500	0.025
		E5645	507	10	490	535	0.02	498.55	10.44	490	535	0.021
Batch NAMD	NAMD	E5-2651	1994	41.9	1952	2036	0.021	1991	39.51	1800	2040	0.02
		E5-2650	2007	28.5	1978	2036	0.014	1963.4	28.41	1900	2015	0.015
		E5645	2043	96.4	1946	2140	0.047	1931.4	93.43	1800	2170	0.048
		E5430	2160	20.7	2135	2189	0.01	2103.6	22.1	2080	2150	0.011
		E5507	2187	18.1	2162	2217	0.008	2191.8	15.69	2150	2200	0.007
Production ST	STREAM	E5430	1446	66	1328	1572	0.045	1404.4	44.33	1328	1572	0.032
		E5507	2348	104	2078	2448	0.044	2346.7	107.21	2078	2448	0.046
		E5645	3395	287	2995	4008	0.085	3388.7	238.22	2995	4008	0.07
		E5-2650	5294	191	4935	5860	0.036	5294.5	197.52	4935	5860	0.037



Experimental work



Findings confirmed: *sensible* allocation better (easier) than consolidation; migrate longer running VMs – but <u>assumes clouds are not opaque</u>. (i.e. provider has knowledge of workload)

Zakarya, M. (2017): A Workload-Specific Performance Brokerage for Infrastructure Clouds (unpub PhD thesis) .



'New' Cloud



A look at so-called 'Function as a Service'

'serverless' computing (yet servers are essential)

- You're not supposed to "worry" about provisioning
- Billing per 100 milliseconds (AWS Lambda, Google Cloud Functions; Azure has at least 2 ways to pay incl. based on memory consumption)
- Functions may be time-constrained 5 mins, Lambda/Azure; though HTTP timeout of 30s (e.g. AWS API Gateway) gives 29s runtime

Are performance/cost questions relevant?







A look at so-called 'Function as a Service' – AWS Lambda

AWS Lambda runs a Function in a Container on a VM ('serverless')

IP address may over time change – 2 functions run gave e.g.: ip-10-23-17-3, ip-10-14-98-122.

Short runtimes good: a small test - 113.44 ms, then 114.34 ms, 102.65 ms, 113.57 ms – all rounded to nearest 100 (200ms) for billing; reasonable consistency.

Underlying process:

USER VSZ PID %CPU %MEM RSS TTY STAT START TIME COMMAND 1 1.3 0.3 212024 15372 ? 490 Ss 16:49 0:00 /usr/bin/python2.7 /var/runtime/awslambda/bootstrap.py 7 0.0 0.0 117208 2476 ? 490 R 16:49 0:00 ps auxw Through several uses, underlying process remains.



A look at so-called 'Function as a Service' – AWS Lambda

Limitations exist, e.g. can't run 'ifconfig', no 'sudo' so can't install as root, and can't get at AWS metadata of VM. But can find out *CPU model* (/proc/cpuinfo [dual 'core' **c4**]) and underlying system (uname [**AWS Linux**]),

model name	:	Intel(R)	Xeon (R)	CPU	E5-2666	v3	G	2.90GHz
cpu MHz	:	2900.066						
cache size	:	25600 KB						

```
Linux ip-10-23-17-3 4.4.35-33.55.amzn1.x86_64 #1 SMP ... x86 64 x86 64 x86 64 GNU/Linux
```

Others have seen:

 Intel(R) Xeon(R) CPU E5-2680 v2 @ 2.80GHz - a <u>c3</u> instance: <u>http://zqsmm.qiniucdn.com/data/20150416152509/index.html</u>.



A look at so-called 'Function as a Service' – AWS Lambda

Can run a [small –time limit!] arbitrary linux application – e.g. a benchmark such as STREAM [~2GB/s], if:

- Precompiled elsewhere, downloaded into local filestore for Function (/tmp), made executable (chmod), executed and output returned
- Variations <u>per execution</u>, with rounding; performance/location/lottery/account remains important

STREAM version \$Revision: 5.10 \$									
Function	Best Rate MB/s	Avg time	Min time	Max time					
Copy:	1979.3	0.091534	0.080836	0.116964					
Scale:	1974.5	0.099291	0.081033	0.117087					
Add:	2370.4	0.122582	0.101247	0.157255					
Triad:	2362.9	0.126896	0.101569	0.157570					



Will Cloud gain an Edge?

Network Edge devices – devices that have sensors (a mobile phone?)

Customer **Edge** (/Edge router) – router on premises

Provider Edge – a provider's router

Edge Datacenter (/*Cloudlet**/Content Delivery Network) – datacenter Multi-access (/mobile) Edge – datacenter + RAN (e.g. 5G) Also, *FOG*

Edge is also Microsoft's web browser, and a Ford vehicle

* Should not be confused with Cloudlet in CloudSim, which is a Task



Fog "complements and extends the Cloud to the edge and endpoints", Bonomi et al

- Fog is additional to and complementing Cloud,
- Example distributed applications: "<u>A</u> has one Cloud component, and two Fog components [..] <u>B</u> has one cloud component, one component in the Core, and a Fog component".

OpenFog Consortium: "a system-level horizontal architecture that distributes resources and services of computing, storage, control and networking anywhere along **the** *continuum* from Cloud to Things".

- OpenFog documents represent Fog diagrammatically as:
 - between Cloud and Things;
 - including Cloud;
 - 'in' Cloud Computing.
- OpenFog Consortium: Fog is "often erroneously called edge computing, but there are key differences. Fog works with the cloud, whereas edge is defined by the exclusion of cloud".

Fog has not proven an entirely helpful notion





Multi-access Edge Computing (until recently, Mobile Edge Computing) – ETSI specification

- provide capabilities of Cloud Computing close to the Radio Access Networks in 4G and 5G telecommunications and converge with other radio access technologies (e.g. WiFi or Satellite).
- "can be seen as a cloud server running at the edge of a mobile network".

ETSI MEC server supports VMs into which "MEC applications from vendors, service providers and third-parties are deployed and executed".

VMs \rightarrow Containers \rightarrow Functions [MEC authors: PaaS for "future releases"].





Not just one MEC server per Edge?

Cloudlet (Carnegie Mellon University (CMU) notion) - middle tier of a 3-tier hierarchy: "mobile device - **cloudlet** – cloud".

A "<u>datacentre in a box</u>", implying multiple servers, local to the user, similar to MEC (and some characterisations of Fog).

CMU formed Open Edge Computing initiative in 2015, together with Intel, Huawei, and Vodafone, intending to synchronize work with ETSI MEC, leading so far to OpenStack++ allowing for **migration between OpenStack clusters**.

Integration with telecommunications per MEC does not appear yet to be paralleled.



Intended benefits of Edge: reduced end-to-end latency; smaller data volumes travelling shorter distances – computational capability and storage is nearer the user.

Intended benefits of Functions: small, fast-executing, provider-scaled capability.

Faster, smaller implies more suitable for cloud-assisted, or cloud-driven, control services.

But: susceptible to hardware variations (performance), including due to contention, as well as provider-driven energy management.

See, also, AWS Lambda at Edge (CDN-related FaaS).



Edges running VMs / Containers / Functions, using various data

Migration when user moves to another edge

Execution would vary with hardware (slowdown/speedup may imply needing more/less resource for equivalence at the target – and need to know this)



Is a user highly likely to keep moving quickly between edges?



An Application







Connected and Autonomous Vehicles





TASCC \rightarrow CARMA



5 year UK research programme - Cloud-Assisted Real-time Methods for Autonomy (CARMA) project





Internet of Things (Tractors, Kettles, Fridges, <u>Cars</u>)

Waymo generates 1GB data **per second** (2 PB/car/year)

Estimate of 2 billion total vehicles by 2020

Some of this 1GB/s may be usefully processed close to the vehicles, with vehicles connecting through the RAN to access or provide e.g. local road information



Some data may aggregated over MEC Servers, or where capability is not needed immediately, or if not available locally





CARMA's vision: design and validate a novel, secure framework to *enable* implementation of safe and robust semi-autonomous and fully autonomous functions

Main objectives

Address key technical research challenges Validate through proof-of-concept demonstrators Evaluate scalability





Multi-access/mobile edge, 5G , Cloud.Security and effects on performanceSecurity and effects on performanceSafetyremains paramount





CARMA Core (Cloud):

- Based on commercially available public cloud resources
- Services where **higher latency** is tolerable, information is coming from a **wider geography**, longer term storage needs, and so on.

CARMA Edge (MEC, 5G):

- Host beneficially off-board (low latency) processes
- Information collected from *around* vehicles to support **cooperativity** and computation beyond capabilities, including sensor ranges, of a given vehicle.

CARMA Vehicle:

- On-board vehicle network across all sensors, infotainment, actuators etc.
- Significant increases in on-board computational capability to be expected.




Cooperativity beneficial for maps where information has varying transience – information around vehicles and beyond sensor ranges



CARMA Platform: Logical Design



Public Cloud Exemplars

IBM Cloud-connected Vehicles

Largely Vehicle to Cloud

- IBM Watson IoT Driver Behavior Service
- IBM Connected-car IoT app with Geospatial Analytics.
- Microsoft Connected Vehicle Platform
- Google Connected Vehicle Platform

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A hint of Edge, in -

- AWS Connected Vehicle Solution
- IBM Edge (Apache Edgent) analytics

None of these address Autonomous





AWS Connected Vehicle Solution

AWS suggests an application stack launchable from a template for multiple services

- Long term data storage
- Treatment of telematics anomalies
- Capture of trip data
- Calculation of a driver safety score
- Diagnostics and reporting

Messages through IoT gateway (Message Queue Telemetry Transport – MQTT, pub/sub): hierarchy of topics, and data payloads; rules on messages trigger execution of Lambda Functions for applications.

Includes AWS Greengrass for the *Edge*, but most of the work in Cloud



Public Cloud Exemplars





IBM Edge Analytics: Edge Agent on a Raspberry Pi and the DGLux tool.





Edge offerings largely not yet replicating Clouds; actions, per provider, required in order to achieve.

- **AWS** IoT rules not yet deployable to (Greengrass) Edge; local versions of other services largely also unavailable.
- **IBM** "Edge Analytic Rule(s) are pushed to the edge device" but no other services
- Microsoft IoT Edge requires a device running Windows 10 or Windows server with Docker
- Google IoT seems not to relate to Edge capability as yet.
- Vendors also promoting their **own** flavour of Function as a Service.

MQTT (Advanced Message Queuing Protocol), tending to be supported – AMQP and others less so.

This is an improving situation.



Summary and take home



Will Cloud gain an Edge?

Which Edge? For us, MEC, but market forces...

<u>'Traditional' Cloud</u> (Big Four) Clouds are big Cost and performance (=cost) variation Performance variation and implications for energy efficiency

<u>'New' cloud</u> 'serverless' and performance Multiplicity of Edges 'serverless' Edges

An application Cloud Cars and exemplars

Summary and take home



Will Cloud gain an Edge?

Provider challenges include:

- What to provide in hardware
 - light lifting (rPis)
 - heavy lifting (servers, stacks)
 - telecommunications connectivity
 - "Moore's law"
- What to provide <u>on</u> hardware
 - Bare metal
 - VMs/Containers/Functions
- Where to locate
- How to maintain
- Support for Migration
- How to secure
- To what quality of service
- At what price





Application/user challenges, as well as services to prioritise, include:

- Which Clouds/Edges to use? Vendor(s) lock-in? (hostage to functionality)
- What runs where in V/E/C, when?
 - Fixed, or dynamic, accounting for limited, heterogeneous, resources under contention? (dynamic reconfiguration)
 - What **<u>Performance</u>** guarantees?
- <u>Migration</u> between V/E/C and/or across Edges? (c/w elastic / scalable)
 - Live migration uninterruptible?
- How to secure? (What to secure? When to secure?)
- How to price across V/E/C? Who pays whom for what?
 - Fixed / dynamic
 - Based on services and/or demand?





An opportunity for an interested researcher

Cloud Assisted Real-time Methods for Autonomy (CARMA)

Research Fellow in Mobile/Multi-access Edge Cloud Computing

Deadline: 8 April

https://jobs.surrey.ac.uk/Vacancy.aspx?id=4701



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Thank You

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https://jobs.surrey.ac.uk/Vacancy.aspx?id=4701

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Journal of Cloud Computing

Advances, Systems and Applications

Cloud Architectures
 Cloud Technologies
 Cloud Services