RAS-Models: A Building Block for Self-Healing Benchmarks

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Overview

- Introduction
- Problem
- Hypothesis
- Experiments & Examples
- Proposed Evaluation Methodology
- Conclusion & Future Work

Introduction

 A self-healing system "...automatically detects, diagnoses and repairs localized software and hardware problems" — The Vision of Autonomic Computing 2003 IEEE Computer Society

Why a Self-Healing Benchmark?

- To quantify the impact of faults (problems)
 - Establish a baseline for discussing "improvements"
- To reason about expected benefits for systems currently lacking self-healing mechanisms
 - Includes existing/legacy systems
- To quantify the efficacy of self-healing mechanisms and reason about tradeoffs
- To compare self-healing systems

Problem

- Evaluating self-healing systems and their mechanisms is non-trivial
 - Studying the failure behavior of systems can be difficult
 - Multiple styles of healing to consider (reactive, preventative, proactive)
 - Repairs may fail
 - Partially automated repairs are possible

Hypotheses

- Reliability, Availability and Serviceability provide reasonable evaluation metrics
- Combining practical fault-injection tools with mathematical modeling techniques provides the foundation for a feasible and flexible methodology for evaluating and comparing the reliability, availability and serviceability (RAS) characteristics of computing systems

Objective

- To inject faults into the components of the popular n-tier web-application
 - Specifically the application server and Operating System
- Observe its responses and any recovery mechanisms available
- Model and evaluate available mechanisms
- Identify weaknesses

Experiment Setup



Target: 3-Tier Web Application

TPC-W Web-application Resin 3.0.22 Web-server and (Java) Application Server Sun Hotspot JVM v1.5 MySQL 5.0.27 Linux 2.4.18

Remote Browser Emulation clients to simulate user loads

Practical Fault-Injection Tools

Kheiron/JVM

- Uses bytecode rewriting to inject faults into Java Applications
- Faults include: memory leaks, hangs, delays etc.
- Nooks Device-Driver Fault-Injection Tools
 - Uses the kernel module interface on Linux (2.4 and now 2.6) to inject device driver faults
 - Faults include: text faults, memory leaks, hangs etc.

Healing Mechanisms Available

- Application Server
 - Automatic restarts
- Operating System
 - Nooks device driver protection framework
 - Manual system reboot

Mathematical Modeling Techniques

- Continuous Time Markov Chains (CTMCs)
 - Limiting/steady-state availability
 - Yearly downtime
 - Repair success rates (fault-coverage)
 - Repair times
- Markov Reward Networks
 - Downtime costs (time, money, #service visits etc.)
 - SLA penalty-avoidance

Example 1: Resin App Server

- Analyzing perfect recovery e.g. mechanisms addressing resource leaks/fatal crashes
 - $S_0 UP$ state, system working
 - S₁ DOWN state, system restarting
 - $\lambda_{\text{failure}} = 1 \text{ every } 8 \text{ hours}$
 - $\mu_{restart}$ = 47 seconds
- Attaching a value to each state allows us to evaluate the cost/time impact associated with these failures.



Results: Steady state availability: 99.838% Downtime per year: 866 minutes

Availability guarantee	Max downtime per year	Expected penalties
99.999	\sim 5 mins	(866 - 5)*\$p
99.99	\sim 53 mins	(866 - 53)*\$p
99.9	\sim 526 mins	(866 - 526)*\$p
99	\sim 5256 mins	\$0

Example 2: Linux w/Nooks

- Analyzing imperfect recovery e.g. device driver recovery using Nooks
 - S₀ UP state, system working
 - S_1 UP state, recovering failed driver
 - S₂ DOWN state, system reboot
 - $\lambda_{driver_{failure}} = 4$ faults every 8 hrs
 - $\mu_{nooks_recovery}$ = 4,093 mu seconds
 - μ_{reboot} = 82 seconds
 - c coverage factor/success rate





Example 3: Resin + Linux + Nooks

- Composing Markov chains
 - $S_0 UP$ state, system working
 - S_1 UP state, recovering failed driver
 - S₂ DOWN state, system reboot
 - S_3 DOWN state, Resin reboot
 - $\lambda_{driver_{failure}} = 4$ faults every 8 hrs
 - $\mu_{nooks_recovery}$ = 4,093 mu seconds
 - μ_{reboot} = 82 seconds
 - c coverage factor
 - $\lambda_{\text{memory_leak}}$ = 1 every 8 hours
 - µ_{restart_resin} = 47 seconds





Min downtime = 866 minutes

Benefits of CTMCs + Fault Injection

- Able to model and analyze different styles of self-healing mechanisms
- Quantifies the impact of mechanism details (success rates, recovery times etc.) on the system's operational constraints (SLA penalties, availability etc.)
 - Engineering view AND Business view
- Able to identify under-performing mechanisms
- Useful at design time as well as post-production
- Able to control the fault-rates

Caveats of CTMCs + Fault-Injection

- CTMCs may not always be the "right" tool
 - Constant hazard-rate assumption
 - True distribution of faults may be different
 - Fault-independence assumptions

16

- Limited to analyzing near-coincident faults
- Not suitable for analyzing cascading faults (can we model the precipitating event as an approximation?)
- Some failures are harder to replicate/induce than others
 - Better data on faults will improve fault-injection tools
- Getting detailed breakdown of types/rates of failures
 - More data should improve the fault-injection experiments and relevance of the results

Real-World Downtime Data*

- Mean incidents of unplanned downtime in a year: 14.85 (n-tier web applications)
- Mean cost of unplanned downtime (Lost productivity #IT Hours):
 - 2115 hrs (52.88 40-hour work-weeks)
- Mean cost of unplanned downtime (Lost productivity #Non-IT Hours):
 - 515.7 hrs** (12.89 40-hour work-weeks)

* "IT Ops Research Report: Downtime and Other Top Concerns,"
StackSafe. July 2007. (Web survey of 400 IT professional panelists, US Only)
** "Revive Systems Buyer Behavior Research," Research Edge, Inc. June 2007

Quick Analysis – End User View

- Unplanned Downtime (Lost productivity Non-IT hrs) per year: 515.7 hrs (30,942 minutes).
- Is this good? (94.11% Availability)

Availability Guarantee	Max Downtime Per Year
99.999	\sim 5 mins
99.99	\sim 53 mins
99.9	\sim 526 mins
99	\sim 5256 mins

- Less than two 9's of availability
 - Decreasing the down time by an order of magnitude could improve system availability by two orders of magnitude

Proposed Data-Driven Evaluation (7U)

- 1. Gather failure data and specify fault-model
- 2. Establish fault-remediation relationship
- 3. Select/create fault-injection tools to mimic faults in 1
- 4. Identify Macro-measurements
 - Identify environmental constraints governing system-operation (SLAs, availability, production targets etc.)
- 5. Identify Micro-measurements
 - Identify metrics related to specifics of self-healing mechanisms (success rates, recovery time, fault-coverage)
- 6. Run fault-injection experiments and record observed behavior
- 9 7. Construct pre-experiment and post-experiment models

The 7U-Evaluation Method



Conclusions

- Dynamic instrumentation and fault-injection lets us transparently collect data and replicate problems
- The CTMC-models are flexible enough to quantitatively analyze various styles of repairs
- The math is the "easy" part compared to getting customer data on failures, outages, and their impacts.
 - These details are critical to defining the notions of "better" and "good" for these systems

Future Work

- More experiments on an expanded set of operating systems using more server-applications
 - Linux 2.6
 - OpenSolaris 10
 - Windows XP SP2/Windows 2003 Server
- Modeling and analyzing other self-healing mechanisms
 - Error Virtualization (From STEM to SEAD, Locasto et. al Usenix 2007)
 - Self-Healing in OpenSolaris 10
- Feedback control for policy-driven repair-mechanism selection

Questions, Comments, Queries?

Thank you for your time and attention

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