

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

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Agenda

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

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

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

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

• 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

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

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

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

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

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

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: $K = \text{dIqr}(A, B, Q, R)$

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

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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 KOMPICS
- Written in Scala
- Publicly available on **github**
	- https://github.com/amir343/ElasticStorage

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Evaluation

• SLO Requirements

- Average CPU Load \leq 55%
- $-$ Response Time ≤ 1.5 seconds
- Average Bandwidth per download > 200 KB/s
- Two Experiments:
	- SLO Experiment
	- Cost Experiment

SLO Experiment

• Workload (interarrival time)

SLO Experiment

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SLO Experiment

SLO Experiment

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SLO Experiment

Cost Experiment

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.