MASSIVIZING COMPUTER SYSTEMS

= MAKING COMPUTER SYSTEMS SCALABLE, RELIABLE, PERFORMANT, ETC., YET ABLE TO FORM AN EFFICIENT ECOSYSTEM

@Large Research Massivizing Computer Systems



http://atlarge.science



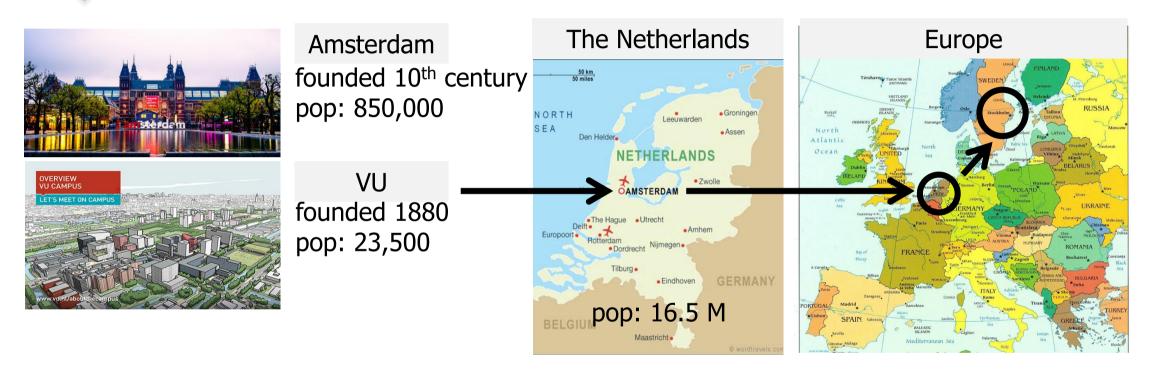




Prof. dr. ir. Alexandru Iosup

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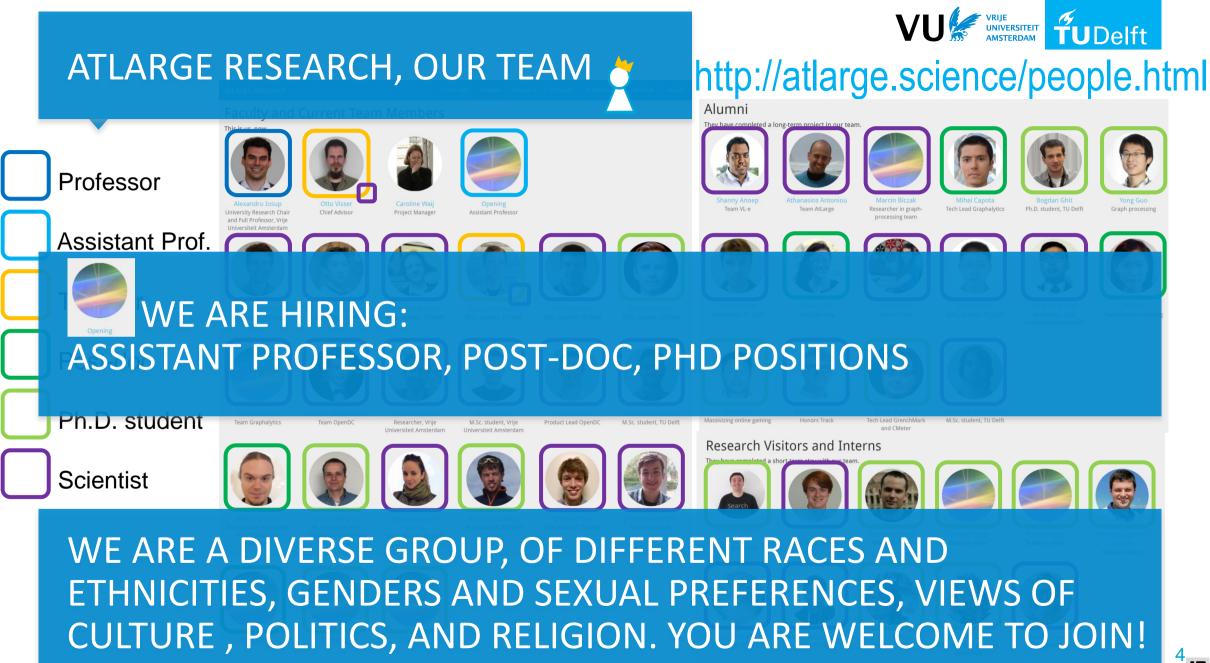


WHO AM I? PROF. DR. IR. ALEXANDRU IOSUP

- Education, my courses:
 - > Systems Architecture (BSc)
 - > Distributed Systems, Cloud Computing (MSc)
- Research, 15 years in DistribSys:
 - > Massivizing Computer Systems
- About me:
 - > Worked in 7 countries, NL since 2004
 - > I like to help... I train people in need
 - > VU University Research Chair
 - > NL ICT Researcher of the Year
 - > NL Higher-Education Teacher of the Year
- **/U** > NL Royal Young Academy of Arts & Sciences







MASSIVIZING COMPUTER SYSTEMS: OUR MISSION



1. Improve the lives of millions through impactful research.

▣

2. Educate the new generation of top-quality, socially responsible professionals.



3. Make innovation available to society and industry.





http://atlarge.science/about.html

Massivizing Computer Systems ~40' A Structured Discussion

~3' — About Our Team \Box

~10' — The Golden Age of Massive Systems ... Yet We Are in a Crisis

- The main challenges
- How we address them

Vote for what you want to hear about

~5'

- Massivizing Computer Systems: Examples \square
 - 1. The Ecosystem Navigation Challenge
 - 2. The New World Challenge
 - 3. The Scheduling Challenge

~2' — Take-Home Message

VISION: WHAT DOES OUR SOCIETY NEED? ISN'T THIS ALREADY HAPPENING?

prosperous society blooming economy inventive academia wise governance



Iosup et al., Massivizing Computer Systems, ICDCS 2018. [Online]

THIS IS THE GOLDEN AGE OF DISTRIBUTED ECOSYSTEMS



THIS IS THE GOLDEN AGE OF DISTRIBUTED COMPUTER SYSTEMS

Do you recognize this App?



Daily Life



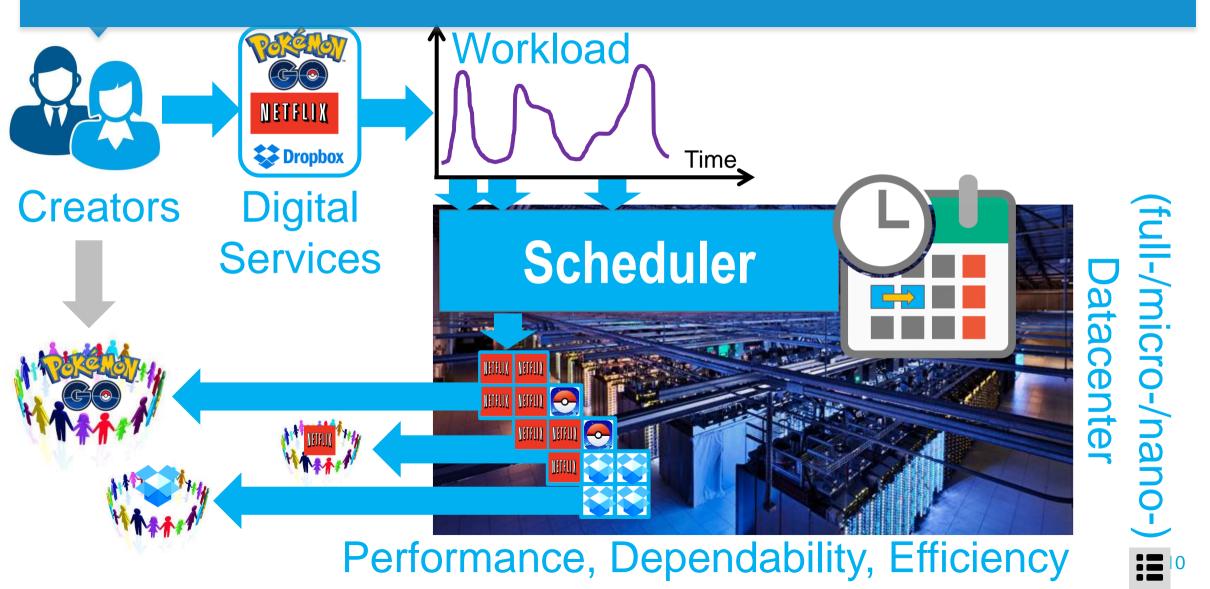




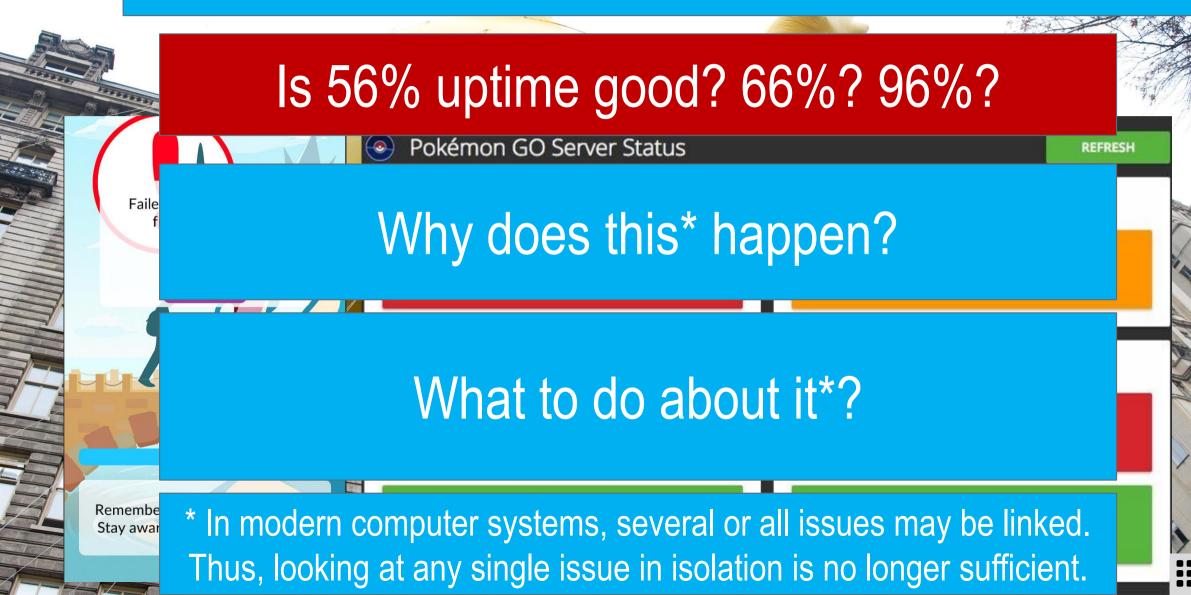




THE CURRENT TECHNOLOGY STACK: DATACENTER, SCHEDULER



My Research: Massivizing Computer Systems



THIS IS THE GOLDEN AGE OF DISTRIBUTED COMPUTER SYSTEMS

YET WE ARE IN A CRISIS – 5 CORE PROBLEMS

1. The Current Laws and Theories Are Built and Tested for Isolated Computer Systems (or Silos, or Narrow Stacks)

> TRADITIONAL DISTRIBUTED SYSTEMS COURSES TEACH YOU ALL ABOUT THIS

2. Need to Understand
How to Maintain Ecosystems
3. Need to Understand

How to Make Ecosystems Automated, Efficient (Smarter)

4. Beyond Tech: Need to Also Be Ethical

5. Need to Address the Peopleware Problems

THIS IS THE GOLDEN AGE OF DISTRIBUTED COMPUTER SYSTEMS

YET WE ARE IN A CRISIS

WHICH WE & YOU CAN HELP SOLVE!

Massivizing Computer Systems Tackles The Challenges of Distributed Systems and Ecosystems...

... and Is Relevant, Impactful, and Inspiring for Many Young Scientists and Engineers OUR DISTRIBUTED SYSTEMS COURSE

THIS IS THE MODERN SCIENCE OF DISTRIBUTED ECOSYSTEMS

MASSIVIZING COMPUTER SYSTEMS IN A NUTSHELL



Iosup et al., Massivizing Computer Systems, ICDCS 2018. [Online]



MASSIVIZING COMPUTER SYSTEMS: MEANINGFUL DISCOVERY

Dependability

Workflows Domain-Specific/Agnostic Space-/Time-Correlation Portfolio, Auto-scaling*

Scheduling

Performance & Failure Analysis* Availability-On-Demand

New World Workload Modeling Serverless **Reference** Architectures

Ecosystem Navigator Performance Variability Grid*, Cloud, Big Data Benchmarking* Longitudinal Studies

Scalability/Elasticity **Delegated Matchmaking*** BTWorld*, POGGI*, AoS

Auto-Scalers* Heterogeneous Systems

Socially Aware

Collaborative Downloads* Groups in Online Gaming **Toxicity Detection* Interaction Graphs**

Education

Social Gamification*

Software Artifacts

Data Artifacts

Graphalytics, OpenDC Distributed Systems Memex*

Fundamental Problems/Research Lines Our Contribution So Far

* Award-level work Competitive personal grants

15

MEANINGFUL DISCOVERY

science + engineering + design

[Iosup et al. ICDCS'18]

16



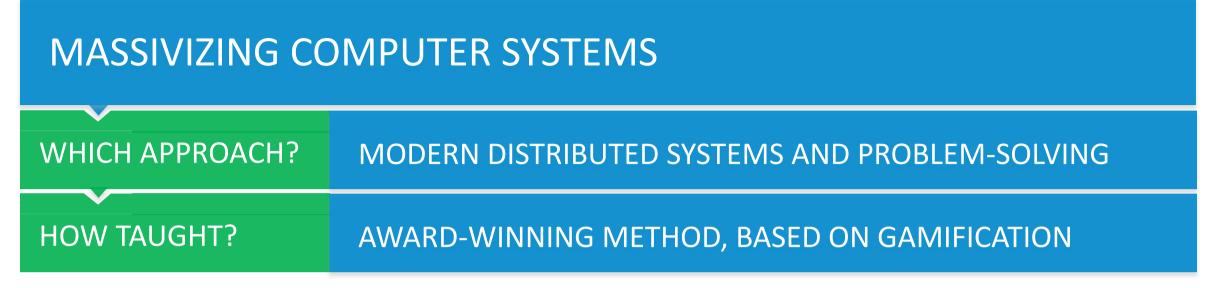
Massivizing Computer Systems ~40' A Structured Discussion

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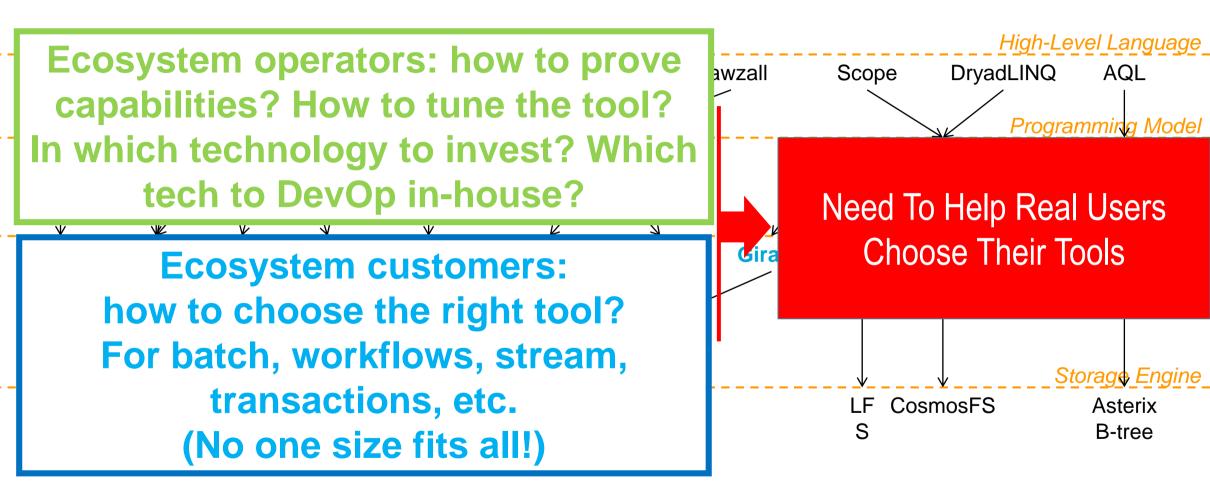
- Massivizing Computer Systems: Examples
 - 1. The Ecosystem Navigation Challenge
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- ~2' Take-Home Message



- Choose your own path:
- > The Ecosystem Navigation Challenge (Understanding + Exp.)
- >The New World Challenge (Abstraction + Design)
- > The Scheduling Challenge (Design + Operation)



The Ecosystem Navigation Challenge

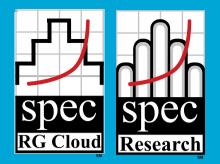


Batch data processing ecosystem in 2011. Our latest update covers ecosystems until 2018.









Alexandru losup Nikolas Herbst Chair Vice-Chair

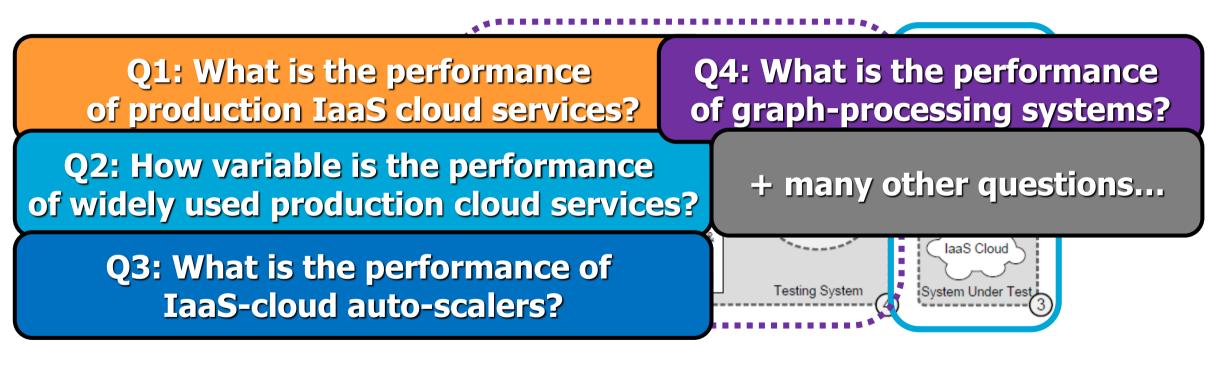
The SPEC RG Cloud Group

Methodology, Benchmarking, and Performance Analysis of Cloud Systems and Applications

"A broad approach, relevant for both academia and industry, to cloud benchmarking, quantitative evaluation, and experimental analysis." "To develop new methodological elements for gaining deeper understanding not only of cloud performance, but also of cloud operation and behavior" "... through diverse quantitative evaluation tools"

http://research.spec.org/working-groups/rg-cloud-working-group.html

Sales team: "\$200/year for research teams" A General Approach for Ecosystem Benchmarking (Works for Cloud and Big Data Benchmarking)





Iosup, Prodan, Epema. IaaS Cloud Benchmarking: Approaches, Challenges, and Experience. Cloud Computing for Data-Intensive Applications 2014: 83-104

EXPERIMENTAL METHOD OF DISCOVERY

UNIQUE OPPORTUNITY TO VALIDATE: WE DRINK OUR OWN CHAMPAGNE (IN VIVO)!



Our Method

 General performance technique, adapted to clouds: model performance of individual components; system performance is performance of workload + model [Saavedra and Smith, ACM TOCS'96]

> Iosup et al., Performance Analysis of Cloud Computing Services for Many Tasks Scientific Computing, IEEE TPDS 2011. Highest cited article in the best journal of the field (2009-2015).



- Performance traces from CloudStatus, to understand variability
 - All Amazon AWS and Google GAE services
 - Periodic performance probes, Sampling rate under 2 minutes
- Simulations to assess impact of performance variability
 - Based on traces collected from other applications



Iosup, Yigitbasi, Epema. On the Performance Variability of Production Cloud Services, IEEE CCgrid 2011. Highest cited study on performance variability in clouds.

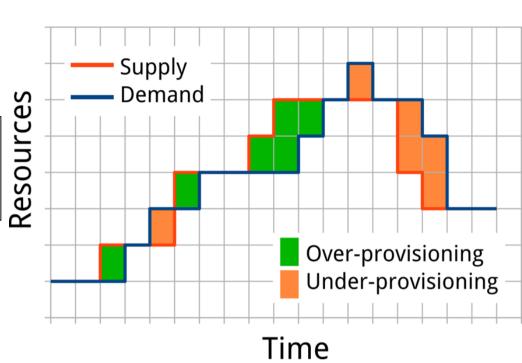
Our Method



- Ilyushkin et al. An Experimental Performance Evaluation of Autoscaling Policies for Complex Workflows. ICPE 2017. Nominated for Best Paper Award.
- Real-world experiments (1 workload)
- Later: simulations (more parameters)
- Many classes of algorithms used in practice
- Diverse real and synthetic datasets
 - Recently, Broido and Clauset found that power-laws are rare in graphs
- Diverse set of experiments representative for practice



Iosup et al. LDBC Graphalytics: A Benchmark for Large-Scale Graph Analysis on Parallel and Distributed Platforms. PVLDB 9(13): 1317-1328 (2016) Collaboration academia-industry.



(intel)

Main Findings



- Lower performance than theoretical peak in laaS services
 - Especially CPU (GFLOPS)
 - Not explained by traditional models, not covered by SLA



- Performance variability in IaaS and PaaS services
 - Explored in longitudinal study of Amazon Web Services and Google App Engine
 - Not captured in traditional models
- Compared performance of IaaS clouds with many commercial alternatives, such as supercomputers and clusters



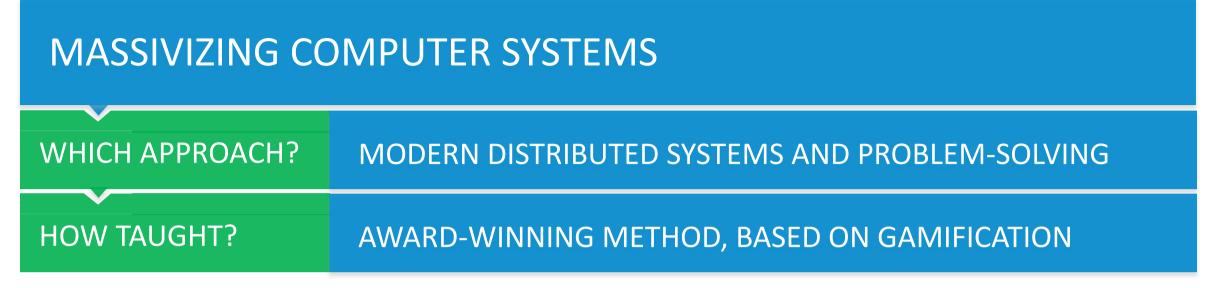
Main Findings

- **Q3**
- Explored impact of auto-scaler (+ ecosystem of schedulers) on over 10 facets of elasticity
 - Findings not explained by traditional models



- The HPAD model for the performance of graph-processing systems ~ replaces previous theories
 - Performance is function of Platform, but also Dataset and Algorithm
 - With configurable many-/multi-cores, the Hardware also is also crucial for performance

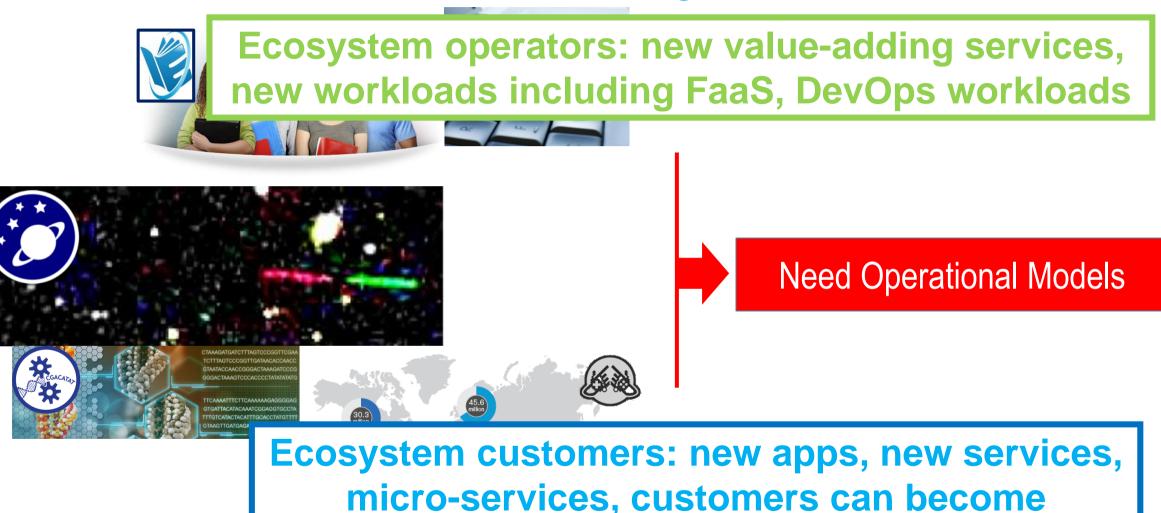




- Choose your own path:
- > The Ecosystem Navigation Challenge (Understanding + Exp.)
- >The New World Challenge (Abstraction + Design)
- > The Scheduling Challenge (Design + Operation)



The New World Challenge



operators (recursive value-chain)









Erwin van Eyk



Serverless / FaaS Execution

Vision and Architecture for Serverless Execution in Cloud Environments

van Eyk, Toader, Talluri, Versluis, Uta, Iosup: Serverless is More: From PaaS to Present Cloud Computing. IEEE Internet Computing Sep/Oct 2018. [<u>Online]</u>

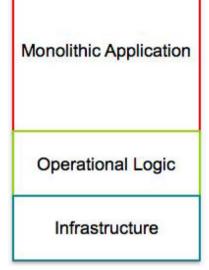
Erwin Van Eyk, Alexandru Iosup, Cristina L. Abad, Johannes Grohmann, Simon Eismann:

A SPEC RG Cloud Group's Vision on the Performance Challenges of FaaS Cloud Architectures. ICPE 2018. [<u>Online</u>]

Erwin van Eyk, Simon Seif (SAP), Markus Thoemmes (IBM Germany), Alexandru Iosup. The SPEC Cloud Group's Research Vision on FaaS and Serverless Architectures. Workshop on Serverless Computing (WoSC'17), held in conjunction with Middleware'17. [<u>Online</u>]



From Monoliths to Microservices to FaaS

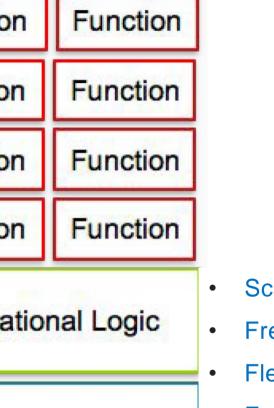


- **Difficult to Scale**
- Infrequent
- Inflexible ٠
- **Complex deployment**
- Tightly coupled stack



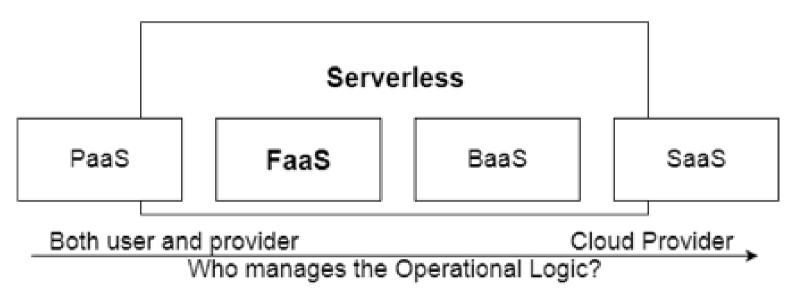
μs	μs	Function	Funct
Operational Logic	Operational Logic		
μs	μs	Function	Funct
Operational Logic	Operational Logic	Function	Funct
Infrastructure		Function	Funct
Scala	ble	100	
Frequent		Operational Logi	
Flexible			
 Complexity: from application logic to operational logic. 		Infrastructure	
Need	Need for DevOps		

Need for DevOps



- Scalable
- Frequent
- Flexible
- Explicit separation of ۲ Business Logic vs. **Operational Logic.**
- Minimal layer coupling, ۲ unit of deployment

Serverless and FaaS



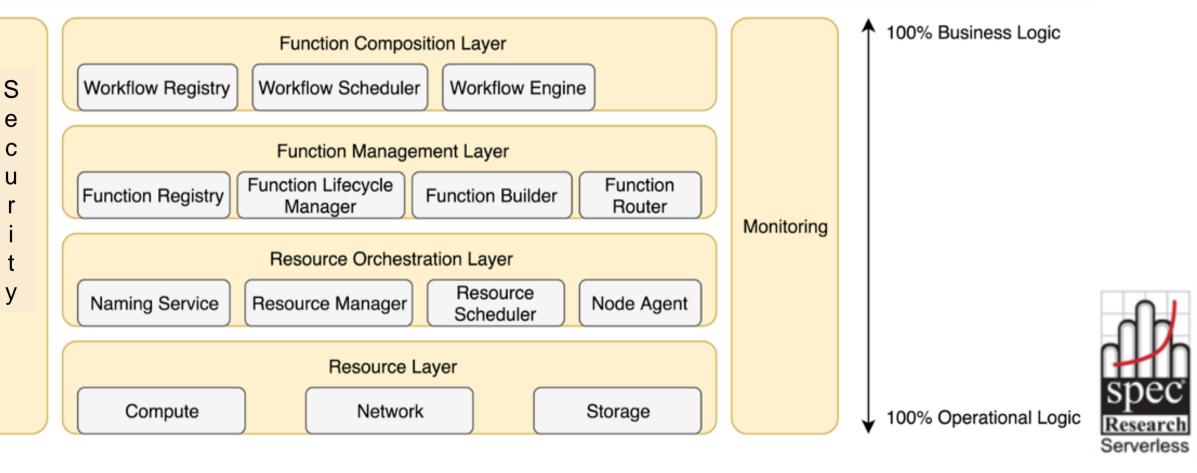
- Serverless
 - (Almost) no operational logic
 - Event-Driven
 - Granular billing

van Eyk et al. [<u>Online]</u>

• FaaS

- A form of serverless computing
- <u>User</u> provides a function
- Function deployed and managed by <u>cloud provider</u>

Reference Architecture for FaaS Management





Van Eyk et al., A SPEC RG Cloud Group's Vision on the Performance Challenges of FaaS Cloud Architectures. ICPE 2018. [<u>Online]</u>

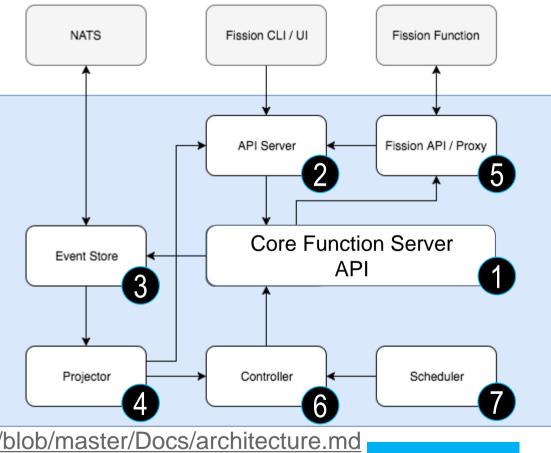
Workflow Management Architecture in Fission.io

Designed by Erwin van Eyk during internship at Platform9, in collaboration w/ Platform9 team and Alexandru losup.

- 1 Core Function / 2 API Server
 - Exposes all actions through API
- 3 Event Store / 4 Projector
 - Events update the workflow
 - Store has Pub/Sub functionality
 - Projector builds current state
- 5 Fission Proxy
 - API access to Fission FaaS
- 6 Controller / 7 Scheduler
 - Workflow manager

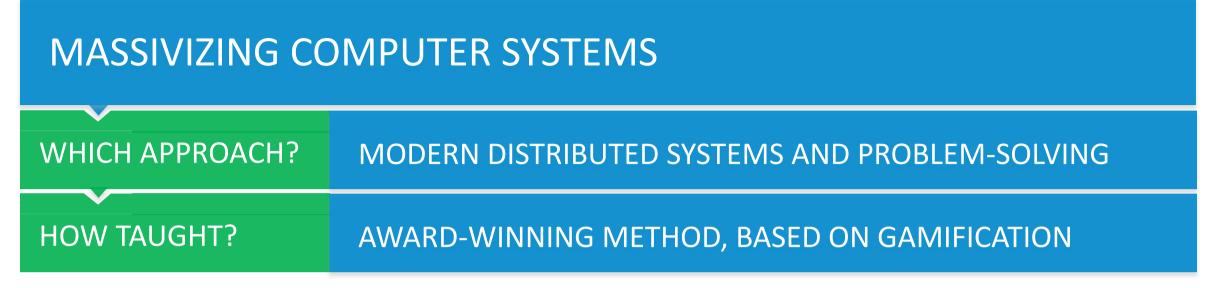
https://github.com/fission/fission-workflows/blob/master/Docs/architecture.md

Oct 3, 2017



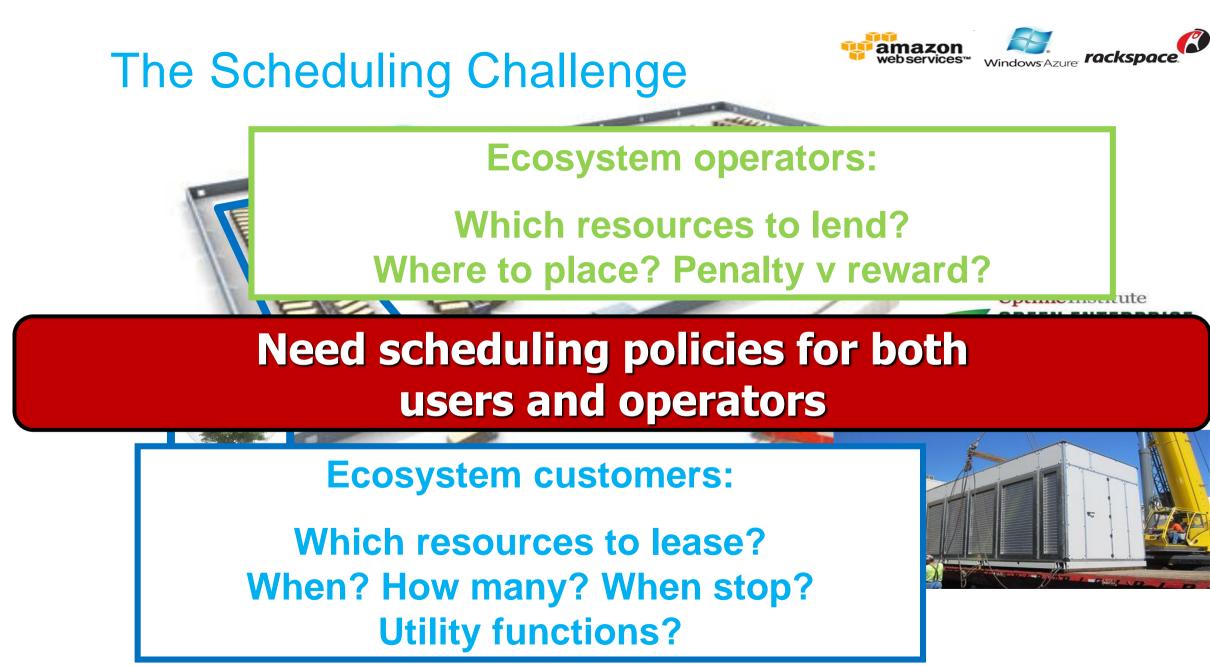
JDelft

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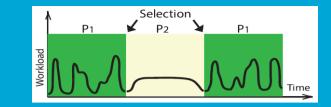


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Vincent van Beek

Tim Hegeman

Jesse Donkervliet Alexandru Iosup

Portfolio Scheduling for DCs

Self-Expressive Management of Business-Critical Workloads in Virtualized Datacenters

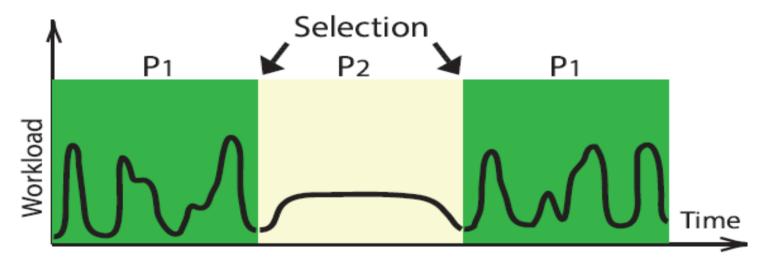
van Beek, Donkervliet, Hegeman, Hugtenburg, Iosup. Self-Expressive Management of Business-Critical Workloads in Virtualized Datacenters. IEEE Computer 48(7): 46-54 (2015)

Deng, Song, Ren, Iosup. Exploring portfolio scheduling for long-term execution of scientific workloads in IaaS clouds. SC 2013: 55:1-55:12

Portfolio Scheduling, In A Nutshell

- Datacenters cannot work without one or even several schedulers
- Instead of ephemeral, risky schedulers, we propose to

(Repeat)



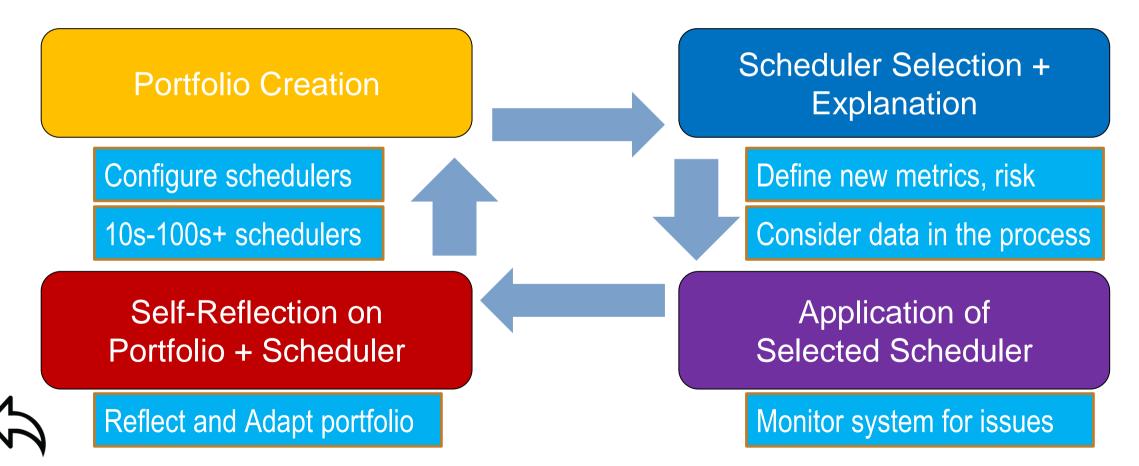
- 1. Create a set of schedulers (resource provisioning and allocation policies)
- 2. Select active scheduler online, apply for the next period, analyze results

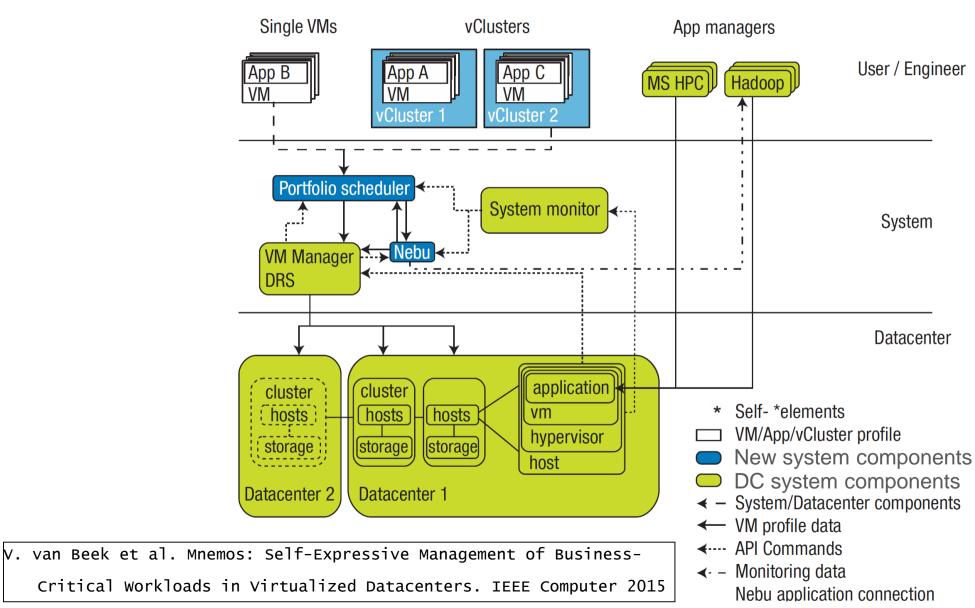
K. Deng et al. Exploring portfolio scheduling for long-term execution of scientific workloads in IaaS clouds. SC|13



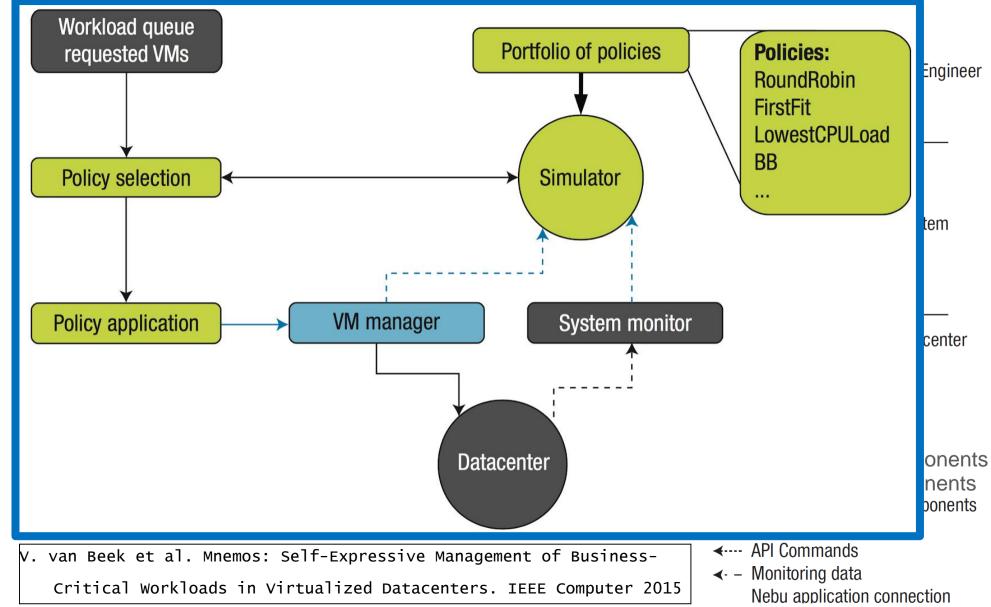
Portfolio Scheduling for Computer Systems

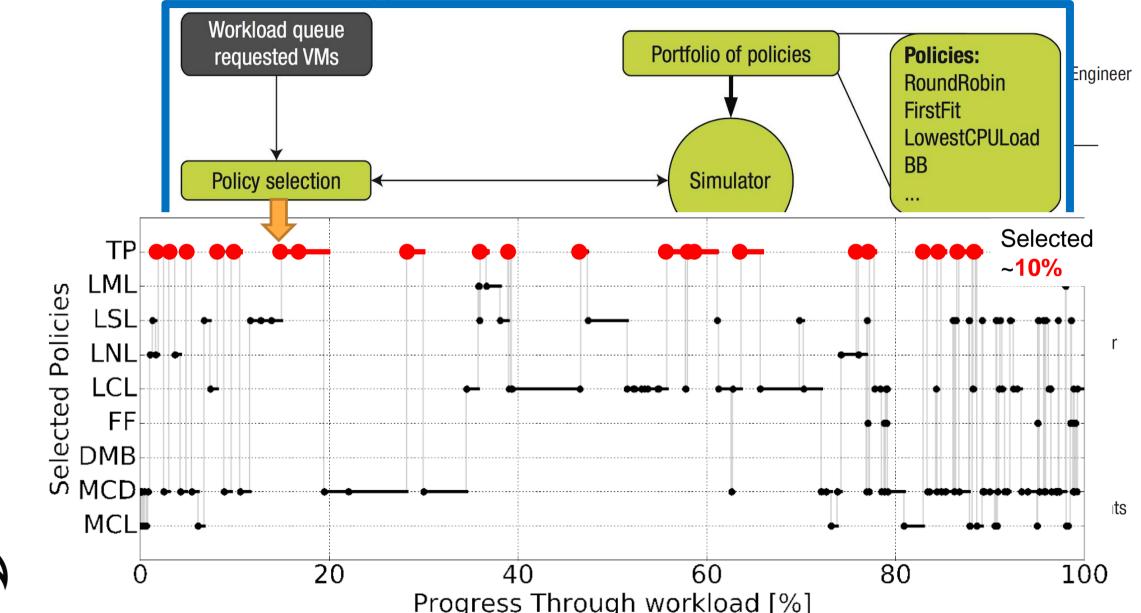
Portfolio Scheduling

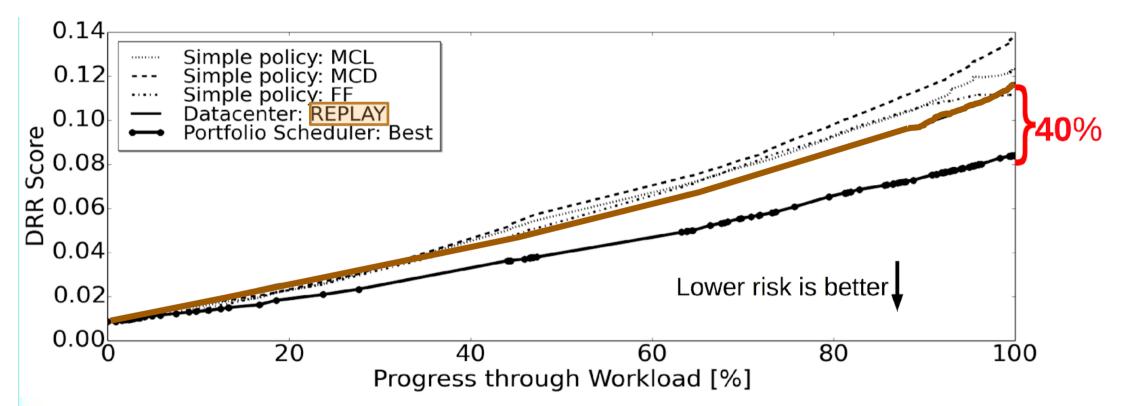












- 1. Portfolio scheduler achieves the lowest risk of all scenarios.
- 2. Portfolio scheduler achieves at least **35%** lower DRR compared to individual policies.
- 3. 40% lower DRR than commercial production system (REPLAY

Not performance-related, but: A portfolio scheduler can explain each decision by presenting its decision data.

Q: Can our sysadmin do this? Can we? (Rhetorical)



Unknown: Fundamental limitations? Use in ecosystem? User studies?

V. van Beek et al. Mnemos: Self-Expressive Management of Business-

Critical Workloads in Virtualized Datacenters. IEEE Computer 2015

Massivizing Computer Systems ~40' A Structured Discussion

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 - The main challenges
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you want to hear
about

- ~5' Massivizing Computer Systems: Examples
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 - 2. The New World Challenge
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~2' — Take-Home Message

MASSIVIZING COMPUTER SYSTEMS

= MAKING COMPUTER SYSTEMS SCALABLE, RELIABLE, PERFORMANT, ETC., YET ABLE TO FORM AN EFFICIENT ECOSYSTEM

- Golden Age of Distirbuted Ecosystems ... Yet a crisis Is looming
- Massivizing Computer Systems means modern distributed systems
 - Think Ecosystems
 - Methods to address key challenges in science, design, and engineering
 - Teaching facilitated by award-winning method
- Much left to do, as we are merely beginning ...
 - You can help! You can make a career, in science / industry.

@Large Research Massivizing Computer Systems

http://atlarge.science



MASSIVIZING COMPUTER SYSTEMS

FURTHER READING

- 1. Iosup et al. Massivizing Computer Systems. ICDCS 2018 (in print)
- 2. Andreadis et al. A Reference Architecture for Datacenter Scheduling: Design, Validation, and Experiments, SC18 (in print)
- 3. Van Eyk et al. Serverless is More: From PaaS to Present Cloud Computing, IEEE IC Sep/Oct 2018 (in print)
- 4. Jiang et al. Mirror: A computation-offloading framework for sophisticated mobile games, CCPE 2018 (in print)
- 5. Ilyushkin et al. An Experimental Performance Evaluation of Autoscaling Policies for Complex Workflows. TOMPECS 2018.
- 6. Iosup et al. The OpenDC Vision: Towards Collaborative Datacenter Simulation and Exploration for Everybody. ISPDC'17.
- 7. Iosup et al. Self-Awareness of Cloud Applications. Self-Aware Computing Systems book, 2017.
- Iosup et al. LDBC Graphalytics: A Benchmark for Large-Scale Graph Analysis on Parallel and Distributed Platforms. PVLDB 2016.
- 9. Guo et al.: Design and Experimental Evaluation of Distributed Heterogeneous Graph-Processing Systems. CCGrid 2016.
- 10. van Beek et al.: Self-Expressive Management of Business-Critical Workloads in Virtualized Datacenters. IEEE Computer 2015.
- 11. Jia et al.: Socializing by Gaming: Revealing Social Relationships in Multiplayer Online Games. TKDD 2015.
- 12. Ghit et al. Balanced resource allocations across multiple dynamic MapReduce clusters. SIGMETRICS 2014.
- 13. Iosup and Epema: Grid Computing Workloads. IEEE Internet Computing 2011.
- 14. Iosup et al.: On the Performance Variability of Production Cloud Services. CCGRID 2011.
- 15. Iosup et al.: Performance Analysis of Cloud Computing Services for Many-Tasks Scientific Computing. IEEE TPDS 2011.

Contact Me or Our Team



Collaboration or discussion about Massivizing Computer Systems:

Understanding, designing, deploying, tuning, analyzing, benchmarking distributed systems and ecosystems, including cloud computing and big data systems. Other topics in large-scale distributed systems and performance engineering are welcome.



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@Alosup

https://atlarge-research.com/aiosup/

https://www.linkedin.com/in/aiosup in

VU University, Faculty FEW/building W&N, Room P4.14 De Boelelaan 1081, 1081HV Amsterdam, The Netherlands



VISION: WHAT DOES OUR SOCIETY NEED? ISN'T THIS ALREADY HAPPENING?

"A world where individuals and human-centered organizations are augmented by an automated, sustainable layer of technology. At the core of this technology is ICT, and at the core of ICT are computer ecosystems, interoperating and performing as utilities and services, under human guidance and control."

- People, good orgs = ICT clients:
 - > Fundamental right to ICT
 - > Understanding
- ICT professionals:
 - > Understand and Create
 - > Experiment and Operate

• ICT = ecosystems:

- > Utilities and services
- > Automated
- > Efficient
- > Controlled
- > Ecosystems
- > Human-guided



RJMetrics 💿 BLUECORE

A AMPLITUDE 😫 granify

retention custora

KNEWTON

Clever

X Recombine

KYRUUS

o@0@zymergen

THE COMPLEXITY CHALLENGE

DataGravity

CipherCloud

VECTRA

splunk>

Danasas

nimblesto

соно

Qumulo

Faculty of Sciences

C

CrowdFlowe

WorkFusio

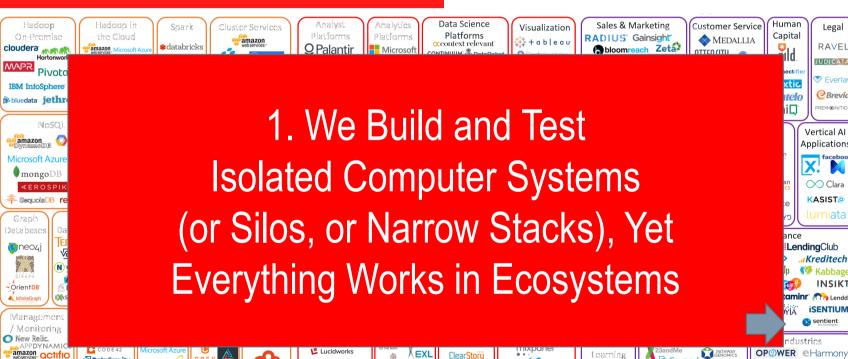
Typesafe

DRIVEN

😓 elastic 💽 ThoughtSp

MAANA 💋 swiftype

Alaolia



ASCIENCE

CIRRO

import io



50

<<1% OF BIG DATA BY MATT TURK (2017)

FLATIRON

Kabbage INSIKT

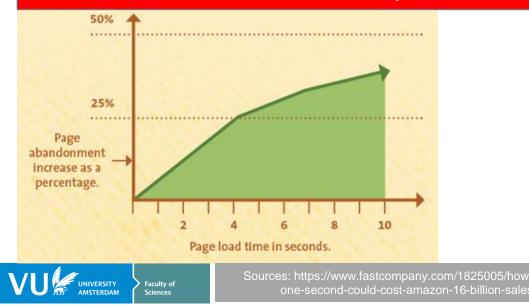
Conductor Conductor

iSENTIUM

duetto

PERFORMANCE, DEPENDABILITY, AND OTHER NON-FUNCTIONAL CHALLENGES

2. We Cannot Even Maintain the Ecosystems we Have Built (and Tested, and Validated)



Google goes dark for 2 minutes, kills 40% of world's net traffic www.theregister.co.uk/2013/08/17/google_outage/

Systemwide outage knocks every service offline



THE RESOURCE MANAGEMENT CHALLENGE



Based on Jav Walker's recent TED talk.

3. Need To Be Much More Efficient,

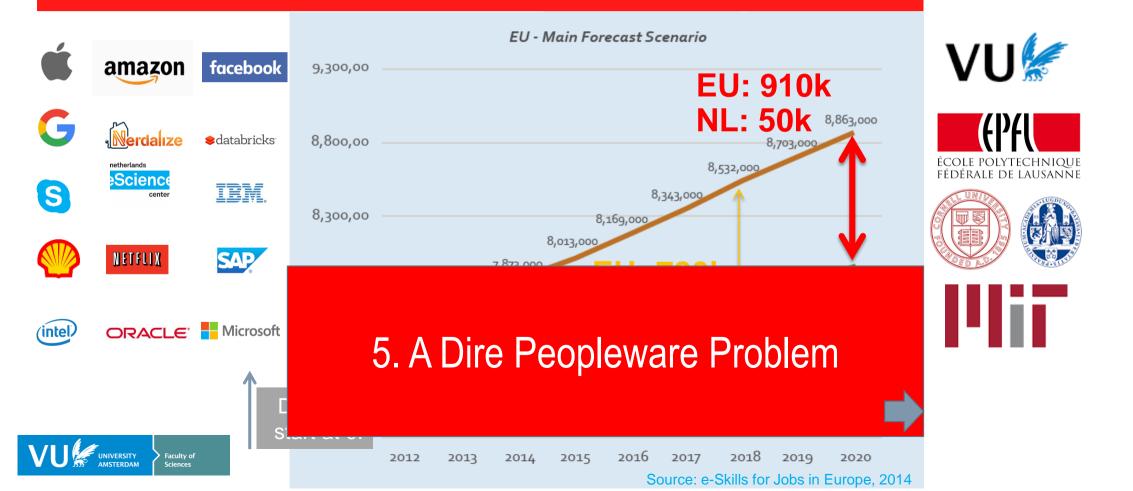
4. Need to Also Be Ethical, and to Also Educate Our Customers

PSY Gangnam consumed ~500GWh

= more than entire countries* in a year (*41 countries),
= over 50MW of 24/7/365 diesel, 135M liters of oil,
= 100,000 cars running for a year, ...

Source: Ian Bitterlin and Jon Summers, UoL, UK, Jul 2013. Note: Psy has >3.5 billion views (last update, May 2018).

THE WORKFORCE GAP, IN THE NETHERLANDS & IN EUROPE



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UNCOVERING THE MYSTERIES OF OUR UNIVERSE

GALILEO GALILEI, 1608-9, 3-8X TELESCOPE

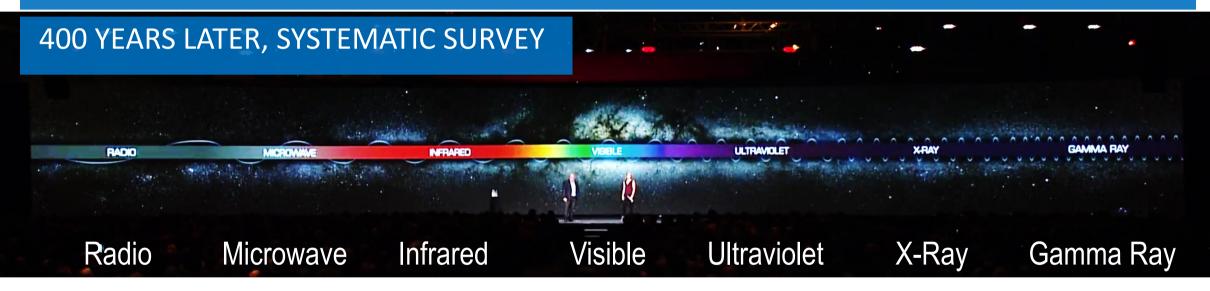
MERELY AN INSTRUMENT?



Garney. The Inquisition's Semicolon: Punctuation, Translation, and Science in the 1616 Condemnation of the Copernican System, ArXiv document 1402.6168. [<u>Online]</u>

Phil Diamond and Rosie Bolton, Life, the Universe & Computing: The story of the SKA Telescope, SC17 Keynote. [Online]

UNCOVERING THE MYSTERIES OF OUR UNIVERSE





Phil Diamond and Rosie Bolton, Life, the Universe & Computing: The story of the SKA Telescope, SC17 Keynote. [Online]



UNCOVERING THE MYSTERIES OF OUR UNIVERSE





James Cordes, The Square Kilometer Array, Project Description, 2009 [<u>Online]</u> The Square Kilometer Array Factsheet, How much will it cost?, 2012 [<u>Online]</u>

Phil Diamond and Rosie Bolton, Life, the Universe & Computing: The story of the SKA Telescope, SC17 Keynote. [Online]

56

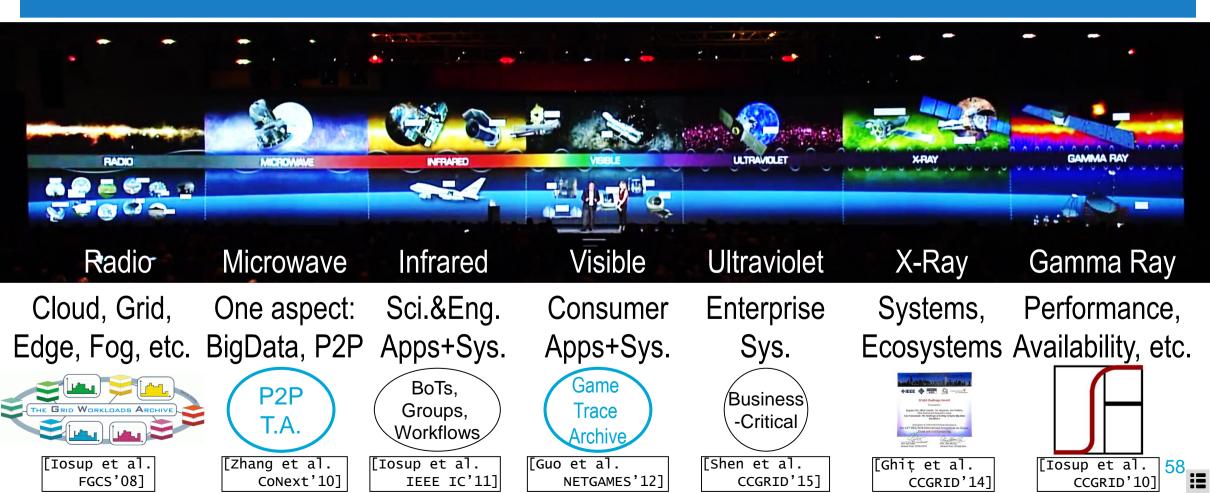
UNCOVERING THE MYSTERIES OF OUR UNIVERSE, PHYSICAL AND DIGITAL



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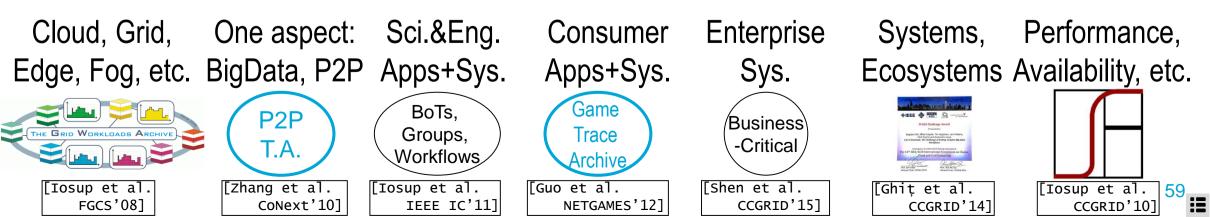
UNCOVERING THE MYSTERIES OF OUR UNIVERSE, PHYSICAL AND DIGITAL





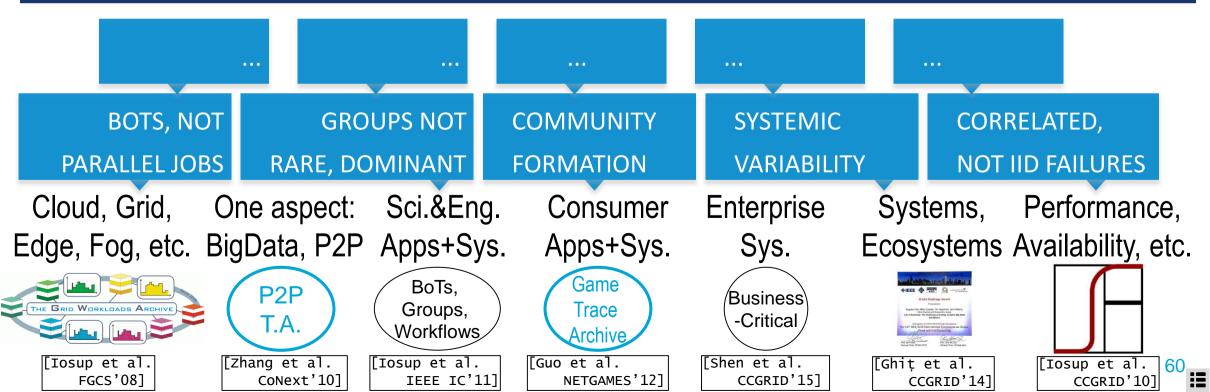
UNCOVERING THE MYSTERIES OF OUR UNIVERSE, PHYSICAL AND DIGITAL

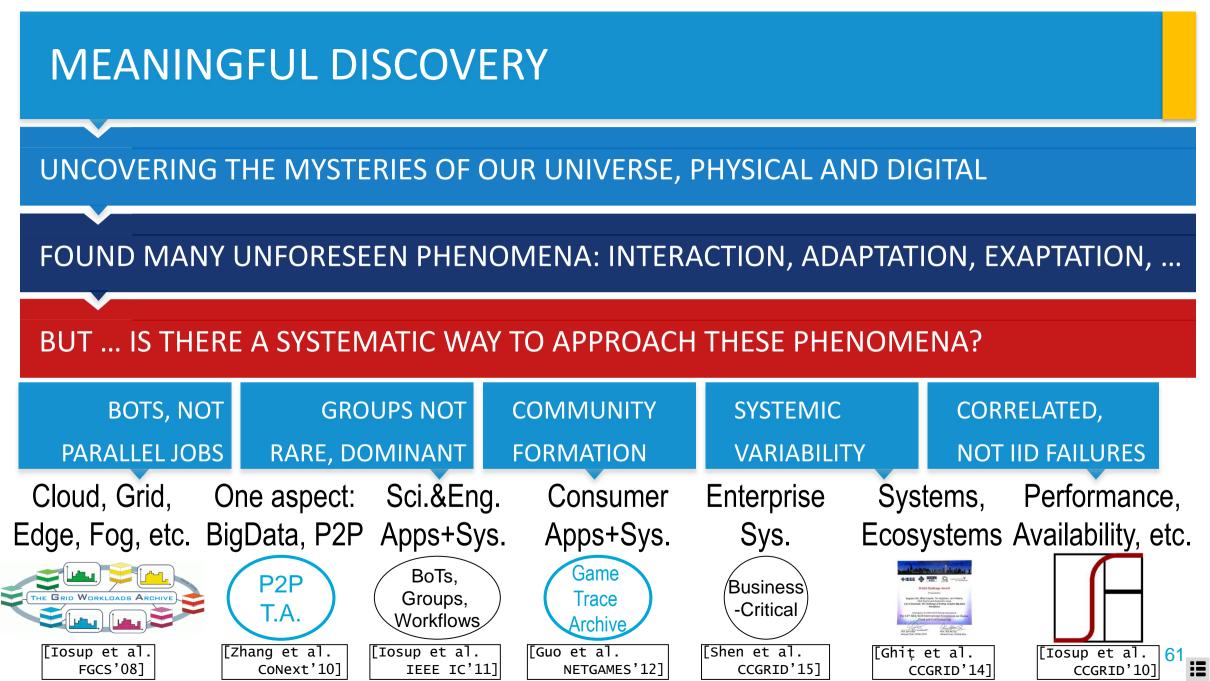
BUT ... WHY WOULD YOU NEED TO UNCOVER AN ARTIFICIAL UNIVERSE?! YOU BUILT IT!



UNCOVERING THE MYSTERIES OF OUR UNIVERSE, PHYSICAL AND DIGITAL

FOUND MANY UNFORESEEN PHENOMENA: INTERACTION, ADAPTATION, EXAPTATION, ...





BUT ... IS THERE A SYSTEMATIC WAY TO APPROACH THESE PHENOMENA?



The Human Genome Project:

FUNDING: > 3B USD

- > Physical map covering >90% human genome
- > Sequence data made available open-access
- Big Science:
 - > Took >10 years to complete
 - > Led by US, work by 20 groups in CN, DE, FR, JP, UK, US
- Big impact:
 - > Decrease cost of sequencing
 - > Facilitate biomedical research

International Human Genome Sequencing Consortium, Initial sequencing and analysis of the human genome, Nature 409, Feb 2011. [<u>Online</u>]

Julie Gould, The Impact of the Human Genome Project, Naturejobs blog, 2015. [Online]

BUT ... IS THERE A SYSTEMATIC WAY TO APPROACH THESE PHENOMENA?

Salos & Markoting luman lustomer Service Legal Capital RADIUS Gainsight **REMEMBER THE COMPLEXITY CHALLENGE?** 🐟 Medallia gild RAVEL Sbloomreach Zeta attensity 🧔 Vevenstring livefyre JUDICATA CLARABRIDGE ker blue vonder **Lattice** CLICKFOX Connectif ©kahuna **√nfer SAILTH** 🖉 Everla STELLAService textic SISENSE ZOOMDAT IBM InfoSphere AZENA TACHYON Core OS pepperdata Digital Reasoning Bottlenose persado AVISO Sente NG 🔅 DATA' 📑 Preact C DOMINO Sense datorama Objection Objective FXU entelo bluedata iethro 🛆 altiscale 🛛 🔲 bole ORBITAL INSIGHT inter ana Stack IQ **û**hat ALGORITHM CHARTIO QUANTIFIND ACTIONI Digital**Genius** hi fuse machines .ENGAGIO appuri *Wise.io* Statistical NoSOL Databases NewSOL Databases BI Platforms log Analytics Ad Optimization Ventical Al Computing Analytics Power BI amazon Clustrix Pivotal SAP splunk Google Cloud Platfor amazon LYLANCE Hootsuite AppNexus ÖRACLE Sas sumologic Ample and a marging the second sec CounterTack cybereasor Microsoft Azure MarkLogic NUODB NETBASE criteo X. Tacebo Wave Analytics 🔘 memsal Threat Metrix. **DATASIFT** kıbana splice **≈**birst DATASTAX SPSS **OpenX** *Procketfuel* AREA 1 mongoDB GoodData SentinelOne VOLTDB Recorded Future tracx bitly CLOUD 👂 Integral 🕐 theTradeDesk MariaDB Clara Couchbase 💫 citusdata 🗖 platfora MATLAB 2 Synthesio Adgorithms dstillerv deapdb Trafodion T Cockroach LAB loggly 🏯 SequciaDB 🛛 redislabs 🚳 influxdata ⊕ atscale KASISTO BISEN simple reach **TFORT**SCALE *****sift**science** LiveIntent DataXU **Oppier** SIGNIFY: MPP Graph Speech & NUE Horizontal Al Real-Time Machine Learnir Database: iransformation Integratio (O) IRI Watson NarrativeScience 🖊 amazon Publisher Govt / Regulation Financo amazon ••••••• 🐣 sentient Teradata alteryx informatica Put potential to work: -Affirm **Lending**Club Google Cloud Pla 🌔 neo4j Tools H₂O C Socrata T METAMARKETS NUANCE W VERTICA VIV amazon vicarious. Microsoft Azure talend Outbrain OnDeck> "Kreditech semantic MuleSof Śstriim Dato Pivotal ARRIA **O OPENGOV** NETEZZA SKYTREE noro 🛃 📩 Numenta **C** TRIFACTA **Tab**²⁰la snapLogic Q **snowflake** 😂 api.ai confluer tamr 👞 p Oction **FN** FiscalNote Cortical.io quantcast \bigcirc BedrockData 🏶 Descartes ຝີສາຳໃສ່ tidemark. Ruc INSIKT deepsenselo Visenz OrientDB **kognitio** DATATORRE StreamSets ذ MindMeld DEXTRO Chartbeat data Artisans 🔽 UOra' 🖪 Dataminr' 👧 Lenddo EXASOL & dremio ctionIO _{glowfils}i 🖌 InfiniteGrad Alation ^a Prec IDIBON (2) **Info**works xplenty MetaMind à yieldbot mark43 AIDŸIÄ **iSENTIUM** KENSHO Search Data Service For Busines Web / Mobile Management Security Storage App Dev Crowd-"" OpenDataSoft 🛯 Quantopian sentient Yieldmo Analysts / Commerce / Monitoring TANIUM amazon webservices sourcing UO OPERA apigee ENDECA Google Analytics 🔀 illumio 🕥 New Relic. S EXALEAD 🔶 OrigamiLogia Life Sciences Industries O Google Cloud mazon mechanical iducation/ Mu Sigma PP DYNAMICS mixpanel CODE42 Microsoft Az Lucidworks 3andMe PATHWAY GENOMICS amazon octifio **EXL OP** WER eHarmony CASK Keen C \$ Learning ClearStory රිසි 🔨 🖻 Counsyl DataGravity Danasas/ RJMetrics 📄 BLUECORE elastic O ThoughtSpot ASCIENCE RetailNext ** Numerify KNEWTON splunk> CipherCloud Recombine duetto 🛸 nimblestora Typesafe A-AMPLITUDE 😫 granify CrowdElowe CIRRO STITCH FIX 🔘 🥏 VECTRA MAANA 🙆 swiftype Clever KYRUUS FLATIRON соно **M** sumAll 🖻 Airtable DATADOG *illi* WorkFusion ©©⊘⊚zymergen HealthTap® BLUEDRIVER K BlueTalon WorkFusio retention custora Trocana DRIVEN import io Qumulo Algolia SINEQU DRIVEN DataKind



<1% OF BIG DATA BY MATT TURK (2017)



D

THE COMPLEXITY CHALLENGE

Focus on Applications, 5 Core Layers:

- 5. Development (Front-end)
- 4. Runtime Engines (Back-end)
- 3. Resources
- 2. Operations Services



		1000									
	Appli	cation									Tools 100% Dev
		igh Level La Iomain-Spe	nguages cific Language	es)				Pig	ŀ	live	2-5'
5	Development (Front-End)	Programn Models	ning				N	lapRedu	ice Mo	odel	
4	Runtime Engines (Back-end)	Execution		Had	оор						Vincent
		lemory &	HD	FS							van Beek
	Netw	ork									
3	Resources			YAF	RN	N	, lesos				
2	Operations Services				Zooke	eper					Tim Hegaman
1	Infrastructure	Physical		tecture/Hierarchy: Pod/Partition, Cluster, R	ack Node	Memory Box	Storage, incl. Tape Robot	Network, incl. F'wall Boxes	Sensor	Virtual,SE	DN/Container

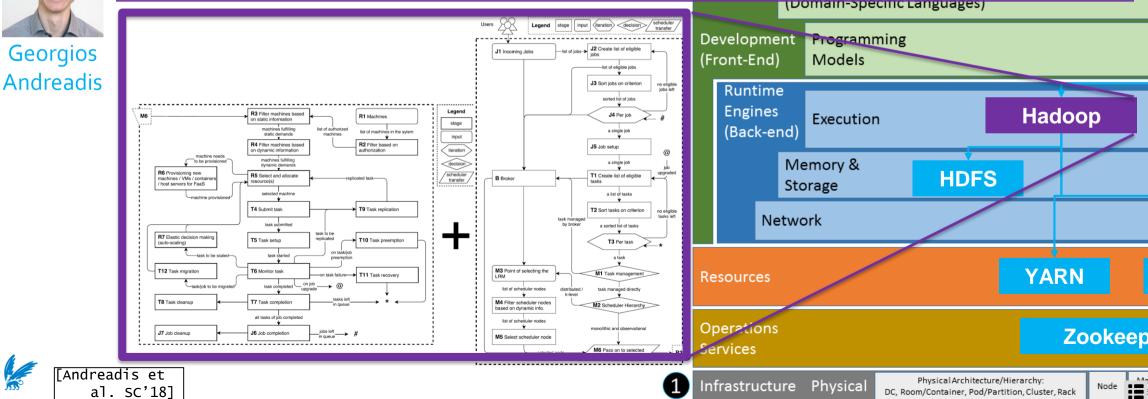
REFERENCE ARCHITECTI

THE COMPLEXITY CHALLENGE

IOSUP ET AL. REFERENCE ARCHITECTURE FOR DCS

B

ANDREADIS ET AL. REFERENCE ARCHITECTURE FOR SCHEDULERS IN DCS



ENGINEERING LDBC GRAPHALYTICS: THE NEED FOR SPEED ... & GRAPHS!



~1 billion vertices ~100 billion connections

Web graph

~100 billion neurons ~100 trillion connections

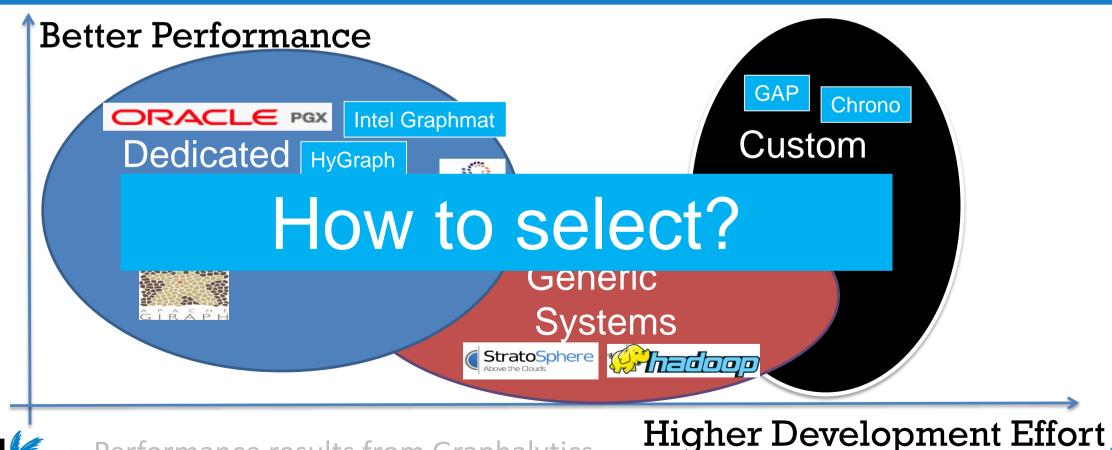
Brain network



Sources: Smith, CHI'10; Blog webpage; Gigandet et al., PLoS ONE 3(12)]



ENGINEERING LDBC GRAPHALYTICS: THE SYSTEMS LANDSCAPE



Performance results from Graphalytics

ENGINEERING LDBC GRAPHALYTICS: BENCHMARKING LEADING TO SCIENCE



The graph & RDF benchmark reference

- Graphalytics:
 - > Benchmark
 - > Many classes of algorithms used in practice
 - > Diverse real and synthetic datasets
 - > Diverse experiments, representative for practice
 - > Renewal process to keep the workload relevant
 - > Enables comparison of many platforms, community-driven and industrial
 - > Global Competition

VU









Wing Lung Ngai

g I

Tim Hegeman









Alex Ută Ahmed Musaafir Mihai Capotã



ENGINEERING LDBC GRAPHALYTICS: BENCHMARKING LEADING TO SCIENCE



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[Iosup et al. PVLDB'16] CCGRID'15] [Guo et al. IPDPS'14]



Community endorsed:

graphalytics.org

Surprising findings:

Performance: orders of magnitude difference due to each of platform, algorithm, dataset, and hardware

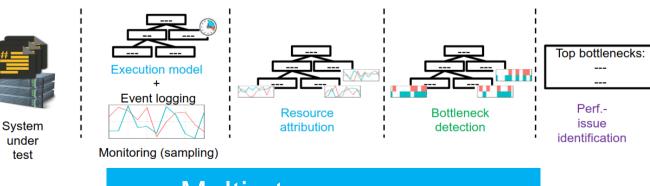
Triggered new research



ENGINEERING LDBC GRAPHALYTICS: MODELING LEADS TO PERFORMANCE ANALYSIS



- Graphalytics Grade10:
 - > Automated bottleneck detection
 - > Automated identification of performance issues





Tim Hegeman

70



Multi-stage process, works in ecosystem

ENGINEERING LDBC GRAPHALYTICS: MODELING LEADS TO PERFORMANCE ANALYSIS



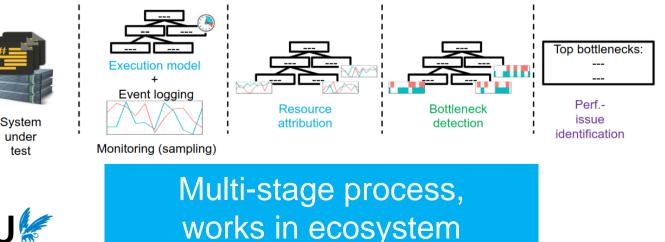
The graph & RDF benchmark reference

Graphalytics Grade10:





- > Automated bottleneck detection
- > Automated identification of performance issues



- No bottleneck at all
- With Grade10:

Always bottleneck Cause: + Message queue full + Garbage collector + CPU + Others

DESIGNING SERVERLESS ARCHITECTURES

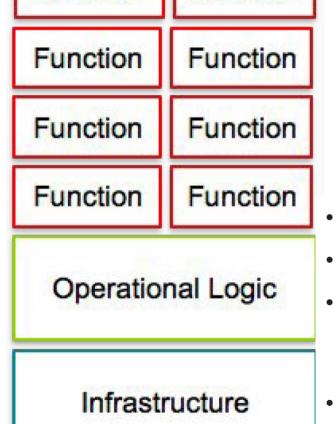
Abstraction: Serverless Design: FaaS systems

Monolithic Application
Operational Logic
Infrastructure

Difficult to Scale Infrequent, Inflexible Complex deployment Tightly coupled stack

μs	μs				
Operational Logic	Operational Logic				
μs	μs				
Operational Logic	Operational Logic				
Infrast	ructure				

- Scalable
- Frequent, Flexible
- Complexity: from application logic to operational logic





Erwin van Eyk

- Lucian Toader
- Scalable
- Frequent, Flexible
- Explicit separation of Business Logic vs. Operational Logic.
- Minimal layer coupling, unit of deployment



DESIGNING SERVERLESS ARCHITECTURES

