

Capturing and Enhancing *In Situ* System Observability for Failure Detection

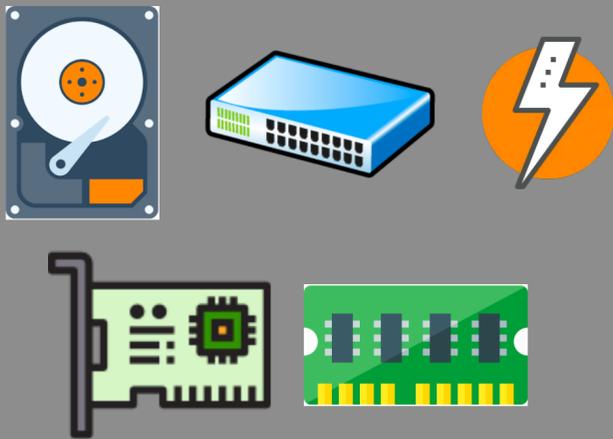
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Chuanxiong Guo[†], Jacob R. Lorch[‡], Lidong Zhou[‡], Yingnong Dang⁺

*Johns Hopkins University [†]ByteDance Inc. [‡]Microsoft Research ⁺Microsoft



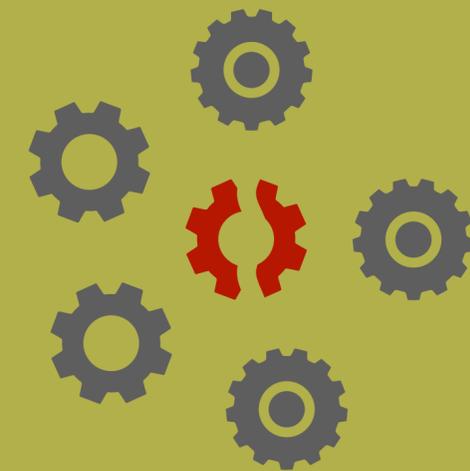
Faults Are Common in Large Systems



Hardware Faults

```
int len = in.readInt();  
if (len == -1)  
    return null;  
byte arr[] = new byte[len];  
in.readFully(arr);  
return arr;
```

Software Bugs



Misconfiguration

Detecting Failure Is Crucial

FD is a fundamental building block for fault-tolerant systems

$$\text{Availability} \uparrow \approx \frac{\text{MTBF} \uparrow}{\text{MTBF} + \text{MTTR} \downarrow}$$

Improve reliability

Speed up recovery^[1]

Detecting Failure Is Crucial

FD is a fundamental building block for fault-tolerant systems

$$\text{Availability} \uparrow \approx \frac{\text{MTBF} \uparrow}{\text{MTBF} + \text{MTTR} \downarrow}$$

Improve reliability

Speed up recovery^[1]

But only if failure can be
detected reliably and rapidly first

Failure Detection is Hard

Unreliable Failure Detectors for Reliable Distributed Systems

TUSHAR DEEPAK CHANDRA

I.B.M. Thomas J. Watson Research Center, Hawthorne, New York

AND

SAM TOUEG

Cornell University, Ithaca, NY

We introduce the concept of an unreliable failure detector for solving Consensus in asynchronous systems. We show that such a detector can be used to solve Consensus in asynchronous systems. We also present a companion paper that introduces a detector for solving Consensus in asynchronous systems. Categories and Subject Descriptors: distributed applications; reliability; programming—distributed systems; reliability; programming—distributed systems; reliability; programming—distributed systems. [Computation by Abstract Data Types] [Computation by Abstract Data Types] [Database Management Systems] General Terms: Algorithms Additional Key Words: Byzantine Generals' problem; fault-tolerance, message

A Gossip-Style Failure Detection Service

Robbert van Renesse, Yaron Minsky, and Mark Hayden*

Perfect Failure Detection In Timed Asynchronous Systems

Christof Fetzer

The φ Accrual Failure Detector

Naohiro Hayashibara*, Xavier Défago*[†] (contact), Rami Yared* and Takuya Katayama*

The Fault Detection Problem

Andreas Haeberlen¹ and Petr Kuznetsov²

Detecting failures in distributed systems with the FALCON spy network

Joshua B. Leners* Hao Wu* Wei-Lun Hung* Marcos K. Aguilera[†] Michael Walfish*

*The University of Texas at Austin [†]Microsoft Research Silicon Valley

ABSTRACT

A common way for a distributed system to tolerate crashes is to explicitly detect them and then recover from them. Interestingly, detection can take much longer than recovery, as a result of many advances in recovery techniques, making failure detection the dominant factor in these systems' unavailability when a crash occurs.

This paper presents the design, implementation, and evaluation of Falcon, a failure detector with several features. First, Falcon's common-case detection time is sub-second, which keeps unavailability low. Second, Falcon is reliable: it never reports a process as down when it is actually up. Third, Falcon sometimes kills to

waiting for the timer to fire. Indeed, we (and everyone else) are personally familiar with the hiccups that occur when a distributed system freezes until a timeout expires. More technically, examples of timeouts in real systems include 60 seconds for GFS [29], at least 12 seconds for Chubby [14], 30 seconds for Dryad [32], and 60 seconds for NFS. Of course, one could set a shorter timeout—and thereby increase the risk of falsely declaring a working node as down. We discuss end-to-end timeouts further in Section 2.2 and for now just assert that there are no good end-to-end timeout values.

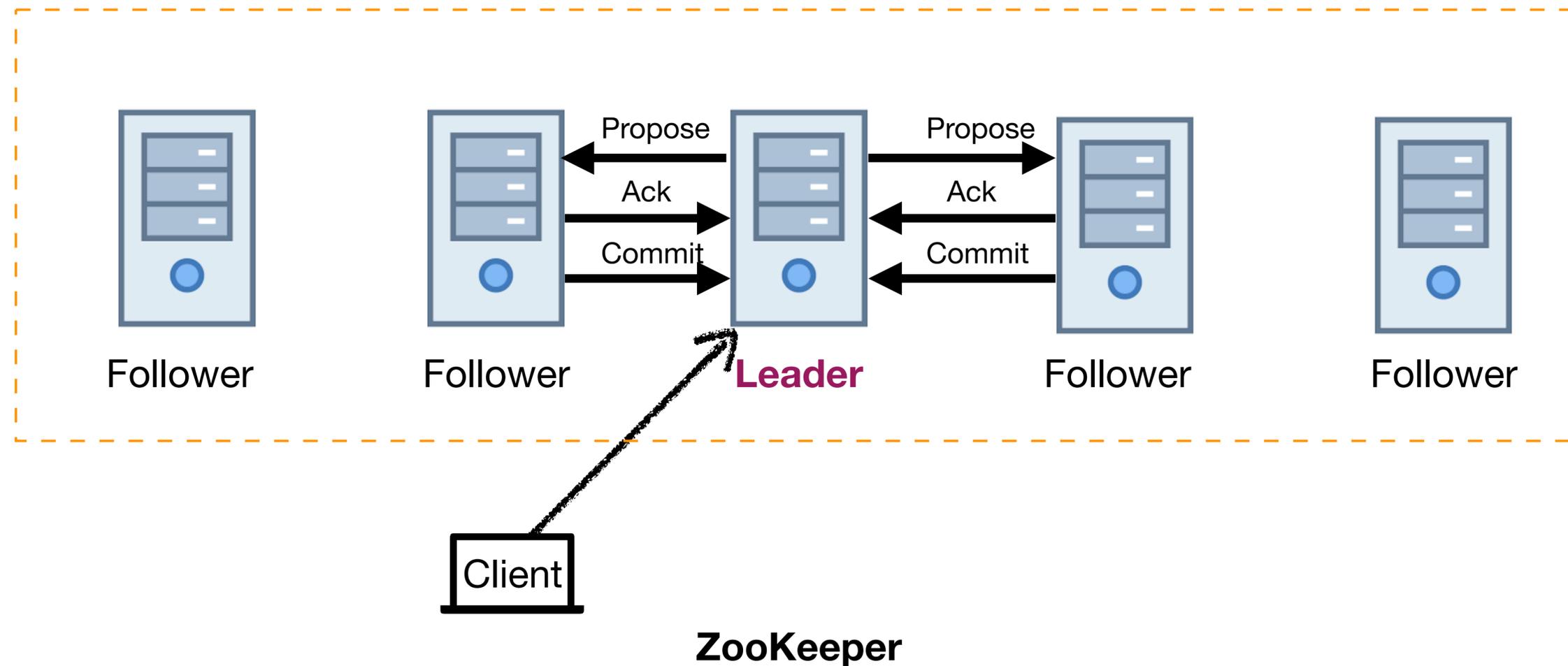
This paper introduces Falcon (Fast And Lethal Component Observation Network), a failure detector that leverages internal knowledge from various system layers to achieve a new combination in

- Extensively studied for several decades
- Focus on detection of fail-stop failures in asynchronous systems
 - ▶ Difficult to reliably determine whether a process has crashed or is merely "very slow"

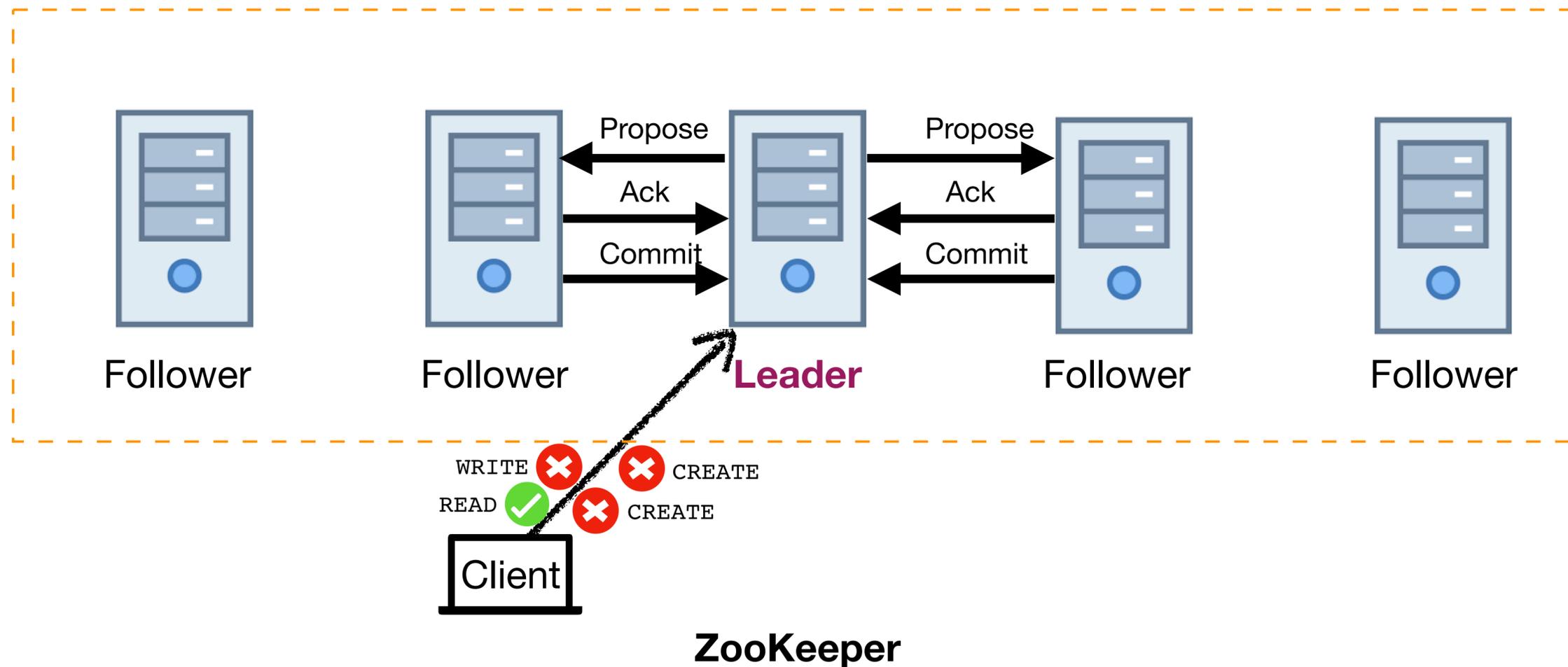
Production Failures Beyond Fail-Stop

- A component status can be between UP and DOWN: **gray failure**
 - ▶ E.g., fail-slow hardware, random packet loss, limp lock, ...
 - ▶ Failure symptoms are very subtle
- Common in production cloud systems
- Pose new challenges for failure detection

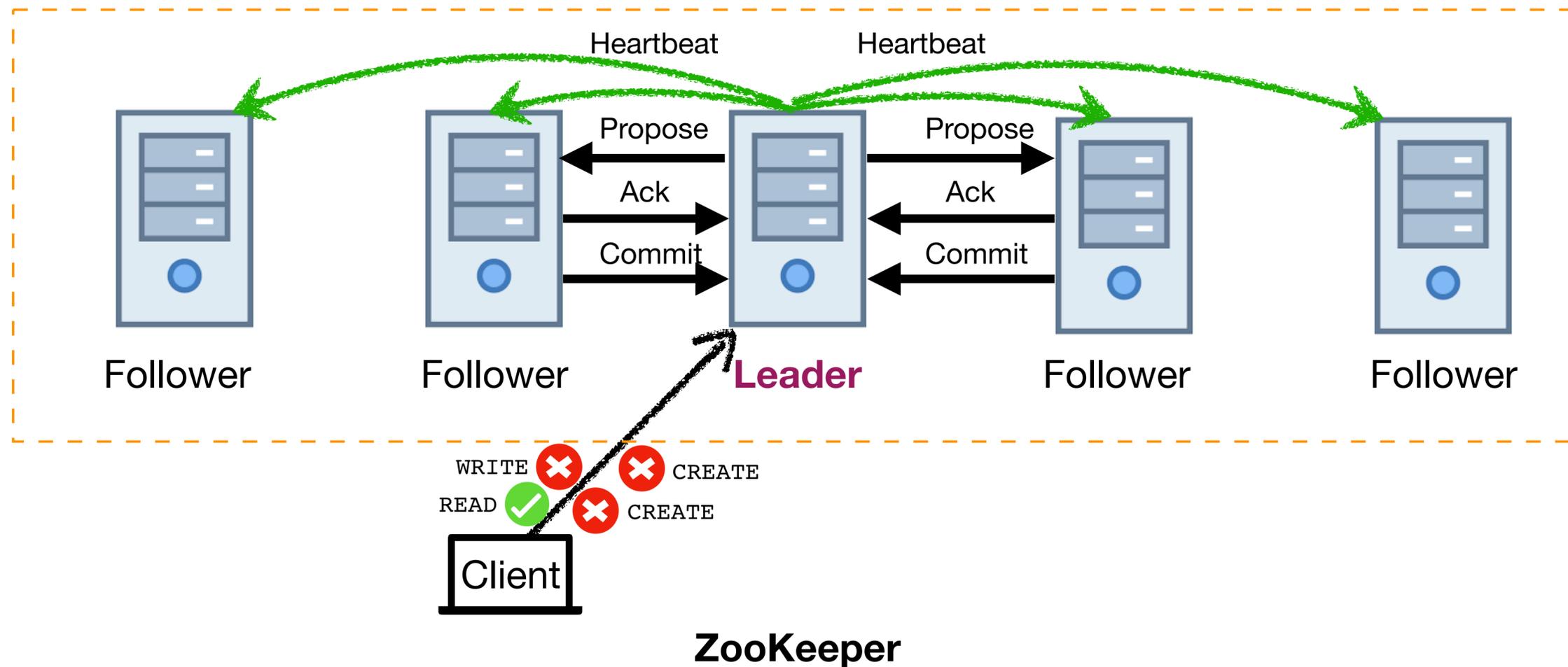
An Example Production Gray Failure



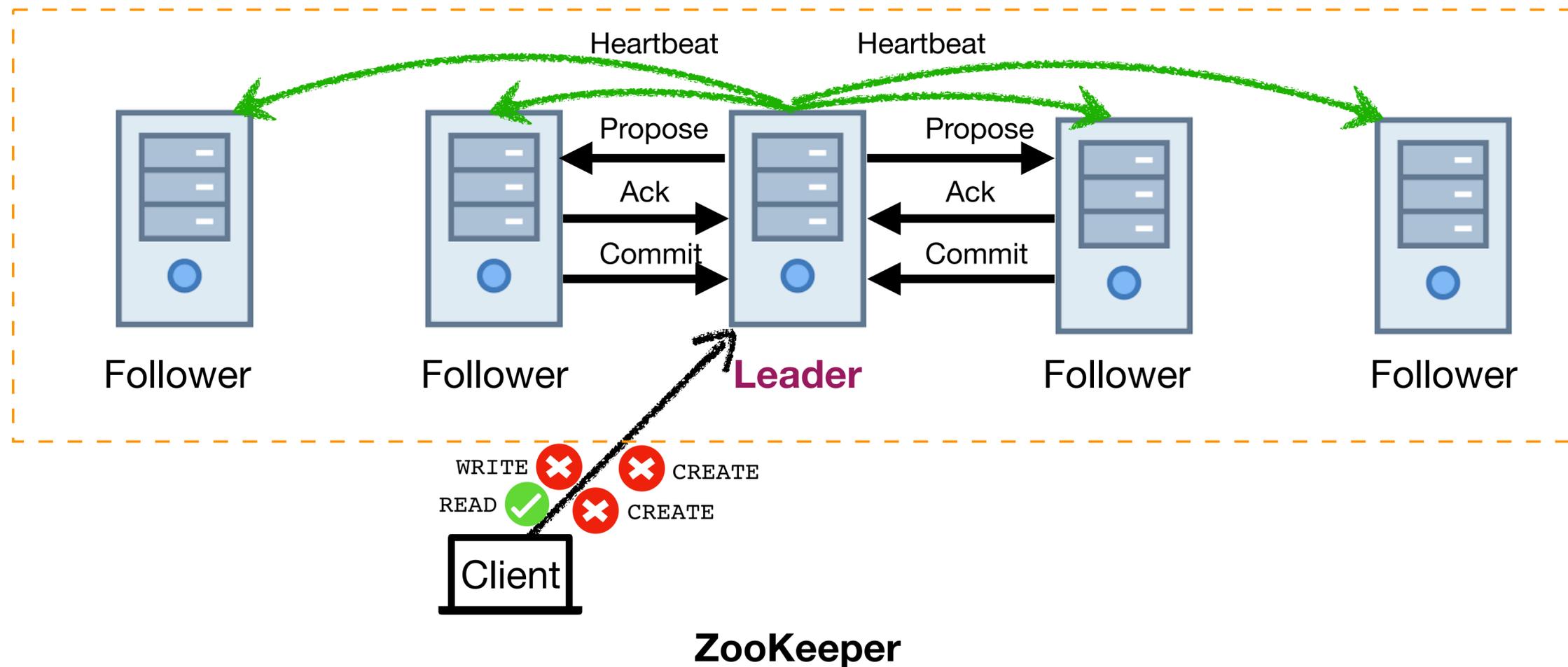
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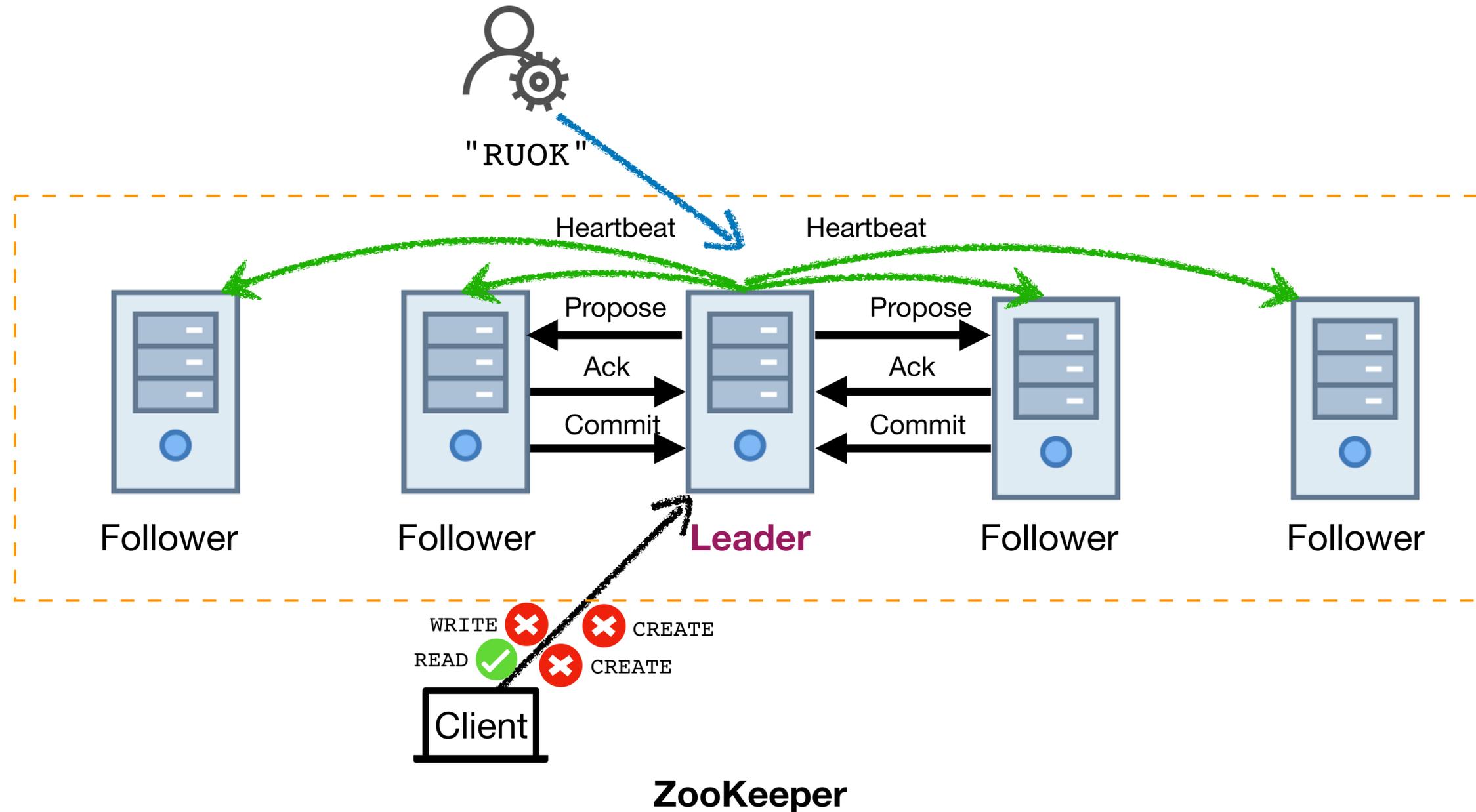
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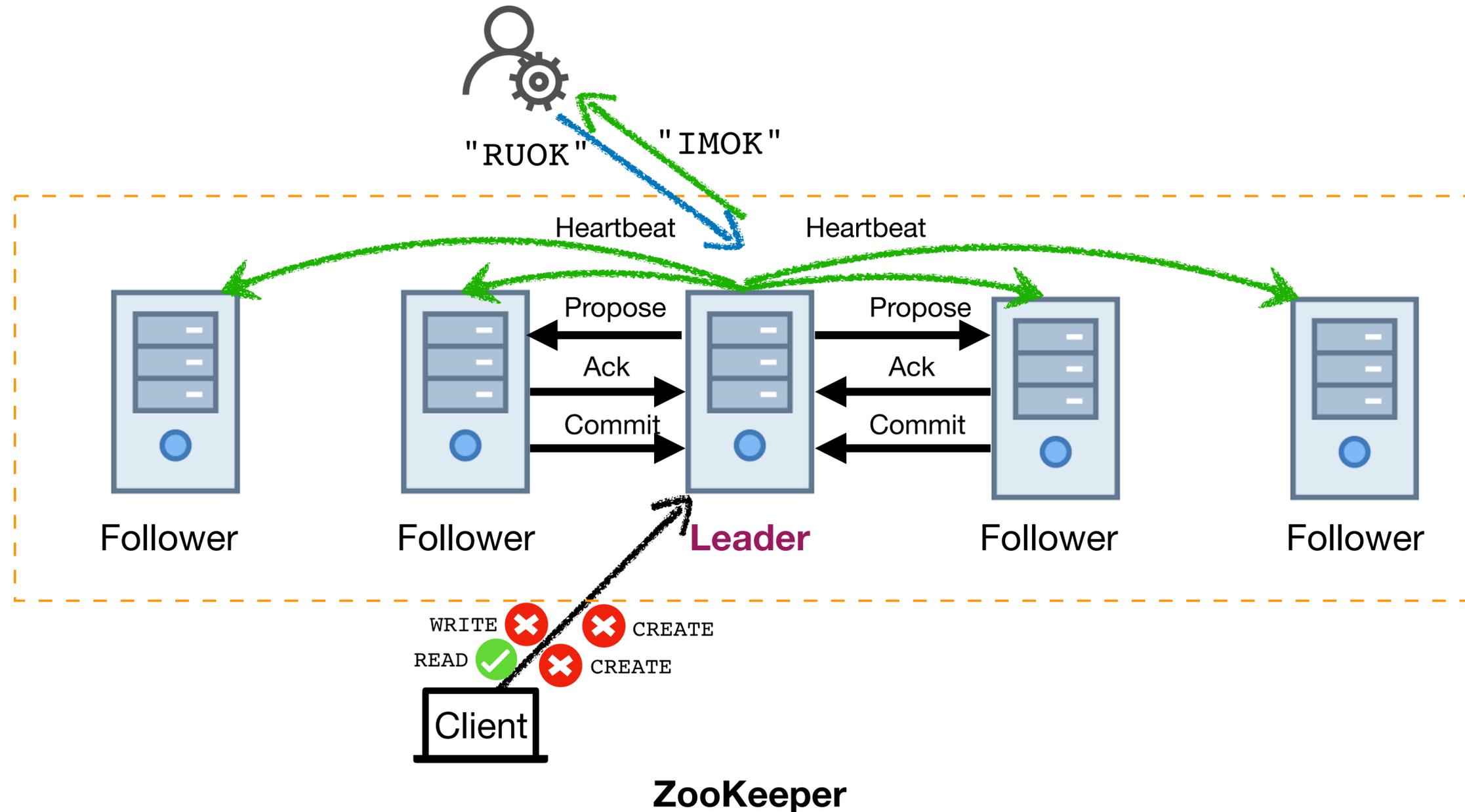
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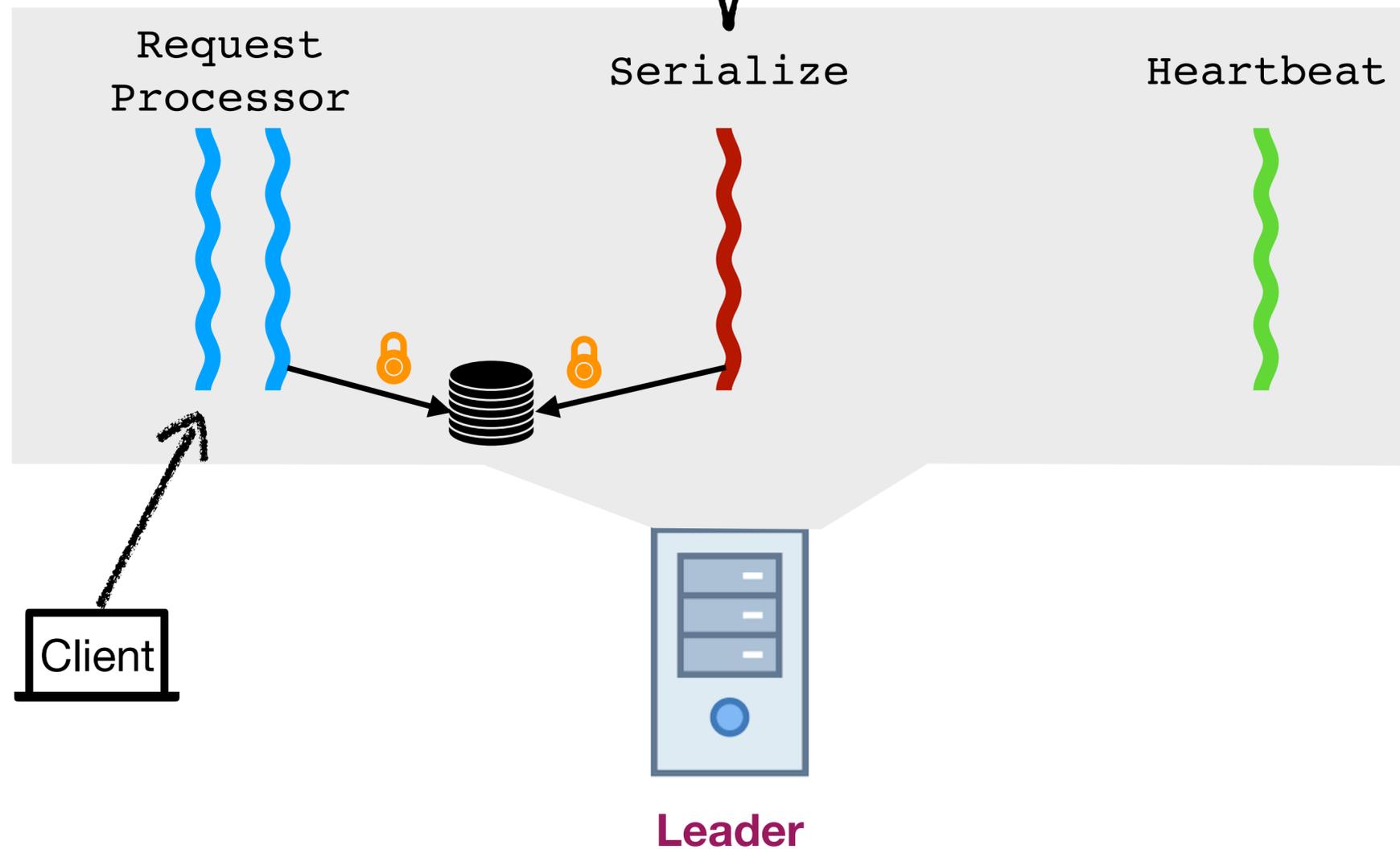
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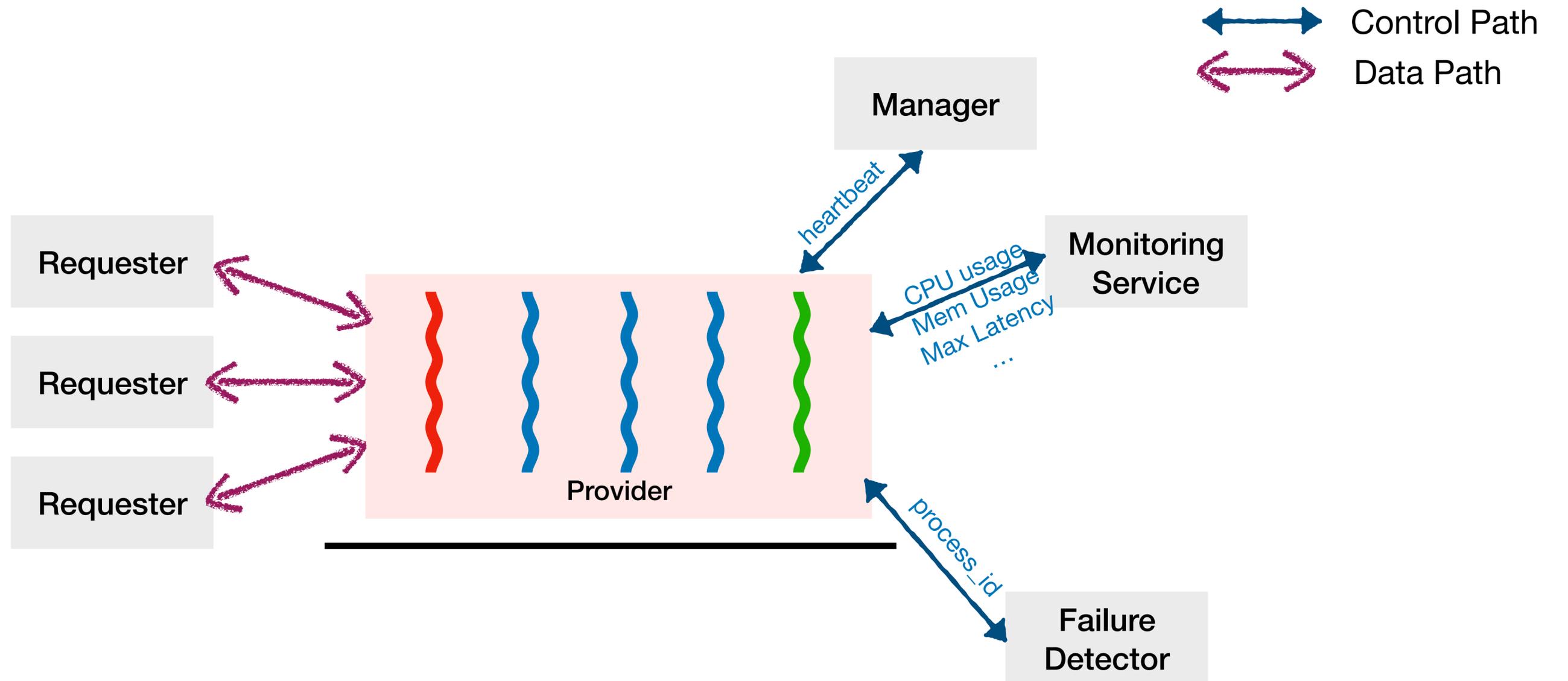


Root Cause of the Gray Failure

```
synchronized (node) {  
    output.writeString(path, "path");  
    output.writeRecord(node, "node");  
}
```

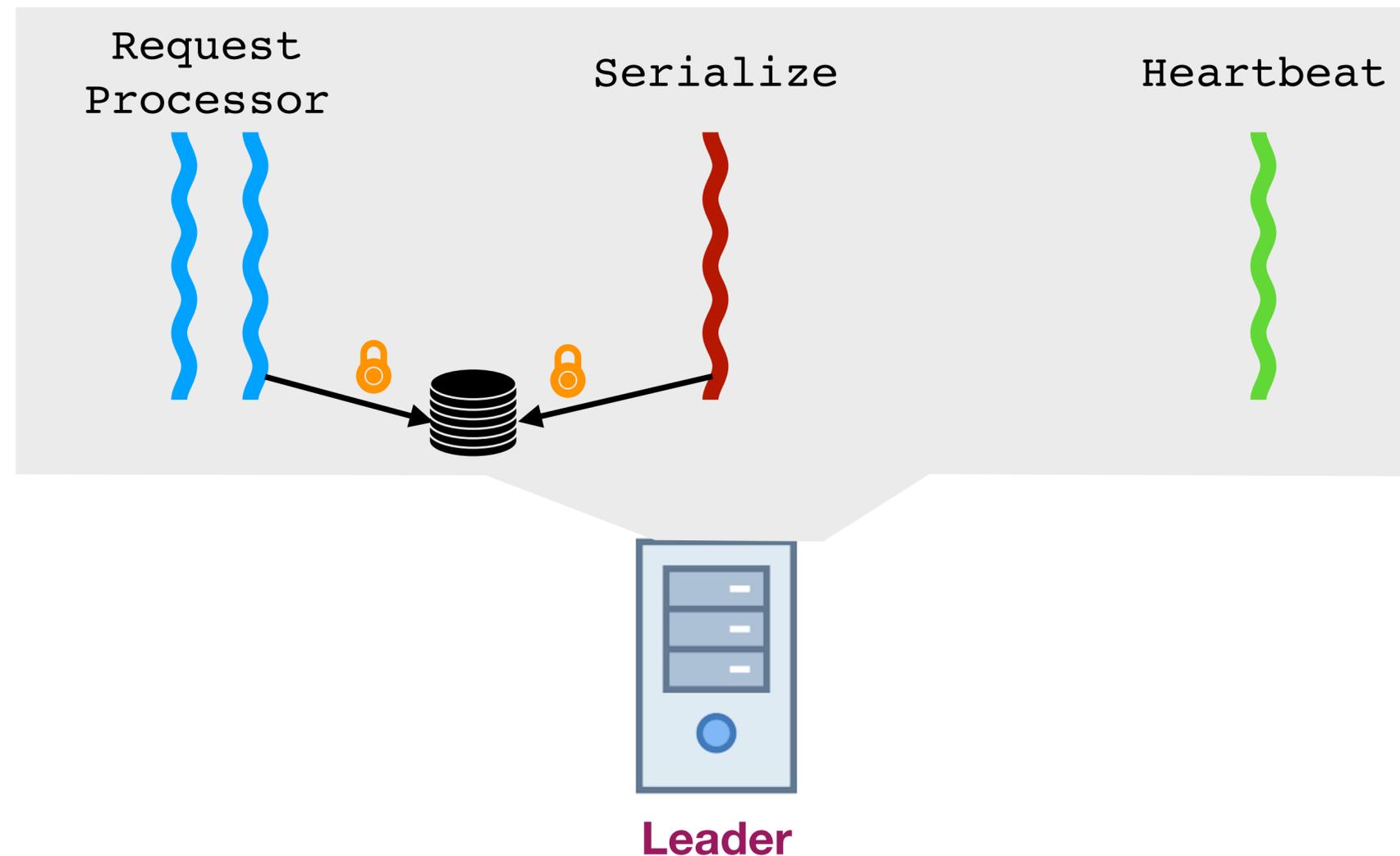
Stuck due to transient network issue



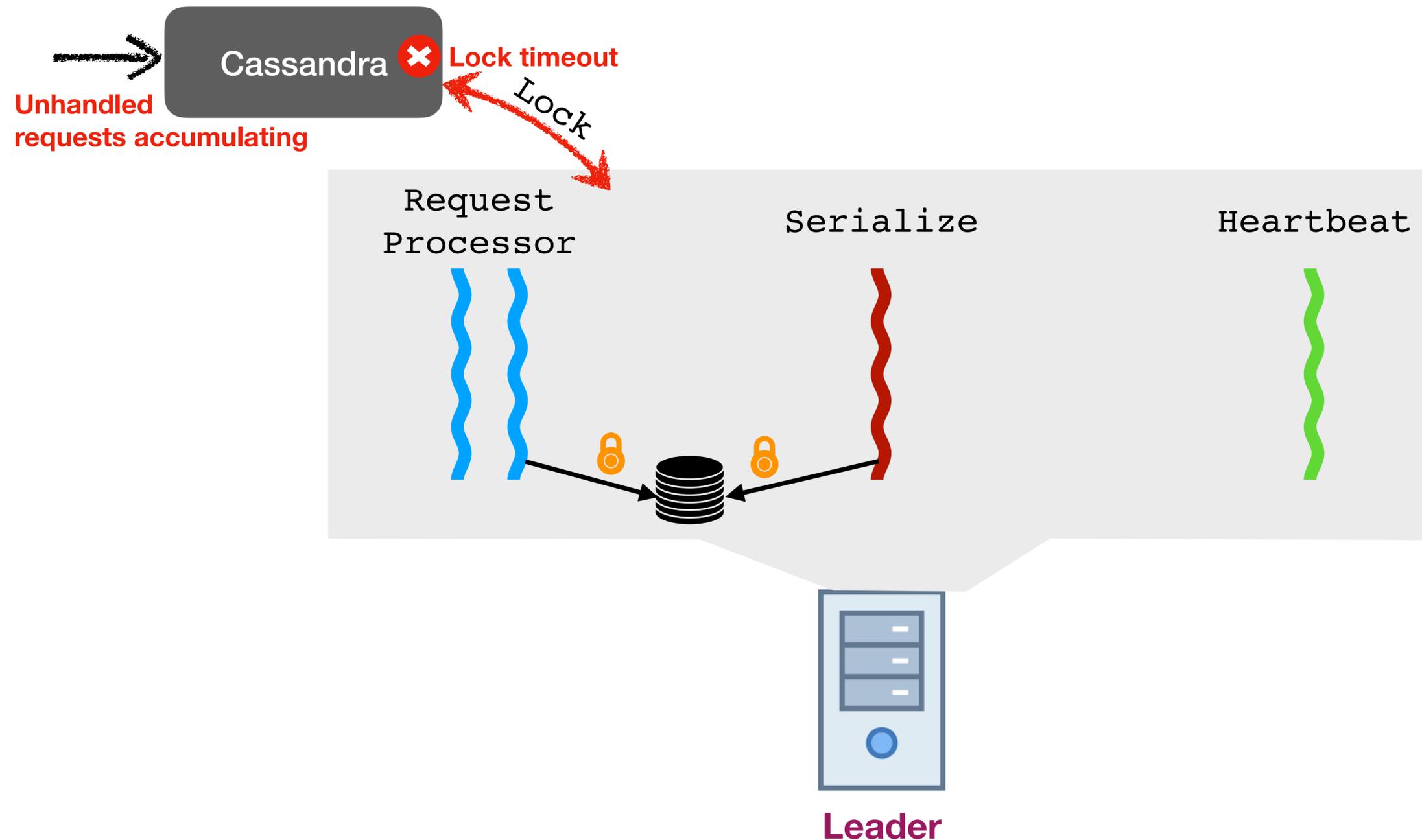


Insight: we should detect what the *requesters* see

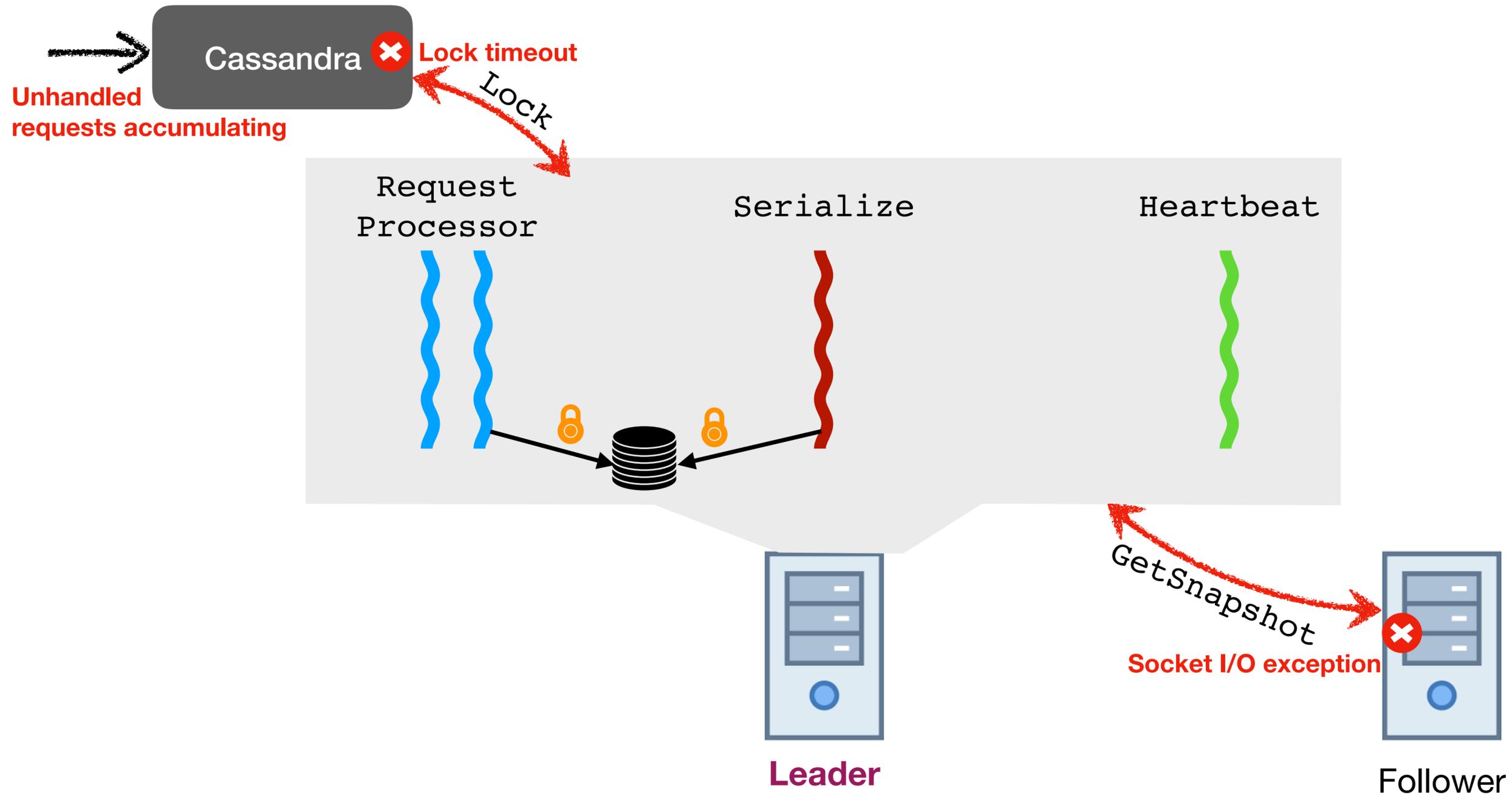
*Insight 1: Critical Gray Failures Are **Observable***



Insight 1: Critical Gray Failures Are *Observable*



Insight 1: Critical Gray Failures Are *Observable*



Insight 2: From Error *Handling* to Error *Reporting*

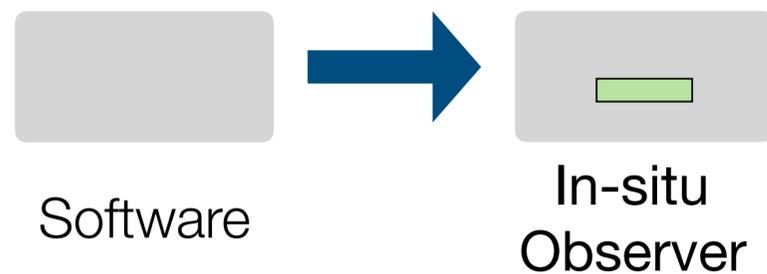
```
void syncWithLeader(long newLeaderZxid) {
    try {
        deserializeSnapshot(leaderIs);
        String sig = leaderIs.read("signature");
        if (!sig.equals("BenWasHere"))
            throw new IOException("Bad signature");
        } else {
            LOG.error("Unexpected leader packet.");
            System.exit(13);
        }
    } catch (IOException e) {
        LOG.warn("Exception sync with leader", e);
        sock.close();
    }
}
```

Insight 2: From Error *Handling* to Error *Reporting*

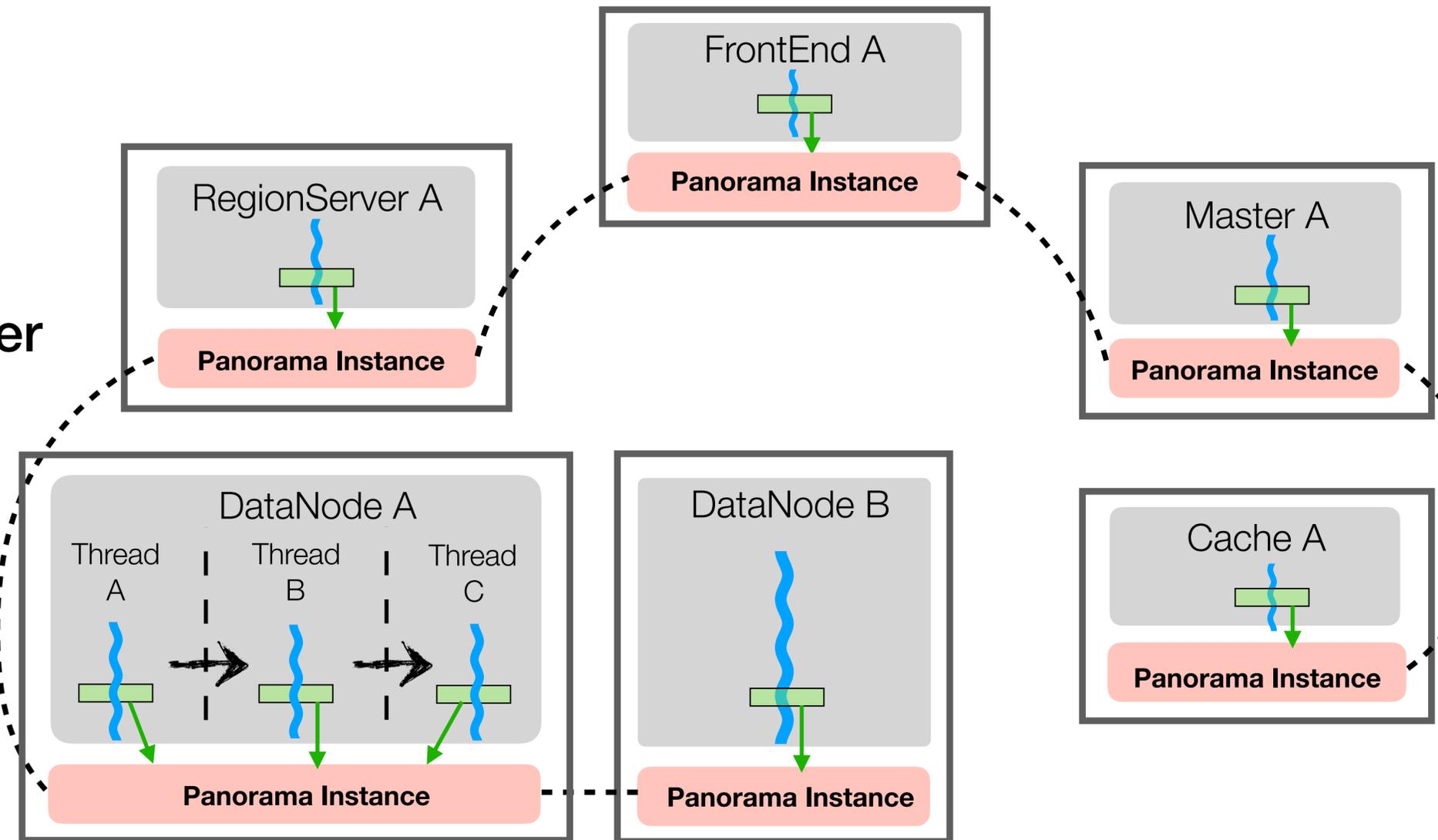
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```



Our Solution: Panorama



- A tool to turn a software into an observer
- Uniform observation abstractions
- A generic failure detection service for any component to participate



Overview of Panorama - Analysis

+ observability hooks

```
...  
void func() {  
  try {  
    sync(t);  
  } catch (RemoteError e) {  
    LOG.error(e);  
    retry();  
  }  
}  
...
```

Original Software

Offline static analysis

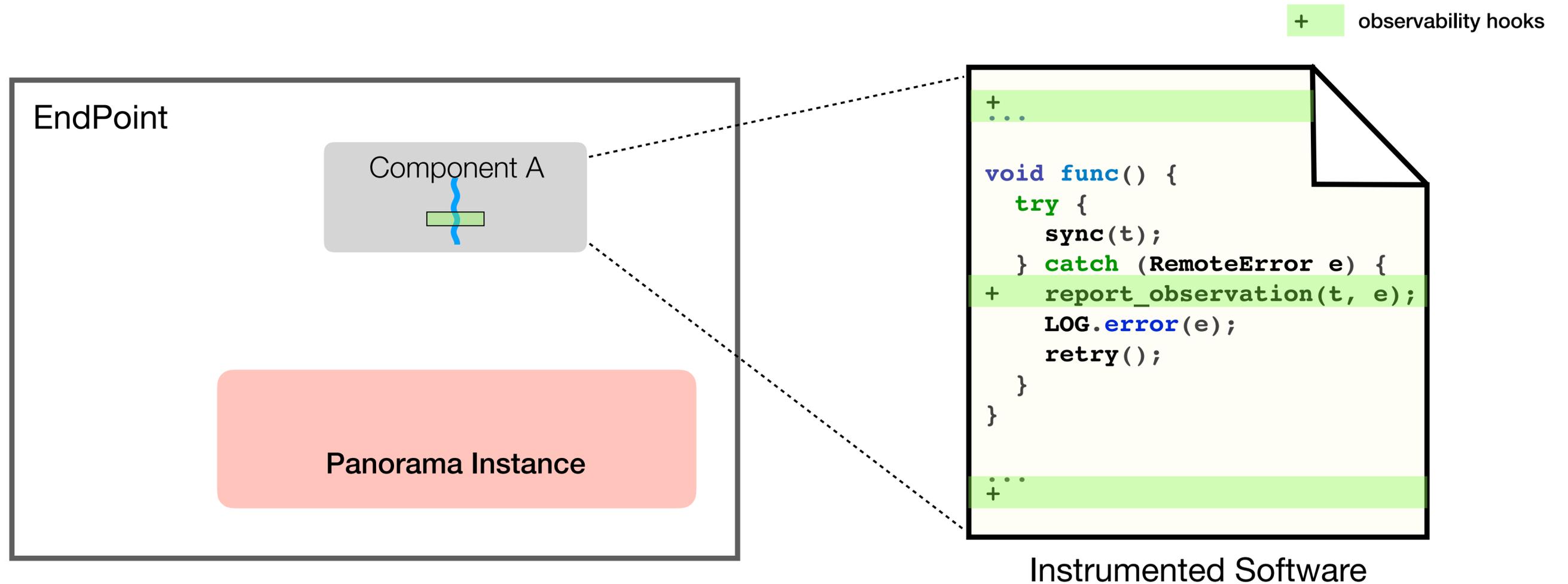
Instrumentation

```
+ ...  
void func() {  
  try {  
    sync(t);  
  } catch (RemoteError e) {  
+ report_observation(t, e);  
    LOG.error(e);  
    retry();  
  }  
}  
+ ...
```

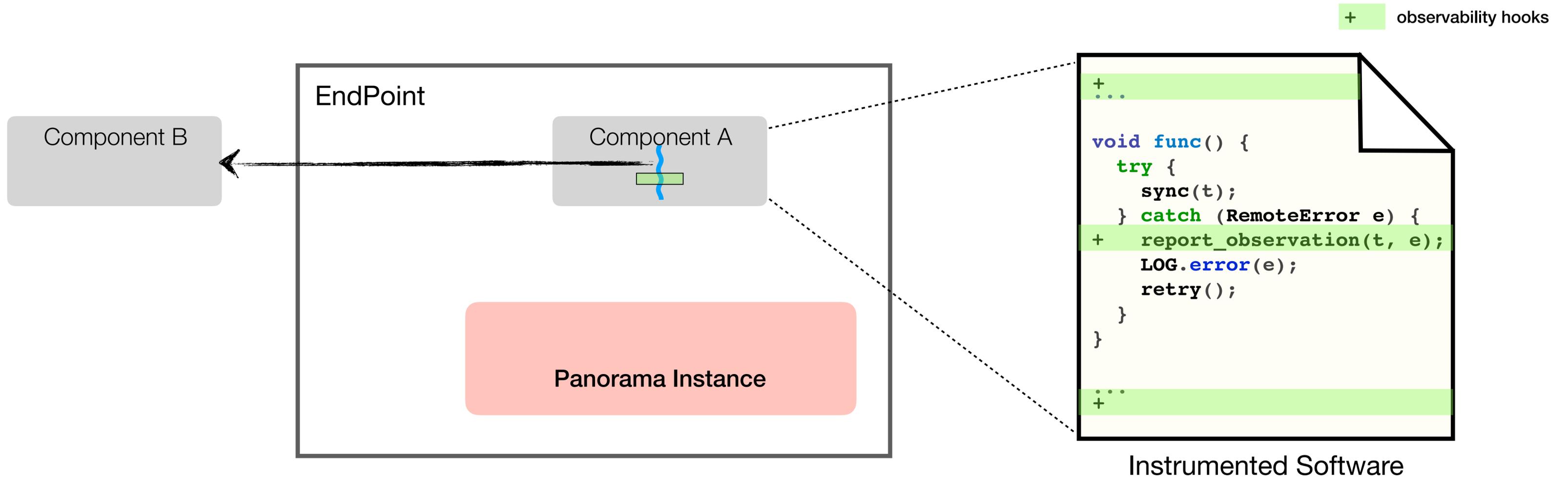
Instrumented Software

Automatically convert a software component into an *in-situ* observer

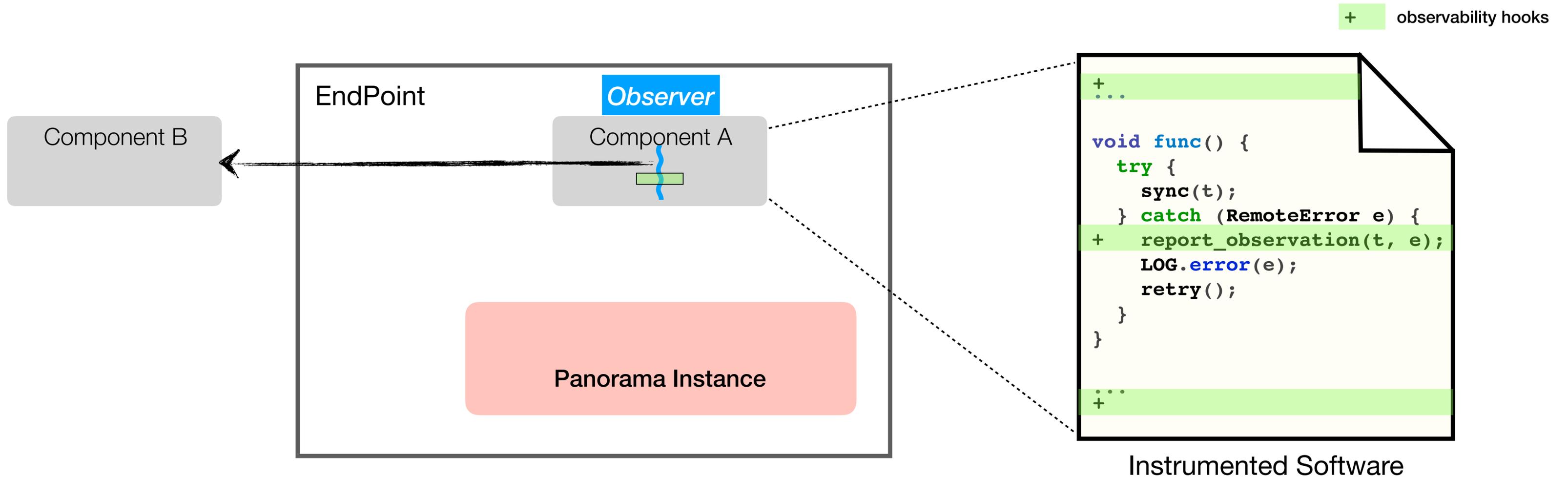
Overview of Panorama - Runtime



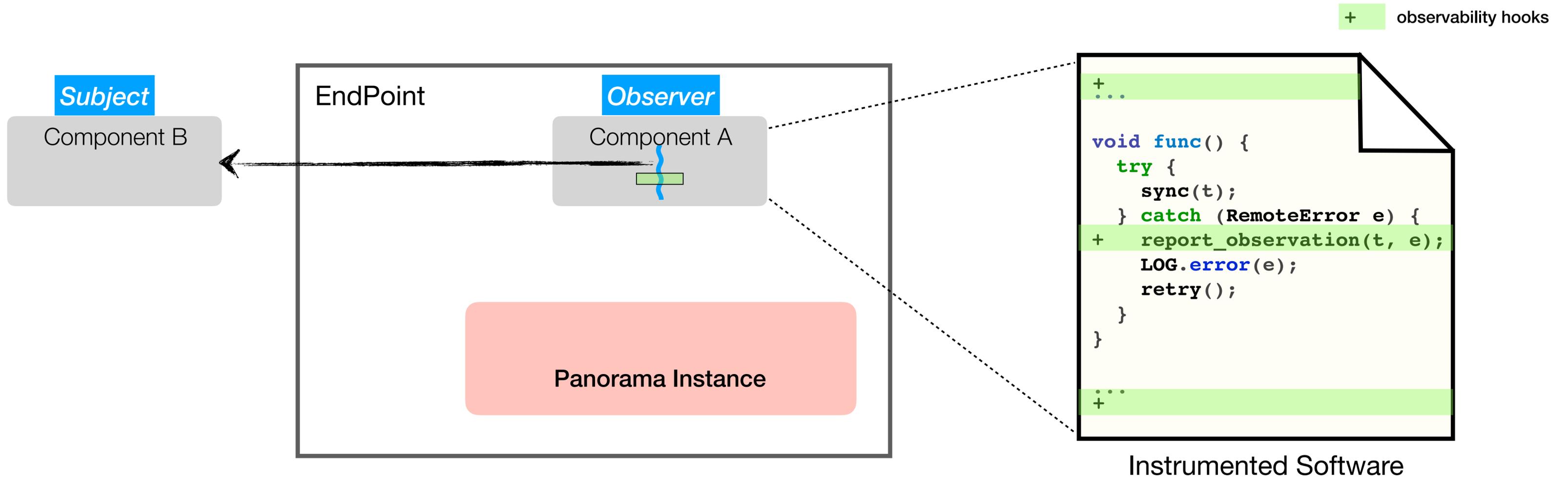
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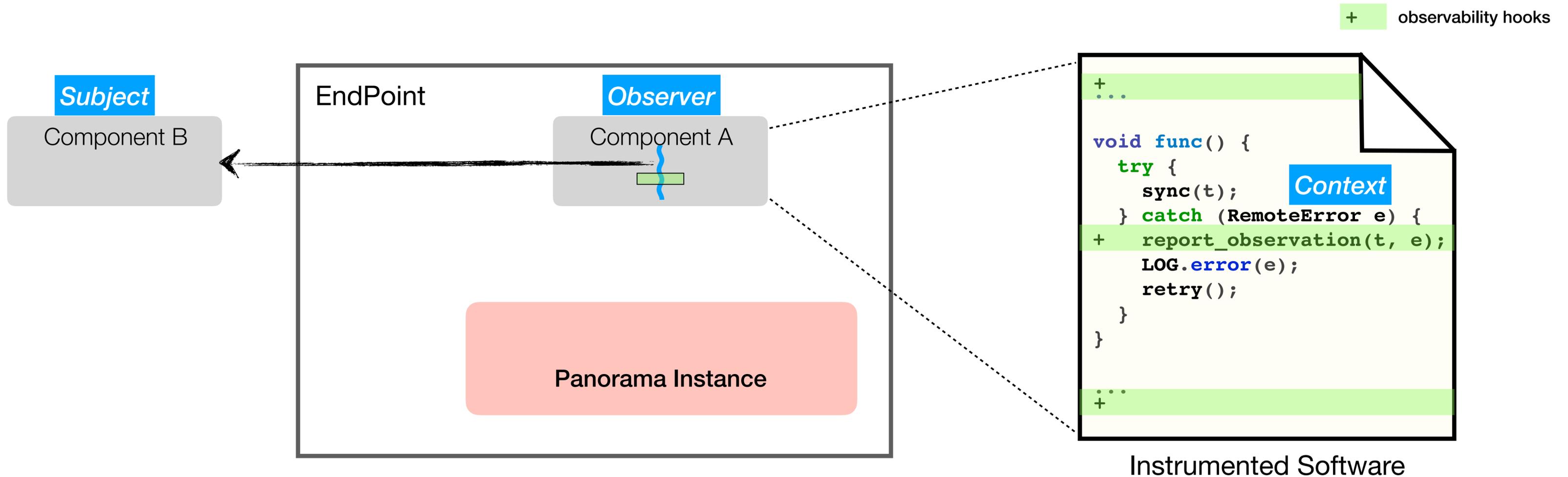
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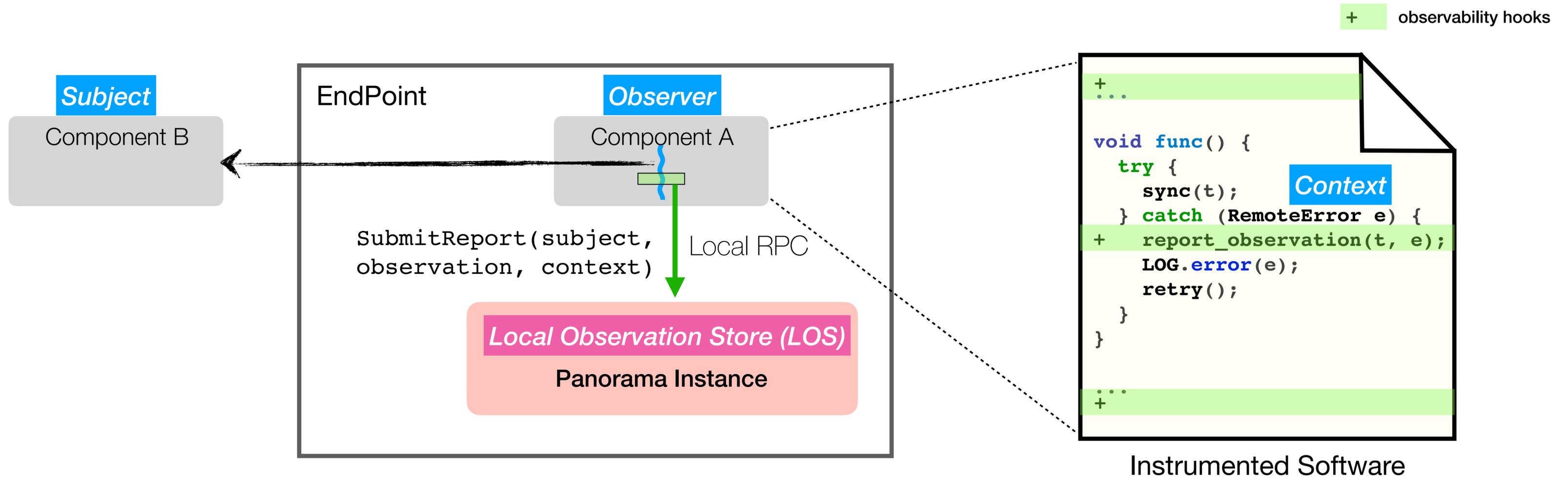
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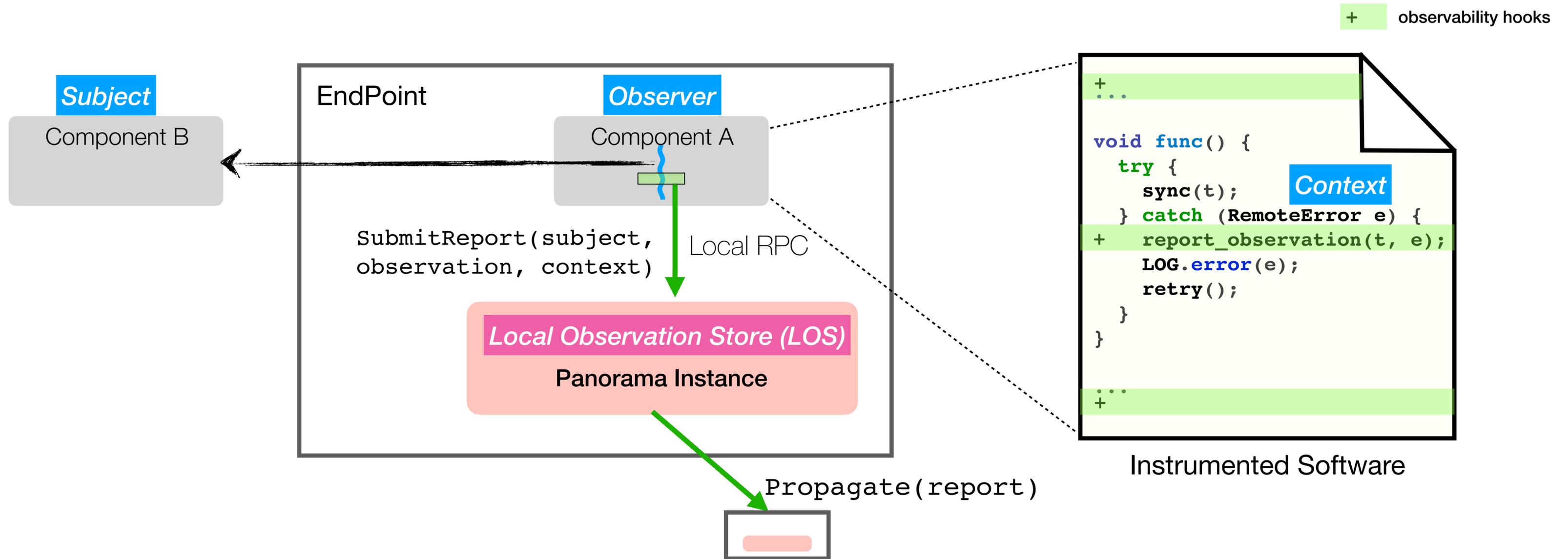
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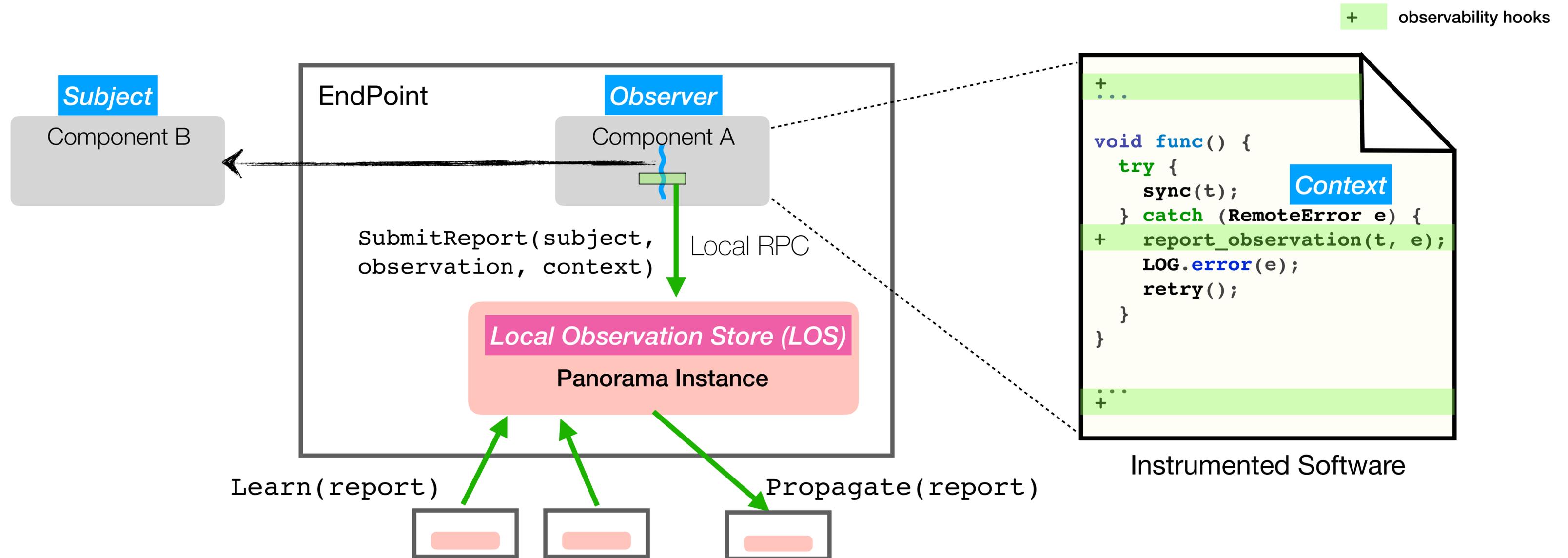
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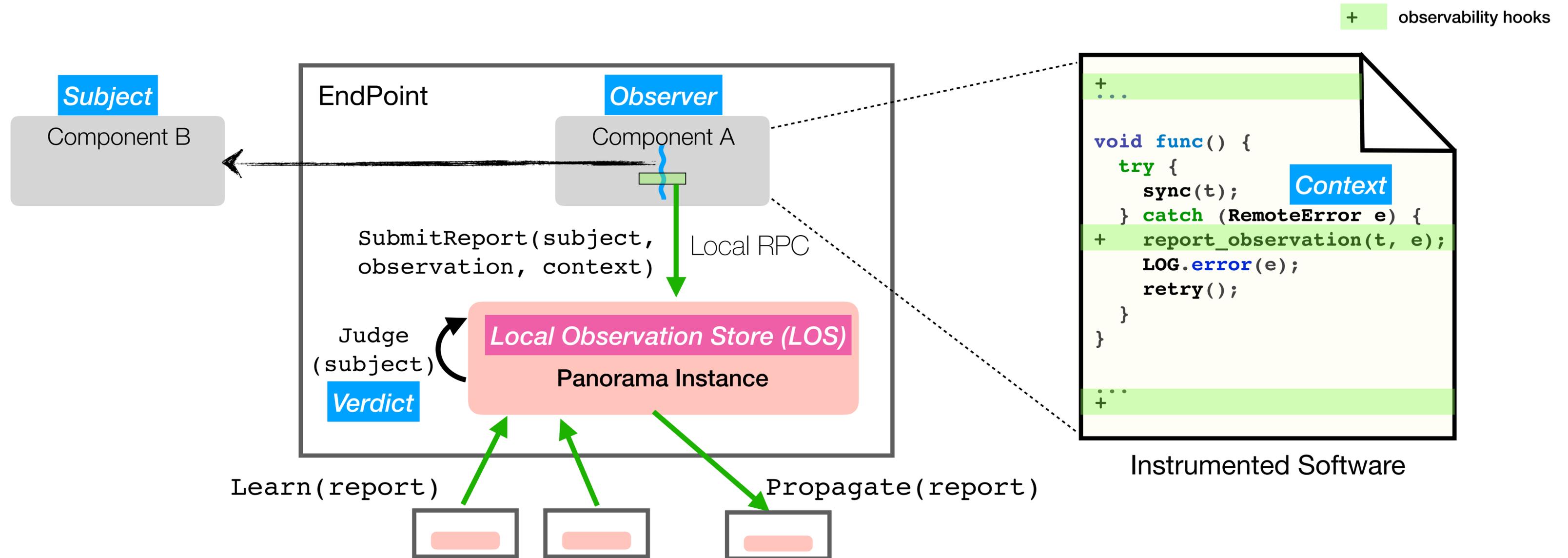
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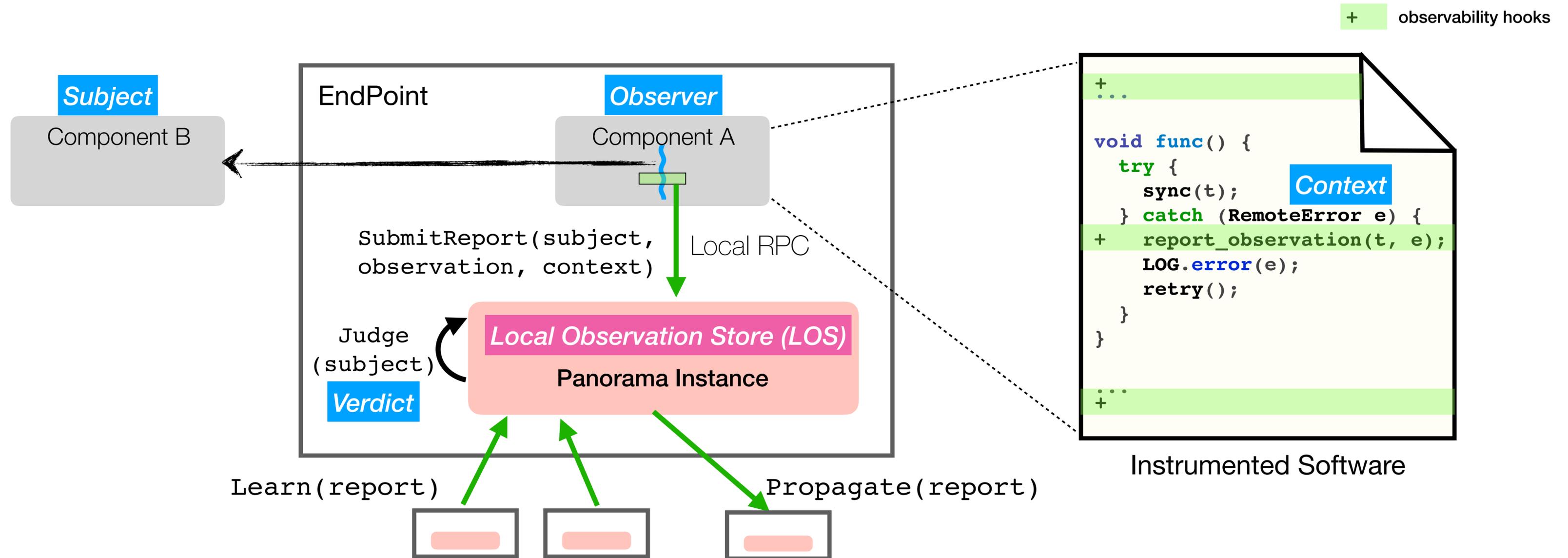
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Overview of Panorama - Runtime



Overview of Panorama - Runtime



This talk focuses on the **analysis stage**; see paper for details of the runtime service

Design Challenges

- Observations dispersed in software's source code
- Observation collection may slow down observer's normal service
- Diverse programming paradigms affect observability

Observability Analysis

- Step ❶ : locate domain-crossing invocations (*ob-boundary*)
- Step ❷ : identify observer and observed
- Step ❸ : extract observation (*ob-point*)

Example of Observability Analysis

```
void deserialize(DataTree dt, InputArchive ia) {
    DataNode node = ia.readRecord("node");
    if (node.parent == null) {
        LOG.error("Missing parent.");
        throw new IOException("Invalid Datatree");
    }
    dt.add(node);
}
void snapshot() {
    InputArchive ia = BinaryInputArchive.getArchive(sock.getInputStream());
    try {
        deserialize(getDataTree(), ia);
    } catch (IOException e) {
        sock.close();
    }
}
```

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1 locate ob-boundary

Example of Observability Analysis

② identify observer and subject

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① locate ob-boundary

③ extract observation (ob-point)
(invalid reply)

Example of Observability Analysis

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    }  
}
```

① locate ob-boundary

③ extract observation (ob-point)
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Example of Observability Analysis

2 identify observer and subject

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    try {  
        deserialize(getDataTree(), ia);  
    } catch (IOException e) {  
        sock.close();  
    }  
}
```

1 locate ob-boundary

3 extract observation (ob-point)
(invalid reply)

3 extract observation (ob-point)
(no reply)

1 Locate Observation Boundary

- Function innovations that cross different components

- ▶ socket I/O, RPCs, messaging, etc.

- `java.io.DataOutput.write*()`

- `org.apache.hadoop.hbase.ipc.RpcCall.sendResponseIfReady()`

- `protobuf.RegionServerStatusService.*()`

② Identify Observer

- **Observer identity is global and tracks the source of observations**
 - ▶ The id/name the process uses in the distributed system
 - e.g., `QuorumServer.id` (value from `myid` file in ZooKeeper service config)
- **One-time registration with Panorama instance when the process starts**
 - ▶ Panorama builds a map between identity and endpoint address for reverse lookup

② Identify Observed (Subject)

- Information is in the argument or object of ob-boundary

▶ `public int sendRR(MessageOut message, InetAddress to, IAsyncCallback cb)`  lookup subject identity from address

▶ `public NNCommand NNProtocolTranslatorPB.startCheckpoint(Registration registration)`

subject identity in the field

`private NNProtocolPB rpcProxy;`

- Sometimes the information is provided by a proxy

▶ `rpc HRegionServer.RegionServerStatusService.*(*)`

subject identity from ZooKeeper

`masterAddressTracker.getMasterAddress(false); // get master address from ZooKeeper`

③ Extract Negative Evidence

- Errors/exceptions that originate from the ob-boundary
 - ▶ instrument exception handlers
 - ▶ needs to distinguish generic exceptions (e.g., `IOException`)
- Unexpected reply content
 - ▶ `if (!reply.sig.equals("BenWasHere")) error("Bad signature");`
 - ▶ intra-procedural analysis to look for errors that have control-dependency on the reply

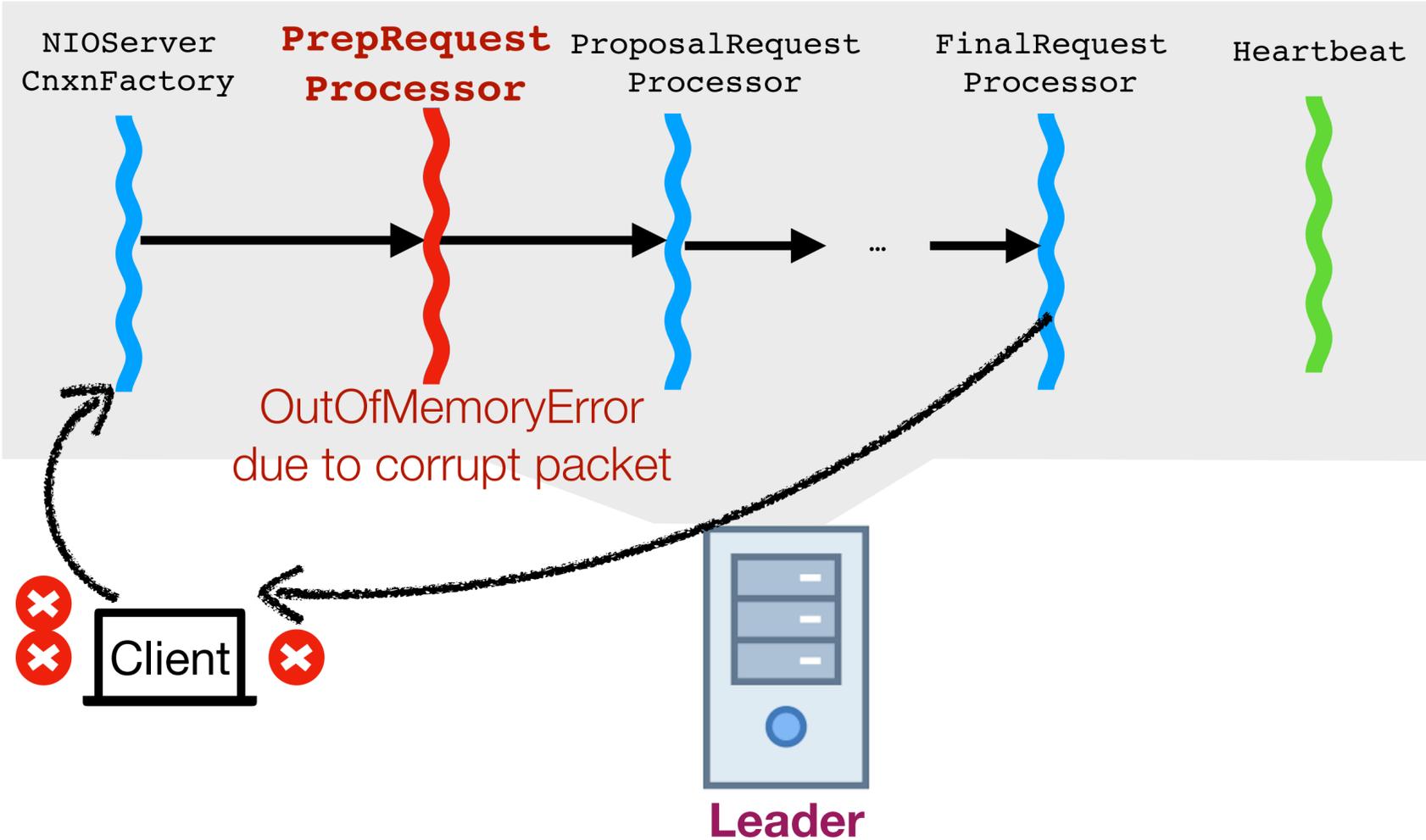
③ Extract Positive Evidence

- **Successful invocation from ob-boundary**
 - ▶ it's OK to submit a positive observation immediately and later find errors in reply content
- **If domain-crossing occurs frequently, positive observations are excessive**
 - ▶ coalesce similar positive ob-points that are located close together
 - ▶ library buffers frequent observations and sends them as one aggregate observation

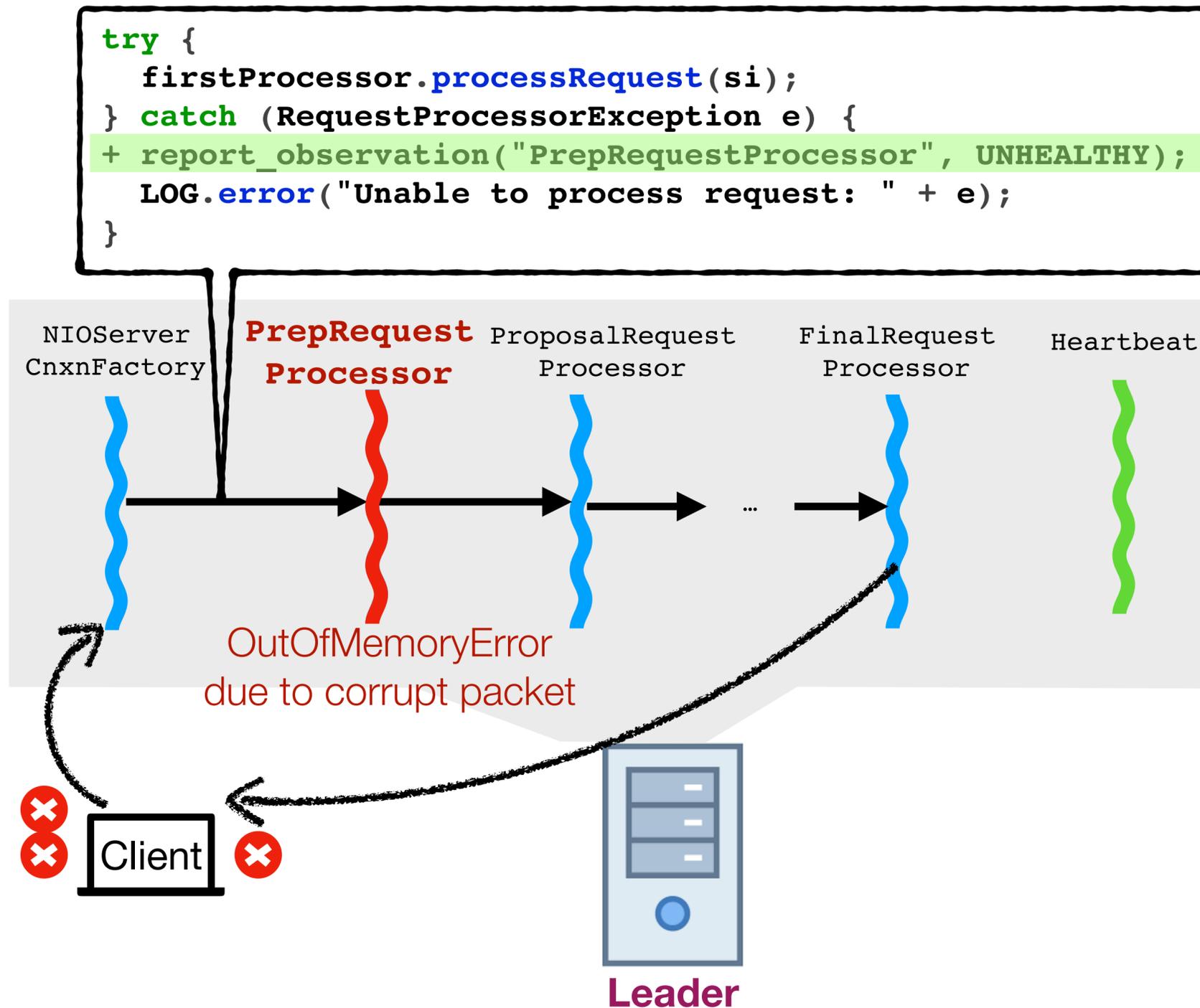
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Learning from a Subtle Case



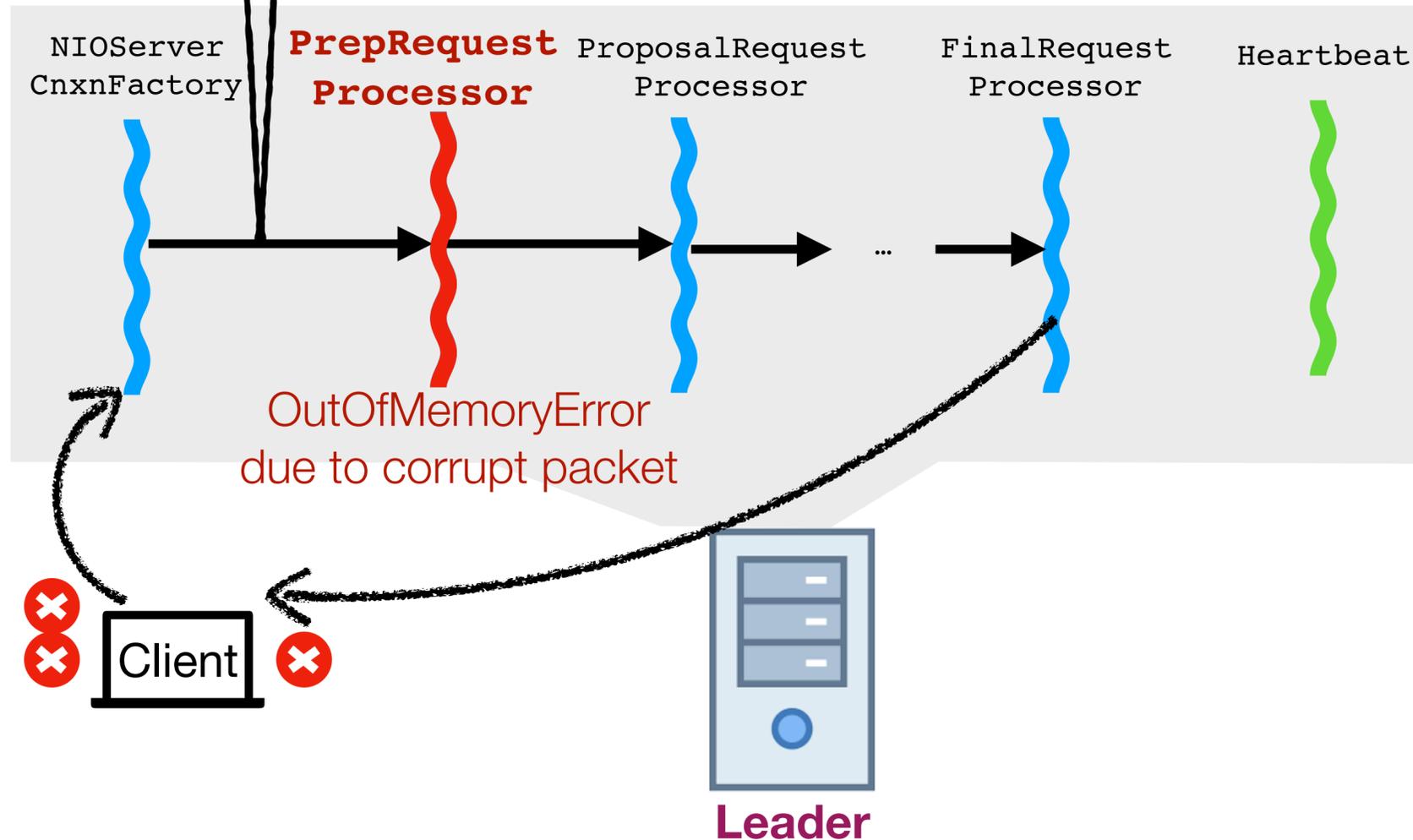
Learning from a Subtle Case



Learning from a Subtle Case

```
try {  
    firstProcessor.processRequest(si);  
} catch (RequestProcessorException e) {  
    + report_observation("PrepRequestProcessor", UNHEALTHY);  
    LOG.error("Unable to process request: " + e);  
}
```

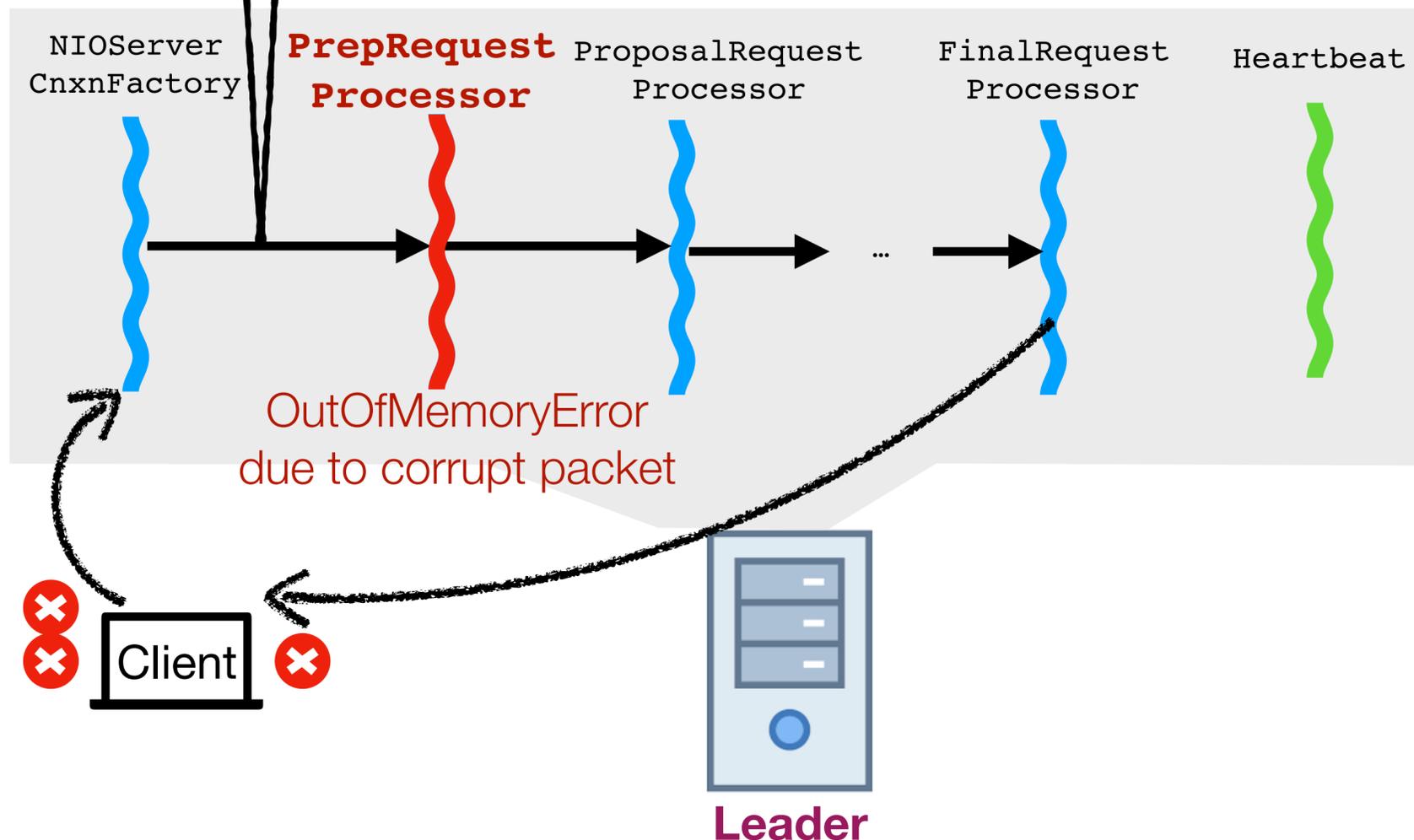
PrepRequest
Processor



Learning from a Subtle Case

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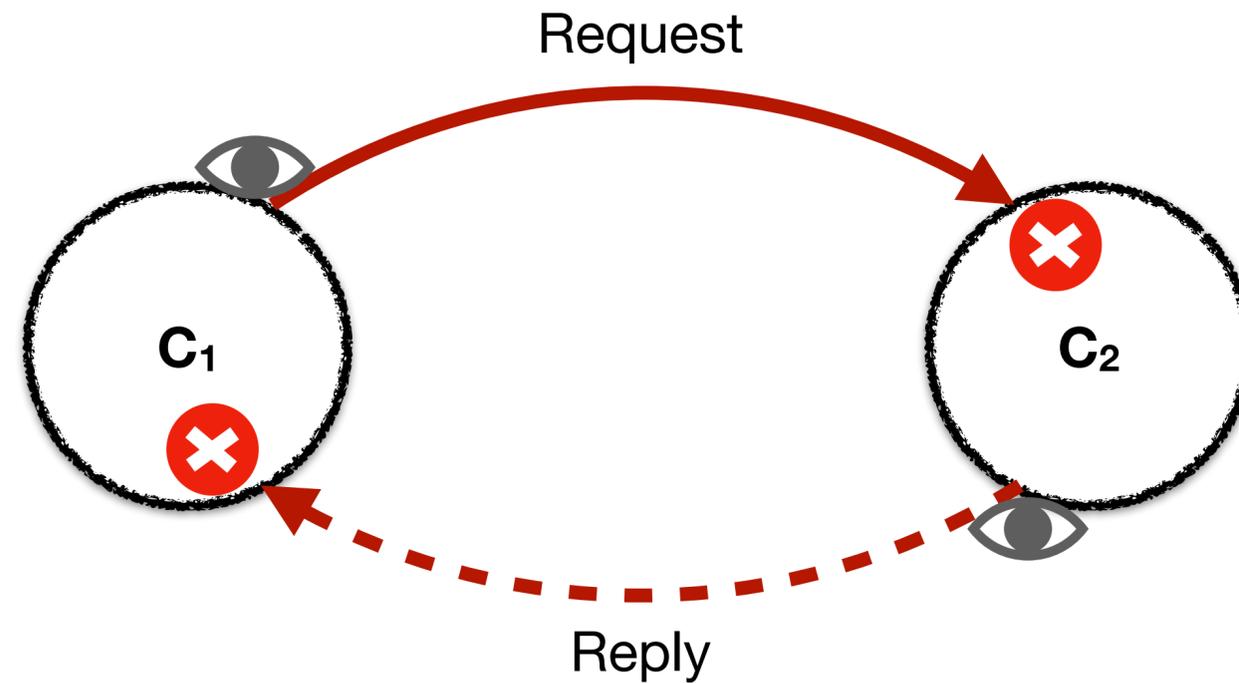
```
public class PrepRequestProcessor extends Thread {  
    BlockingQueue<Request> requests = new ...  
    void run() {  
        while (true) {  
            Request r = requests.take();  
            pRequest(r); // the OOM occurred inside  
        }  
    }  
    public void processRequest(Request request)  
        throws RequestProcessorException {  
        requests.add(request);  
    }  
}
```



Learning from a Subtle Case

- Indirection can *reduce* failure observability
- Evidence is made about the indirection layer instead of actual component

Observability Design Patterns

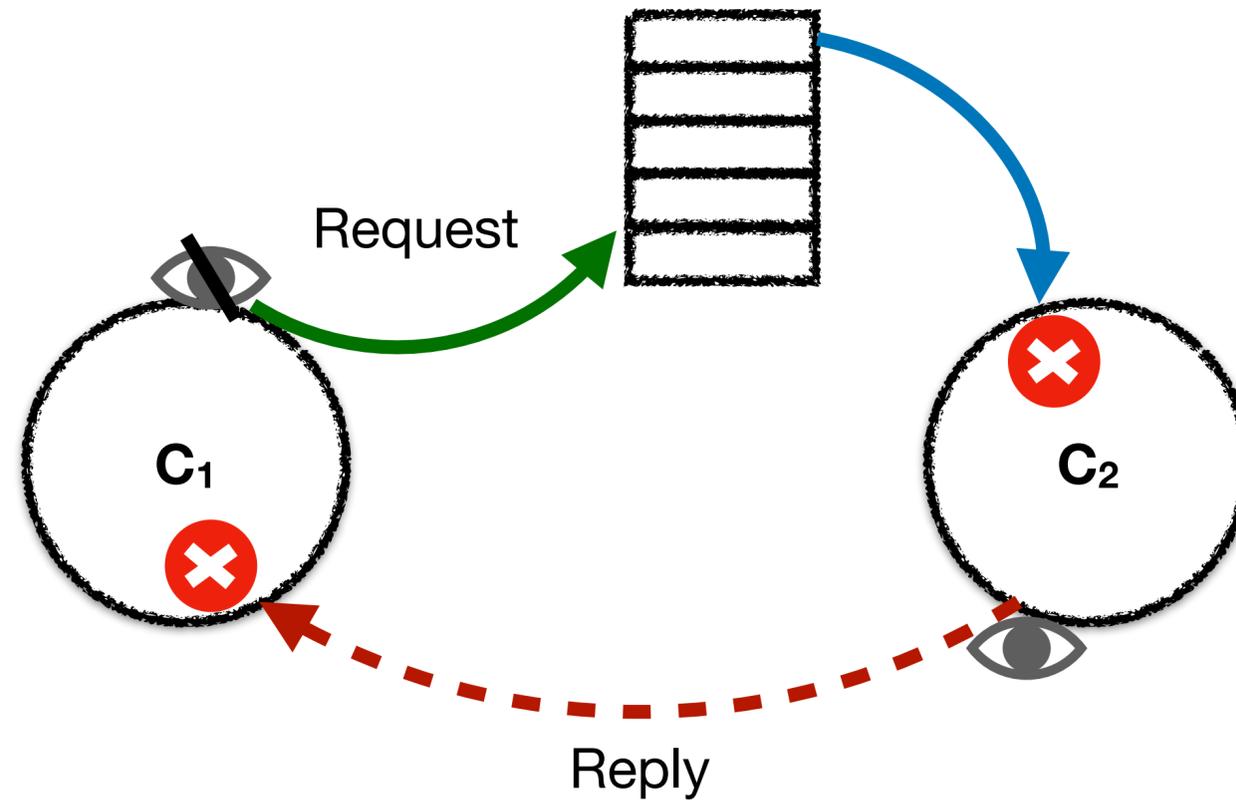
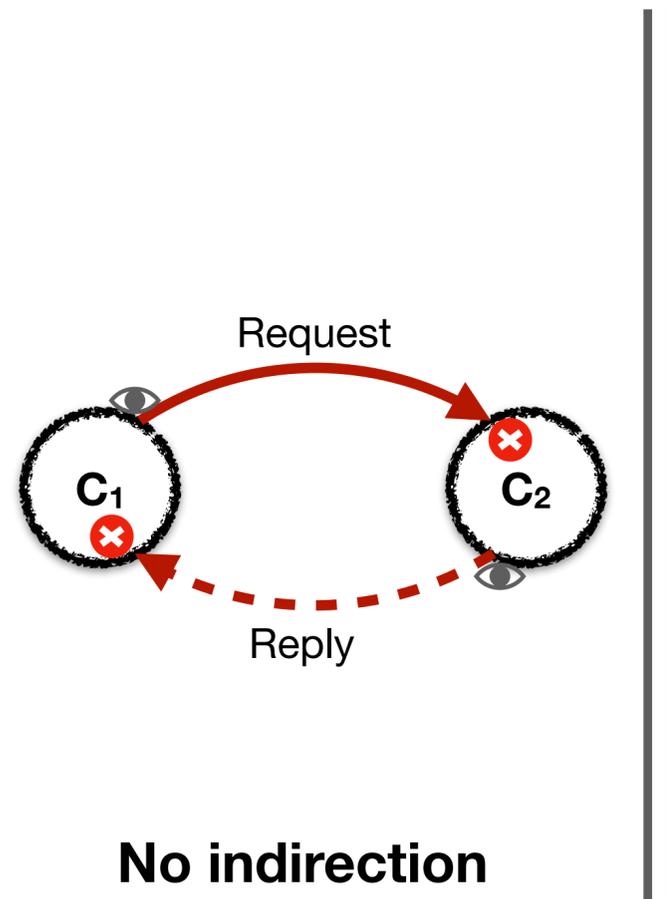


C₂'s failure is observable to C₁

C₁'s failure is observable to C₂

Interaction of two components C₁ and C₂

Observability Design Patterns

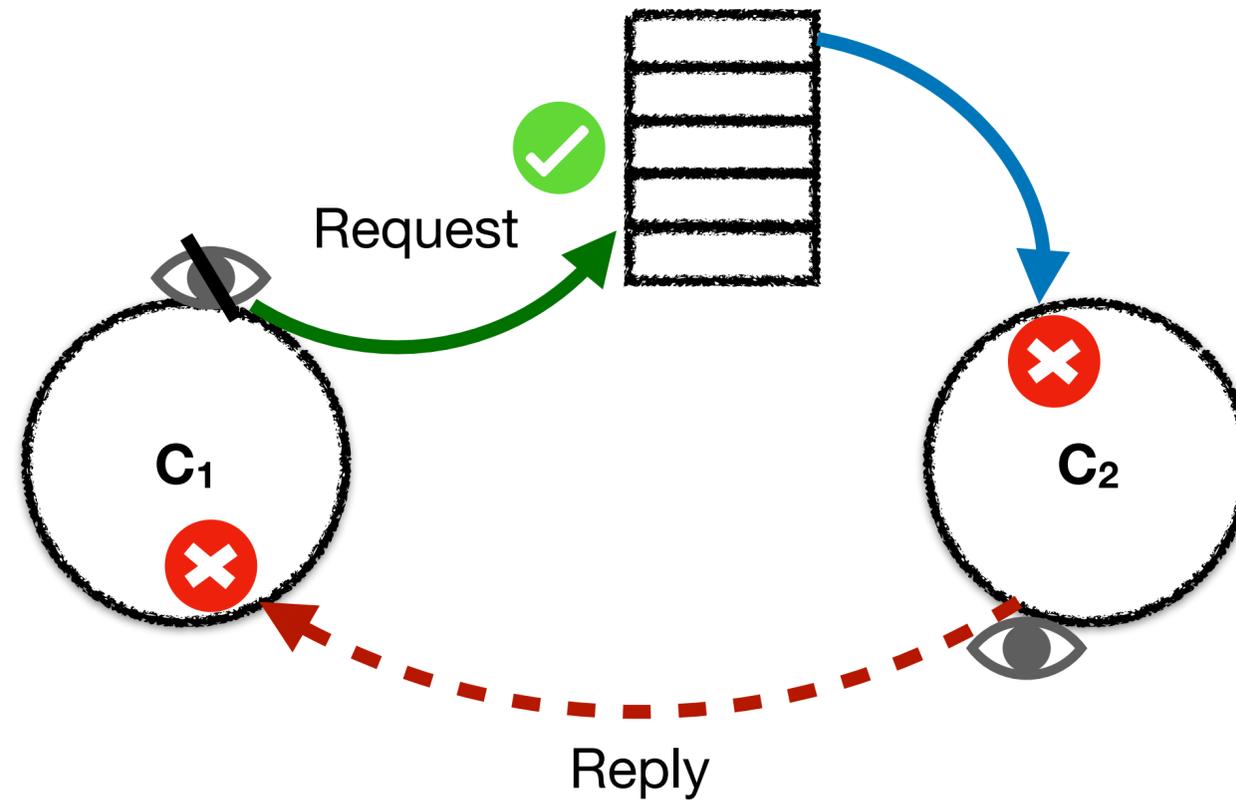
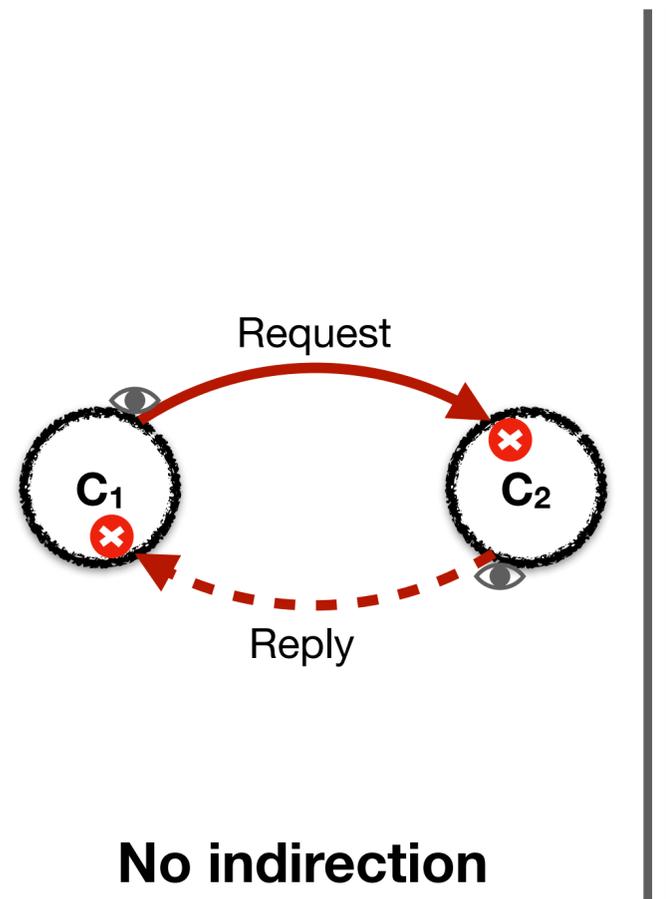


C2's failure is unobservable to C1

C1's failure is observable to C2

Interaction of two components C₁ and C₂

Observability Design Patterns

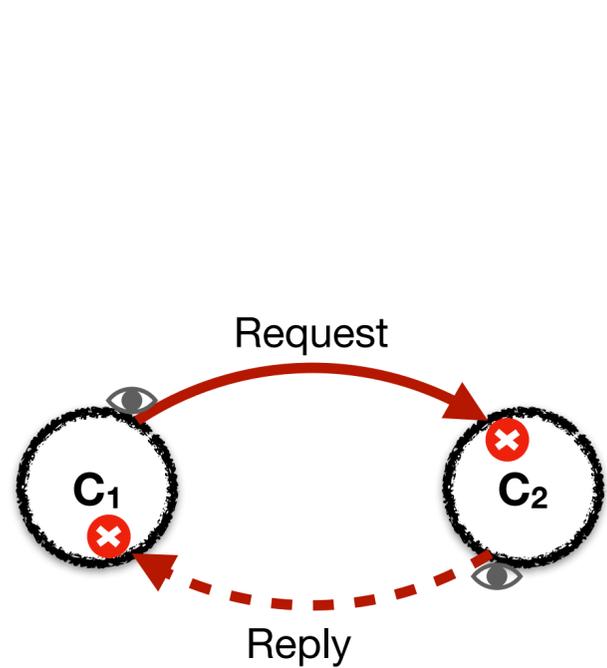


C2's failure is unobservable to C1

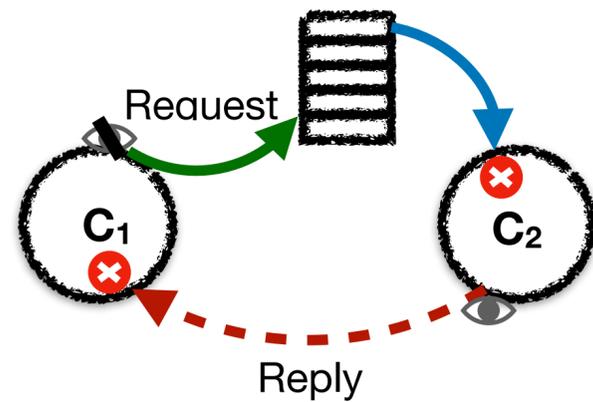
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Interaction of two components C₁ and C₂

Observability Design Patterns



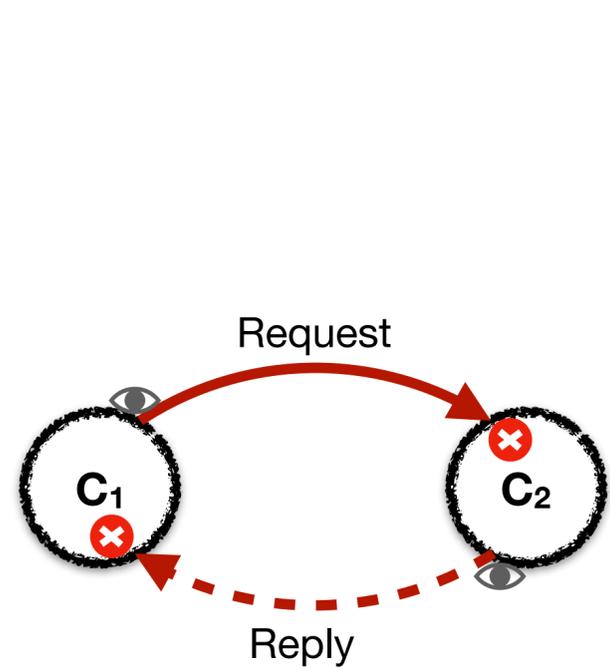
No indirection



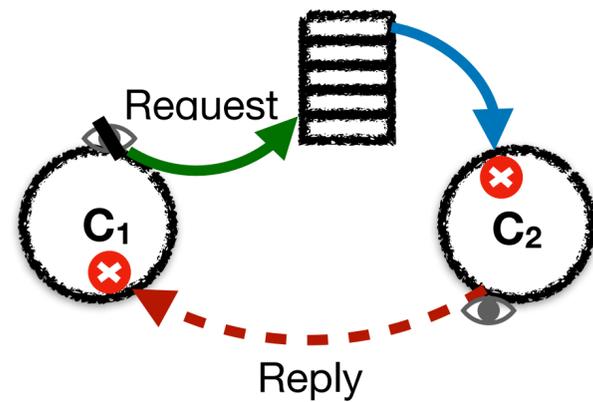
Request indirection

Interaction of two components C₁ and C₂

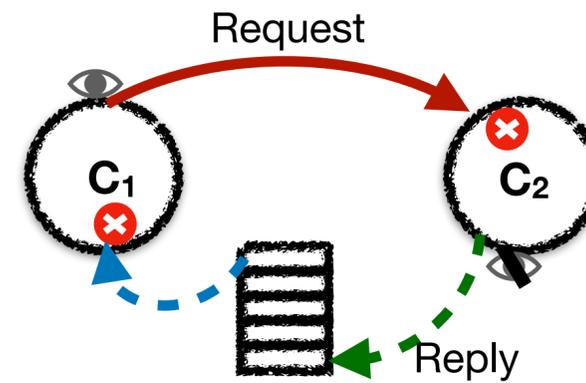
Observability Design Patterns



No indirection



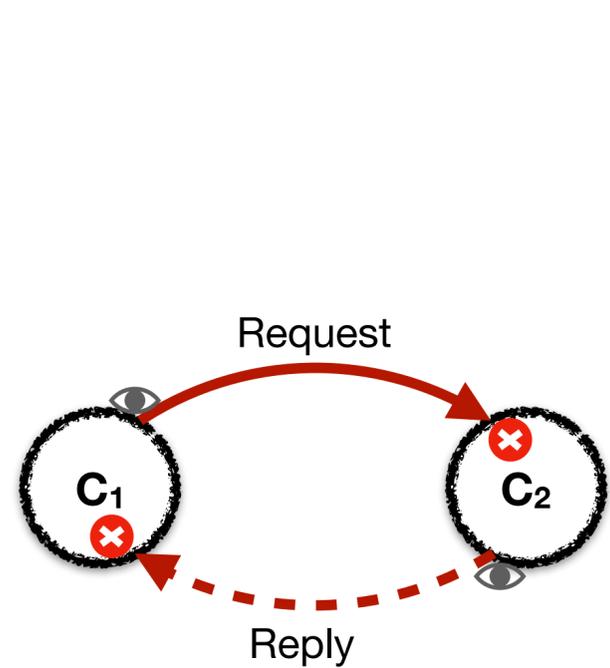
Request indirection



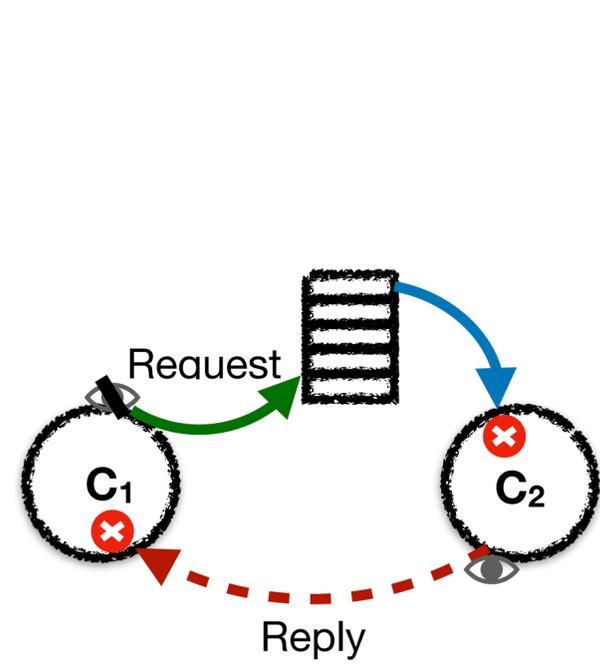
Reply indirection

Interaction of two components C₁ and C₂

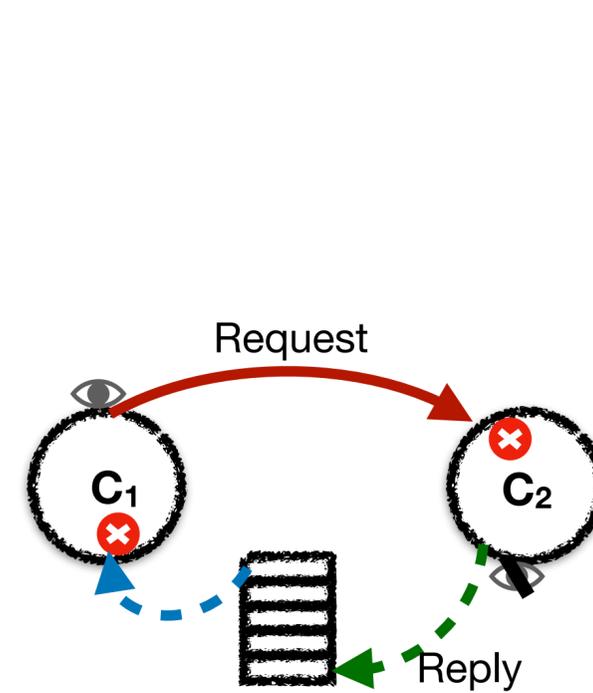
Observability Design Patterns



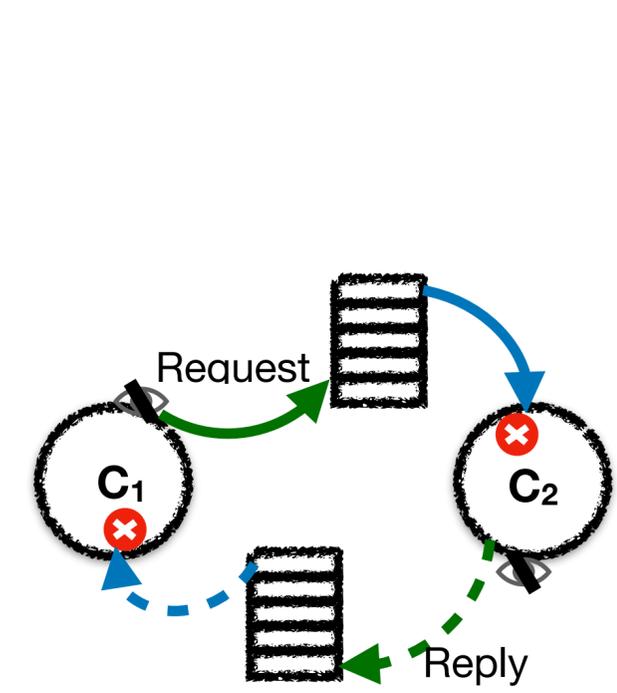
No indirection



Request indirection



Reply indirection



Full indirection

Interaction of two components C₁ and C₂

Capture Observation w/ Indirection

- Fundamental issue is that an observation becomes split
 - ▶ We call them ob-origin and ob-sink

positive, but
temporary and
weak evidence

complete and
strong evidence

Identify Ob-Origin and Ob-Sink

- Ob-origin is ob-boundary in asynchronous method
 - ▶ Submit a **PENDING** observation after ob-origin
- Ob-sink lies in notification mechanism
 - ▶ (1) callback set invocation; (2) blocking wait callback get; (3) checking a completion flag
 - ▶ Submit a **HEALTHY/UNHEALTHY** observation before ob-sink (1) or after ob-sink (2), (3)

Match Ob-Origin and Ob-Sink

- Match based on `<subject, context, [request_id]>`
- A later **HEALTHY** observation merges a past **PENDING** observation
 - the positive observation is now complete
- A later **UNHEALTHY** observation overrides a past **PENDING** observation
- If **PENDING** observations outstanding for too long, degrade to **UNHEALTHY**

Implementation

- **Panorama system implemented in Go + gRPC**
 - ▶ Easy to plug-in with different clients
- **A thin library provides async reporting, buffering, etc.**
 - ▶ Most reporting does not directly trigger local RPCs
- **Panorama analyzer implemented with Soot and AspectJ**

Evaluation

Analysis

Integration with
real-world
systems

1. Practicality and Effort

Detection Service

Catching
crash and
gray failures

2. Detection Time

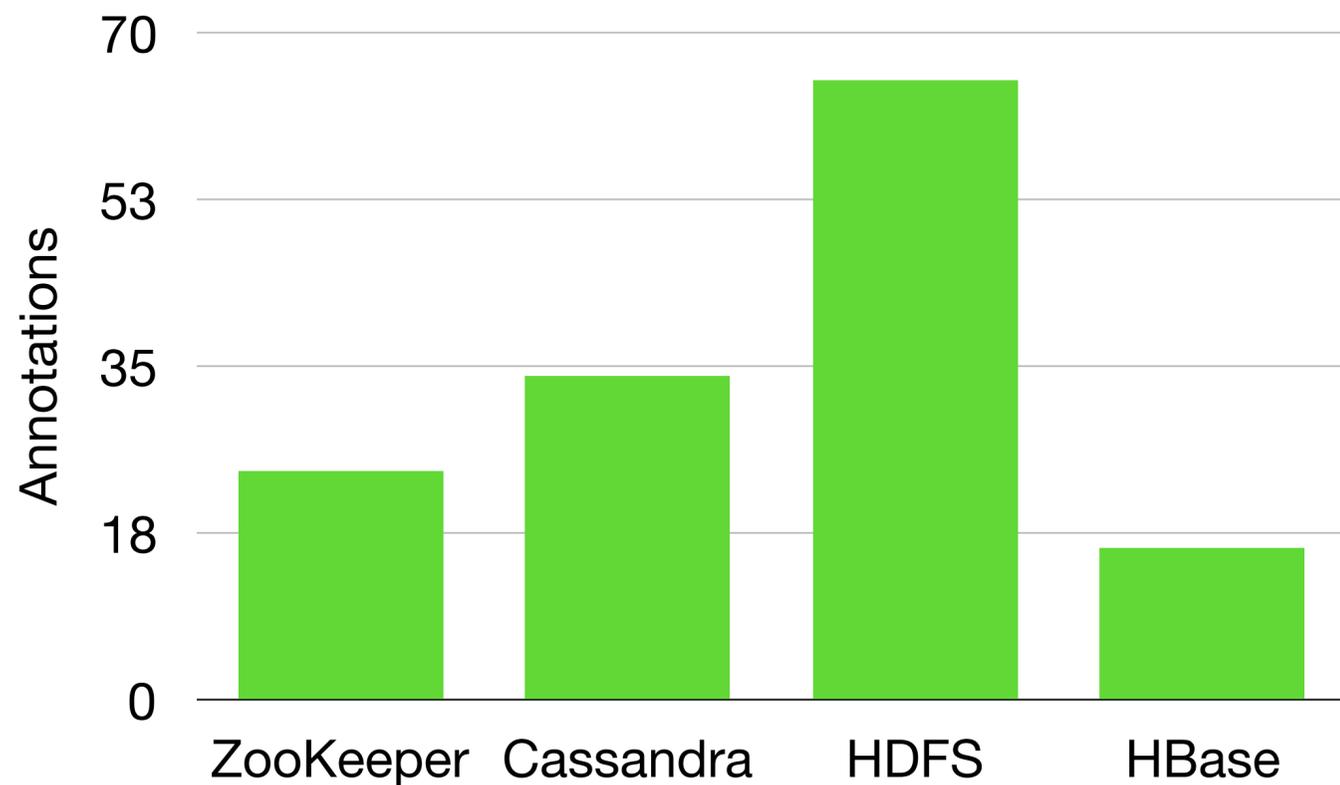
Reacting to
transient failures

3. Accuracy

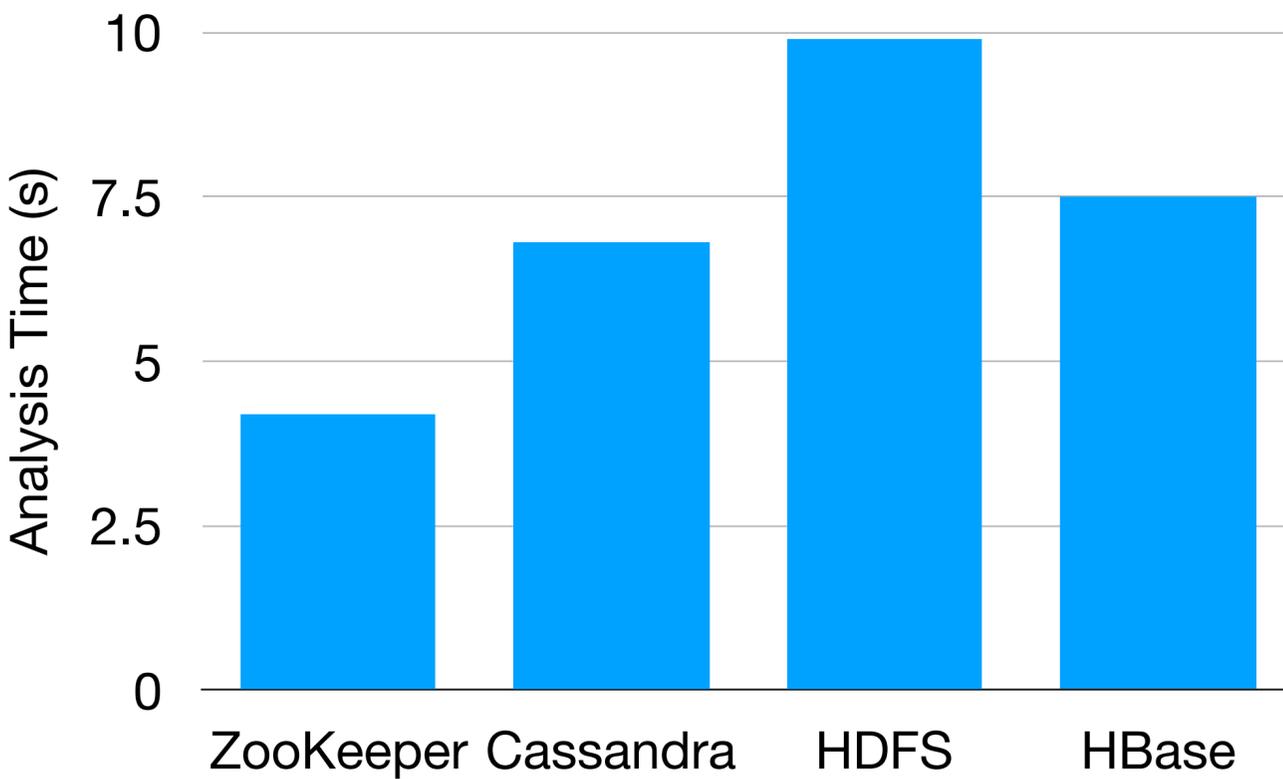
Cost for observers

4. Overhead

Integration with Real-world Systems

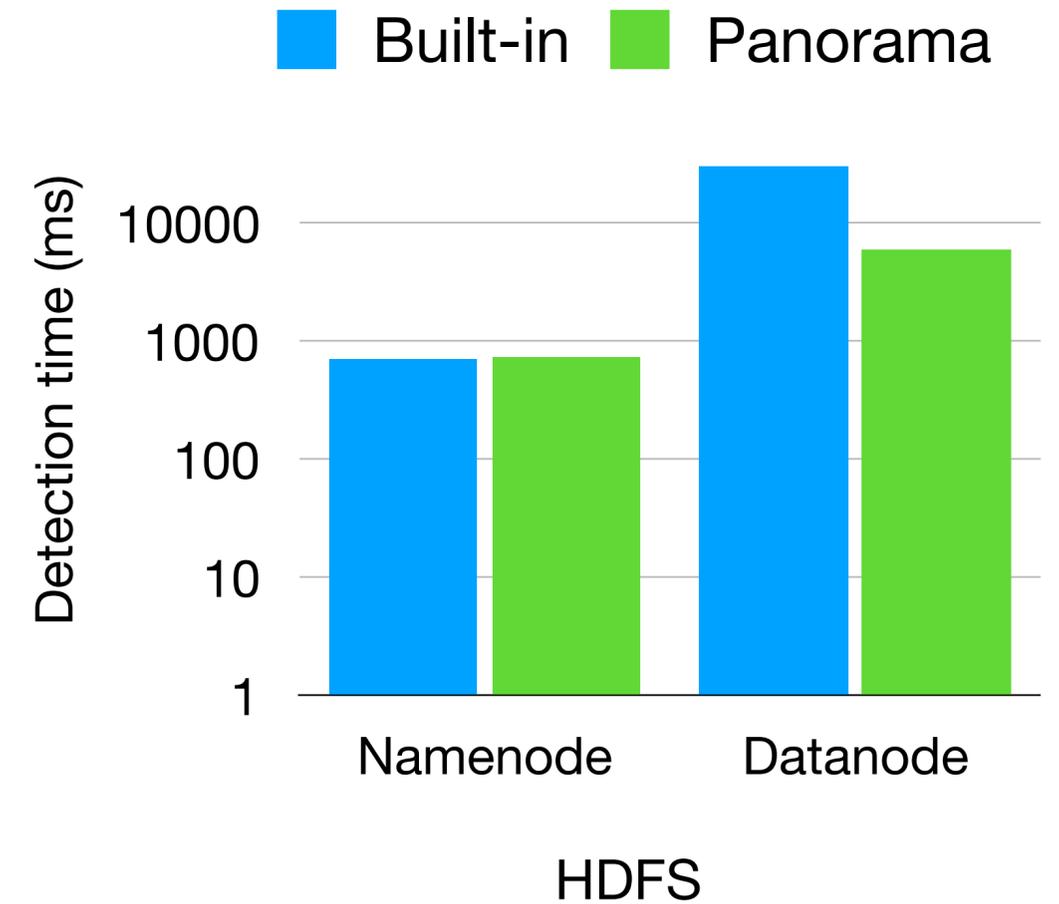
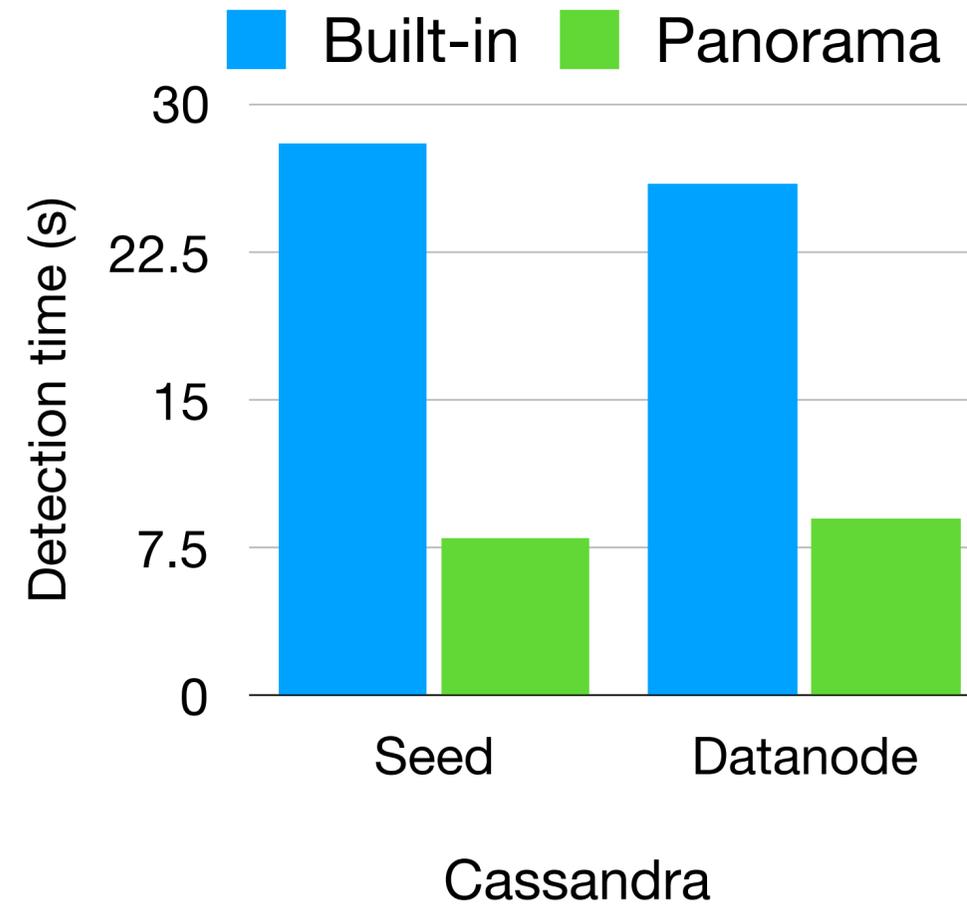
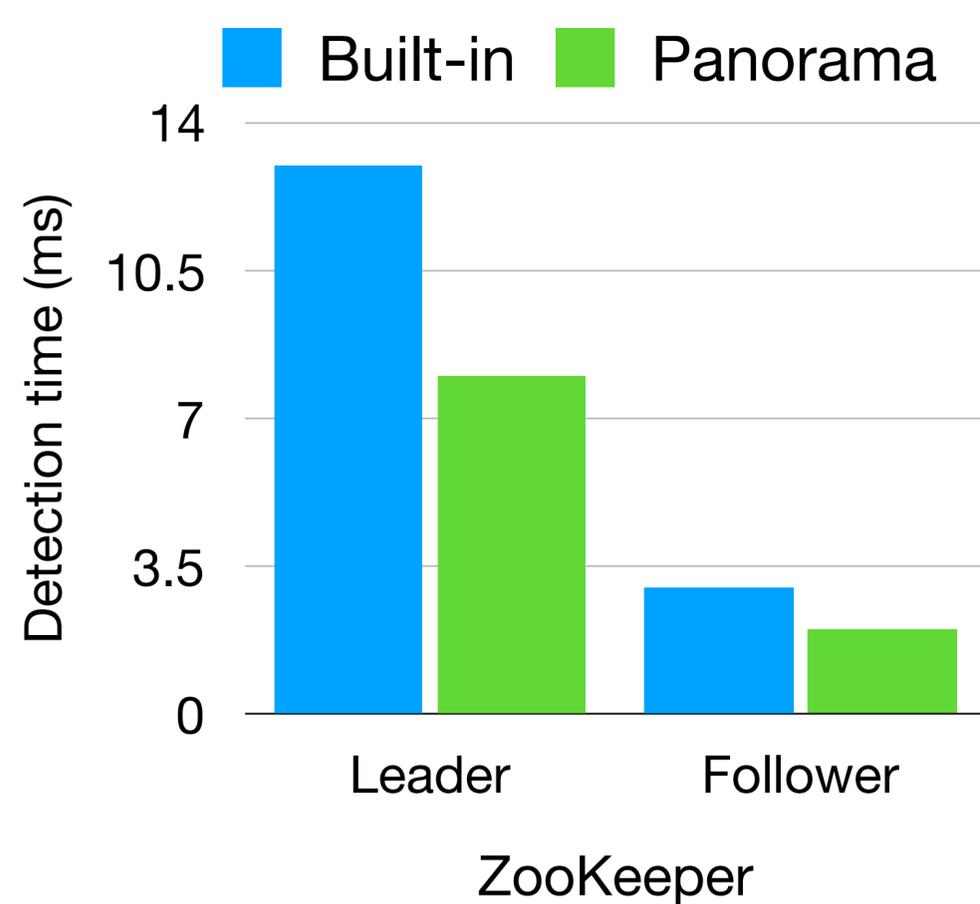


**Information about custom interfaces and async mechanisms.
Most of them are reusable across different versions.**



Analysis and instrumentation are fast.

Detect Injected Crash Failures



Built-in crash failure detectors have to be conservative to deal with asynchrony.

Real-World Gray Failures

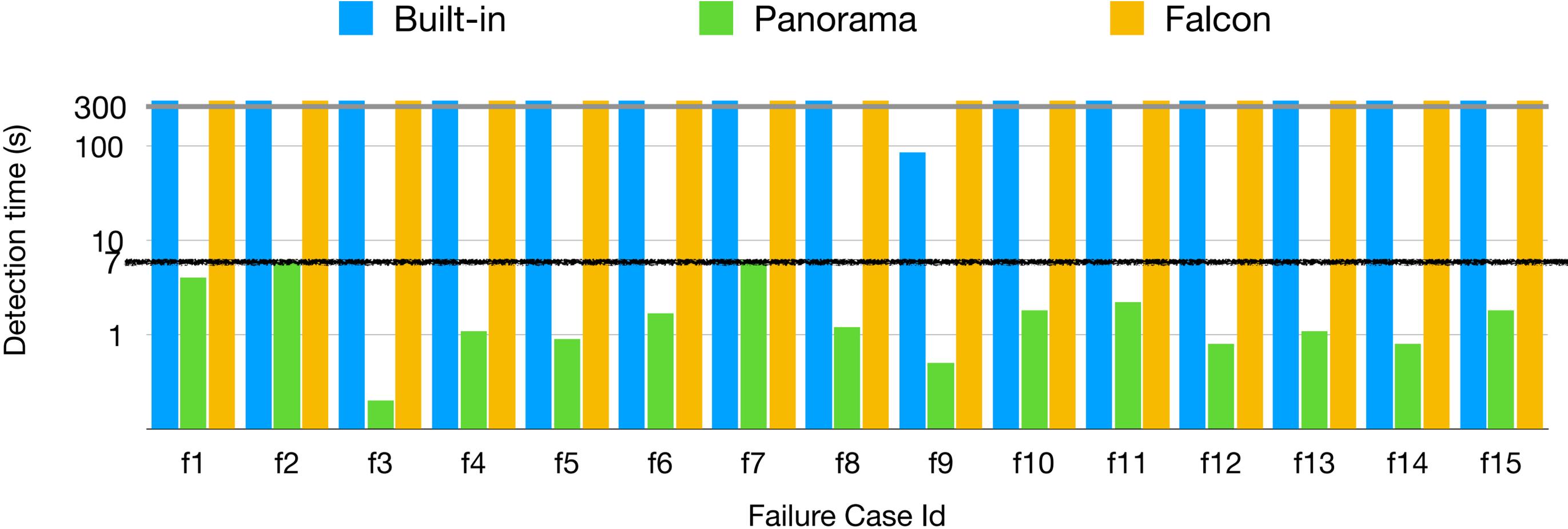
ID	System	Fault Synopsis
f ₁	ZooKeeper	faulty disk in leader causes cluster lock-up
f ₂	ZooKeeper	transient network partition leads to pro-longed failures in serving
f ₃	ZooKeeper	corrupted packet in de-serialization
f ₄	ZooKeeper	transaction thread exception
f ₅	ZooKeeper	leader fails to write transaction log
f ₆	Cassandra	response drop blocks repair operations
f ₇	Cassandra	stale data in leads to wrong node states
f ₈	Cassandra	streaming silently fail on unexpected error
f ₉	Cassandra	commitlog executor exit causes GC storm
f ₁₀	HDFS	thread pool exhaustion in master
f ₁₁	HDFS	failed pipeline creation prevents recovery
f ₁₂	HDFS	short circuit reads blocked due to death of domain socket
f ₁₃	HDFS	blockpool fails to initialize but continues
f ₁₄	HBase	dead root drive on region server
f ₁₅	HBase	replication stalls with empty WAL files

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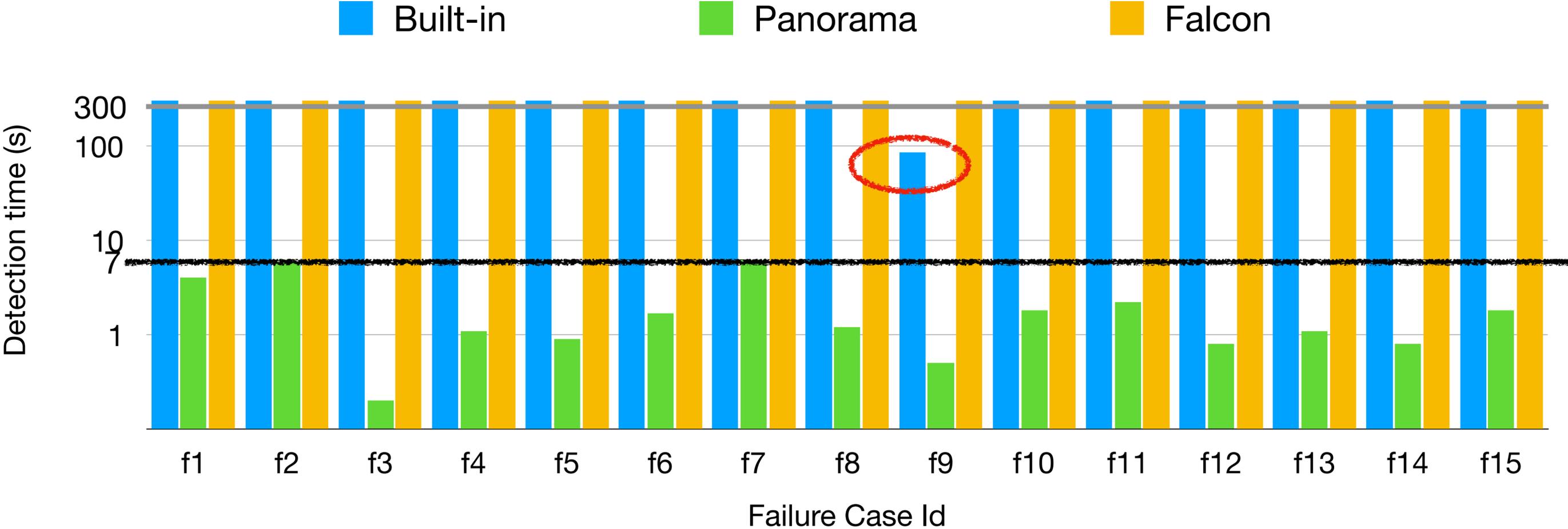
In all cases, severe service disruption occurred (e.g., create requests failed) while the failing component was perceived to be healthy.

Detection Time of Gray Failures



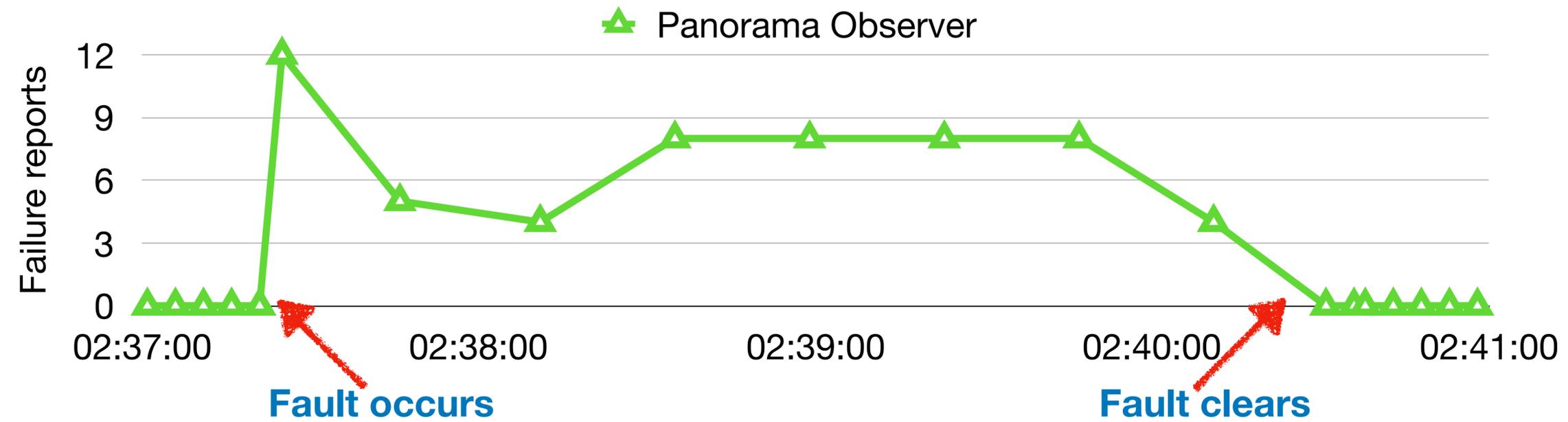
Panorama detects all 15 failures in under 7 seconds; Built-in detectors only detect one case within 300 seconds.

Detection Time of Gray Failures

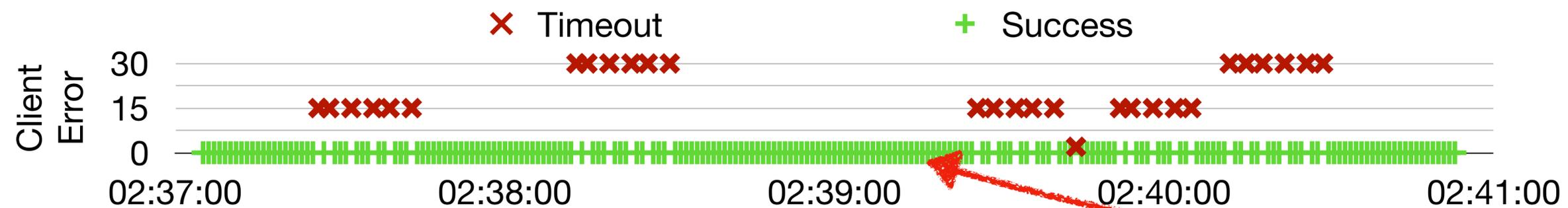
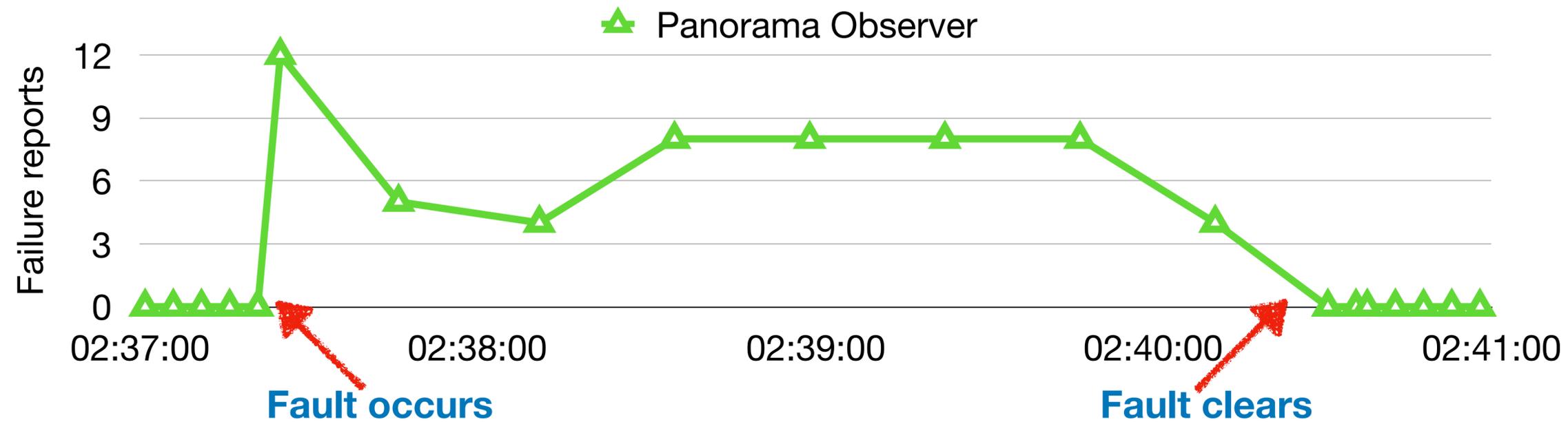


Panorama detects all 15 failures in under 7 seconds; Built-in detectors only detect one case within 300 seconds.

Timeline in Detecting Gray Failure f_1

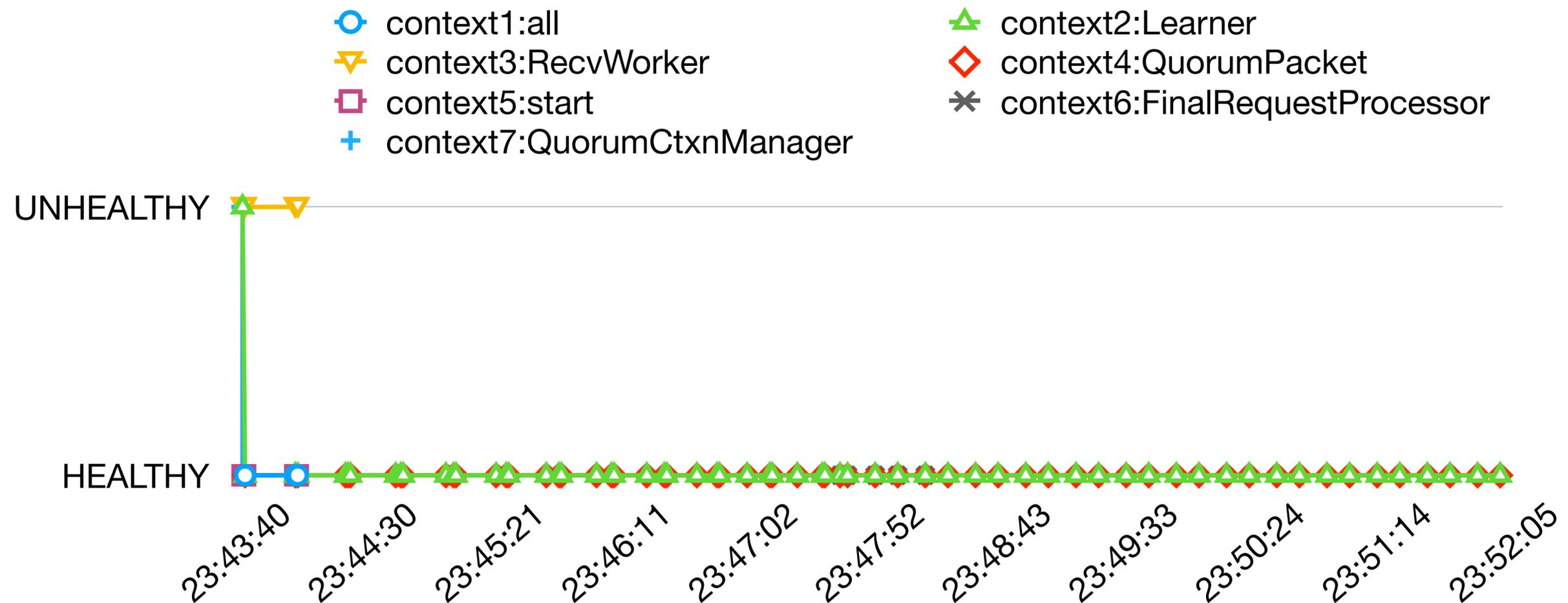


Timeline in Detecting Gray Failure f_1



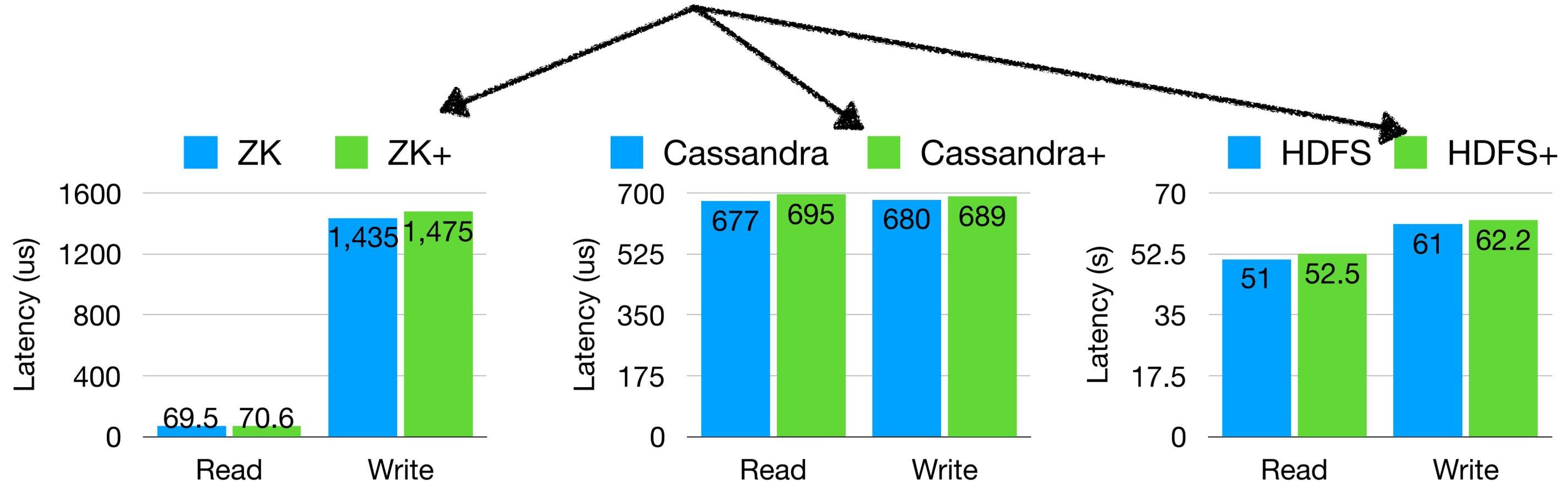
Client executing mixed workloads

Transient Failures



Latency Overhead to Observers

System instrumented with Panorama reporting hooks



Less than 3% latency and throughput (not shown) overhead

Limitations & Future Work

- **Panorama relies on dependencies and interactions in large systems**
 - ▶ if observability is inherently low, have to improve the built-in FDs.
- **Panorama currently focuses on failure detection**
 - ▶ need to integrate detection results to the subject to take actions
- **Panorama currently does not identify root cause of observations**
 - ▶ useful for localizing failing component in cascading failures

Related Work

- Failure Detection

- ▶ Gossip [Middleware '98], ϕ [SRDS '04], Falcon [SOSP '11], Pigeon [NSDI '13]

- Monitoring and Tracing

- ▶ Magpie [OSDI '04], X-Trace [NSDI '07], Dapper [Google TR '10], Pivot Tracing [SOSP '15]

- Accountability

- ▶ PeerReview [SOSP '07]

Panorama's major contribution: a new way of constructing failure detectors for gray failures

- Gray Failures

- ▶ Fail-Stutter [HotOS '01], Limplock [SoCC '13], Fail-Slow Hardware [FAST '18], Gray Failure [HotOS '17]

Conclusion

Detect what the “requesters” see

- Failures that matter are observable to requesters
 - Turn **error *handlers*** into **error *reporters***
 - But must deal with design patterns that reduce observability
- Panorama enables construction of *in-situ* observers
 - Detects 8 crash failure and 15 gray failures in 4 real-world systems faster



<https://github.com/ryanphuang/panorama>

Backup Slides

Performance of Panorama Core API

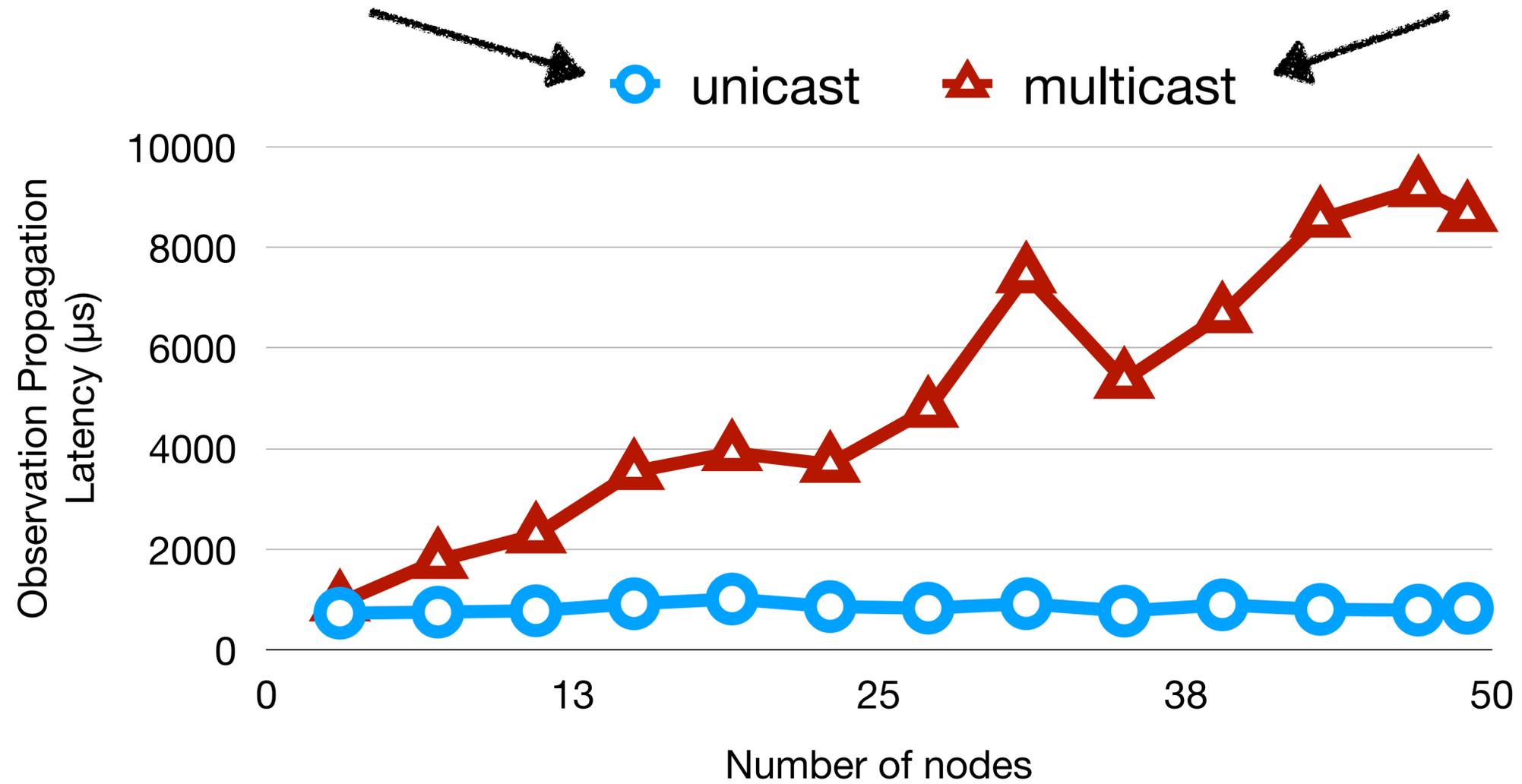
Local RPC	Library Call		RPC
Report	ReportAsync	Judge	Propagate
114.6 μ s	0.36 μ s	109.0 μ s	776.3 μ s

**Main overhead perceived
by the *in-situ* observers**

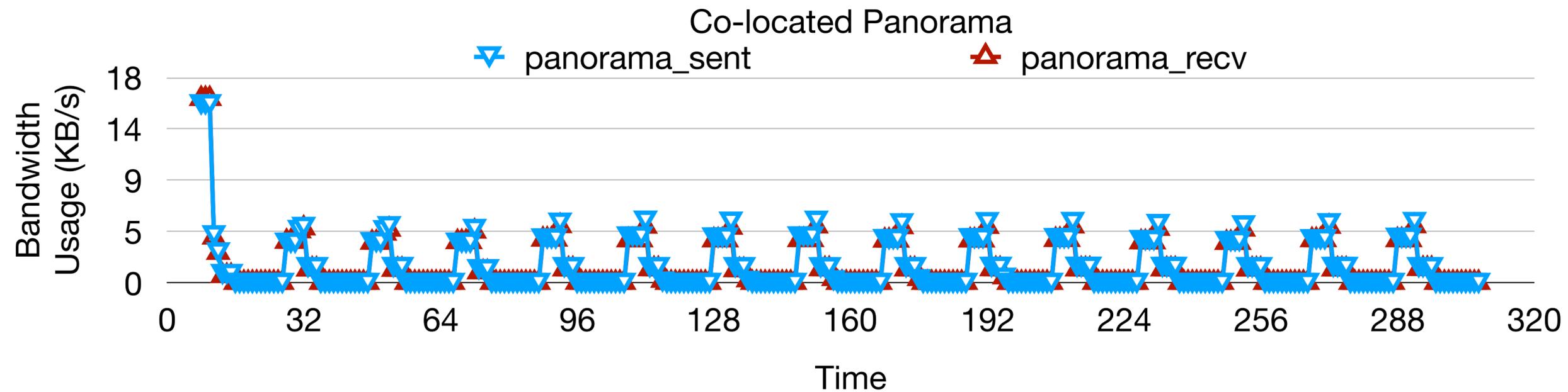
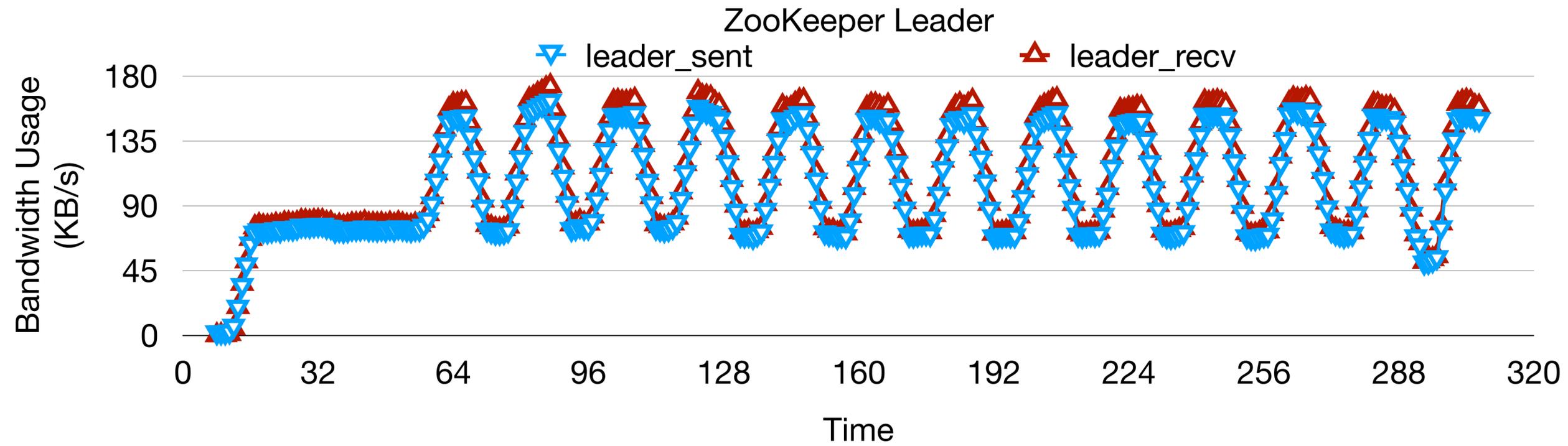
Scalability

Exchange an observation with a single Panorama instance

Exchange an observation with a clique of Panorama instance

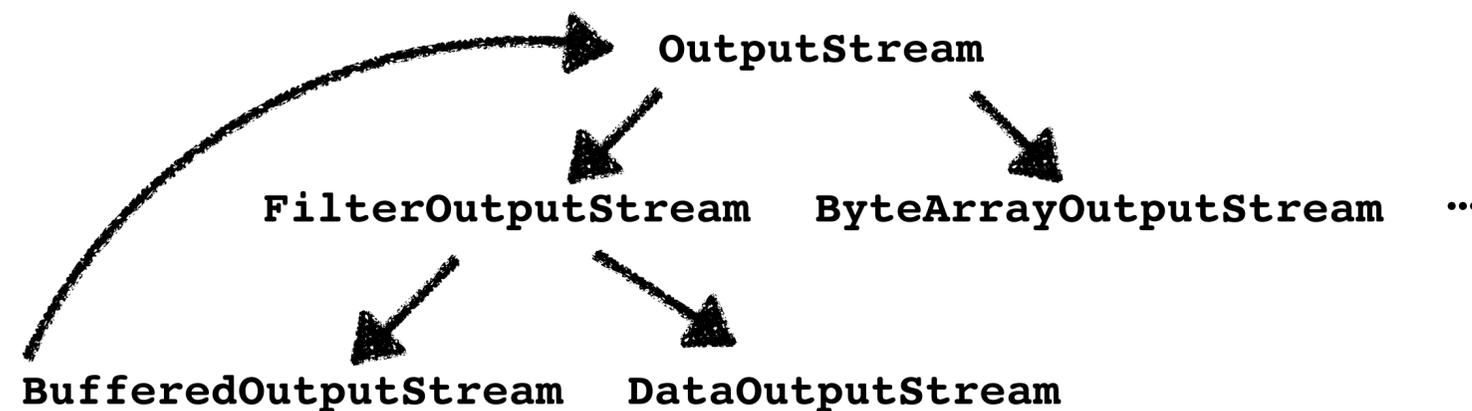


Network Bandwidth Usage

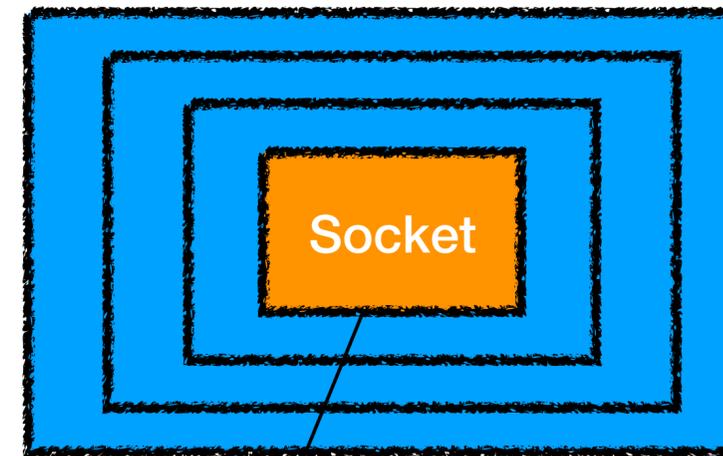


Challenge: Polymorphism and Interface

- May not distinguish ob-boundary and normal invocations
 - ▶ e.g., `void java.io.OutputStream.write(byte[] b)` could be local write or remote write



```
if (strm instanceof BufferedOutputStream) {
    ???
}
```



```
public OutputStream Socket.getOutputStream()
```

Challenge: Polymorphism and Interface

- May not distinguish ob-boundary and normal invocations
- Interface does not contain enough information (lack of fields)

```
/**  
 * A BlockReader is responsible for reading a single block from a single datanode.  
 */  
public interface BlockReader {  
    int read(byte[] buf, int off, int len) throws IOException;  
    long skip(long n) throws IOException;  
    int available() throws IOException;  
}
```

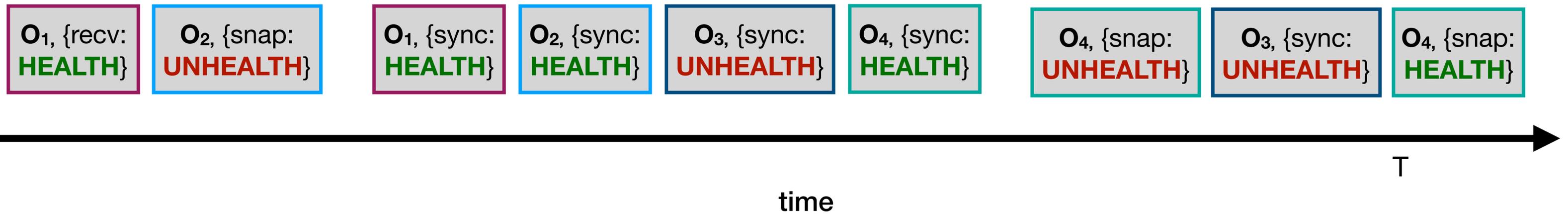
Solution

- **Change the constructors of subclasses for the abstract type**
 - ▶ return a compatible wrapper with additional subject identity field
 - ▶ set the field for remote types (e.g., `java.net.Socket.getOutputStream()`)
 - ▶ check this field at runtime to distinguish ob-boundary from normal types
- **Extend the interface with additional subject related methods**
 - ▶ call setter in the implementors (e.g., `BlockReaderFactory`)

Design Challenges

- Observations dispersed in software's source code
- Observation collection may slow down observer's normal functionality
- Diverse real-world programming paradigms affect system observability
- **Observations may have biases**

Reach Verdict from Observations



Reach Verdict from Observations

Verdict policy #1: latest observation is the verdict

- ▶ Miss intermittent failure



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time

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Reach Verdict from Observations

Verdict policy #1: latest observation is the verdict

- ▶ Miss intermittent failure

Verdict policy #2: unhealthy if *any* recent observation is unhealthy

- ▶ One biased observer mislead others



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time

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Reach Verdict from Observations

Our solution: a simple bounded look-back majority algorithm

- ▶ Group by context and observer, summarize and vote; see paper for more detail

Verdict policy #1: latest observation is the verdict

- ▶ Miss intermittent failure

Verdict policy #2: unhealthy if *any* recent observation is unhealthy

- ▶ One biased observer mislead others



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