

# Understanding and Detecting Timing Bugs in Cloud Systems Haopeng Liu, Shan Lu

Background & Motivation

1. Timing bugs in distributed systems:

- Unexpected timing among dist. events:
- -- Message: time-of-execution bugs
- -- Fault: time-of-fault bugs
- Common in dist. systems [1, 2]
- Difficult to avoid and tackle
- Little tool support

2. State of the art – Model checking

Challenge: Complex manual specifications Q1: Can we judge what are timing bugs without manual specifications?

2. State of the art – Random fault injection

Challenge: Many fault injection runs Q2: Can we predict time-of-fault bugs based on just one fault injection, instead of many?



How to estimate distributed impact? How to handle the huge # of accesses? •

## Logical time model for Distributed Systems

#### **Distributed HB Rules** Distributed RPC Sno while-loop Synchron **HBASE HBASE** ZooKeeper ronous Socket service HBASE

## Local HB Rules



## Fault-aware logical time model

### 1. What is new data flow introduced by timing of faults?



Nregular Ncrash

Crash-regular TOF bug □ Fault timing: before W

**Gault-tolerance: timeout** 

obj.wait(long timeout); //R

2. How about data flow between recovery and crash nodes?





Data flow is totally determined by TOF!

	Fault timing: after W
	Fault-tolerance: sanity check
	<pre>//Recovery node if (f.valid()) { //R1: sanity check     dt = f read(): //R2: read</pre>
     	}

## DCatch: predict TOE bugs

**Overview:** a. Customized for distributed systems and TOE bugs; b. Aims Scalability, Coverage, and Accuracy.

- 1. Runtime Tracing [S, C, A]
- Heap accesses related to dist. computation/communication
- Happens-before related operations following HB model

#### 2. HB Analysis [S, C, A]

- HB graph construction following the HB model above
- Data race detection from HB graph<sup>[3]</sup>



## FCatch: predict TOF bugs

**Q.** Only fault-free correct run is enough?

W & R are from different nodes •

#### **Crash-recovery TOF bugs:**

□ Fault timing: after W

Analyze fault-free & faulty traces

Crash-recovery TOF bug

□ Fault-tolerance: sanity check

- W is from *Ncrash* in the fault-free
- R is from *Nrecovery* in the faulty

#### 3. Static Pruning [S, A]

• Statically estimate distributed & local impact of each race Prune out races unlikely to cause failures  $\bullet$ 

Intolerant ops		<ul><li>Crash-regular TOF bugs:</li><li>Timeout (statically check R)</li></ul>			<ul><li>Crash-recovery TOF bugs:</li><li>Sanity check + impact analysis</li></ul>					
			E	valu	atior	1				
APACHE		FCacth	CA1&2	HB-1	HB-2	MR-1	MR-2	ZK	Total	
Hibase here		#. Bench (harmful)	2 + /	1 + /	/+1	/ + 1	/+2	/+1	3+5	
		#. Unknown (harmful)	1 + 0	0 + 0	2 + 2	1 + 1	1 + 1	0 + 0	4 + 4	
Apache Zookooper		<b>#. False positives</b>	0 + 2	3+6	2 + 0	0 + 0	0 + 0	0 + 2	5 + 10	
<b>cassandra</b>		More det	ails: <b>ht</b>	tp://	fcatch	.cs.uc	hicago	o.edu		

#### Evaluation

<b>DCatch</b>	CA	HB-1	HB-2	MR-1	<b>MR-2</b>	<b>ZK-1</b>	<b>ZK-2</b>	Tota	l
Detected?	<b>~</b>	<b>~</b>	~	~	~	<b>~</b>	~		
#. Harmful bugs	3	3	4	2	1	5	6	20	Inc
#. Benign bugs	0	0	1	0	2	1	2	5	IIE
<b>#. False positives</b>	0	1	0	4	4	1	0	7	

[1]. T. Leesatapornwongsa, J. Lukman, S. Lu, and H. Gunawi. TaxDC: A Taxonomy of Non-Deterministic Concurrency Bugs in Datacenter Distributed Systems. In ASPLOS, 2016 [2]. Zhenyu Guo, Sean McDirmid, Mao Yang, Li Zhuang, Pu Zhang, Yingwei Luo, Tom Bergan, Peter Bodik, Madan Musuvathi, Zheng Zhang, and Lidong Zhou. Failure recovery: When the cure is worse than the disease. In HotOS, 2013 [3]. V. Raychev, M. Vechev, and M. Sridharan. Effective Race Detection for Event-Driven Programs. In OOPSLA, 2013