

State-Space Feedback Control for Elastic Distributed Storage in a Cloud Environment

M. Amir Moulavi Ahmad Al-Shishtawy

Vladimir Vlassov

KTH Royal Institute of Technology, Stockholm, Sweden ICAS 2012, March 26, St. Maarten, Netherlands Antilles



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Agenda

- Introduction & Problem Definition
- System Identification
- Controller Design
- EStoreSim: Elastic Store Simulator
- Evaluation
- Conclusions



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Motivation

- Web 2.0 applications
 - WiKis, social networks, media sharing
 - Rapidly growing number of users and the amount of user generated data



You Tube

flickr



facebook

- Challenges for a storage service
 - Growing number of users and the amount of data (scalability)
 - Uneven load, user geographically scattered (low request latency, load balancing)
 - Partial failures, very high load (high availability)
 - Acceptable data consistency guarantees (e.g., eventual consistency)



Cloud-Based Services

- Cloud computing offers an efficient and effective solution to the challenges of scale and the (highly) dynamic load
- Provides the illusion of infinite amount of resources
- "Pay-as-you-go": pay for a service only when/if you use it
- End-user does not need to be involve in the configuration and maintenance of the cloud-based system
- Enables development of Cloud-based Elastic
 Services and Applications



Need for Elasticity

- Web services, e.g. storage, frequently experience high workloads
 - A service can become popular in just an hour
- The high level load does not last for long and keeping resources in the Cloud costs money
- This has led to *Elastic Computing*
 - Ability of a system to grow and shrink at run-time in response to changes in workload
- Cloud computing allows on-the-fly requesting and releasing VM instances to scale the service in order to meet SLOs at a minimal cost



Elasticity versus Static Provisioning

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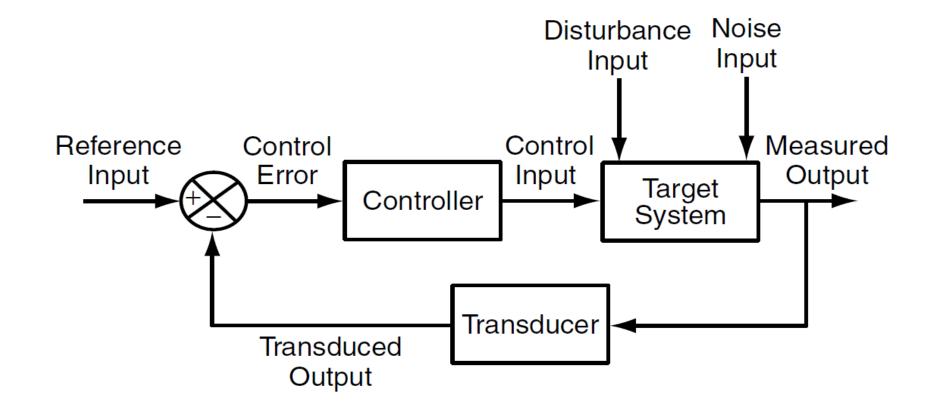


Automation of Elasticity

- Elasticity can be done either manually (by the sys-admin) or automatically (by a autonomic manager)
- Automation of elasticity can be achieved by providing an Elasticity Controller
 - Helps to avoid SLO violations while keeping the cost low
 - Automatically adds/removes VMs (servers, service instances) in response to changes in some SLO metrics, e.g., request latency, caused by changes in workload
 - Can be built using elements of Control Theory
 - Feedback-loop (a.k.a. closed-loop) control
 - Model Predictive Control (MPC)



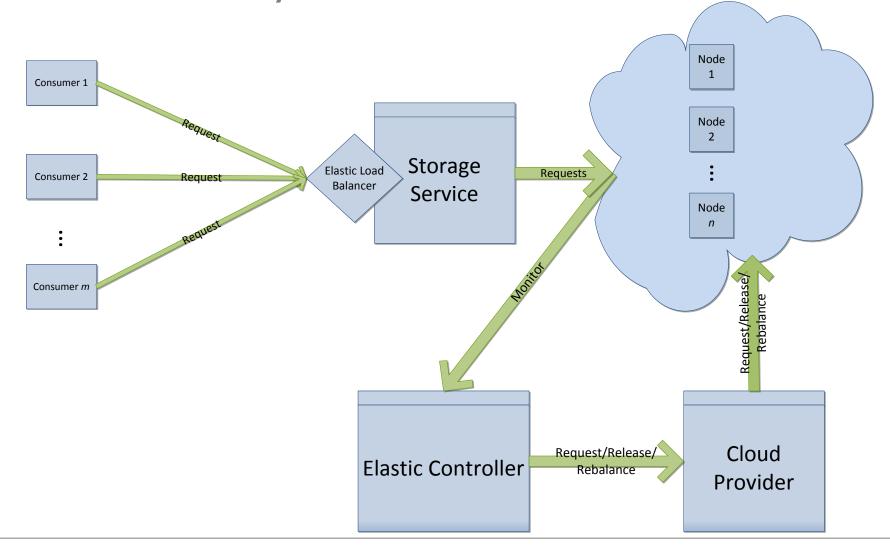
Feedback (Closed Loop) Control [Hel2004]





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Automatic Control of Storage Elasticity in the Cloud





Two Phases of a Feedback Controller Design

System identification

• Building a mathematical model of a dynamic system

- How control outputs depend on control inputs
- Two main approaches:
 - First-principle (e.g., using queuing theory)
 - Black-box (e.g., state-space)

Controller design

- Choose a controller type (e.g., PID, State-Space)
- Determine controller gains based on the system model



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State-Space Model

Advantages

- Provides scalable approach to model systems with large number of inputs/outputs
- Can be extended easily

State variables

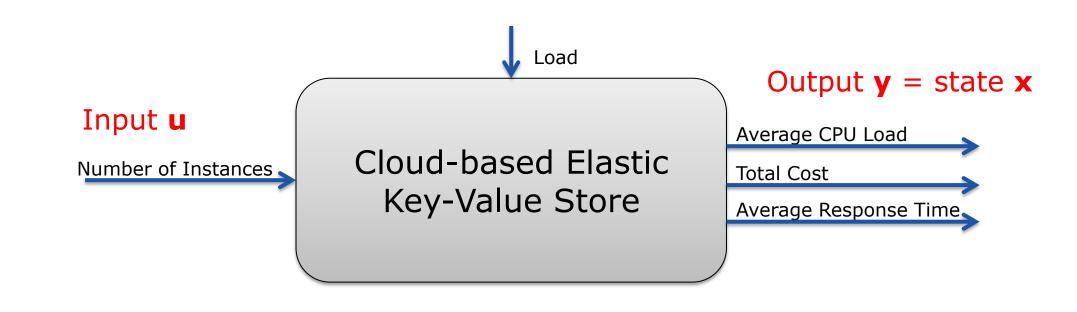
- Express the dynamics of the system

Main Steps

- Study system behavior in order to identify the control inputs, control outputs, and state variables of the system
- Construct the characteristic equations
- Design an experiment to estimate the parameters of the characteristic equations



Control Inputs/Outputs of an Elastic Storage System





Characteristic Equations

- A state space model considers relationship between inputs u, outputs y, and state variables x
- State variables used in two ways
 - Describe dynamics (state changes)
 - Determine the measured output from the state

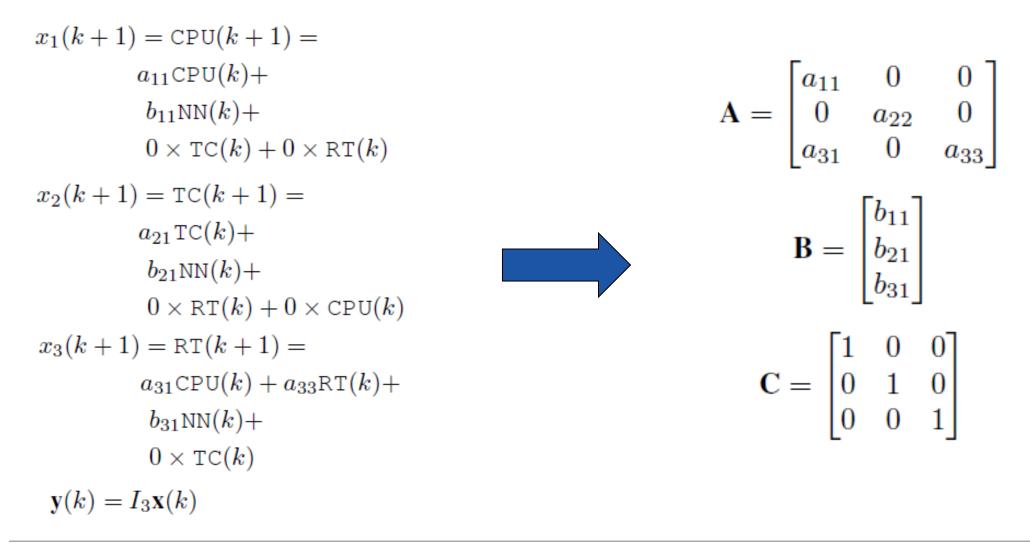
 $\begin{aligned} \mathbf{x}(k+1) &= \mathbf{A}\mathbf{x}(k) + \mathbf{B}\mathbf{u}(k) \\ \mathbf{y}(k) &= \mathbf{C}\mathbf{x}(k) \end{aligned}$

• Allows modeling of a MIMO system with multiple inputs and outputs



Characteristic Equations

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Parameter Estimation

Identification: Estimate the coefficient matrices **A** , **B** and **C** from experimental data

- Feed the system with an input signal and observe outputs and internal state variables periodically.
- Compute the matrices from the collected data using the multiple linear regression method
 - The Matlab regress(y,X) function can be used to calculate matrices

$$\mathbf{A} = \begin{bmatrix} 0.9 & 0 & 0\\ 0 & 0.724 & 0\\ 5.927 & 0 & 0.295 \end{bmatrix} \qquad \mathbf{B} = \begin{bmatrix} 2.3003\\ 0.0147\\ 77.8759 \end{bmatrix}$$



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Controller Design

- Dynamic State Feedback
 - a State-Space analogous to PI (Proportional Integral) control
- Has good disturbance rejection properties
- the control error is

e(k) = r - y(k)

• The integrated control error is

$$x_I(k+1) = x_I(k) + e(k)$$

The control law is

$$u(k) = - \begin{bmatrix} \mathbf{K}_p & K_I \end{bmatrix} \begin{bmatrix} \mathbf{x}(k) \\ x_I(k) \end{bmatrix}$$



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LQR Controller Design

- LQR: Least Quadratic Regulation
- An approach to controller design is to focus on the tradeoff between control effort and control errors
- Minimizing control errors (Defined by matrix R):
 - Improve accuracy and reduce both settling times and overshoot
- Minimizing control effort (Defined by matrix Q):
 - Sensitivity to noise is reduced

$$\mathbf{Q} = \begin{bmatrix} 100 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \qquad \qquad \mathbf{R} = \begin{bmatrix} 1 \end{bmatrix}$$



LQR Controller Design

 Given: A and B (from system identification), weighting matrices R and Q (state/input and output, respectively), and the quadratic cost function J;

$$J = \frac{1}{2} \sum_{k=0}^{\infty} \left[\mathbf{x}^{\top}(k) \mathbf{Q} \mathbf{x}(k) + \mathbf{u}^{\top}(k) \mathbf{R} \mathbf{u}(k) \right]$$

- Find: The controller gain vector K (for three outputs) that minimizes the quadratic cost function J for given R and Q;
- Use Matlab dlqr() function: K = dlqr(A, B, Q, R)

$$\mathbf{K} = \begin{bmatrix} 0.134 & 1.470162e - 06 & 0.00318 \end{bmatrix}$$



EStoreSim: Elastic Key-Value Store Simulator

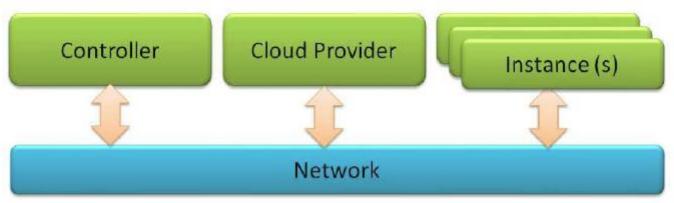
- Simulates a Cloud environment
 - VMs (CPU & Memory)
 - Network (Upload bandwidth)
 - ...
- Generates various workload patterns
- Supports controller design
 - Run system identification experiments and gather data
 - Experiment with different controller designs



EStoreSim: Elastic Key-Value Store Simulator

Implementation

- Based on KOMPIES
- Written in 🔮 Java Scala
- Publicly available on
 - https://github.com/amir343/ElasticStorage



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Evaluation

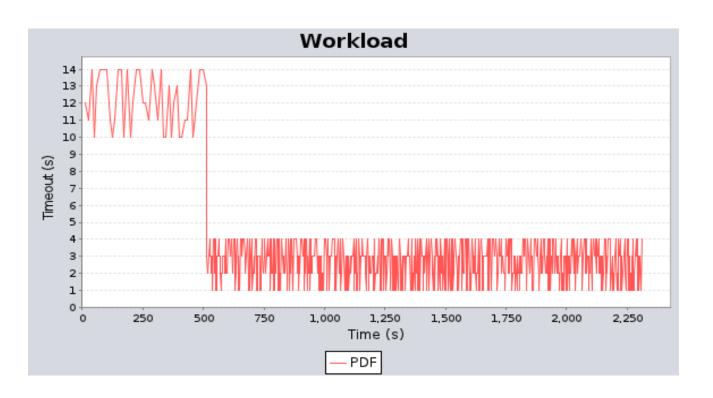
SLO Requirements

- Average CPU Load \leq 55%
- Response Time \leq 1,5 seconds
- Average Bandwidth per download > 200 KB/s
- Two Experiments:
 - SLO Experiment
 - Cost Experiment



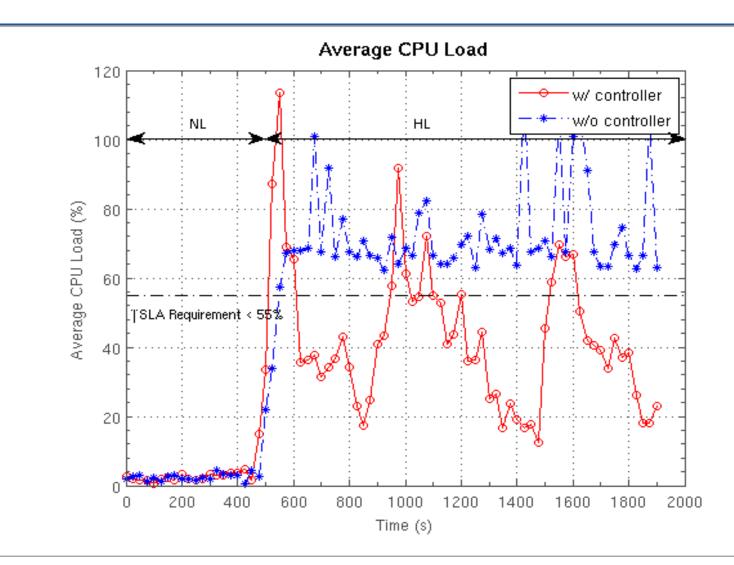
SLO Experiment

Workload (interarrival time)



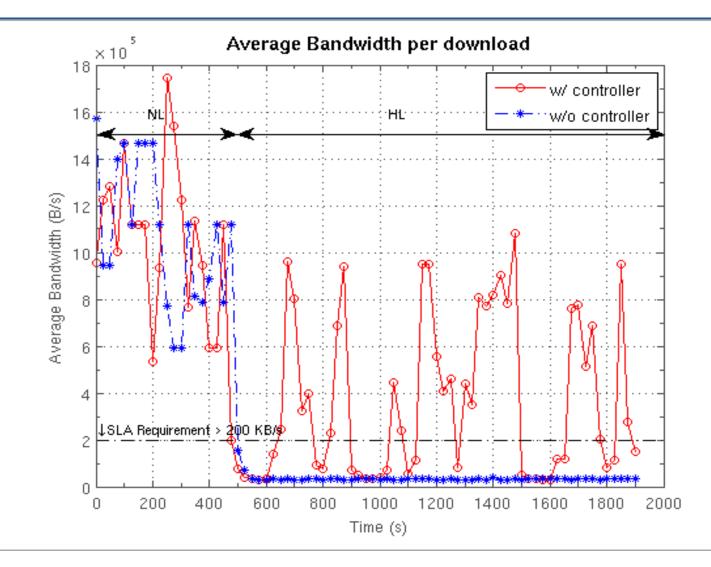


SLO Experiment





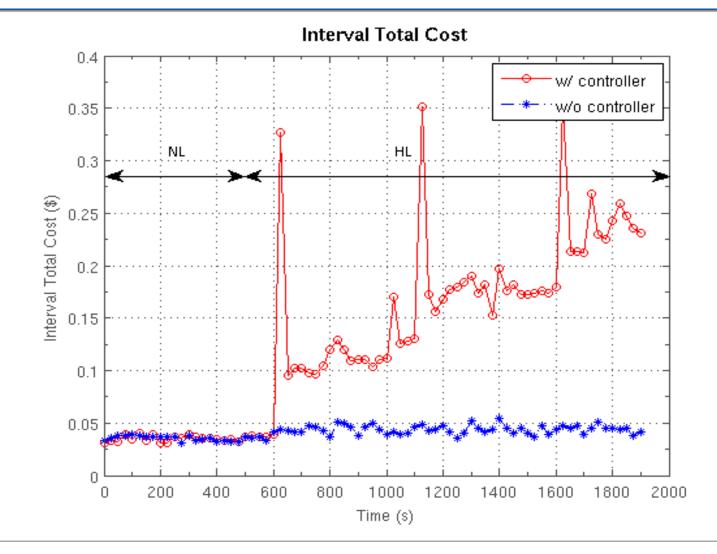
SLO Experiment





SLO Experiment

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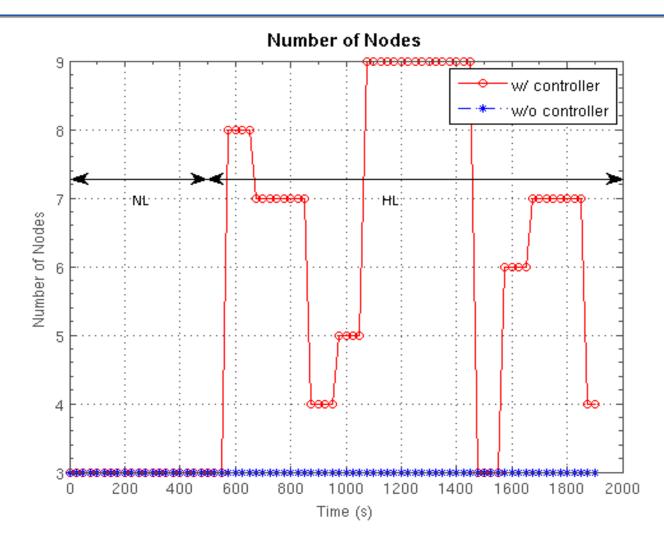


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SLO Experiment





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Cost Experiment



	w/ controller	w/o controller
Total Cost (\$)	10,5	16,5



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Conclusions

- Elasticity in Cloud computing is an ability of a system to scale up and down in response to changes in its environment and workload
 - Improves Cloud-based systems by reducing the total cost for the system while meeting SLOs
- Described the steps in designing an elasticity controller for a Cloud-based key-value store
- EStoreSim: Open source simulation framework for Cloud systems
- Experiments have shown the feasibility of our approach to automate elasticity control of storage services using state-space feedback control.