

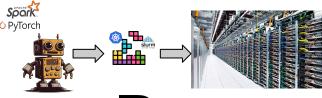
The Cost of Simplicity:



lease/release

migrate

Understanding Datacenter



Scheduler

Programming Abstractions

Aratz Manterola Lasa¹, **Sacheendra Talluri**², Tiziano De Matteis², Alexandru Iosup²

¹WarpStream Labs ²Vrije Universiteit Amsterdam





WHERE

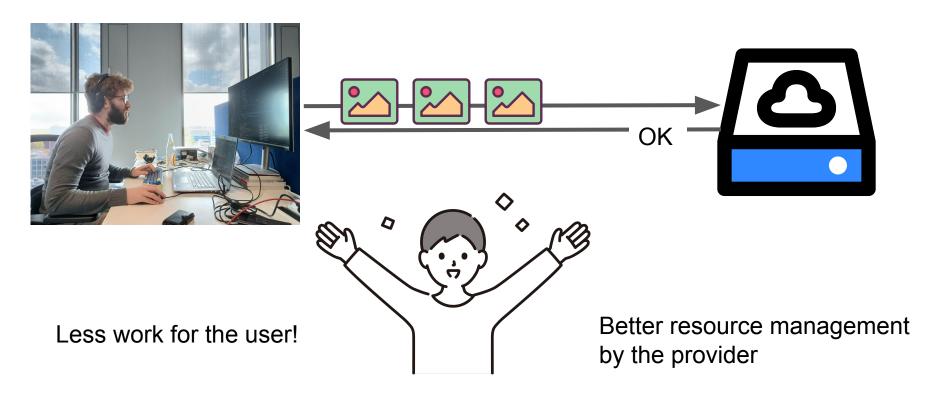
Scheduler (

resource

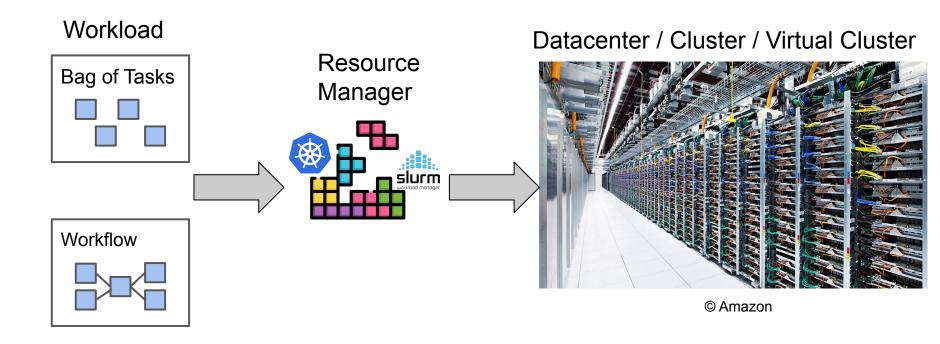
Why APIs/abstractions?



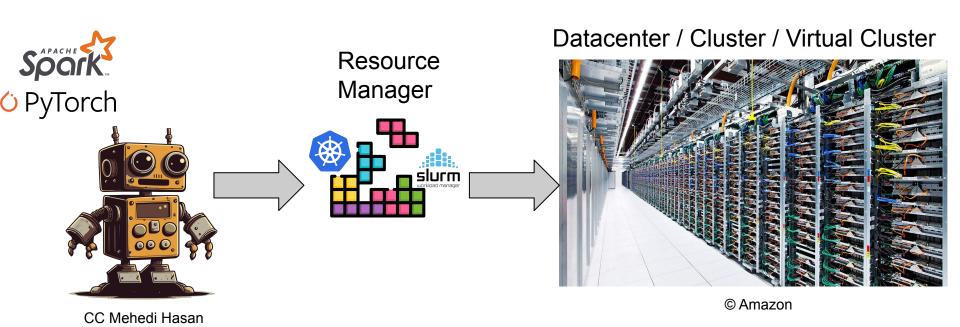
Why APIs/abstractions?



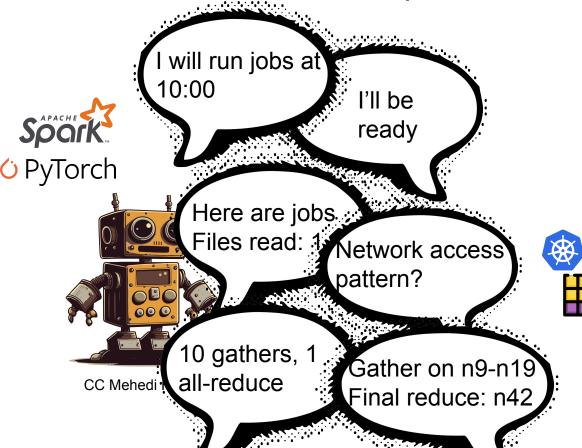
Datacenter System Model



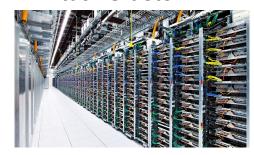
Runtimes Can Use Complex Resource Management APIs



Runtimes Can Use Complex Resource Management APIs



Datacenter /
Cluster /
Virtual Cluster



© Amazon

slurm

Runtimes Can Use Complex Resource Management APIs

Paragon Paragon: QoS-Aware Heterogeneous Datacenters Availabili Remand Data Grid " : {ccurino, cdoug, subru, raghu, sriramra}@microsoft.com, ": djelleleddine.difallah@unifr.cl Decoupling Computation and Data Scheduling in Distributed Data-Intensive Applications Siqi Shen†, Alexandru Iosu ontinuous shift towards data-driven approaches to busi-† Delft University of Technol and a growing attention to improving return on in-Kayitha Ranganathan Jan Foster # The Technion tents (ROI) for cluster infrastructures is generating new Department of Computer Science, University of Chicago, Chicago, IL 60637, USA inges for big-data frameworks. Systems originally de-Math and Computer Science Division, Argonne National Laboratory, Argonne, IL 60439, USA d for big batch jobs now handle an increasingly com-{krangana,foster}@cs.uchicago.edu mix of computations. Moreover, they are expected to ntee stringent SLAs for production jobs and minimize Abstract-Datacenters are at the core of a wide variety daily ICT utilities, ranging from scientific computing to online gaming. Due to the scale of today's datacenters, the failure of computing resources is a common occurrence that may disrupt y for best-effort jobs. CERN [1], which will generate petabytes of scientific Abstract this paper, we introduce reservation-based scheduling, data by 2006. In those experiments, a community of Figure 1: Performance degradation for 5,000 applications of In high energy physics, bioinformatics, and other hundreds of physicists around the world will submit, individually and collectively, ultimately millions of approach to this problem. We develop our solution disciplines we encounter applications involving the availability of ICT services, leading to revenue loss. Although oblivious and baseline least-loaded schedulers compared to ideal scheduling (application runs alone on best platform). Results are d four key contributions: 1) we propose a reservation numerous, loosely coupled jobs that both access and generate large data sets. So-called Data Grids seek to y high availability (HA) techniques have been proposed to k resource failures, datacenter users—who rent datacenter jobs, each accessing some subset of that data. tion language (RDL) that allows users to declaratively harness peopraphically distributed resource. large-scale data-intensive problems. Ye scheduling in such environments is challen a need to address a variety of metrics and (e.g., resource utilization, response time, local allocation policies) while dealing w However, all open-source resource managers offer only simple APIs potentially independent sources of jobs as number of storage, compute, and network res We describe a scheduling framework that these problems. Within this framework, data operations may be either tightly bound scheduling decisions or, alternatively, perfore decounted assucknowns process on the Raajay Viswanathan*

Reservation-based Scheduling: If You're Late Don't Blame Us!

Carlo Curinom, Diellel E. Difallahm, Chris Douglasm, Subru Krishnanm, Raghu Ramakrishnan^m, Sriram Rao^m

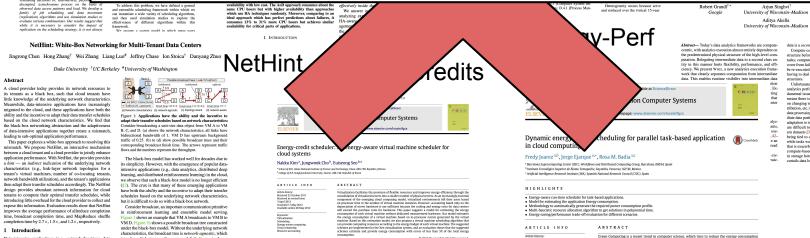
Cloud Information Services Lab (CISL) - Microsoft Corp

Yahoo! [32]. As these architectures and tools become ubio uitous, maximizing cluster utilization and, thus, the return on investment (ROI) is increasingly important.

Characterizing a "typical" workload is nuanced, but the hundreds of thousands of daily jobs run at these sites [36, 40] can be coarsely classifed in two groups

1. Production jobs: These are workflows submitted periodically by automated systems [27, 38] to process data feeds, refresh models, and publish insights. Production jobs are often large and long-running, consuming tens of TBs of data and running for hours. These (DAGs of) jobs are central to the business, and come with strict service level agreements (SLA) (i.e., completion deadlines).

effort jobs: These are ad-hoc, exploratory comput ubmitted by data scientists and engineers engaged ng/debugging ideas. They are typically numerous saller in size. Due to their interactive nature, bestt jobs do not have explicit SLAs, but are sensitive to



Received 1 December 2015

and carbon footprint produced by computers on distributed platforms such as clusters, grids, and clouds. Traditional scheduling solutions attempt to minimize processing times without taking into account the

energetic cost. One of the methods for reducing energy consumption is providing scheduling policies

data is a second-class citizen

Compute-centricity was a natural early choice: knowing job structure beforehand simplifies carving containers to execute tasks: compute-centricity provided clean mechanisms to re cover from failures - only tasks on a failed machine needed to be re-executed; and job scheduling became simple because of having to deal with static inputs, i.e., fixed tasks/dependency

Uber Technologies Inc

Unfortunately, today, compute-centricity severely hinders

analytics performance and cluster efficiency due to four fundamental issues (§2, §9): (1) Intermediate data-unawareness means there is no way to quickly adapt job execution based on changing run-time data properties (e.g., volume, key distribution, etc.) to ensure performance- and resource-optima data processing. (2) Likewise, static parallelism and intermediate data partitioning inherent to compute-centricity preven adaptation to intermediate data skew and resource flux which are difficult to predict ahead of time, yet, common to modern datasets [30] and multi-tenancy. (3) Execution schedules being tied to compute structure can lead to resource waste while tasks wait for input data to become available - an effect that is exacerbated under multi-tenancy. (4) The skew due to compute-based organization of intermediate data can result in storage hotspots and poor cross-job I/O isolation: it also

Research Questions

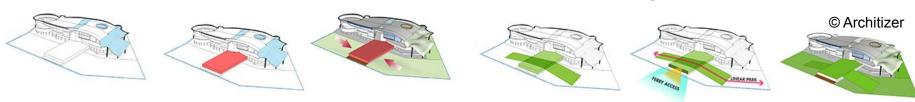
RQ1. What API features are missing from commercial open-source resource managers?

RQ2. What is the performance cost of these missing features?

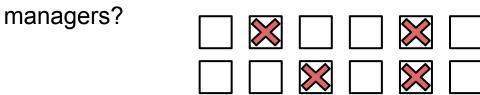


Research Questions

RQ0. What is the reference architecture for resource manager APIs?



RQ1. What API features are missing from commercial open-source resource



RQ2. What is the performance cost of these missing features?



Resource Manager API Ref. Arch. Design Process

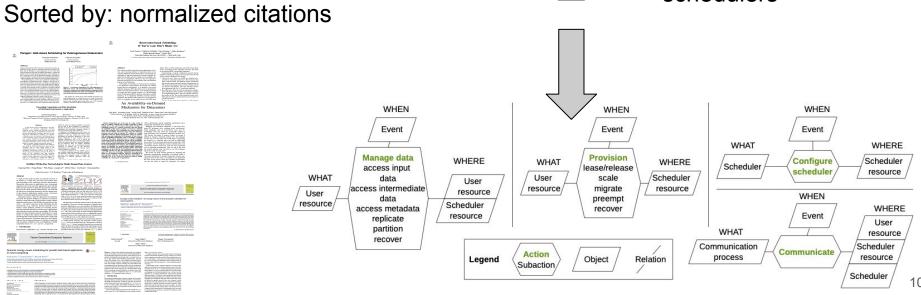
Systematic Literature Survey Keywords eg.: scheduler AND (datacenter OR API)

First 15 papers



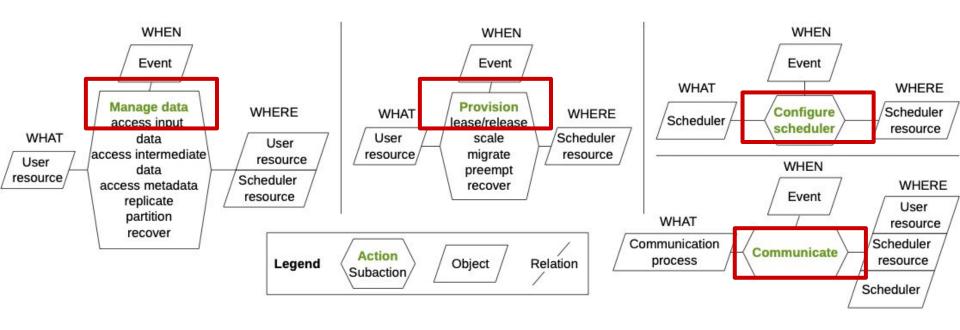
Features from 5 industrial schedulers



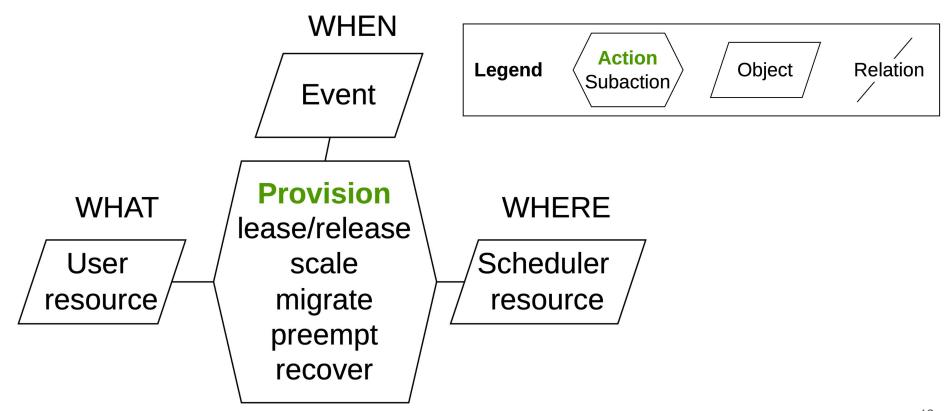


h.ps.//amarge_rescurch.com/pdfs/2024-icpe-datacenter-scheduler.pdf

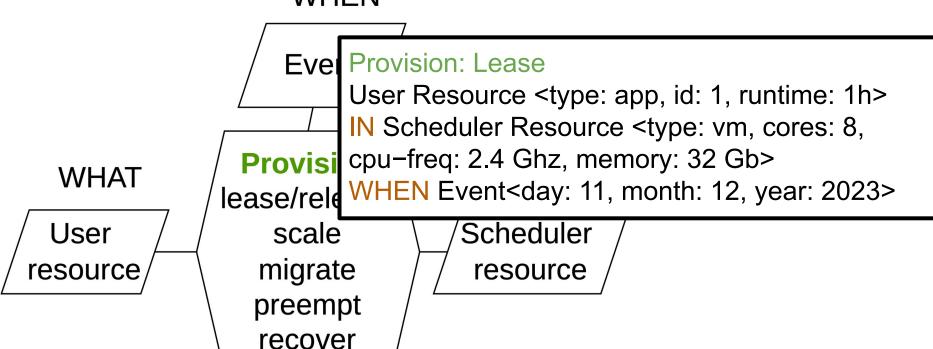
Resource Manager API Reference Architecture



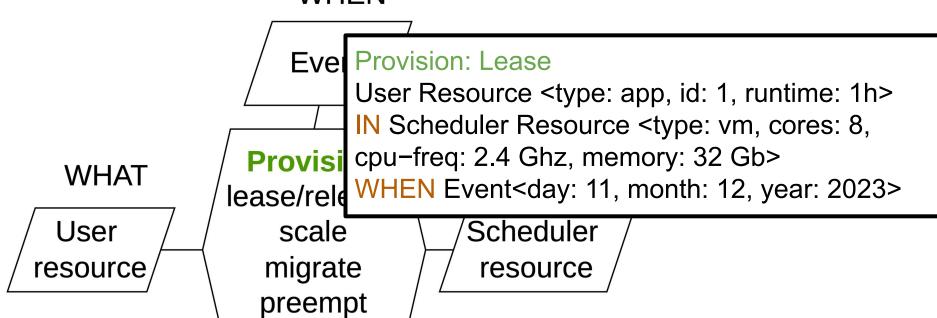
Resource Manager API Reference Architecture - Provision



Resource Manager API Reference Architecture - Syntax WHEN



Resource Manager API Reference Architecture - Syntax WHEN



RQ0. What is the reference architecture for resource manager APIs?

Mapping Industrial Scheduler API Actions to Ref. Arch.

Action	Sub-Action	Schedulers					Ku = Kubernetes
Action		Ku	Sl	Sp	Co	Ai	SI = SLURM
	Lease / release		•			•	Sp = Spark
Provision	Scale		\mathbf{O}		O	O	Co = HTCondor
	Migrate	\circ	\circ	\circ	\circ	\circ	Ai = Airflow
	Preempt			\circ		O	
	Recover						
Configure scheduler		•		•	•	•	■ = Full support
	Access input	•		•	•	•	■ = Partial support
	Access interm.	O	O		O	O	○ = No support
Manage	Access metadata	O	\circ	O	\circ	O	
data	Replicate	O	\circ		O	O	
	Partition	O	\circ		\circ	O	
	Recover		\circ			\circ	
Communicate			•				

APIs Missing in Industrial Schedulers

Action	Sub-Action	Schedulers					Ku = Kubernetes
Action		Ku	Sl	Sp	Co	Ai	SI = SLURM
	Lease / release		•			•	Sp = Spark
	Scale		\mathbf{O}		O	O	Co = HTCondor
Provision	Migrate	O	\circ	O	O	O	Ai = Airflow
	Preempt			O		O	
	Recover						
Configure scheduler Little Su		pport for Data Management					■ = Full support
	Access input						● = Partial support
	Access interm.	O	O		O	O	○ = No support
Manage	Access metadata	O	\circ	\circ	O	O	
data	Replicate	O	\circ		O	O	
	Partition	O	O	•	O	O	
	Recover		O			O	
Communicate			•				

APIs Missing in Industrial Schedulers

Action	Sub-Action	Schedulers						
Action	Sub-Action	Ku Sl		Sp	Co	Ai		
	Lease / release	•	•	•	•	•		
	Scale		O		O	O		
Provision	Migrate	O	O	O	O	O		
	Preempt		•	O	•	O		
	Recover	No Support for Migration						
Configure	scheduler	<u> </u>						
	Access input							
	Access input Access interm.	• •		•	•	•		
Manage	_	• • •) ()		OO	• • •		
Manage data	Access interm.	OOO		• •	OOO	9		
C	Access interm. Access metadata		0 0 0	• • • • • • • • • • • • • • • • • • •		0		
C	Access interm. Access metadata Replicate					0		

Ku = Kubernetes SI = SLURM Sp = Spark Co = HTCondor

● = Full support

Ai = Airflow

- **▶** = Partial support
- = No support

APIs Missing in Industrial Schedulers

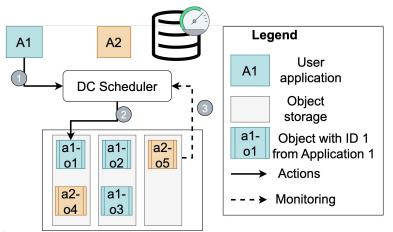
Action	Sub-Action	Schedulers					Ku = Kubernetes
Action		Ku	S1	Sp	Co	Ai	SI = SLURM
	Lease / release	•	•			•	Sp = Spark
	Scale		O		O	O	Co = HTCondor
Provision	Migrate	O	O	O	O	O	Ai = Airflow
	Preempt		•	O	•	O	
	Recover	No S	uppo	rt for	Migra		
Configure scheduler							■ = Full support
	Access input	•		•	•	•	▶ = Partial support
	Access interm.	O	O		O	O	○ = No support
Manage	Access metadata	O	O	O	O	O	
data	Replicate	\circ	\circ		O	O	

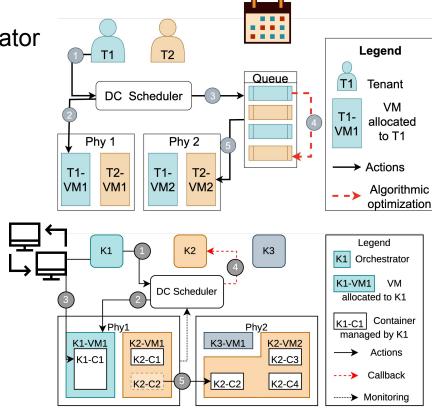
RQ1. What API features are missing from commercial open-source resource managers?

Cost of Missing APIs - Experiments

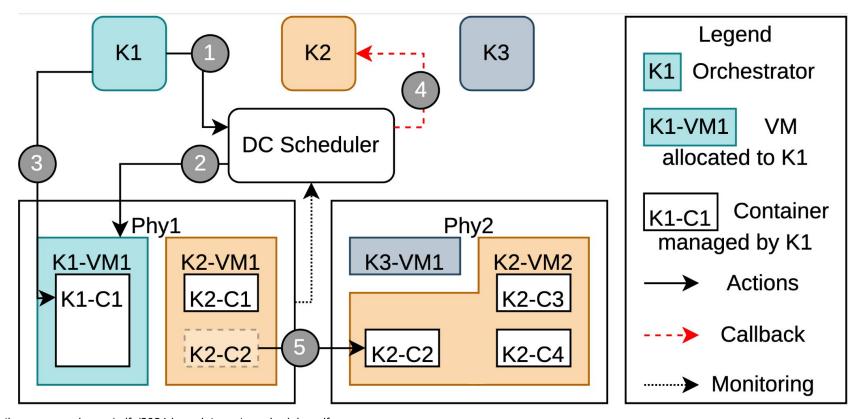
3 Experiments Using the OpenDC Simulator

- Resource Reservation
- Container Migration Storage Metadata
- 3.

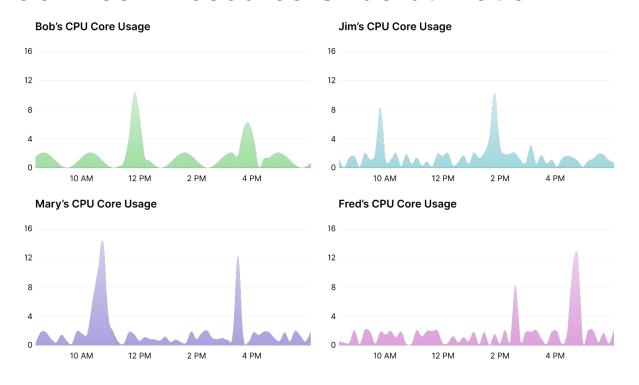




Container Migration Experiment - Setup



Virtual Machines - Resource Underutilization

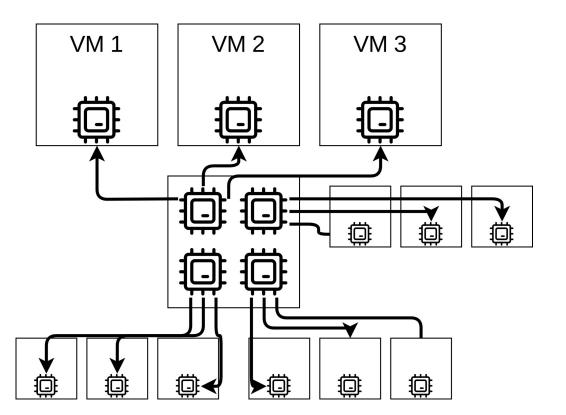


Oversubscription

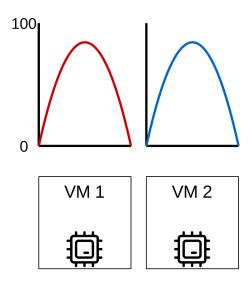
1 CPU for 3 VMs

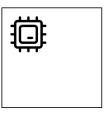
Oversubscription ratio = 3

Time sharing

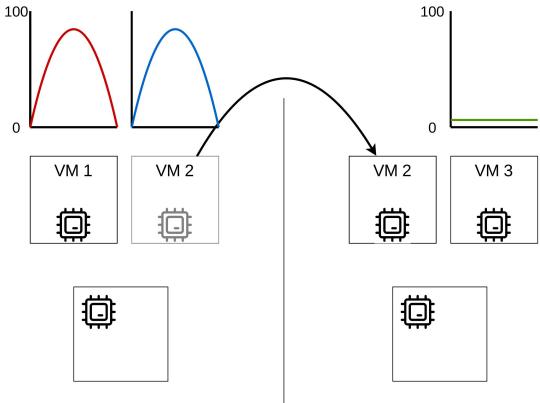


Oversubscription - Resource Contention

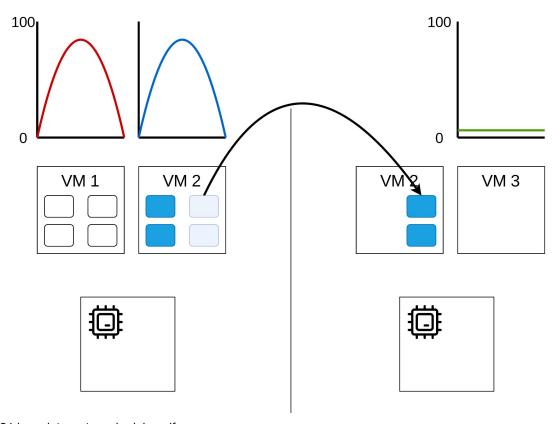




Migration



Container Migration



Container Migration Experiment - Setup

Cluster Setup:

1 Physical Cluster

5 Kubernetes Virtual Clusters

512 Mbps migration speed

80% average CPU utilization target

Physical cluster size calibrated for each trace

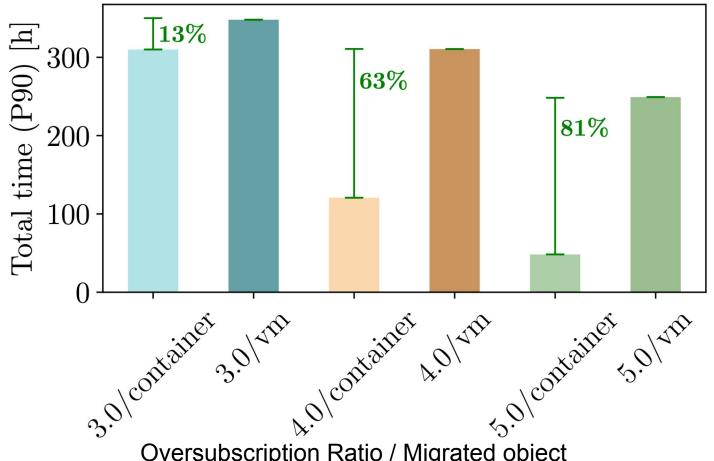
3 Traces:

Google 2011 (~25 machines)

Azure 2017 (~65 machines)

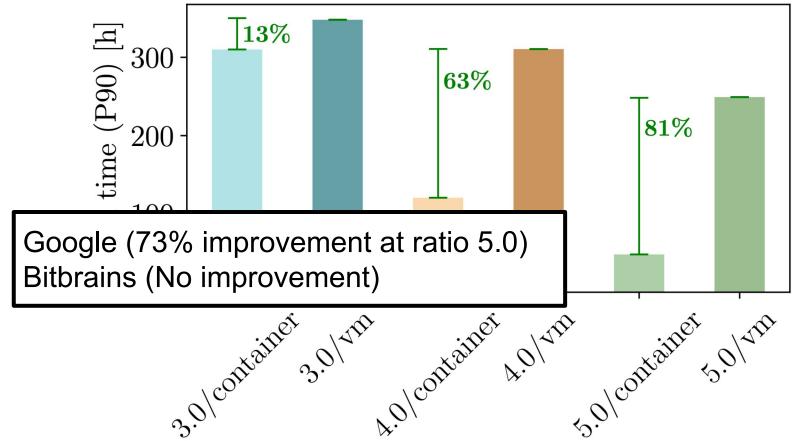
BitBrains 2015 (~100 machines)

Container Migration - Results - Azure P90 App Runtime



Oversubscription Ratio / Migrated object

Container Migration - Results - Azure P90 App Runtime



Oversubscription Ratio / Migrated object

Key Takeaways

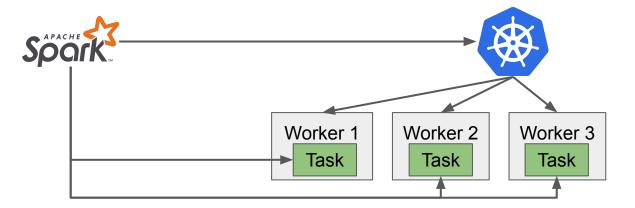
- Distributed Systems Runtimes can take advantage of complex APIs for resource management
- Current open-source resource managers are missing complex resource management abstractions
 - Migrations, Data Management
- Complex abstractions offer performance benefits for some workloads, but not all
 - 17% to 81% improve in 90th percentile app runtime using container migration callbacks for Azure trace, but not the Bitbrains trace

Why didn't these APIs make it into industrial schedulers?

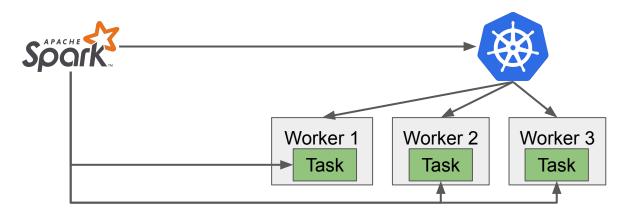
- 1. New APIs/scheduling techniques only work for specific workloads
 - a. Widely-deployed schedulers need generality

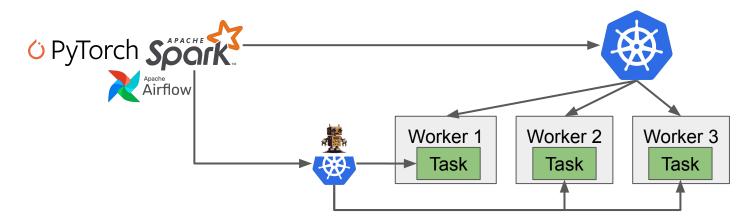
- Need to implement separately for different systems and system versions
 - a. Can't use Spark 3.2 scheduler with Spark 2.4
 - b. Cannot use a SLURM scheduler with Kubernetes

Future Work - Kubernetes-based Embeddable Scheduler

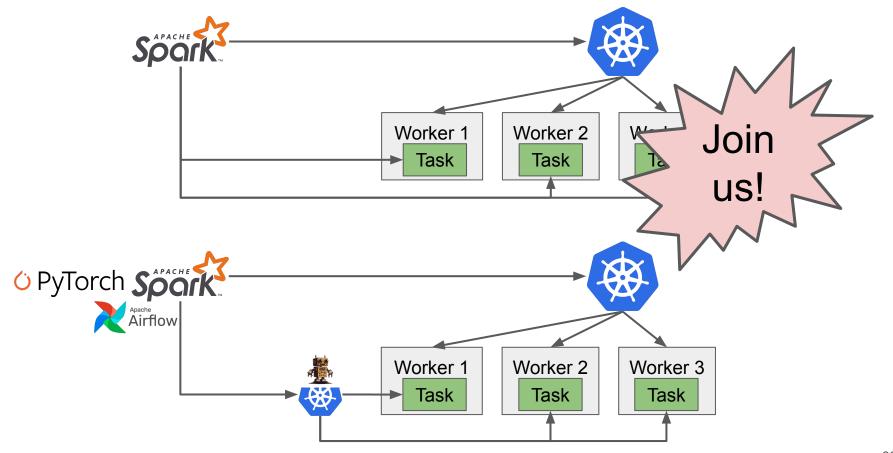


Future Work - Kubernetes-based Embeddable Scheduler





Future Work - Kubernetes-based Embeddable Scheduler



Key Takeaways

- Distributed Systems Runtimes can take advantage of complex APIs for resource management
- Current open-source resource managers are missing complex resource management abstractions
 - Migrations, Data Management
- Complex abstractions offer performance benefits for some workloads, but not all
 - 17% to 81% improve in 90th percentile app runtime using container migration callbacks for Azure trace, but not the Bitbrains trace

Further Reading

[this work] Talluri, S., Herbst, N., Abad, C., De Matteis, T., & Iosup, A. (2024). ExDe: Design space exploration of scheduler architectures and mechanisms for serverless data-processing. Future Generation Computer Systems, 153, 84-96.

[related work on scheduler APIs] Manterola Lasa, A., Talluri, S., De Matteis, T., & Iosup, A. (2024, May). The Cost of Simplicity: Understanding Datacenter Scheduler Programming Abstractions. In Proceedings of the 15th ACM/SPEC International Conference on Performance Engineering (pp. 166-177).

[using the tools] Talluri, S., Herbst, N., Abad, C., Trivedi, A., & Iosup, A. (2023, May). A Trace-driven Performance Evaluation of Hash-based Task Placement Algorithms for Cache-enabled Serverless Computing. In Proceedings of the 20th ACM International Conference on Computing Frontiers (pp. 164-175).

[reference architecture] Andreadis, G., Versluis, L., Mastenbroek, F., & Iosup, A. (2018, November). A reference architecture for datacenter scheduling: design, validation, and experiments. In SC18: International Conference for High Performance Computing, Networking, Storage and Analysis (pp. 478-492). IEEE.