Observe-Analyze-Act Paradigm for Storage System Resource Arbitration

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Outline

• Observe-analyze-act in storage system: CHAMELEON

- **O** Motivation
- O System model and architecture
- O Design details
- O Experimental results
- Observe-analyze-act in other scenarios
 - O Example: network applications
- Future challenges

Need for Run-time System Management



- Static resource allocation is not enough
 - O Incomplete information of the access characteristics: workload variations; change of goals
 - O Exception scenarios: hardware failures; load surges.

Approaches for Run-time Storage System Management

- Today: Administrator observe-analyze-act
- Automate the observe-analyze-act:
 - O Rule-based system
 - Complexity
 - Brittleness
 - O Pure feedback-based system
 - Infeasible for real-world multi-parameter tuning
 - O Model-based approaches
 - Challenges:
 - How to **represent** system details as models?
 - How to create/evolve models?
 - How to **use** models for decision making?

System Model for Resource Arbitration



- Input:
 - O SLAs for workloads
 - O Current system status (performance)
- Output:
 - O Resource reallocation action (Throttling decisions)

Our Solution: CHAMELEON



Knowledge Base: Component Model

•Objective: Predict service time for a given load at a component (For example: storage controller).

Service_time_{controller} = L(request size, read write ratio, random sequential ratio, request rate)

An example of component model

O FAStT900, 30 disks, RAID0

O Request Size 10KB, Read/Write Ratio 0.8, Random Access



Component Model (cont.)

- Quadratic Fit
 O S = 3.284, r = 0.838
- Linear Fit
 - O S = 3.8268, r = 0.739
- Non-saturated case: Linear Fit
 O S = 0.0509, r = 0.989





Knowledge Base: Workload Model

 Objective: Predict the load on component i as a function of the request rate j

Component_load_{i,i} = $W_{i,i}$ (workload j request rate)

• Example:

O Workload with 20KB request size,

0.642 read/write ratio and 0.026 sequential access ratio



Knowledge Base: Action Model

 Objective: Predict the effect of corrective actions on workload requirements

Workload J request Rate = A_i (Token Issue Rate for Workload J)

• Example:



Analyze Module: Reasoning Engine

•Formulated as a constraint solving problem

O Part 1: Predict Action Behavior:

For each candidate throttling decision, predict its performance result based on knowledge base

O Part 2: Constraint Solving: Use linear programming technique to scan all feasible solutions and choose the optimal one

Reasoning Engine: Predict Result

- Chain all models together to predict action result
- Input: Token issue rate for each workloads
- Output: Expected latency



Reasoning Engine: Constraint Solving

- -atency(%) Formulated using Linear Programming
- Formulation:
 - O Variable: Token issue rate for each workload
 - O Objective Function:
 - Minimize number of workloads violating their SLA goals
 - Workloads are as close to their SLA IO rate as possible
 - Example: Minimize $\sum p_{ai}p_{bi}$ [SLA_i – T(current_throughput_i, t_i)]

SLA_i

where p_{ai} = Workload priority p_{bi} = Quadrant priority

1

FAILED

 $0^{0}0^{0}$ MEET

0

EXCEED

 \bigcirc

LUCKY

IOps(%)

()

- Constraints: \bigcirc
 - Workloads should meet their SLA latency goals

Act Module: Throttling Executor



Experimental Results

• Test-bed configuration:

- O IBM x-series 440 server (2.4GHz 4-way with 4GB memory, redhat server 2.1 kernel)
- O FAStT 900 controller
- O 24 drives (RAID0)
- O 2Gbps FibreChannel Link
- Tests consist of:
 - O Synthetic workloads
 - O Real-world trace replay (HP traces and SPC traces)

Experimental Results: Synthetic Workloads



Effect of model errors on output of the constraint solver



Experiment Result: Real-world Trace Replay

- Real-world block-level traces from HP (cello96 trace) and SPC (web server)
- A phased synthetic workload acts as the third flow
- Test goals:
 - O Do they converge to SLAs?
 - O How reactive the system is?
 - O How does CHAMELEON handle unpredictable variations?

Real-world Trace Replay

• Without CHAMELEON:





• With throttling



Real-world Trace Replay



Handling system changes



Other system management scenarios?

- Automate the observe-analyze-act loop for other selfmanagement scenarios
- Example: CHAMELEON for network applications O Example: A proxy in front of server farm



Future Work

- Better methods to improve model accuracy
- More general constraint solver
- Combining with other actions
- CHAMELEON in other scenarios
- CHAMELEON for reliability and failure

References

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Questions?

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