ABS: Modeling and analysis with resource-sensitive actors

Silvia Lizeth Tapia Tarifa

University of Oslo, Norway sltarifa@ifi.uio.no

Oslo, 13 November 2023







http://www.sirius-labs.no

Main idea:

- Describe a system using a language.
- Describe the properties or analysis of the system that one wants to do.
- Use a systematic method (with tool support) to check the properties or do some other kind of analysis (e.g., correctness, reachability analysis, time-related analysis, resource analysis, what-if scenarios, optimization, planning, etc.).

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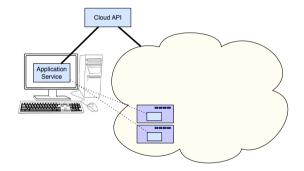
Main topic of the day:

- Modeling with an actor-based language.
- Elasticity/scalability of the deployment of a system: performance vs. cost
- Resource-usage predictions using simulations.

• Question: Can we use models to predict resource usage for applications running on the cloud?

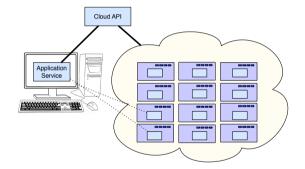
- Question: Can we use models to predict resource usage for applications running on the cloud?
- Starting point: actors a computation model which decouples execution from synchronization
- ABS: modeling language which adds resource-sensitive behavior to actors
- We look at how to enrich this model to make simulation of models of cloud-deployed services
- ABS has been used to model Hadoop clusters, industrial cloud applications, Kubernetes clusters, ...

We want to make effective use of cloud computing to meet service requirements



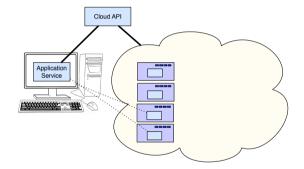
- Virtualization makes elastic amounts of resources available to application-level services
- Metered resources: Resources on the Cloud are pay-on-demand

We want to make effective use of cloud computing to meet service requirements



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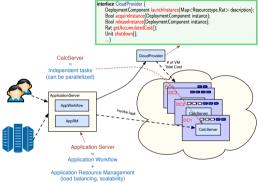
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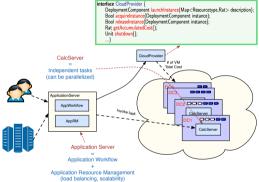
Resource-aware design

• **Goal:** Build software that can dynamically modify its own deployment to improve performance and/or reduce cost



Resource-aware design

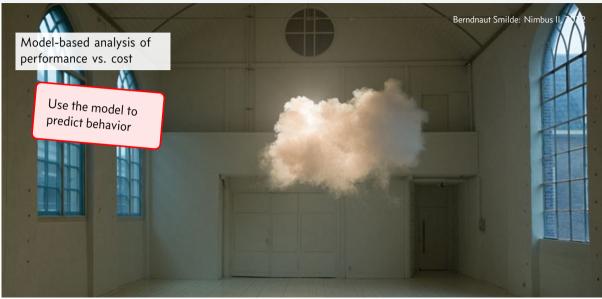
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Discovering **bad design** after deployment on the Cloud can be a very costly (wasting both time and money!)









- 1. How will response time and cost of running my system change if I double the number of servers?
- 2. Can I meet my performance requirements with my current deployment strategy? What about fluctuations in client traffic?
- 3. Can I control the performance of my system better by using a custom resource manager?



Client Layer

Provisioning Layer



Executable Model of Client Layer

Provisioning Layer

SLA	Legal Contract
	Service Contract

Executable Model of Client Layer Cloud API Provisioning Layer

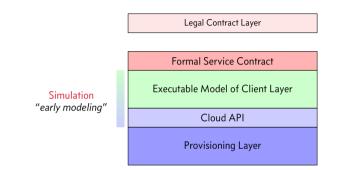
Legal Contract Layer

Formal Service Contract

Executable Model of Client Layer

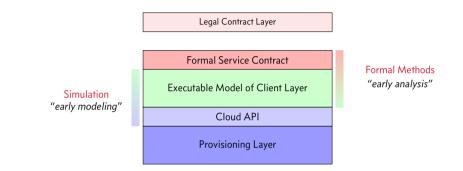
Cloud API

Provisioning Layer



Combine techniques based on abstract executable models

• Modeling using Abstract Behavioral Specifications (ABS)



Combine techniques based on abstract executable models

- Modeling using Abstract Behavioral Specifications (ABS)
- Formal methods: Performance Analysis, Cost Analysis

Actors and ABS

Actors: Software for a concurrent world

• Actor model [Hewitt, Baker 77]: *actors* are the universal primitive of concurrent computation.



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- When receiving a message, an actor can react by making local decisions, creating actors, sending messages, and deciding how to respond to the next message.



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• Actors may modify their own private state, but can only affect each other indirectly through messaging.

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ABS explores the asynchronous features of loosely coupled actors in the setting of formal models

- State-of-art programming language concepts
 - ADTs + functions + objects + interfaces
 - Formal semantics, type-safety, data race-freeness by design

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Upper tier: asynchronous, no shared state, actor-based Lower tier: shared state, cooperative scheduling

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- Modeling of variability/resources with first-class language support
 - Variability: feature models, delta-oriented programming
 - Deployment: abstract resources and cost annotations

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ABS is designed with analysis/code generation tools in mind

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 Models follow the execution flow of OO programs, but abstract from implementation details using ADTs

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ABS is a formal, tool-supported modeling language

- Formal language: operational semantics, type system
- Backends: compiles into Java COGs, Erlang Actors
- Maude interpreter formalizes semantics as a rewrite system

Asynchronous Method Calls

ABS decouples communication and synchronization

• ABS supports asynchronous method calls, using futures.

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f1=x!m(); f2=y!p(f1);

Flexible synchronization: blocking or suspending activities

- Blocking the object: get-operation on a future: v=f1.get;
- Suspending the process: **polling** a future: **await** f1?
- Polling as part of a (per object) guarded command: await b & f1? & f2?

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Cooperative scheduling of method activations

• Easy to combine active and reactive behavior

Cooperative Scheduling in ABS

Important Consequences

- Task suspension is a syntactically explicit decision of the modeller
- No preemptive scheduling \Rightarrow **no data races**
 - Scheduling cannot arbitrarily interfere with the computation
- Non-deterministic scheduling otherwise
 - User-defined specification of schedulers via annotations
 - Analysis results for ABS programs are valid for any scheduler

Cooperative Scheduling in ABS

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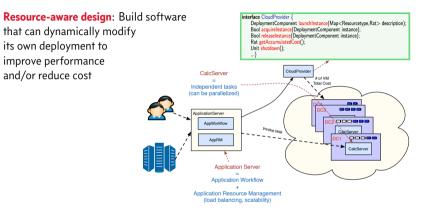
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Reading Futures

- f.get blocks execution until result is available, then reads future f
- Deadlocks possible (use static analyzer for detection)
- Programming idiom: use await f? to prevent blocking (safe access)

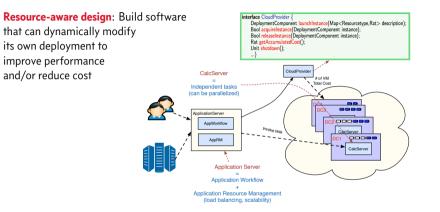
Modeling and Analysis of Deployment on the Cloud Using Real Time ABS

Services deployed on the cloud: Predicting behavior from models



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Services deployed on the cloud: Predicting behavior from models



- Model-based deployment decisions at design time using service models
- Formal semantics: Architects and developers can simulate and analyze at design time how an application runs on the cloud

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To express and compare resource usage, we need a notion of time in our models!

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Timed semantics

- Let us assume that ABS programs execute under a **maximal progress** timed semantics
- Our model has a global clock
- Assumption: Execution is "infinitely fast" unless we say otherwise



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duration(x,y);

await duration(x,y);

while $(timeValue(now()) - x < 60) \{ skip \} ;$

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Time x = **now**();

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duration(60,60);

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ABS: Modeling with resource-sensitive actors

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Time
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duration(x,y);

await duration(x,y);

while $(timeValue(now()) - x < 60) \{ skip \} ;$

duration(60,60);

await duration(60,70);

Deployment components are abstract "cost centers"

Ĵ,	Ĵ,		
	Server object 1 [cost] Task1 [cost] Task2	 object n	objectEnv [cost] Task1 [cost] Task2

Deployment components are abstract "cost centers"

- Each deployment component has a given resource capacity
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- *How* resources are assigned and consumed, depends on the kind of resource
- For cloud computing, let us focus on processing capacity

Example: Phone Services - Abstract Behavioral Model

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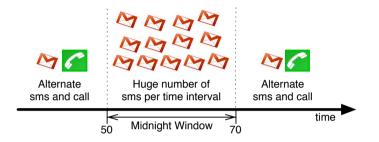


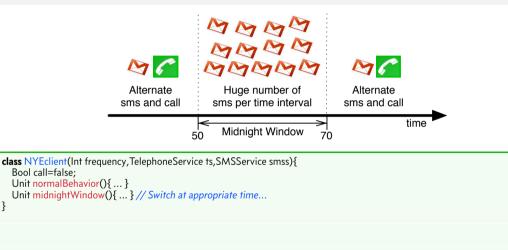
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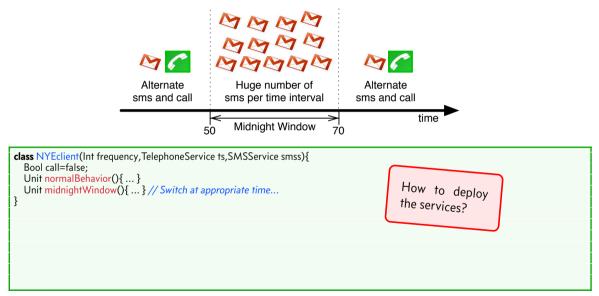


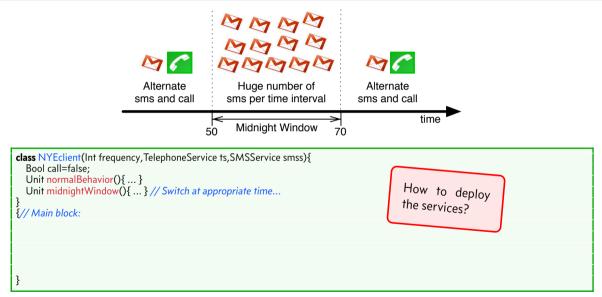
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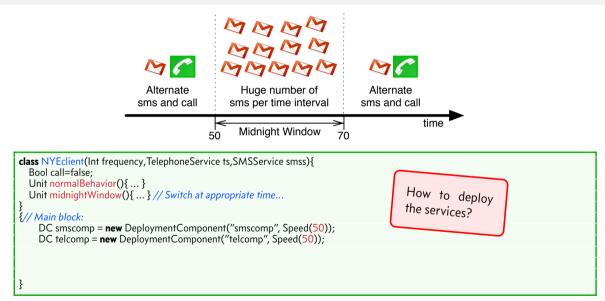


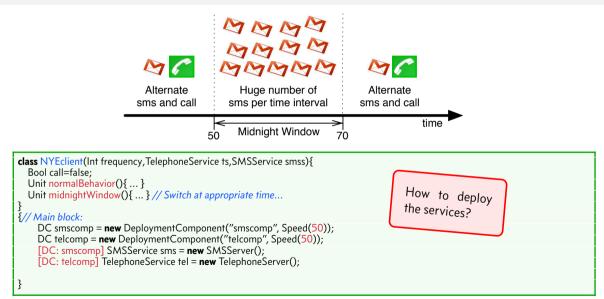


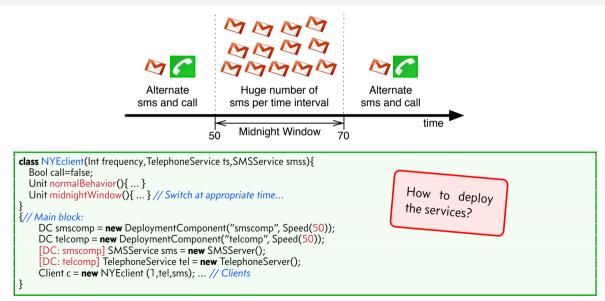
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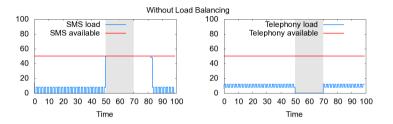




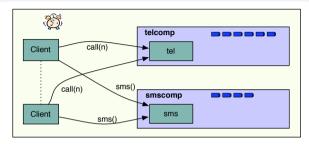
Example: NYEclient

```
class NYEclient (Int frequency, TelephoneServer ts, SMSServer smss) {
  Bool call = False:
  Unit normalBehavior() {
     if (timeValue(now()) > 50 && timeValue(now()) < 70) { this!midnightWindow(); }
     else {
       if (call) { await ts!call(1); } else { smss!sendSMS(); }
       call = \sim call:
       await duration(frequency, frequency); this!normalBehavior(); }
  3
  Unit midnightWindow() {
     if (timeValue(now()) >= 70) { this!normalBehavior(); }
     else {
       Int i = 0:
       while (i < 10) { smss!sendSMS(): i = i + 1: }
       await duration(1,1); this!midnightWindow(); }
  }
  Unit run(){ this!normalBehavior(); }
}
```

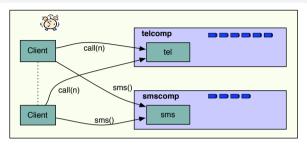
Example: Simulation Results



Load Balancing with Vertical Scaling



Load Balancing with Vertical Scaling

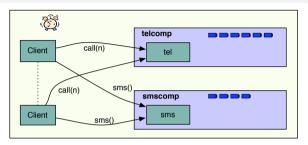


Resource awareness: monitor & react

- dc.**load**(rtype,e): average load on dc during the last e time intervals
- dc.total(rtype): currently allocated resources on dc
- dc.**transfer**(dc2, r, rtype): transfer r resources to dc2

• thisDC: reference to my deployment component

Load Balancing with Vertical Scaling

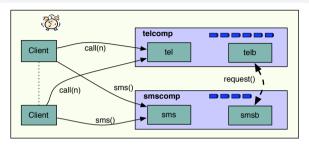


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Load Balancing Strategy for the Phone Services

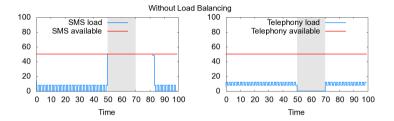
Reallocate $1/2 \times total$ resources upon request from partner

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- Reaction: resource reallocation, object mobility, job distribution

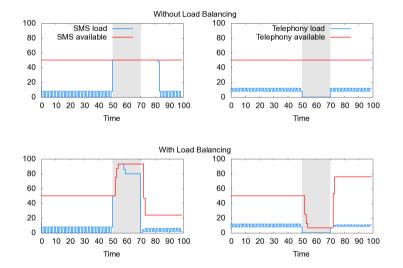
Exercise 4: Solution

```
class Balancer() implements Balancer {
  Balancer partner = null;
  Unit run() {
    await partner != null;
    while (True) {
        await duration(1, 1);
        Rat Id = await thisDC()!load(Speed, 1);
        if (Id > 90) { await partner!requestdc(thisDC()); }
  Unit requestdc(DC comp) {
    InfRat total = await thisDC()!total(Speed);
    Rat Id = await thisDC()!load(Speed, 1);
    if (Id < 50) { // we know total will not be InfRat
      thisDC()!transfer(comp, finvalue(total) / 2, Speed);
  Unit setPartner(Balancer p) { partner = p; }
}
```

Example: Simulation Results



Example: Simulation Results





BREAK: 15 minutes

ABS modeling abstractions

- Dynamically created objects
- Very flexible synchronization: asynchronous method calls and futures
- Execution with time and resources
- Deployment components
 - can be dynamically created

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All the building blocks we need to model an abstract **cloud provider** class

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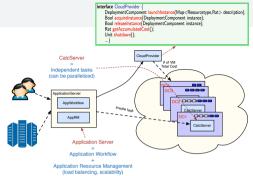
All the building blocks we need to model an abstract cloud provider class

- DC launchInstance(Rat speed){...}
- Bool acquireInstance(DC instance){ ... } // start dc
- Bool releaseInstance(DC instance){ ... }
- Rat getAccumulatedCost() { ... }
- Unit shutdown(DC instance){ ... }

// create dc // start dc // stop dc

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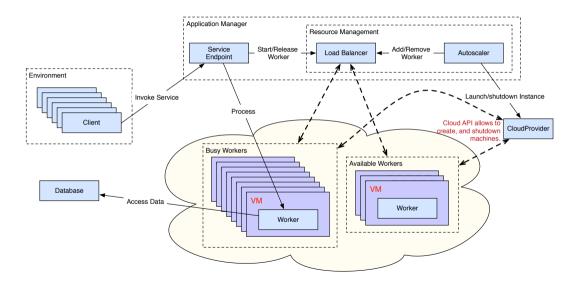


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Elastic Cloud Computing Architecture



Scenario

We assume given a class Worker. Define a Load Balancer class which implements round robin load balancing

- A field: List<Worker> available = list[];
- A field: List<Worker> inuse = list[];
- Implement a method getWorker(){...} which returns a Worker object
- Implement a method Unit releaseWorker(Worker w){...}
- Implement a method Unit addWorker(Worker w){...}

You can assume operations on lists: head, tail, isEmpty, appendright, without

```
class RoundRobinLoadBalancer() implements LoadBalancer {
  List<Worker> available = list[]:
  List<Worker> inuse = list[]:
  Worker aetWorker(){
    await (!isEmpty(available));
    Worker w = head(available):
    available = tail(available);
    inuse = appendright(inuse,w);
    return w;
  }
  Unit releaseWorker(Worker w){ available = appendright(available,w); inuse = without(inuse,w); }
  Unit addWorker(Worker w) { available = appendright(available, w); }
}
```

Example: Workers

Let us define tasks and workflows

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Let us define tasks and workflows

type TaskID=Int; data Workflow=Workflow(List<Task> tasks); data Task=Task(TaskID tID, Rat cost, List<TaskID> dependencies); Let us define tasks and workflows

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Scenario

Define a Worker class

- Implements a method processTask which receives a task
- The Worker has already been started,
- It should stop when execution is completed (i.e., returned to the load balancer)

Workers

Here, we assume given

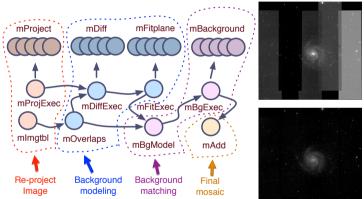
- a load balancer lb,
- futures for all tasks in the dependencies,
- a deadline for task execution, and
- a starting time.

```
class WorkerObject(LoadBalancer Ib) implements Worker {
    Bool processTask(Rat taskCost, Time started, Duration deadline, List<Fut<Bool>> dependencies){
    while(dependencies != Nil) {
        Fut<Bool> dep = head(dependencies);
        dependencies = tail(dependencies);
        await dep?;
    }
    [Cost: taskCost] skip; // This is the part we abstract from!
    Rat spentTime = timeDifference(now(),started);
    Ib!releaseWorker(this);
    return (spentTime <= durationValue(deadline));
</pre>
```

Validating the Models

Case Study: Montage (1)

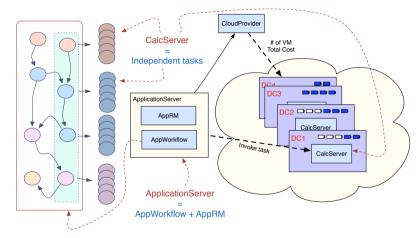
Montage is a toolkit for assembling astronomical images into customized mosaics http://montage.ipac.caltech.edu/



Partly ordered workflow and highly parallelizable tasks

SIRUS SIRUS S. L. Tapia Tarifa (U. Oslo)

Case Study: Montage (2)

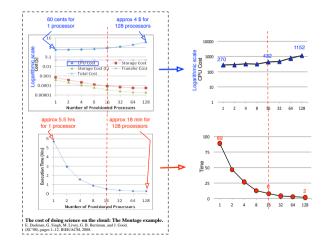


Model the Montage toolkit using the **Cloud Provider API**, run simulations varying the deployment scenario, and compare the results.

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Case Study: Montage (3)

Cost vs. Time Tradeoff: can we reproduce the results of other informal cloud simulation tools (GridSim)?



Case Study: Big Data Processing Frameworks (1)

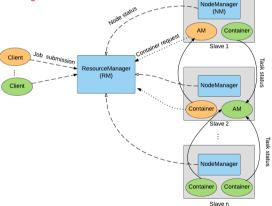


Open-source software framework that implements a cluster management technology for distributed processing.

Popular cloud framework for big data processing:

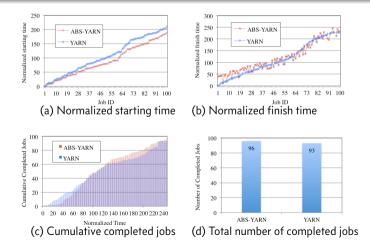
Hadoop YARN and SPARK Clusters:

- Resource allocation
- Code distribution
- Distributed data processing
- Streaming data



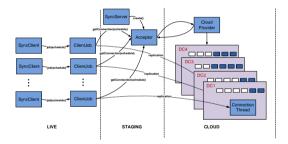
Case Study: Big Data Processing Frameworks (2)

How does the ABS YARN compare to the actual YARN implementation?



Case Study: Fredhopper Replication Server (1)

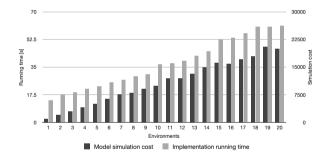
The Fredhopper Access Server (FAS) is a distributed, concurrent OO system providing search and merchandising services to e-Commerce companies. The Replication Server is one part of FAS.



- Very detailed model: consists of 5000 lines of ABS
- Model API: use actual system logs to run simulations
- SLA monitoring framework for failure metrics at end-points

Case Study: Fredhopper Replication Server (2)

How does the accumulated cost in our model compare to the actual Java implementation?



Measured execution time of the implementation (left scale) Accumulated cost of the simulation (right scale)

The deviation roughly seems to correspond to the start-up time of JVM

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Kubernetes

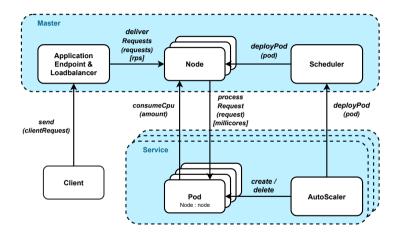


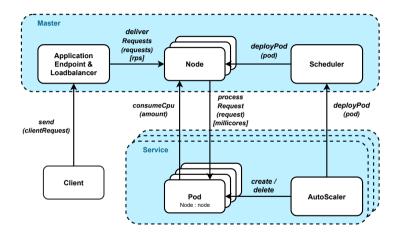


- Service: logical collection of load-balanced pods, with given service end-point & load balancer
- Container: units of deployment, corresponds to Workers in the previous slides
- *Pods*: scheduling unit, group of containers (one service container + sidecar and helper containers for health probing etc)
- Nodes: Explicit resource capabilities (CPU, memory, etc), contain pods
- *Scheduler* assigns pods to nodes based on resource requirements and capacities

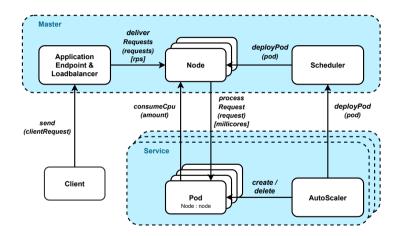


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- *Load balancing*: choice of loadbalancer crucial to the performance of a Kubernetes cluster (layer-4, layer-7)

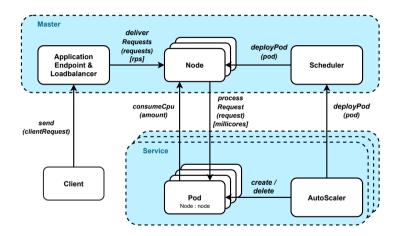




Challenge: A cluster can support many workflows.



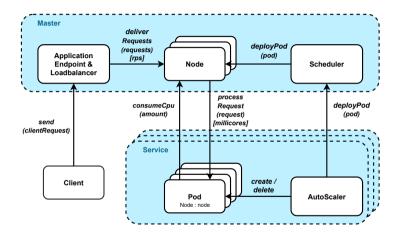
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Problem: Pods are not fully isolated.

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Challenge: A cluster can support many workflows. Solution: Costs depend on wf and node configuration

Problem: Pods are not fully isolated. Solution: Costs modelled in tables, depend on wf/node config

Methodology

- 1. Instrument the cluster
- 2. Identify suitable workflows
- 3. Identify node configurations
- 4. Define a sampling strategy for service loads to derive cost tables
- 5. Perform model-based predictions by means of simulations

Building a Kubernetes Model from an Application

Methodology

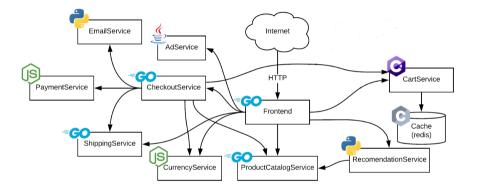
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Cost tables

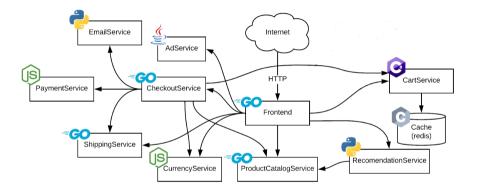
- The cost table captures resource consumption on a node for specific configurations.
- The cost table stores information about resource consumption for each workflow and service for different RPS entries

(workflow, serviceName,RPS) \mapsto cost

Example: Google's Online Boutique Microservices Demo



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- Workflows: get index page, change currency, view random product, ...
- Workflows like these invovle a surprising number of services!
- We obtain very good prediction models for static deployments!
- Source: Google microservices demo

Work with executable actor models!

- ABS models are **easy to understand** and **very efficient** for exploring designs and deployment decisions for distributed software
- Actors, asynchronous method calls: express flexible synchronization patterns
- Basic building blocks for resource-aware models: Cost annotations and deployment components
- Service end-points, load balancers, autoscaling etc etc easy to model
- Methodology for how to model of a real system: a concrete Kubernetes cluster with diverse workflows

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- Service end-points, load balancers, autoscaling etc etc easy to model
- Methodology for how to model of a real system: a concrete Kubernetes cluster with diverse workflows
- ABS is open source and well documented: http://www.abs-models.org

Einar Broch Johnsen, Rudolf Schlatte, Silvia Lizeth Tapia Tarifa: Modeling Resource-Aware Virtualized Applications for the Cloud in Real-Time ABS. ICFEM 2012: 71-86 [Preprint]

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Gianluca Turin, Andrea Borgarelli, Simone Donetti, Ferruccio Damiani, Einar Broch Johnsen, Silvia Lizeth Tapia Tarifa: Predicting resource consumption of Kubernetes container systems using resource models. J. Syst. Softw. 203: 111750 (2023) [Preprint]