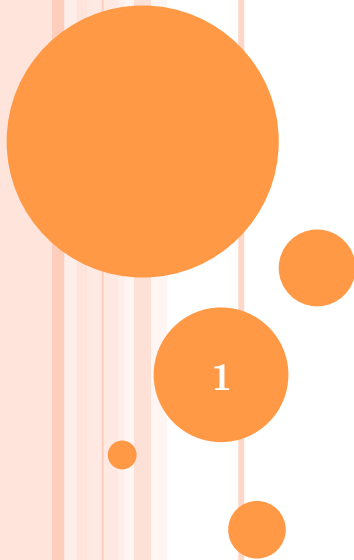


RESOURCE PROVISIONING IN HYBRID CLOUD COMPUTING IN THE PRESENCE OF RESOURCE FAILURES

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AGENDA

- Introduction
- Hybrid Cloud Architecture
- Proposed Approaches
- Proposed Provisioning Policies
- Performance Evaluation
- Simulation Results
- Conclusions
- References



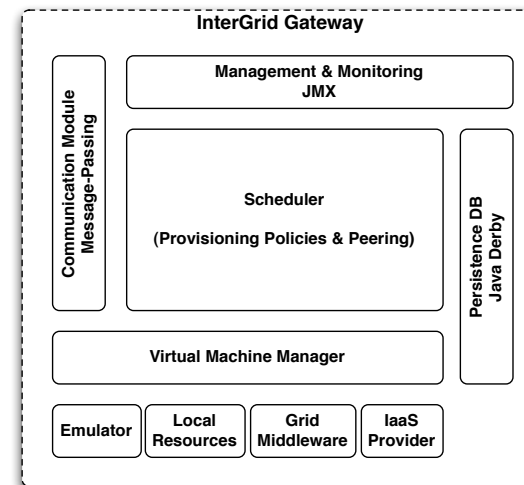
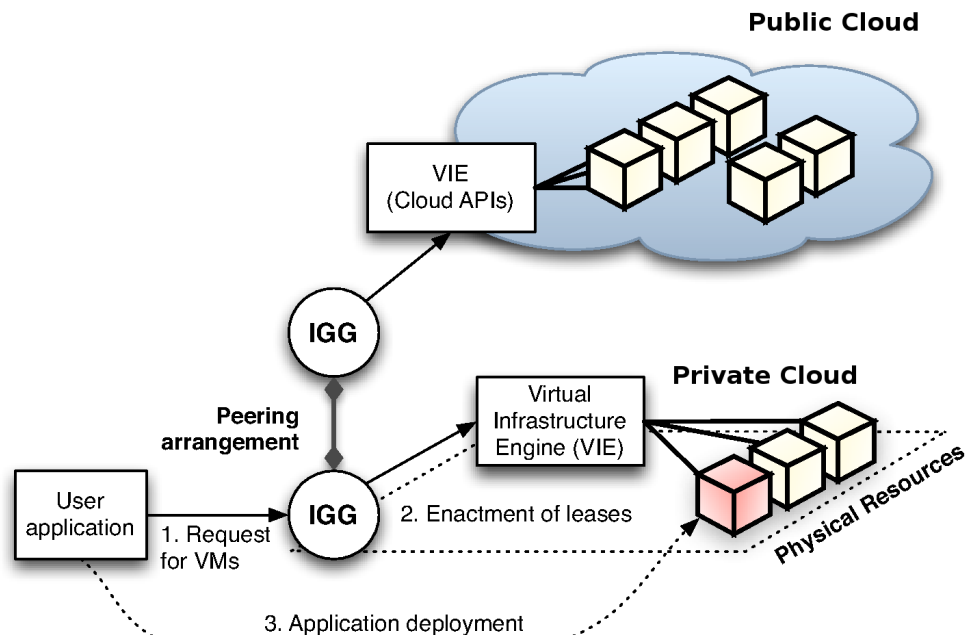
INTRODUCTION

- Hybrid Cloud Systems
 - Public Clouds
 - Private Clouds
- Resource Provisioning in Hybrid Cloud
 - Users' QoS (i.e., deadline)
 - Resource failures
- Taking into account
 - Workload model
 - Failure characteristics
 - Failure correlations
 - Failure model



HYBRID CLOUD ARCHITECTURE

- Based on InterGrid components
- Using a Gateway (IGG) as the broker



IGG



WORKLOAD MODEL

- Scientific Applications
 - Potentially large number of resources over a short period of time.
 - Several tasks that are sensitive to communication networks and resource failures (*tightly coupled*)
- User Requests
 - Type of virtual machine;
 - Number of virtual machines;
 - Estimated duration of the request;
 - Deadline for the request (optional).

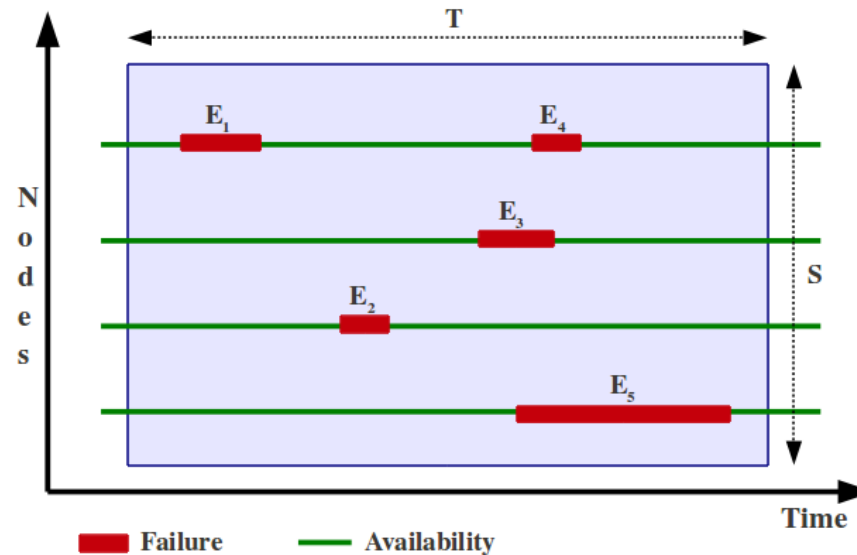


FAILURES IN USER REQUESTS

- Resource failure is inevitable
 - Redundant components in public Clouds
 - highly reliable service
 - Leads to service failure in private Clouds
- Correlation in Failures → *overlapped failures*
 - *Spatial* correlation means multiple failures occur on different nodes within a short time interval.
 - *Temporal* correlation is the skewness of the failure distribution over which means failure events exhibit considerable **autocorrelation** at small time lags, so the failure rate changes over time.



FAILURES IN USER REQUESTS (CONT.)



- The sequence of overlapped failures

$$H = \{F_i \mid F_i = (E_1, \dots, E_n), T_s(E_{i+1}) \leq T_e(E_i)\}$$

- Downtime of the service

$$D = \sum_{\forall F_i \in H} (\max\{T_e(F_i)\} - \min\{T_s(F_i)\})$$



PROPOSED APPROACHES

- Knowledge-free Approach
 - No Failure Model
 - Using failure correlation
 - Three brokering policies
- Knowledge-based Approach
 - Failure Model
 - Generic resource provisioning model
 - Two brokering policies (cost-aware)
- Workload model
 - Request size
 - Request duration



PROPOSED POLICIES

○ Size-based Strategy

- **Spatial correlation** : multiple failures occur on different nodes within a short time interval
- *Strategy*: sends wider requests to more reliable public Cloud systems
- Mean number of VMs per request
 - P_1 : probability of one VM
 - P_2 : probability of power of two VMs

$$\bar{S} = P_1 + 2^{\lceil k \rceil} (P_2) + 2^k (1 - (P_1 + P_2))$$

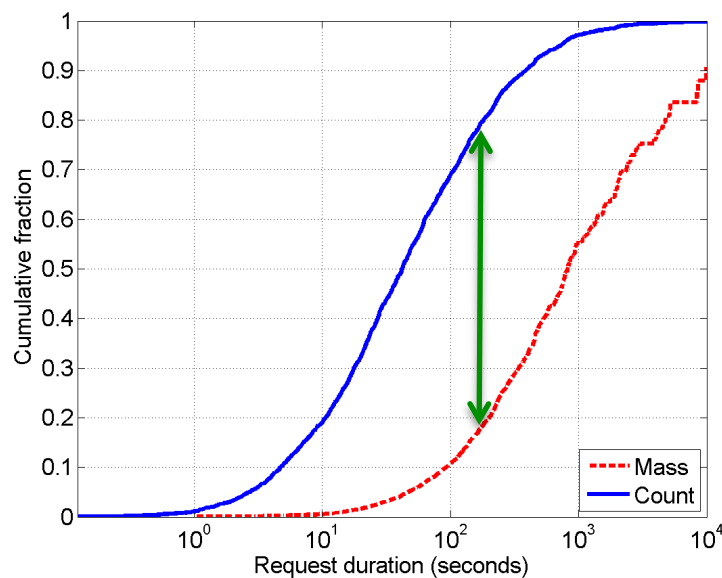
- Request size: two-stage uniform distribution (l, m, h, q)

$$k = \frac{ql + m + (1 - q)h}{2}$$



PROPOSED POLICIES (CONT.)

- Time-based strategy
 - **Temporal correlation**: the failure rate is time-dependent and some periodic failure patterns can be observed in different time-scales
 - **Request duration**: are *long tailed*.



- The mean request duration
 - Lognormal distribution in a parallel production system

$$\bar{T} = e^{\mu + \frac{\sigma^2}{2}}$$



PROPOSED POLICIES (CONT.)

- Area-based strategy
 - Making a **compromise** between the size-based and time-based strategy
 - The mean area of the requests

$$\bar{A} = \bar{T} \cdot \bar{S}$$

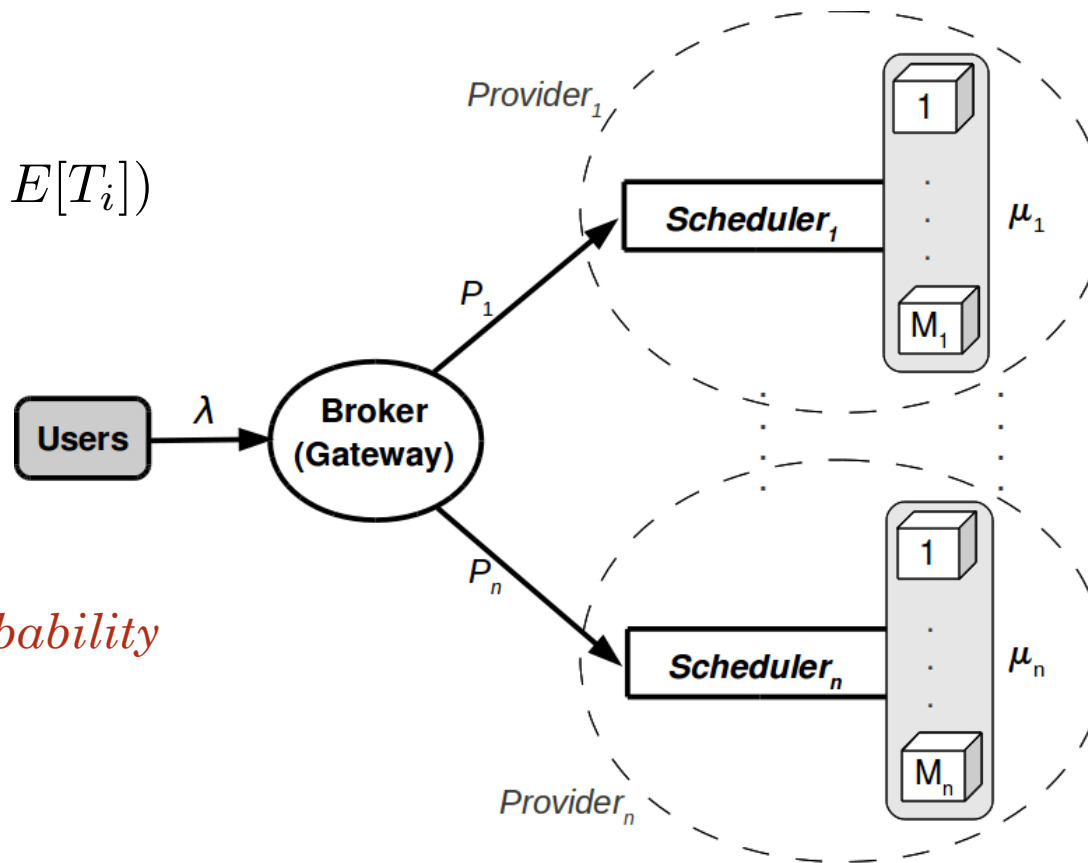
- This strategy sends long *and* wide requests to the public Cloud,
- It would be more conservative than a *size-based* strategy and less conservative than a *time-based* strategy.

KNOWLEDGE-BASED APPROACH: GENERIC RESOURCE PROVISIONING MODEL

- Model based on routing in distributed parallel queue

$$\min \sum_{i=1}^n (K_i \cdot E[T_i])$$

P_i: routing probability



K_i: price of provider i



MODEL PARAMETERS

- Using Lagrange multipliers methods, we obtained the routing probability as follows:

$$P_i = \frac{\mu_i}{\lambda} - \frac{\sum_{i=1}^n \mu_i - \lambda}{\lambda} \cdot \frac{\sqrt{K_i \eta_i}}{\sum_{i=1}^n \sqrt{K_i \eta_i}}$$

- Private Cloud service rate

$$\mu_s = \left(\frac{\bar{W}}{M_s \cdot \tau_s} \frac{t_a + t_u}{t_a} + L_s \right)^{-1}$$

- Public Cloud service rate

$$\mu_c = \left(\frac{\bar{W}}{M_c \cdot \tau_c} + L_c \right)^{-1}$$



ADAPTIVE POLICIES

- Adaptive with Random Sequence (ARS)
 - Routing probabilities (P_i)
 - Dispatch using *Bernoulli* distribution
- Adaptive with Deterministic Sequence (ADS)
 - Routing probabilities (P_i)
 - Dispatch using *Billiard* sequence

$$i_b = \min_{\forall i} \left\{ \frac{X_i + Y_i}{P_i} \right\}$$





SCHEDULING ALGORITHMS

- Scheduling the request across private and public Cloud resources
- Two well-know algorithms where requests are allowed to leap forward in the queue
 - Conservative backfilling
 - Selective backfilling

$$XFactor = \frac{W_i + T_i}{T_i}$$

- VM Checkpointing
 - VM stops working for the unavailability period
 - The request is started from where it left off when the node becomes available again



PERFORMANCE EVALUATION

- CloudSim Simulator
- Performance Metrics

- Deadline violation rate

- Slowdown
$$Slowdown = \frac{1}{M} \sum_{i=1}^M \frac{W_i + \max(T_i, bound)}{\max(T_i, bound)}$$

- Cloud Cost on EC2

$$Cost_{pl} = (H_{pl} + M_{pl} \cdot H_u) C_n + (M_{pl} \cdot B_{in}) C_x$$

- Workload Model

- Parallel jobs model of a multi-cluster system (i.e., DAS-2)

Input Parameters	Distribution/Value
Inter-arrival time	Weibull ($\alpha = 23.375, 0.2 \leq \beta \leq 0.3$)
No. of VMs	Loguniform ($l = 0.8, m, h = \log_2 N_s, q = 0.9$)
Request duration	Lognormal ($2.5 \leq \mu \leq 3.5, \sigma = 1.7$)
P_1	0.02
P_2	0.78



PERFORMANCE EVALUATION (CONT.)

○ Failures from Failure Trace Archive (FTA)

- <http://fta.scem.uws.edu.au/>

• Grid'5000 traces

- 18-month
- 800 events/node

Parameters	Description	Value (hours)
t_a	Mean availability length	22.25
σ_a	Std of availability length	41.09
t_u	Mean unavailability length	10.22
σ_u	Std of unavailability length	40.75

○ Synthetic Deadline

$$d_i = \begin{cases} st_i + (f \cdot ta_i), & \text{if } [st_i + (f \cdot ta_i)] < ct_i \\ ct_i, & \text{otherwise} \end{cases}$$

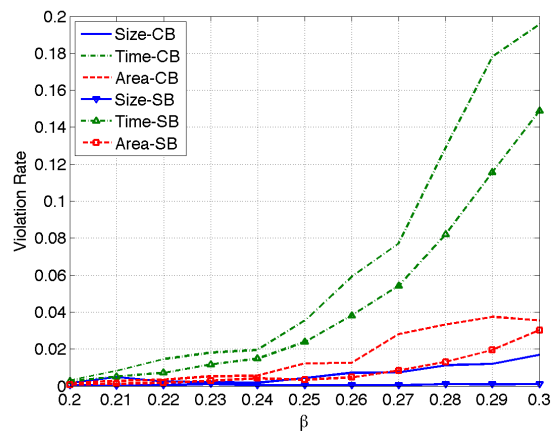
- f : stringency factor
- $f > 1$ is normal deadline (e.g., $f=1.3$)

○ $N_s = N_c = 64$

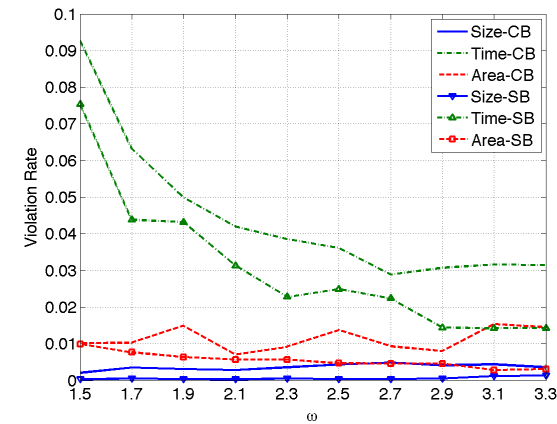


SIMULATION RESULTS

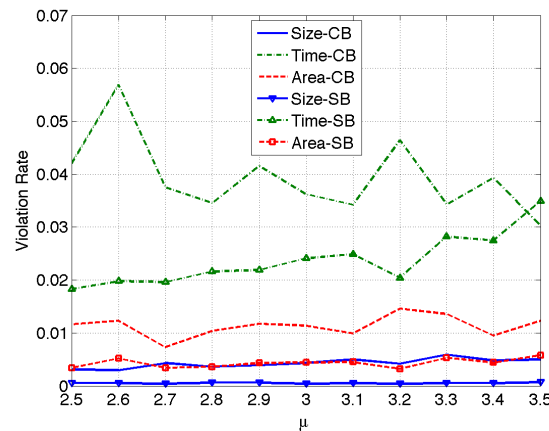
○ Violation rate (knowledge-free policies)



Request arrival rate



Request size

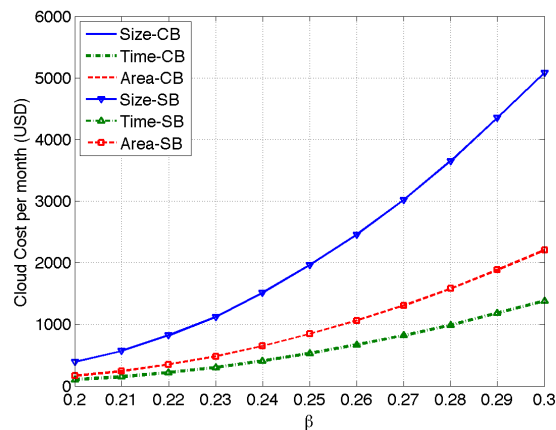


Request duration

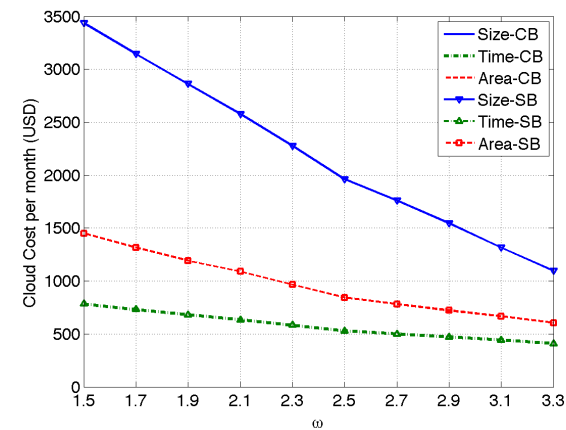


SIMULATION RESULTS (CONT.)

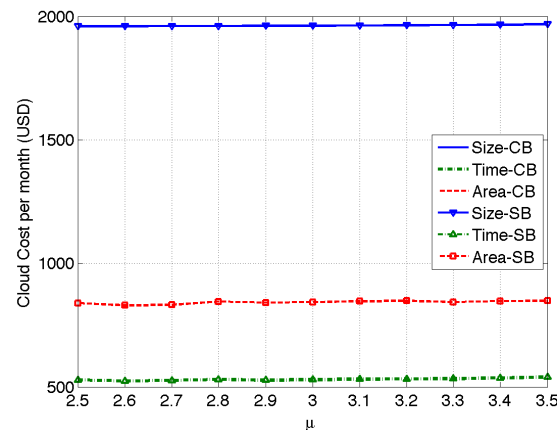
Cloud Cost on EC2 (knowledge-free policies)



Request arrival rate



Request size

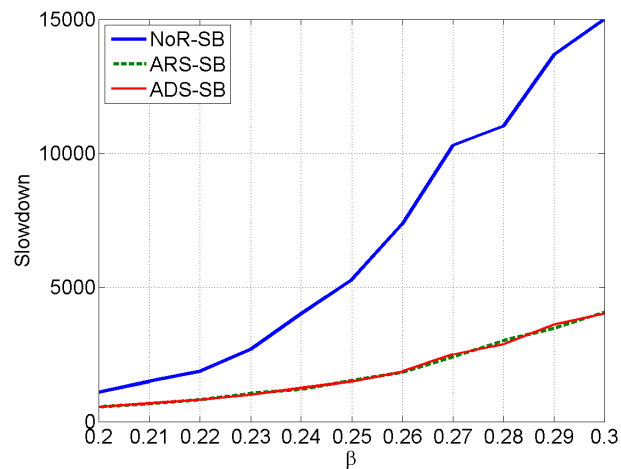


Request duration

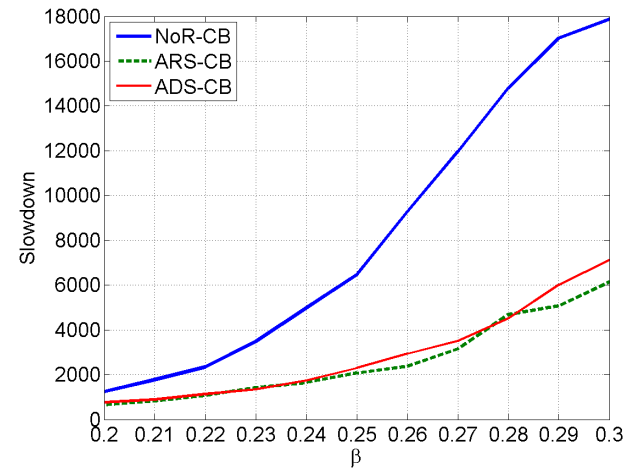


SIMULATION RESULTS (CONT.)

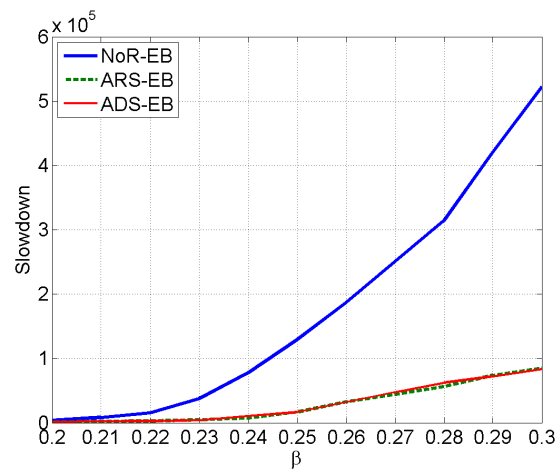
○ Slowdown (Knowledge-based policies)



Request arrival rate (SB)



Request arrival rate (CB)



Request arrival rate (EB)

FAILURE TRACE ARCHIVE (FTA)



- 27 Failure Traces
 - Supercomputers, HPC, Grid, P2P
- FTA Format
- Simulator and Scripts

FAILURE TRACE ARCHIVE

FOR IMPROVING THE RELIABILITY OF DISTRIBUTED SYSTEMS

MAR 09, 2015



ABOUT

PURPOSE
PEOPLE
NEWS

TRACES

FTA FORMAT
DATA SETS
DOWNLOAD
PUBLICATIONS

TOOLS

PARSING
ANALYSIS
SIMULATORS

HOMEPAGE

The **Failure Trace Archive (FTA)** is centralized public repository of availability traces of parallel and distributed systems, and tools for their analysis. The purpose of this archive is to facilitate the design, validation, and comparison of fault-tolerant models and algorithms.

In particular, the FTA contains the following:

- availability traces of parallel and distributed systems, differing in scale, volatility, and usage
- a standard format for failure traces
- scripts and tools for analyzing these traces

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<http://fta.scem.uws.edu.au/>



CONCLUSIONS

- Adaptive resource provisioning in a failure-prone hybrid Cloud system
- *Flexible* brokering strategies based on failure correlation/model as well as workload model
- Improve performance of hybrid Cloud
 - Knowledge-free approach: 32% in terms of deadline violation and 57% in terms of slowdown while using 135\$/month on EC2
 - Knowledge-based approach: 4.1 times in terms of response time while using 1200\$/month on EC2



OPEN QUESTIONS

- Recourse Failures vs. Energy Consumption for Cloud Systems
 - How they are related?
- Reliability-as-a-Service (RaaS) in Cloud Computing
 - Providing reliability on demand based on the users' requirements (e.g., Amazon Spot Instances)
- Cost Model for Resource Failures in Cloud Systems
 - Repair Replacement



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- Bahman Javadi, Jemal Abawajy, and Richard O. Sinnott , “*Hybrid Cloud Resource Provisioning Policy in the Presence of Resource Failures*” 4th IEEE International Conference on Cloud Computing Technology and Science (CloudCom 2012), Taipei, Taiwan, December 2012, pp. 10-17. **Best Paper Finalist Award.**
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Thank You

