COMPARISON OF RESOURCE PLATFORM SELECTION APPROACHES FOR SCIENTIFIC WORKFLOWS

Science Cloud 2010 Yogesh Simmhan, MSR & Lavanya Ramakrishnan, LBL

# Workflows for Modeling eScience

- Workflows common for in silico experiments
  - DAGs & dataflows
- Allow easy composition
  - Change often
- Loosely-coupled tasks, Tighlycoupled MPI
  - Different task characteristics

- Compute & Data intensive
  - Challenge of eScience problem sizes
- Resource needs often exceed available ones
  - Scale-out beyond current

resources









# Flourishing Space of Resource Platforms

- Cluster, Cloud, HPC, Desktop
- Local, captive resources
- Batch systems
- On-demand platforms
- Different characteristics of resource platforms



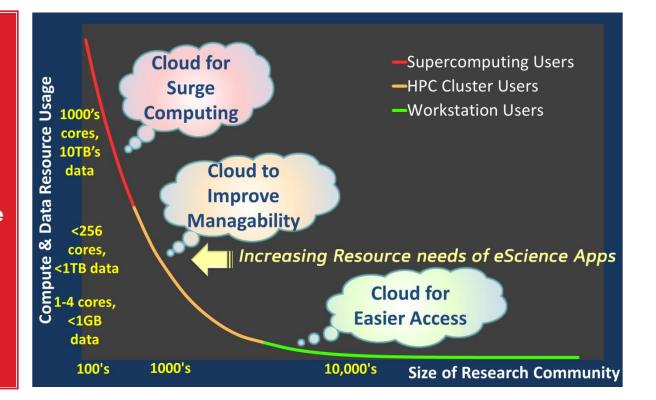




# **Cloud Platform for eScience**

On-demand

- Scale out
- Available
- Management Ease
- **Economical (TCO)**
- **Simple APIs**



# Resource Selection for Workflows

- Scientists need to select from existing & emerging resources
- Ad hoc, Rule of thumb, based on familiarity
  - Can be sub-optimal, punitive
- Different characteristics of resource platforms
  - Dynamic over short, long term
- Different goals
  - Makespan, usage, co\$t

#### DAG Scheduling Algorithms

- Tasks/WF scheduled to a platform
- Automatic WF Scheduling
  - Pegasus, Swift, Trident, etc. schedule WFs to remote resources
  - Support various platforms: Cluster, HPC, EC2.
  - Mandal, et al. Perf-based advance reservation for GrADS
  - Batch Queue Prediction Service
  - Blythe, et al. Task level greedy algorithm v. WF level optimization

# **Resource Selection for Workflows**

- These need information about WFs
  - Structural, task level details, data flow
- Fine grained details hard to determine & specify
  - Provenance mining, perf models
- Different granularity of WF details
  - Blackbox, Graybox, Whitebox

Fine grained workflow specs, evolving resource platforms pose overhead for users

Hypothesis —

Can we make intelligent resource platform selection with limited workflow information?

- Length
- Width
- Data In/Out

What are trade-offs of running applications on different platforms?



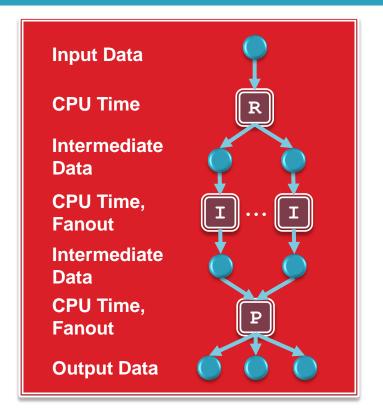
# **Workflow Characteristics**

### **Structural Information**

- Pattern: Sequential, Fork-Join, Control flow
- Length: # of stages, length per stage, total length
- Width: Fanout

### **Resource Usage**

- Data: In/Out
- Compute: Cores required



# **Resource Platforms & Characteristics**

#### Desktop

- Full application control
- Growth of multi-core
- eScience beats Moores Law

#### Cluster

- Small-Mid Clusters (~256 core)
- Under-subscribed, instant use
- Large science apps don't fit

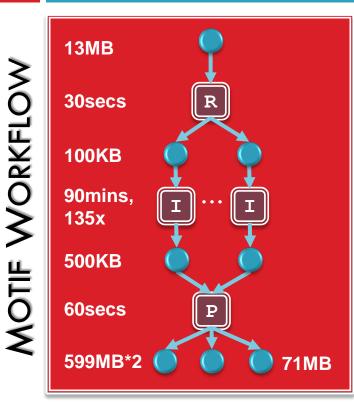
### HPC

- Shared, national centers
- Large # of cores (>1000)
- Over-subscribed queues, policiesCloud
- Infra. & Platform as Service
- On-demand, customizable
- Virtualization impact, Bandwidth

• Available cores • Queue/VM Latency • N/W Bandwidth • Core Speed

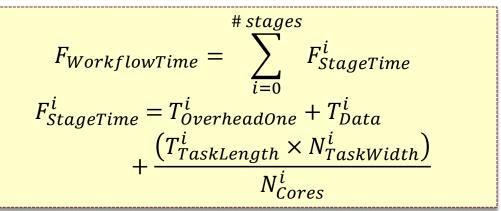
## PLATFORM SELECTION FOR WORKFLOWS

# Whitebox Selection (Fine Grained)

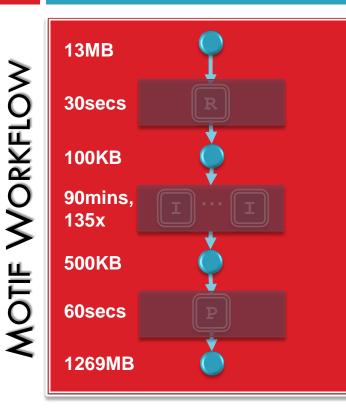


 Full workflow & Task details available
 Total runtime due to CPU Time I/O data transfer Queue/VM Overhead

Time is from by each independent task

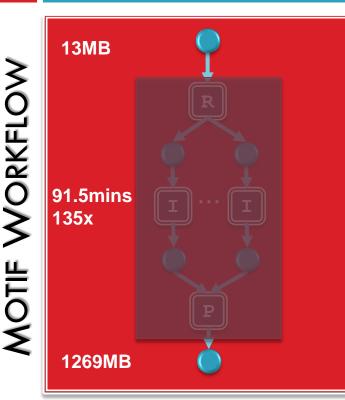


# Graybox Selection (Hybrid)



Workflow Stage details available Stages opaque. No task details available. Time is from by each independent stage Overhead time only for longest stage Queue/VM times pipelined *# stages*  $F_{WorkflowTime} = T_{OverheadMax} + \sum_{i=0}^{i} F_{StageTime}^{i}$  $F_{StageTime}^{i} = T_{Data}^{i} + \frac{\left(T_{StageLength}^{i} \times N_{StageWidth}^{i}\right)}{N_{Cores}^{i}}$ 

# Blackbox Selection (Coarse Grained)



Only Workflow outline details available

- Workflow internals opaque
- Total CPU Time Data xfer at boundary
- Overhead time for entire workflow
  - All required cores for workflow acquired

F<sub>WorkflowTime</sub>  $= T_{OverheadMax} + T_{Data}$  $(T_{WorkflowLength} \times N_{WorkflowWidth})$ 



# eScience Workloads for Evaluation

#### **MOTIF Network Workflow**

- Gene regulation dependency networks
- Compute & data intensive
  - 13MB input, 1300MB output
  - 90mins long, 135 task wide
  - 3 Stages: Fork, Compute fanout, Join

#### **GWAS Workflow**

- Genome wide association study
- Compute intensive & wide
  - 150MB input, 160MB output
  - 19mins long, 1100 task wide
  - 6 stages: Two compute fanouts 1100 and 150 tasks wide

## **Resource Platforms for Evaluation**

#### Local Workstation

- **1** Core, 2.5GHz
- All data local
- Local Cluster
- Up to 256 cores of 2.5GHz
- Data remote on client
- 1Gbps LAN bandwidth

#### **Teragrid HPC Clusters**

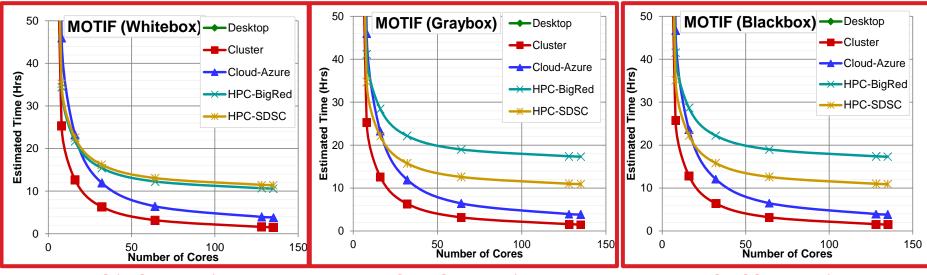
- SDSC & BigRed clusters
- 1 2048 cores of 2.5GHz
- NWS Batch Queue Prediction Service (95% Quantile)
- Data remote. 10Mbps WAN.

### Azure Cloud

- Small VM, 1 core, 1.6GHz
- **v** W start time  $\sim 200 + 20c$  secs
- Data remote. 10Mbps WAN.

### **Results: Motif Workflow**

Black & Graybox ordering of platforms same for different # cores
 Black & Whitebox ordering similar ... except for the two HPC's



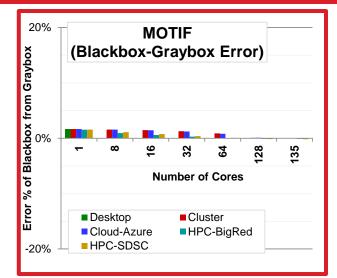
Whitebox Estimate

**Graybox Estimate** 

**Blackbox Estimate** 

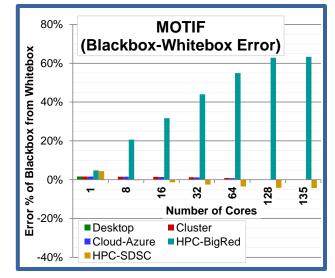
# **Results: Motif Workflow**

### Black & Graybox absolute difference is small across platforms



#### Blackbox – Graybox Absolute Difference

#### Black & Whitebox absolute difference for BigRed is large

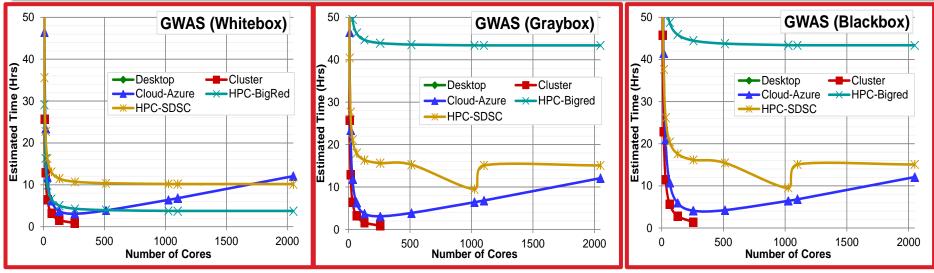


Blackbox – Whitebox Absolute Difference

## Results: GWAS Workflow

#### Blackbox & Whitebox ordering similar

- BigRed: 1 core job queue time in WBox faster than width-core job for BBox
- SDSC/Azure: Azure linear time; SDSC has step at 1024 cores for BBox



Whitebox Estimate

**Graybox Estimate** 

#### **Blackbox Estimate**



## Conclusions & Future work

- Runtime estimated from
  Blackbox good enough for relative comparison
  - Absolute values vary
- Queue overhead for task
  v. WF as a job has impact
  Azure linear, HPC step times
  Graybox ~= Blackbox

- More complex workflows
  - Simulation v. Calculation
- Synthetic workflow runs
  - Effect of each workflow attribute on estimate
- Other WF features that have impact
  - E.g. Min required cores per stage





Comparison of Resource Platform Selection Approaches for Scientific Workflows Yogesh Simmhan, Microsoft Research Lavanya Ramakrishnan, Lawrence Berkeley Lab Science Cloud Workshop, 2010

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