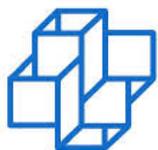


Optimizing Resource Allocation for Workflow Execution in Multi-site Clouds

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Motivation

Comparative Genomics (CG)

- A bioinformatics domain used for exploring complete genomes.
- Phylogenetics, phylogenomics and evolutionary analysis play an important role in CG.
 - composed by the execution of several programs in a coherent flow of activities:
 - (i) identify homologues sequences
 - (ii) construct phylogenetic trees
 - (iii) construct phylogenomic trees
 - (iv) infer hypotheses about evolutionary relationships.



Motivation

- Most of scientific workflows execute on traditional HPC environments.
- Many users do not have access to large clusters or grids.
 - They have adopted clouds to execute their experiments in parallel.



Main Advantage of Clouds

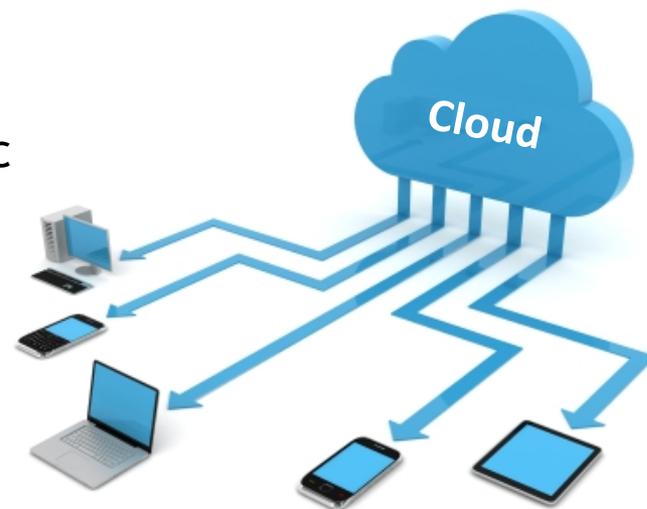
- huge variety of resources for the general public offered via Internet:

virtual machines (VMs)

storage

as-a-service model

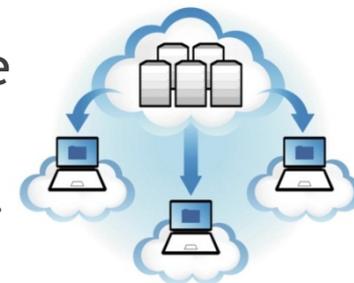
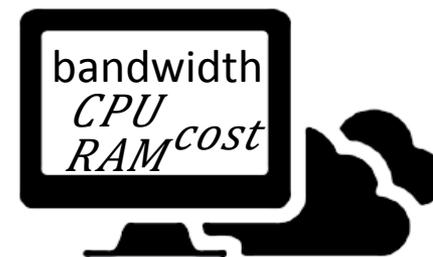
on demand model



Motivation

Infrastructure-as-a-Service

- Large number of offered VM types
 - each VM type has its own characteristics such as bandwidth, CPU, memory and financial cost.
- These VMs can be used to create a **virtual cluster** in the cloud to execute the workflow in parallel.
 - new VMs can be instantiated according to an estimated workload.
- **Problem: what is the amount of VMs to instantiate of each type and for how long?**



Instâncias on demand padrão	
Pequena (padrão)	\$0.060 por hora
Médio	\$0.120 por hora
Grande	\$0.240 por hora
Extragrande	\$0.480 por hora
Instâncias on demand padrão de segunda geração	
Extragrande	\$0.500 por hora
Dupla extragrande	\$1.000 por hora
Microinstâncias on demand	
Micro	\$0.020 por hora
Instâncias on demand com mais memória	
Extragrande	\$0.410 por hora
Dupla extragrande	\$0.820 por hora
Quádrupla extragrande	\$1.640 por hora
Instâncias on demand com CPU de alta performance	
Médio	\$0.145 por hora
Extragrande	\$0.580 por hora
Instâncias de computação em cluster	
Quádrupla extragrande	\$1.300 por hora
Ótupla extragrande	\$2.400 por hora
Instâncias on demand de cluster com mais memória	
Ótupla extragrande	\$3.500 por hora
Instâncias de GPU de cluster	
Quádrupla extragrande	\$2.100 por hora
Instâncias on demand com E/S elevada	
Quádrupla extragrande	\$3.100 por hora

Motivation

- Some solutions for dimensioning the virtual cluster for scientific applications were proposed

SciDim

- Based on genetic algorithms
- Non-optimal estimations
- Embedded existing SWfMS

GraspCC

- Near-optimal estimations
- HPC applications
- Not designed for scientific workflows

- Non-optimal estimations may produce over and under dimensioning.

Goal

- To analyze the feasibility of the GraspCC approach for dimensioning the virtual cluster for this class of bioinformatics experiments.



Applying GraspCC in Workflows

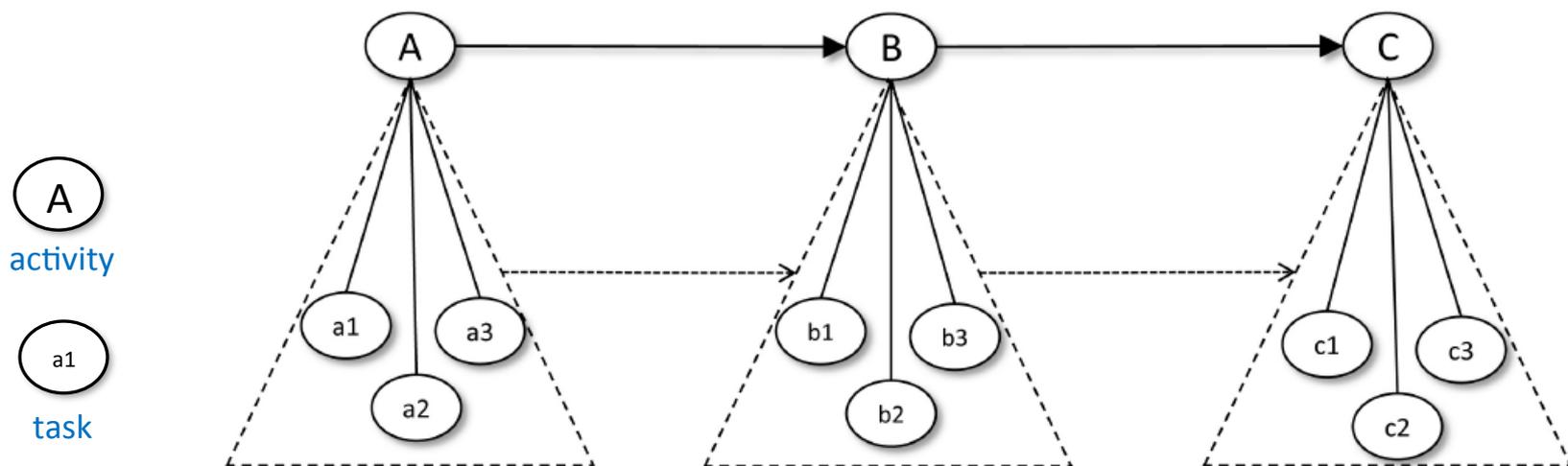
- Underestimation or Overestimation



- It is fundamental to find the **best tradeoff** between performance and financial cost.
- **GraspCC**
 - provides **optimal or near-optimal estimations** of the amount VMs to be instantiated for general application executions in a **reasonable time**.
 - but... no data dependency among programs is considered.

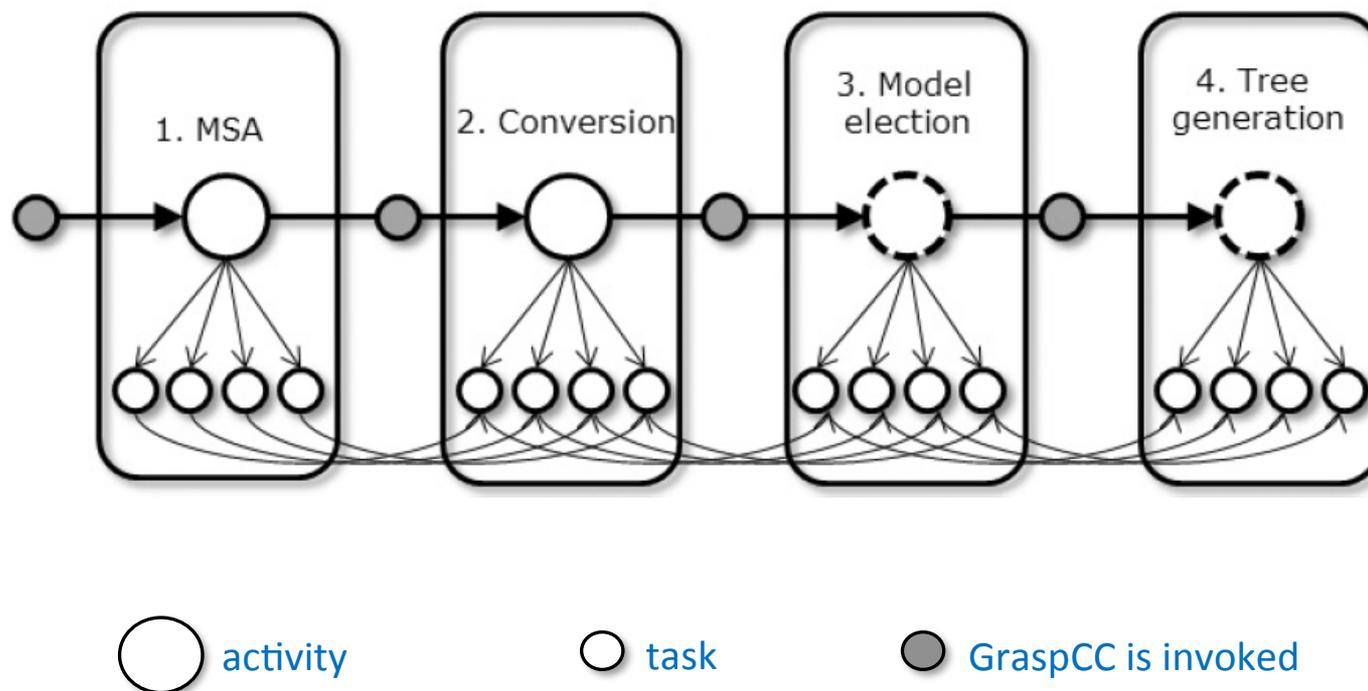
Applying GraspCC in Workflows

- The use of GraspCC was **adapted** for scientific workflow engine.
- GraspCC was invoked immediately before each activity execution
 - First Activity First (FAF) dataflow strategy.
 - each activity is divided into several tasks that can be executed in parallel in different VMs.



Applying GraspCC in Workflows

- SciPhy workflow execution with the invocations of GraspCC



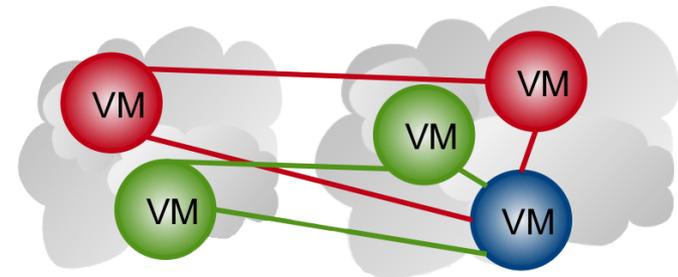
Applying GraspCC in Workflows

Modeling the problem...

➤ Let \mathcal{P} be the set of VMs types offered by a cloud provider during a set of time periods.

➤ Each VM type $p \in \mathcal{P}$ has:

- cost $c \downarrow p$
- computing resources:
 - disk storage $d \downarrow p$
 - memory capacity $m \downarrow p$
 - processing power of $g \downarrow p$ Gflop per period of time



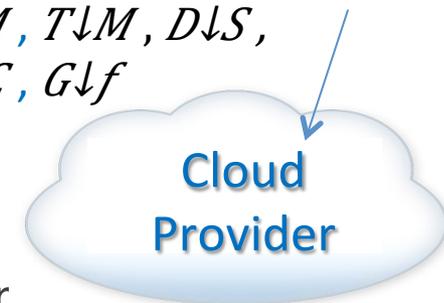
Applying GraspCC in Workflows

Modeling the problem...

- The set of user requirements is defined by:
 - maximum cost
 - maximum time
 - disk storage
 - memory capacity
 - processing demand of Gflop
- Let τ , the maximum limit of VM that can be purchased in each period of time.
- Let t be the last period that a VM was selected.



$C \downarrow M, T \downarrow M, D \downarrow S,$
 $M \downarrow C, G \downarrow f$



er in

Modeling

<i>Notation</i>	<i>Description</i>
P	the set of virtual machine types
C_M	the maximum financial cost requirement
T_M	the maximum execution time requirement
D_S	the disk storage requirement
M_C	the memory capacity requirement
G_f	the processing demand requirement
c_p	the cost of purchasing the virtual machine p for one period of time
d_p	the disk storage of virtual machine p
m_p	the memory capacity of virtual machine p
g_p	the processing power of virtual machine p
N_M	the maximum limit of allocated virtual machine for each scientist in each period of time
x_{pit}	$x_{pit} = 1$, if and only if virtual machine i of type p is allocated at time t
t_m	the last time period that a virtual machine was allocated by the scientist

$$\min(\alpha_1 \sum_{p \in P} \sum_{i=1}^{N_M} \sum_{t \in T} c_p x_{pit} + \alpha_2 t_m) \quad (1)$$

$$\sum_{p \in P} \sum_{i=1}^{N_M} \sum_{t \in T} c_p x_{pit} \leq C_M \quad (2)$$

$$\sum_{p \in P} \sum_{i=1}^{N_M} d_p x_{pit} \geq D_S x_{p'it}, \quad \forall t \in T, \forall p' \in P, \quad (3)$$

$$\forall i' \in \{1, \dots, N_M\}$$

$$\sum_{p \in P} \sum_{i=1}^{N_M} m_p x_{pit} \geq M_C x_{p'it}, \quad \forall t \in T, \forall p' \in P, \quad (4)$$

$$\forall i' \in \{1, \dots, N_M\}$$

$$\sum_{p \in P} \sum_{i=1}^{N_M} \sum_{t \in T} g_p x_{pit} \geq G_f \quad (5)$$

$$\sum_{p \in P} \sum_{i=1}^{N_M} x_{pit} \leq N_M, \quad \forall t \in T \quad (6)$$

$$t_m \geq t x_{pit}, \quad \forall t \in T, \forall p \in P, \quad (7)$$

$$\forall i \in \{1, \dots, N_M\}$$

$$x_{pit+1} \leq x_{pit}, \quad \forall t \in T, \forall p \in P, \quad (8)$$

$$\forall i \in \{1, \dots, N_M\}$$

$$x_{pi+1t} \leq x_{pit}, \quad \forall t \in T, \forall p \in P, \quad (9)$$

$$\forall i \in \{1, \dots, N_M - 1\}$$

$$x_{pit} \in \{0, 1\}, \quad \forall t \in T, \forall p \in P, \quad (10)$$

$$\forall i \in \{1, \dots, N_M\}$$

$$t_m \in \mathbb{Z} \quad (11)$$

Applying GraspCC in Workflows

Modeling the problem...

- Define a *cost function* which will measure the quality of the solution:

$$F(s) = (\alpha_1 \sum_{(p,i,t) \in s} c_p + \alpha_2 t_m(s))$$
$$+ \lambda_1(\max\{0, t_m(s) - T_M\}) + \lambda_2(\max\{0, \sum_{(p,i,t) \in s} c_p - C_M\})$$

Applying GraspCC in Workflows

GraspCC

- The heuristic *GraspCC* is composed of two phases:
 - a construction phase *coCC*
 - a local search phase *lsCC*
- GraspCC consists to perform the *coCC* following by the *lsCC* until the maximum number of iterations without improvement in the best solution found is satisfied.

Modeling Comparative Genomic Workflows

SciHm: a workflow for homologue sequence identification

- First workflow that has to be executed in order to identify candidate drug targets for a specific disease.
- Performs a cross validation procedure to evaluate the specificity and sensibility of each Multiple Sequence Alignment (MSA) method.

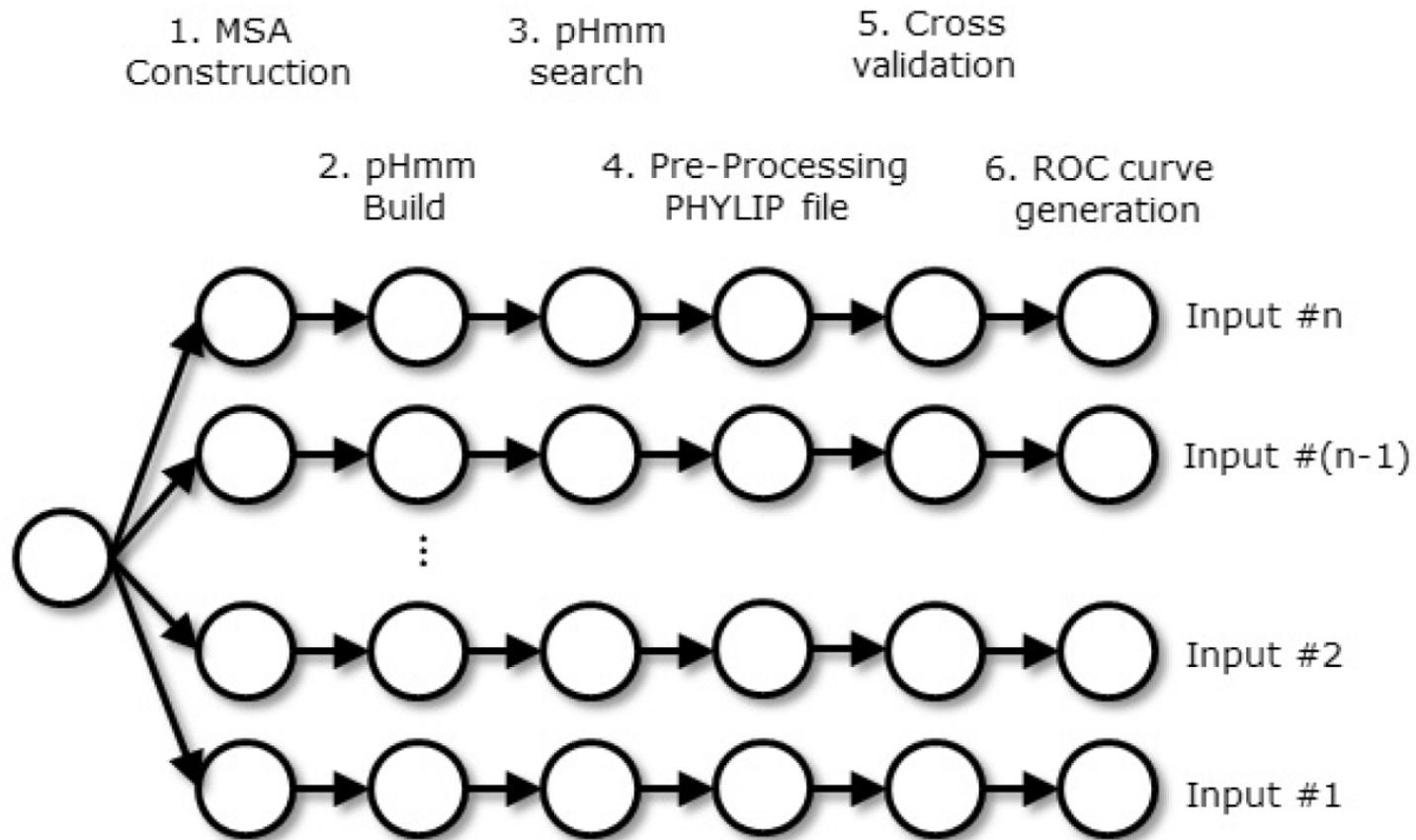
OCAÑA, K. ; OLIVEIRA, D. ; DIAS, J. ; OGASAWARA, E. ; MATTOSO, M. L. Q. . Optimizing Phylogenetic Analysis Using SciHm Cloud-based Scientific Workflow. In: The seventh IEEE e Science conference, 2011, Estocolmo. Proceedings of the seventh IEEE e Science conference. New York: IEEE Computer Society, 2011.

Modeling Comparative Genomic Workflows

SciHmm: a workflow for homologue sequence identification

- Composed by five activities:
 1. MSA Construction
 2. Construction of Profile Hidden Markov Model (pHMM)
 3. pHMM search against a target database
 4. Cross-validation analysis
 5. Generation of Receiver-Operating Characteristic (ROC) curves.

Modeling Comparative Genomic Workflows



Modeling Comparative Genomic Workflows

SciPhy: a workflow for phylogenetic analysis

- A phylogenetic analysis workflow that produces phylogenetic trees.
- First step towards identifying candidate drug targets enzymes in genomes.

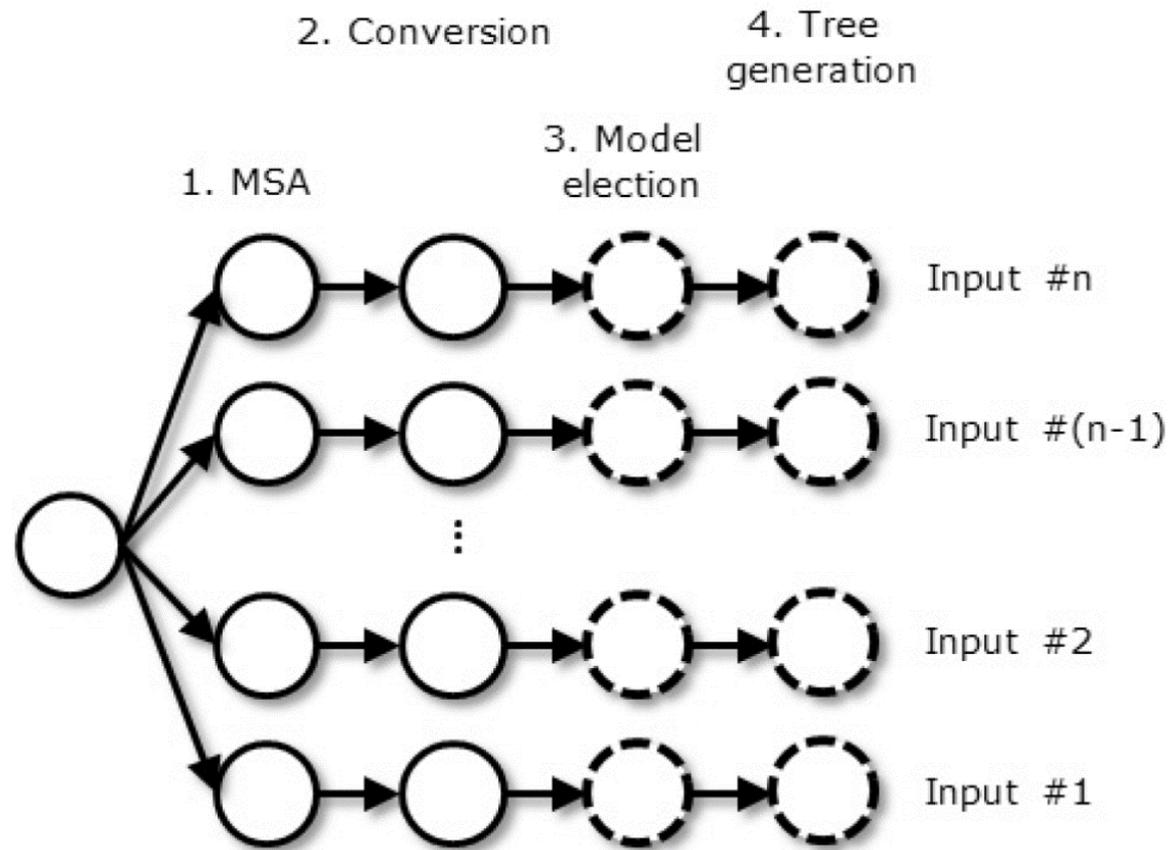
OCAÑA, K. ; OLIVEIRA, D. ; OGASAWARA, E. ; DAVILA, A. M. R. ; LIMA, A. A. B. ; MATTOSO, M. L. Q. .
SciPhy: A Cloud-based Scientific Workflow for Phylogenetic Analysis of Drug Targets in Protozoan Species. In: Brazilian Symposium of Bioinformatics 2011, 2011, Brasilia, DF. Proceedings of the Brazilian Symposium of Bioinformatics 2011, 2011.

Modeling Comparative Genomic Workflows

SciPhy: a workflow for phylogenetic analysis

- Composed by four activities:
 1. Sequence alignment
 2. Sequence conversion
 3. Search for the best evolutionary model
 4. Construction of the phylogenetic trees.

Modeling Comparative Genomic Workflows



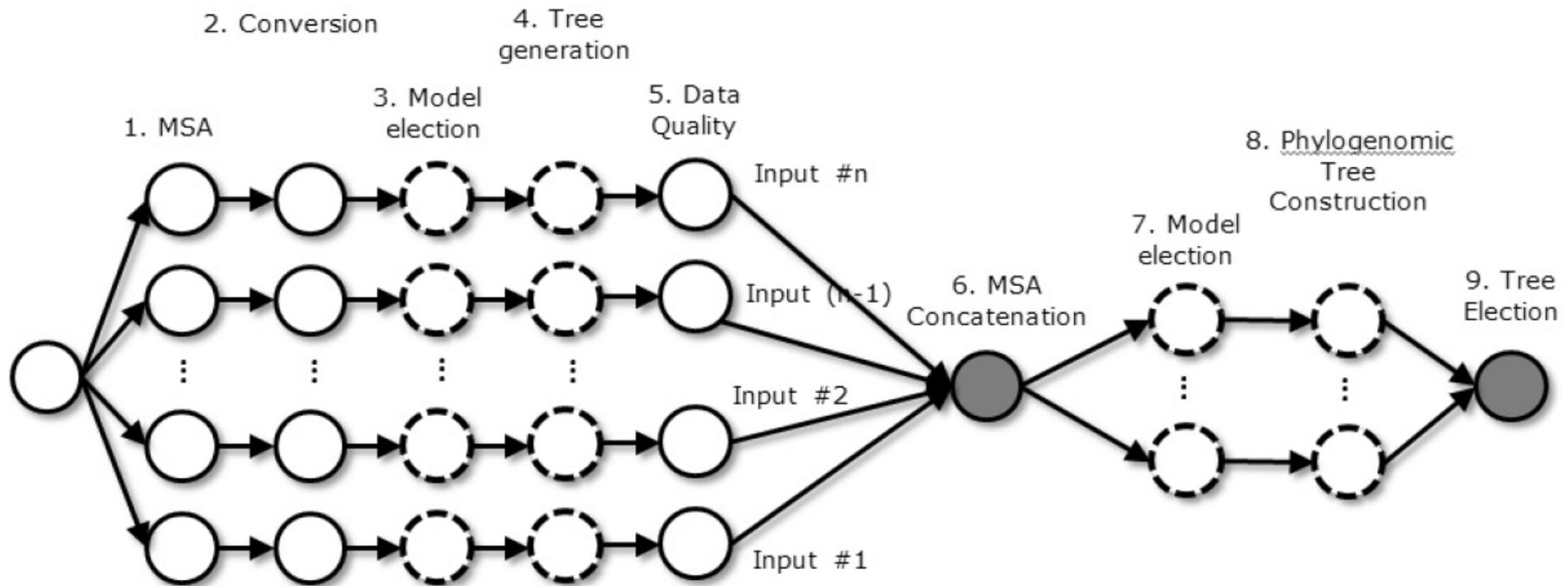
Modeling Comparative Genomic Workflows

SciPhylomics: a workflow for phylogenomic analysis

- A scientific workflow that aims inferring evolutionary relationships between homologous genes of different species.
- Composed by nine activities:
 - 1-4. Sub-workflow SciPhy
 5. Filter results that do not comply with a given quality criteria.
 - 6-9. Phylogenomic analysis

OLIVEIRA, D. ; OCAÑA, K. ; OGASAWARA, E. ; DIAS, J. ; GONCALVES, J. ; BAIAO, F. ; MATTOSO, M. L. Q. .
Performance evaluation of parallel strategies in public clouds: A study with phylogenomic workflows. Future Generation Computer Systems, v. 29, p. 1816-1825, 2013.

Modeling Comparative Genomic Workflows

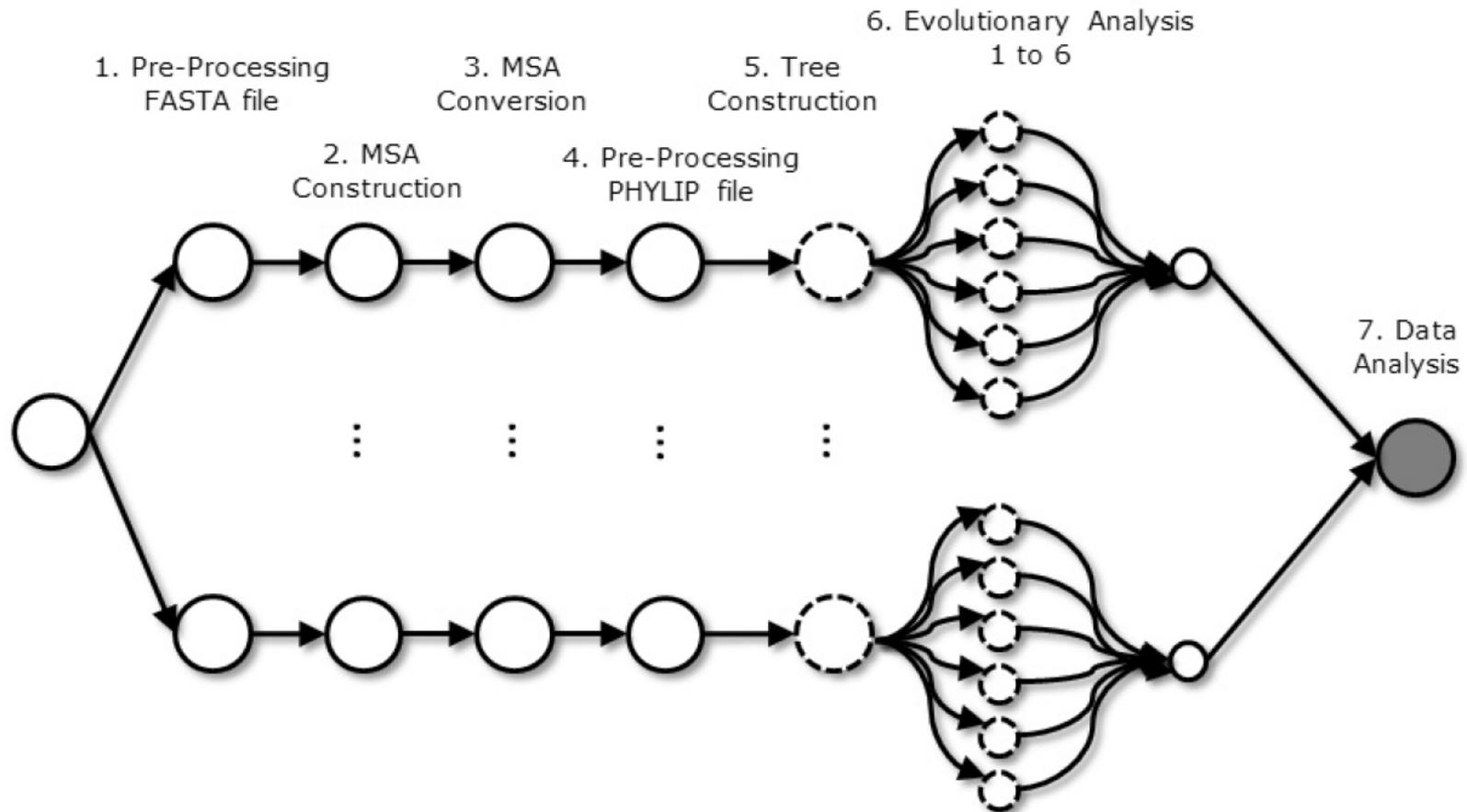


Modeling Comparative Genomic Workflows

SciEvol: a workflow for molecular evolutionary analysis

- A scientific workflow that aims detecting positive Darwinian selection on genomic data.
- Composed by eleven activities:
 1. Stop codons removal
 2. MSA construction
 3. MSA format conversion to the PHYLIP format
 4. Phylogenetic tree construction
 - 5-10. Evolutionary analysis execution
 11. Evolutionary data analysis.

Modeling Comparative Genomic Workflows



Experimental Results

- Evaluation of GraspCC to estimate the amount of VM for instantiating four real comparative genomics workflows executed in parallel in a cloud environment.
- Comparison the real performance and financial costs with the estimated by GraspCC:
 - Is it suitable to dimension the virtual cluster for this class of bioinformatics workflows?



Experimental Results

Cloud Environment Setup



- 5 types of VM: m1.small, m3.medium, m3.large, m3.xlarge and m3.2xlarge
- Linux Cent OS 5 (64-bit) and Secure Shell (SSH)
- Amazon image ami-6e1a8907 contains all programs installed and it is stored in the cloud.
- SciCumulus creates a virtual cluster to execute the experiment based on this image.
- GraspCC implemented in ANSI C
 - Executed in an isolated computer with processor equivalent to Intel Core i5 2.5GHz and 6Gb of RAM under
 - Linux (Ubuntu 12.04) operating system.

Experimental Results

Experiment Setup



- **Focus:** cysteine proteases (CPs) as a candidate drug target for protozoan diseases.
 - malaria are candidates for novel approaches utilizing CPs as targets.
- **Input:** a dataset of 100 fasta files of target protein sequences of falcipain CPs from Plasmodium species to identify CPs.
 - using scientific workflows that were executed using cloud with SciCumulus engine.
- **Workflows:** SciHmm, SciPhy, SciPhylomics and SciEvol.
- BLAST 2.2.18, MAFFT v7.012b, Muscle v3.7, MODELLER 9-11, NEST (Jackal 1.5), CONGEN Version 2.2.1, SEGMOD v1.0, and PROCHECK v.3.5.4, RAXML 7.2.8-ALPHA, Mega 5.1, ReadSeq 2.1.26, Molsoft ICM browser 3.7-2c and PAML 4.7.
 - using default parameters.

Experimental Results

Performance Results



- **First step to use GraspCC:** calculate the necessary GFlops for each workflow
 - historical execution of all previous executions.
- Requirements for each workflow:
 - RAM, disk space, maximum execution time allowed and maximum financial cost allowed.

- The maximum number allowed is 20.

Workflow	RAM (GB)	Disk Storage (GB)	GFLOP Number	Time (hours)	Cost (\$)
SciPhy	4	10	1175040	15	100
SciHm	4	10	829440	15	100
SciPhylomics	4	10	4147200	15	100
SciEvol	4	10	6635520	15	100

Experimental Results

Performance Results



➤ GraspCC produced the following estimations (with $\alpha \downarrow 1 = 0.5$ and $\alpha \downarrow 2 = 0.5$):

1. SciPhy - 1 m1.small and 4 m3.xlarge for 1 hour
2. SciHmm - 12 m1.small virtual machines for 1 hour
3. SciPhylomics - 1 m3.xlarge and 7 m3.2xlarge virtual machines for 1 hour
4. SciEvol - 12 m3.2xlarge virtual machines for 1 hour.

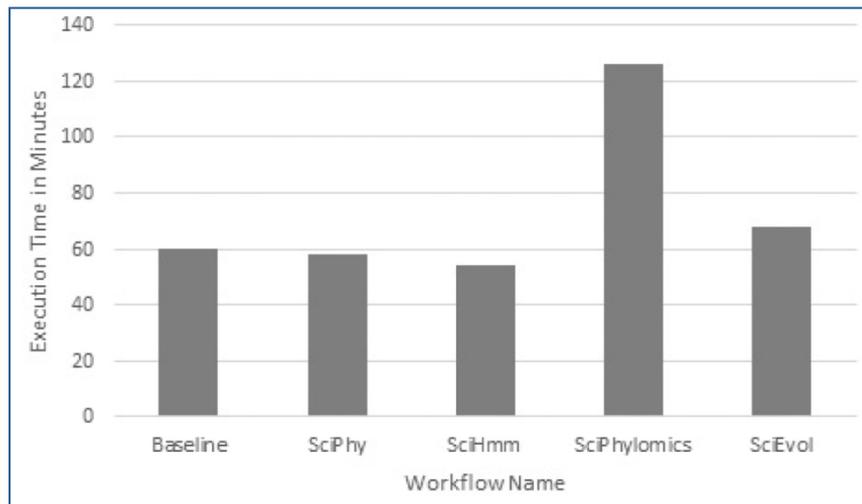
Workflow	Function Cost	Solution Value		Total Time (s)
		Time	Financial	
SciPhy	0.0368	1	1.16	0.12
SciHmm	0.0349	1	0.53	0.05
SciPhylomics	0.0458	1	4.20	0.20
SciEvol	0.0533	1	6.72	0.64

Experimental Results

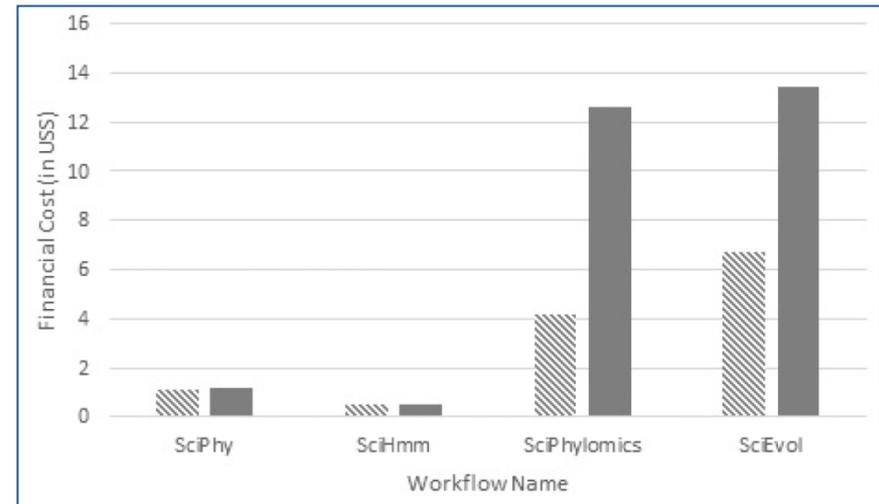
Performance Results



- After the GraspCC estimation, the workflows were executed using the suggested configuration.



Total execution time per workflow



Total financial cost per workflow

<i>Notation</i>	<i>Description</i>
P_j	the set of virtual machines types offered by cloud provider j
P	the set generated by the union of all sets P_j $P = \{P_1 \cup P_2 \cup \dots \cup P_q\}$
N_M^p	the maximum number of allocated virtual machines by the scientist in each cloud provider $P_j \mid p \in P_j$.
$\vec{c}_{pip' i'}$	the communication cost from a virtual machine i of type p to another virtual machine i' of type p'
up_p	the upload cost from a virtual machine p
$down_p$	the download cost to a virtual machine p
$cost_{storage}$	the storage cost of the transmitted data
$size_{data}$	the average size of the transmitted data
cs_p	the communication cost of virtual machine type p with others virtual machine types of the same cloud provider
y_{pi}	$y_{pi} = 1$, if and only if virtual machine i of type p is allocated in some period of time; otherwise, $y_{pi} = 0$
$\vec{z}_{pip' i'}$	$\vec{z}_{pip' i'} = 1$, if and only if $y_{pi} * y_{p' i'} = 1$; otherwise, $\vec{z}_{pip' i'} = 0$

(CC-IP-fed)

$$\begin{aligned} \min(\alpha_1(\sum_{p \in P} \sum_{i=1}^{N_M^p} \sum_{t \in T} \vec{c}_p x_{pit} + \\ \sum_{p \in P} \sum_{p' \in P} \sum_{i=1}^{N_M^p} \sum_{i'=1}^{N_M^{p'}} \vec{c}_{pip'i'} \vec{z}_{pip'i'}) + \alpha_2 t_m) \end{aligned} \quad (17)$$

$$\sum_{p \in P_j} \sum_{i=1}^{N_M^p} x_{pit} \leq N_M^p, \quad \forall j = 1 \dots q, \forall t \in T \quad (18)$$

$$\begin{aligned} y_{pi} \geq \vec{z}_{pip'i'}, \quad \forall p, p' \in P, \\ \forall i \in \{1, \dots, N_M^p\}, i' \in \{1, \dots, N_M^{p'}\} \end{aligned} \quad (19)$$

$$\begin{aligned} y_{p'i'} \geq \vec{z}_{pip'i'}, \quad \forall p, p' \in P, \\ \forall i \in \{1, \dots, N_M^p\}, i' \in \{1, \dots, N_M^{p'}\} \end{aligned} \quad (20)$$

$$\begin{aligned} y_{pi} + y_{p'i'} - 1 \leq \vec{z}_{pip'i'}, \quad \forall p, p' \in P, \\ \forall i \in \{1, \dots, N_M^p\}, i' \in \{1, \dots, N_M^{p'}\} \end{aligned} \quad (21)$$

$$\begin{aligned} \sum_{t \in T} x_{pit} \leq y_{pi} |T|, \quad \forall p \in P, \\ \forall i \in \{1, \dots, N_M^p\} \end{aligned} \quad (22)$$

$$\begin{aligned} y_{pi} \in \{0, 1\}, \quad \forall p \in P, \\ \forall i \in \{1, \dots, N_M^p\} \end{aligned} \quad (23)$$

$$\begin{aligned} \vec{z}_{pip'i'} \in \{0, 1\}, \quad \forall p, p' \in P, \\ \forall i \in \{1, \dots, N_M^p\}, i' \in \{1, \dots, N_M^{p'}\} \end{aligned} \quad (24)$$

“Conclusions”

- It was possible measure the accuracy of the GraspCC estimations in large-scale and parallel scientific workflow scenarios.
- The results indicate that the real executions of comparative genomics workflows with virtual cluster dimensioned by GraspCC have 100% of precision in 2 of the 4 executed workflows (SciHmm and SciPhy).
- The experiments with multi-site clouds are ongoing work...