

## Alerting for Distributed Systems A Tale of Symptoms and Causes, Signals and Noise

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O(100) engineers ~5% is ProdEng



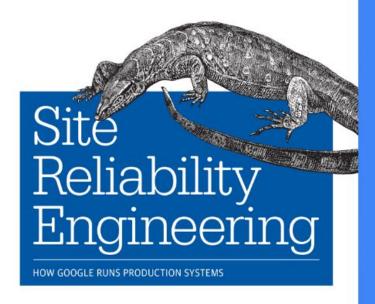
"You build it, you run it." "True DevOps" "NoOps"

O(10k) engineers

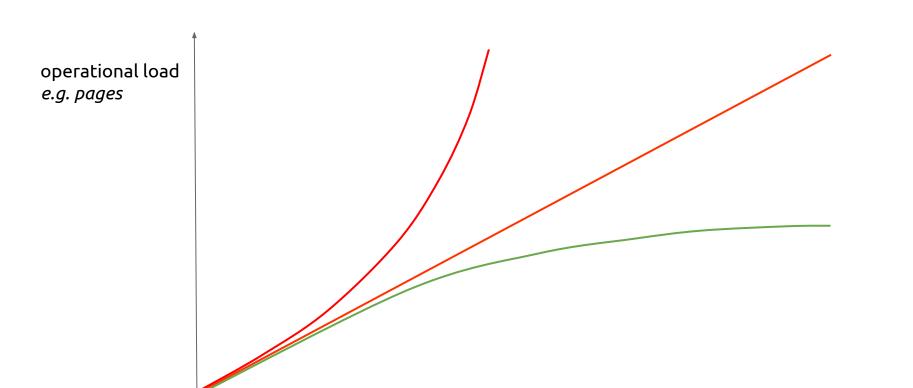
~5% is SRE

SRE "by the book"

#### O'REILLY"



Edited by Betsy Beyer, Chris Jones, Jennifer Petoff & Niall Murphy



traffic \* complexity

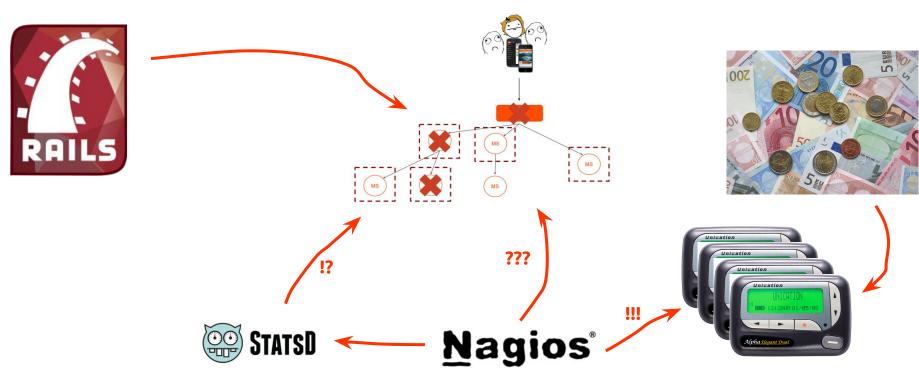
Our SRE organization has an advertised goal of keeping operational work (i.e., toil) below 50% of each SRE's time. At least 50% of each SRE's time should be spent on engineering project work that will either reduce future toil or add service features. [...] We share this 50% goal because toil tends to expand if left unchecked and can quickly fill 100% of everyone's time.

Chapter 5: Eliminating Toil



## SoundCloud's trajectory 3 years ago

All started with a healthy growth in traffic and features...



## The three kinds of "alerts"

SRE book calls them *monitoring output*.

Alerts is Prometheus terminology.

Expected response	SRE book	SoundCloud lingo	Delivered to
Act immediately	Alerts	Pages	Pager
Act eventually	Tickets	Tickets / "email alerts"	Issue tracker / email :-(
None (for diagnostics only)	Logs	Informational alerts	Nowhere / dashboards



## Sense of urgency and action

Every time the pager goes off, I should be able to react with a sense of urgency. I can only react with a sense of urgency a few times a day before I become fatigued.

Every page should be *actionable*.

Chapter 6: Monitoring Distributed Systems

*True story:* One day, SoundCloud was down, and a single page fired...

"The outage was so bad, more pages should have fired." (From a SC Postmortem)



## Symptoms vs. causes

How to make pages more meaningful?

"What" versus "why" is one of the most important distinctions in writing good monitoring with maximum signal and minimum noise.

Chapter 6: Monitoring Distributed Systems

Symptom	Cause	
I'm serving HTTP 500s or 404s	Database servers are refusing connections	
My responses are slow	CPUs are overloaded by a bogosort, or an Ethernet cable is crimped under a rack, visible as partial packet loss	
Users in Antarctica aren't receiving animated cat GIFs	Your Content Distribution Network hates scientists and felines, and thus blacklisted some client IPs	
Private content is world-readable	A new software push caused ACLs to be forgotten and allowed all requests	

#### Causes and symptoms are loosely bound in distributed systems.

At the scale our systems operate, being alerted for *single-machine failures* is unacceptable because such data is *too noisy to be actionable*.

Chapter 10: Practical Alerting from Time-Series Data

What was thought to be good signals for problems might just be noise today (or worse, you can't say if it is noise or not):

- A machine is down. *Happens all the time*.
- Load average is high. *Really?*
- My network uplink / CPUs / disk / RAM ... are fully utilized. *Good or bad?*



## Black-box vs. white-box

Black-box is just perfect for symptom-based alerting, isn't it?

We combine heavy use of white-box monitoring with modest but critical uses of black-box monitoring. The simplest way to think about black-box monitoring versus white-box monitoring is that black-box monitoring is symptom-oriented and represents active—not predicted—problems. [...]

For paging, black-box monitoring has the key benefit of forcing discipline to only nag a human when a problem is both *already ongoing and contributing to real symptoms*. On the other hand, for *not-yet-occurring but imminent* problems, black-box monitoring is *fairly useless*.

Chapter 6: Monitoring Distributed Systems



#### Pros & cons

#### Black-box:

- End-to-end test "as the user sees it".
- Probes may be different from current user traffic.
- Tail latency and rare failures only visible over a long time.

#### White-box:

- Reported latency serving the frontend might be a lie, but reported latency of requests to the backend is "live-traffic probing".
- Must resist temptation to alert on countless internal details.
- Indispensable to detect imminent problems and to investigate causes.



## Imminent problems

White-box and time-series based monitoring FTW.

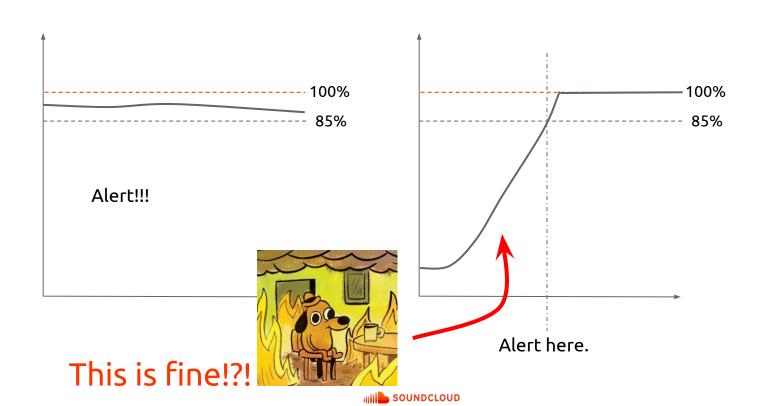
- Loss of redundancy (going from N+1 to N+0).
- More complex reasoning based on insights into a system.
- "Nearly full" scenarios.

[...] the idea of treating time-series data as a data source for generating alerts is now accessible to everyone through those open source tools like Prometheus, Riemann, Heka, and Bosun [...]

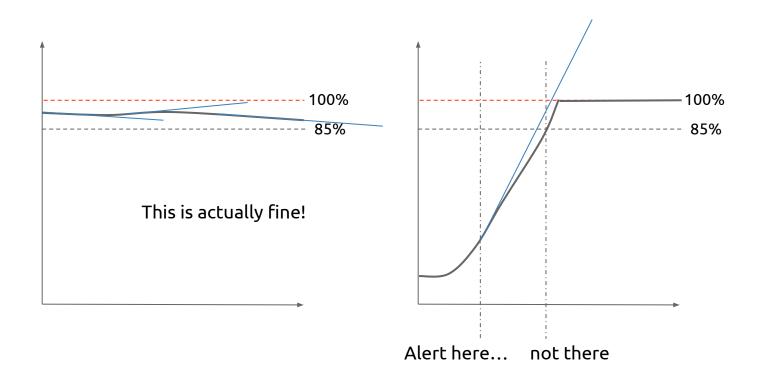
Chapter 10: Practical Alerting from Time-Series Data



### Static disk-full alert (e.g. Nagios)



#### Time-series based disk-full alert (e.g. Prometheus)





## Causes are important, too.

*True story:* Google's stats processing pipeline.

Informational alerts and sometimes tickets are great for causes.



## Symptom or cause?

Note that in a multilayered system, one person's symptom is another person's cause.

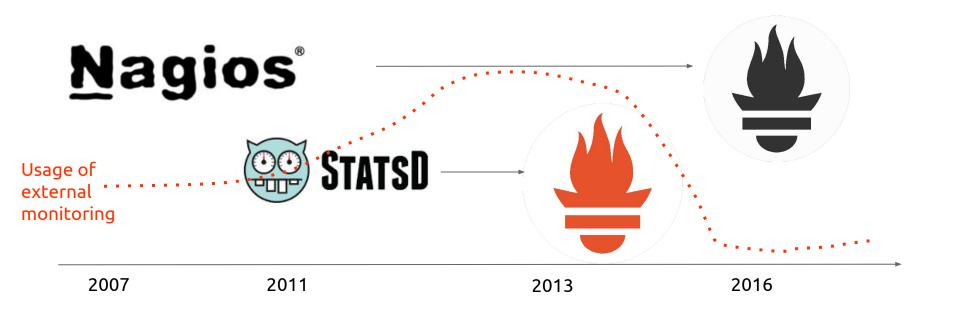
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To achieve the decoupling desired in a microservice architecture, teams become users of each other (in addition to the "real" user in the big picture).

Tuesday, July 12 • 16:00 - 16:20







We need monitoring systems that allow us to alert for high-level service objectives, but retain the granularity to inspect individual components as needed.

Chapter 10: Practical Alerting from Time-Series Data



## But what about anomaly detection?

Neither symptom nor cause.

In general, Google has trended toward *simpler and faster* monitoring systems, with better tools for post hoc analysis. We avoid "*magic*" systems that try to *learn thresholds or automatically detect causality*. [...]

Similarly, to keep noise low and signal high, the elements of your monitoring system that direct to a pager need to be *very simple and robust*. Rules that generate alerts for humans should be simple to *understand* and represent a *clear failure*.

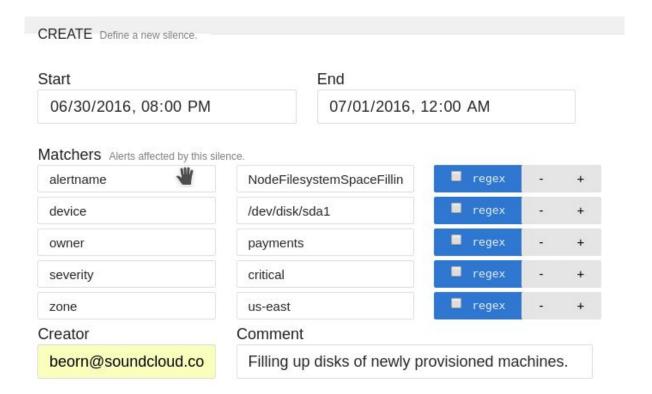
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Anomaly detection for pages should be simple and robust.

More complex systems can be great under circumstances, but not for pages.



## Silencing for humans





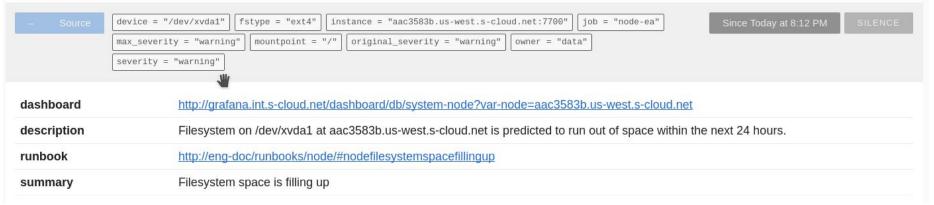
## Runbooks and robotic responses

Google SRE relies on on-call playbooks, in addition to exercises such as the "Wheel of Misfortune," to prepare engineers to react to on-call events.

Chapter 1: Introduction

Every page response should require intelligence. If a page merely merits a robotic response, it shouldn't be a page.

#### Chapter 6: Monitoring Distributed Systems



## Perfectly self-healing systems?

Caveats of automation

Being on-call for a quiet system is blissful, but what happens if the system is too quiet or when SREs are not on-call often enough? *An operational underload is undesirable for an SRE team.* 

Chapter 11: Being On-Call

Do gamedays, DiRT, "Wheel of Misfortune", whatever you call it...



## The End

# May the queries flow, and the pager stay silent.

Chapter 10: Practical Alerting from Time-Series Data



## **Bonus Slides**

#### Muting for machines.

Google SRE has experienced only limited success with complex dependency hierarchies. [...] Dependency-reliant rules usually pertain to very stable parts of our system, such as our system for draining user traffic away from a datacenter. For example, "If a datacenter is drained, then don't alert me on its latency" is one common datacenter alerting rule. Few teams at Google maintain complex dependency hierarchies because our infrastructure has a steady rate of continuous refactoring.

Chapter 6: Monitoring Distributed Systems



## Alert grouping

Thousands of nodes suddenly cried out in terror...

Noisy alerts that systematically generate more than one alert per incident should be tweaked to approach a 1:1 alert/incident ratio. Doing so allows the on-call engineer to focus on the incident instead of triaging duplicate alerts.

Chapter 11: Being On-Call





#### PrometheusOutOfOrderSamplesDiscarded

#### Meaning

Prometheus is an append-only time series database. If you try to insert samples into a time series with a time stamp older than the most recent sample in that series, or with a time stamp equal to the time stamp of the most recent sample but with a different value, the insert will be discarded and this alert will fire.

#### Diagnosis

The logs will usually only tell you about the number of discarded samples. If you run version 0.19 or above, the prometheus\_local\_storage\_out\_of\_order\_samples\_total metric will have a reason label. Also, by activating DEBUG level logging, you can see the exact dropped samples in the logfile.

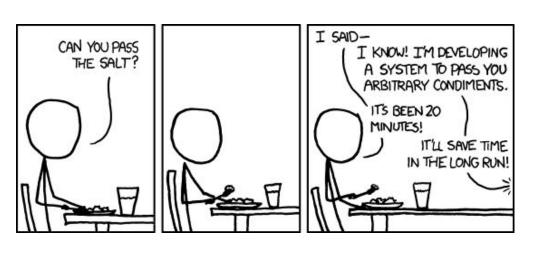
If you see this alert on prometheus-mysql, the reason is probably this bug O.

#### Resolving

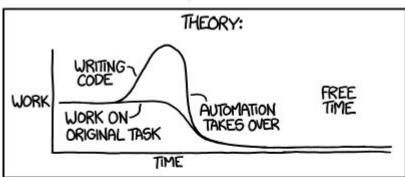
Out-of-order samples are usually created by configuration errors. Multiple rules funnel into the same time series, or a broken federation setup results in the same. Another suspect to look for is a target that exports explicit time stamps.

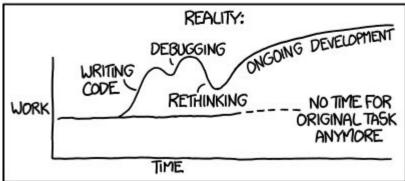


#### Automation might not pay off.



"I SPEND A LOT OF TIME ON THIS TASK.
I SHOULD WRITE A PROGRAM AUTOMATING IT!"

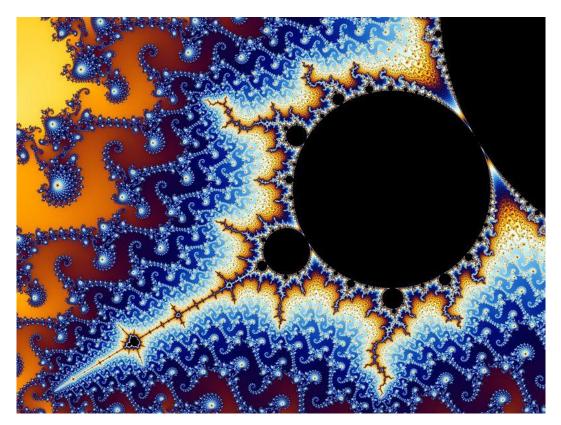




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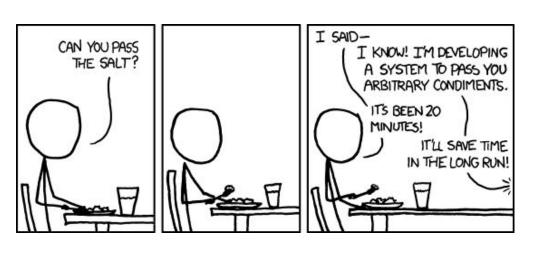


#### Automation introduces a feedback loop...

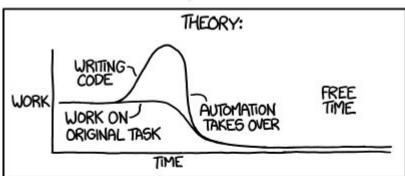


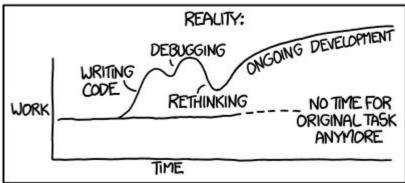
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