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Web Performance  
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CONFERENCE



Santa Clara, CA



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[velocityconf.com](http://velocityconf.com)

#velocityconf

# Stop the Guessing

## Performance Methodologies for Production Systems

Brendan Gregg

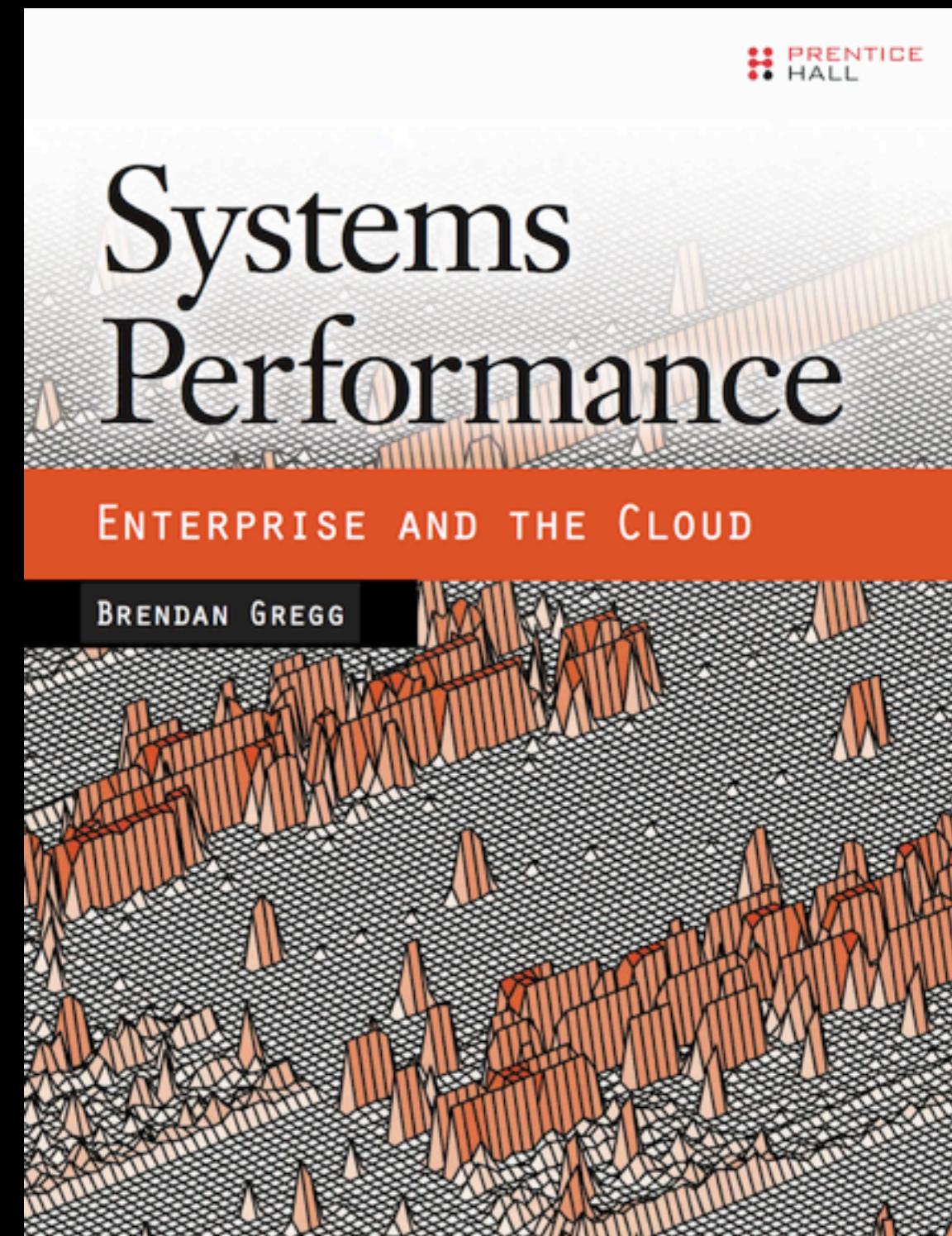
Lead Performance Engineer, Joyent

# Audience

- This is for developers, support, DBAs, sysadmins
- When perf isn't your day job, but you want to:
  - Fix common performance issues, quickly
  - Have guidance for using performance monitoring tools
- Environments with small to large scale production systems

# whoami

- Lead Performance Engineer: analyze everything from apps to metal
- Work/Research: tools, visualizations, methodologies
- Methodologies is the focus of my next book





- High-Performance Cloud Infrastructure
  - Public/private cloud provider
- OS Virtualization for bare metal performance
- KVM for Linux and Windows guests
- Core developers of SmartOS and node.js



# Performance Analysis

- Where do I start?
- Then what do I do?

# Performance Methodologies

- Provide
  - Beginners: a starting point
  - Casual users: a checklist
  - Guidance for using existing tools: pose questions to ask
- The following six are for production system monitoring

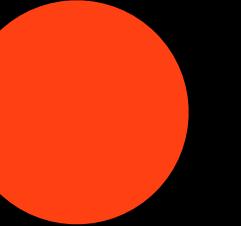
# Production System Monitoring

- Guessing Methodologies
  - 1. Traffic Light Anti-Method
  - 2. Average Anti-Method
  - 3. Concentration Game Anti-Method
- Not Guessing Methodologies
  - 4. Workload Characterization Method
  - 5. USE Method
  - 6. Thread State Analysis Method

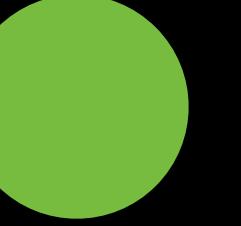
# Traffic Light Anti-Method

# Traffic Light Anti-Method

- 1. Open monitoring dashboard
- 2. All green? Everything good, mate.



= **BAD**



= **GOOD**

# Traffic Light Anti-Method, cont.

- Performance is subjective
  - Depends on environment, requirements
  - No universal thresholds for good/bad
- Latency outlier example:
  - customer A) 200 ms is bad
  - customer B) 2 ms is bad (an “eternity”)
- Developer may have chosen thresholds by guessing

# Traffic Light Anti-Method, cont.

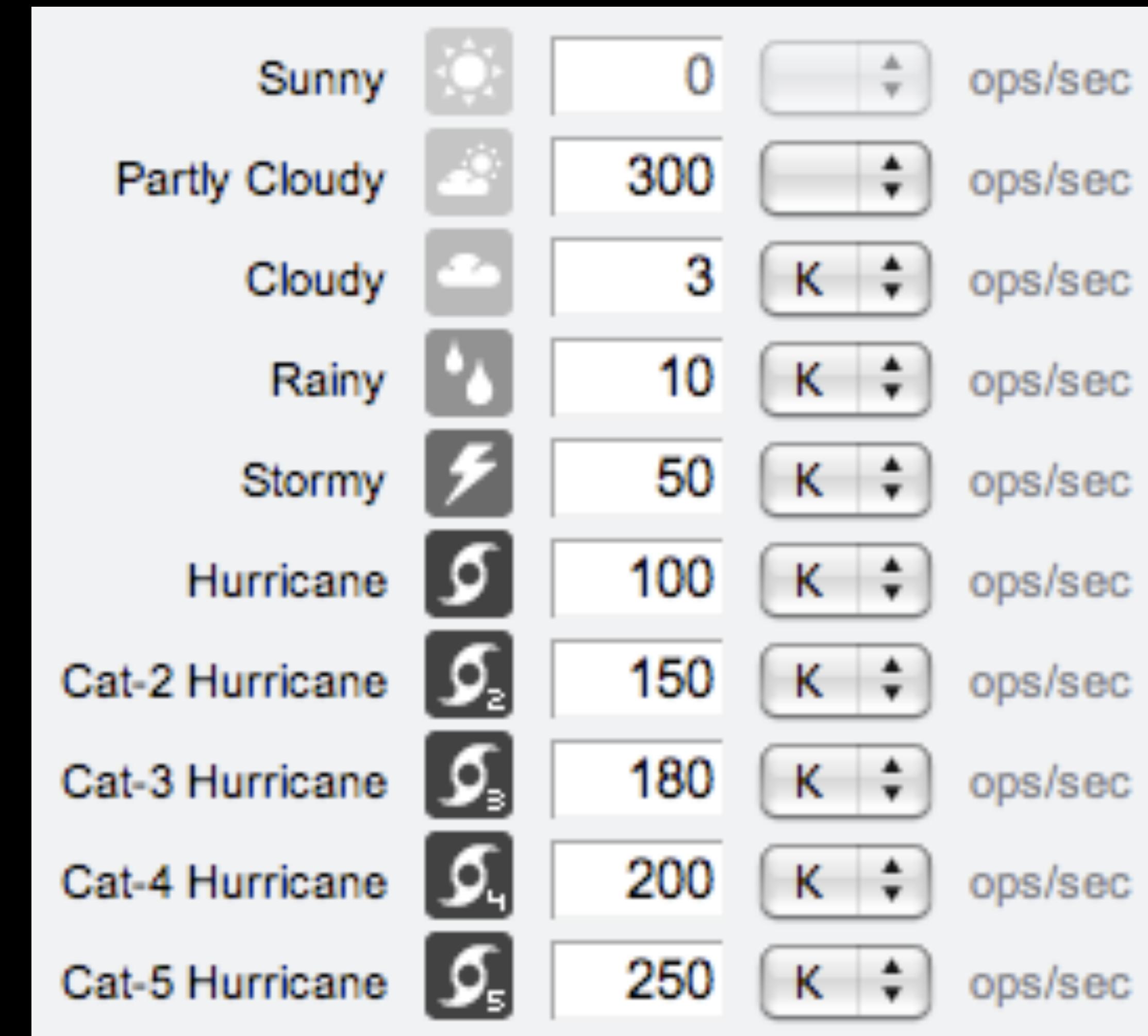
- Performance is complex
  - Not just one threshold required, but multiple different tests
- For example, a disk traffic light:
  - Utilization-based: one disk at 100% for less than 2 seconds means **green** (variance), for more than 2 seconds is **red** (outliers or imbalance), but if all disks are at 100% for more than 2 seconds, that may be **green** (FS flush) provided it is async write I/O, if sync then **red**, also if their IOPS is less than 10 each (errors), that's **red** (sloth disks), unless those I/O are actually huge, say, 1 Mbyte each or larger, as that can be **green**, ... etc ...
  - Latency-based: I/O more than 100 ms means **red**, except for async writes which are **green**, but slowish I/O more than 20 ms can **red** in combination, unless they are more than 1 Mbyte each as that can be **green** ...

# Traffic Light Anti-Method, cont.

- Types of error:
  - I. False positive: **red** instead of **green**
    - Team wastes time
  - II. False negative: **green** instead of **red**
    - Performance issues remain undiagnosed
    - Team wastes *more* time looking elsewhere

# Traffic Light Anti-Method, cont.

- Subjective metrics (opinion):
  - utilization, IOPS, latency
- Objective metrics (fact):
  - errors, alerts, SLAs
- For subjective metrics, use weather icons
  - implies an inexact science, with no hard guarantees
  - also attention grabbing
- A dashboard can use both as appropriate for the metric



# Traffic Light Anti-Method, cont.

- Pros:

- Intuitive, attention grabbing
- Quick (initially)

- Cons:

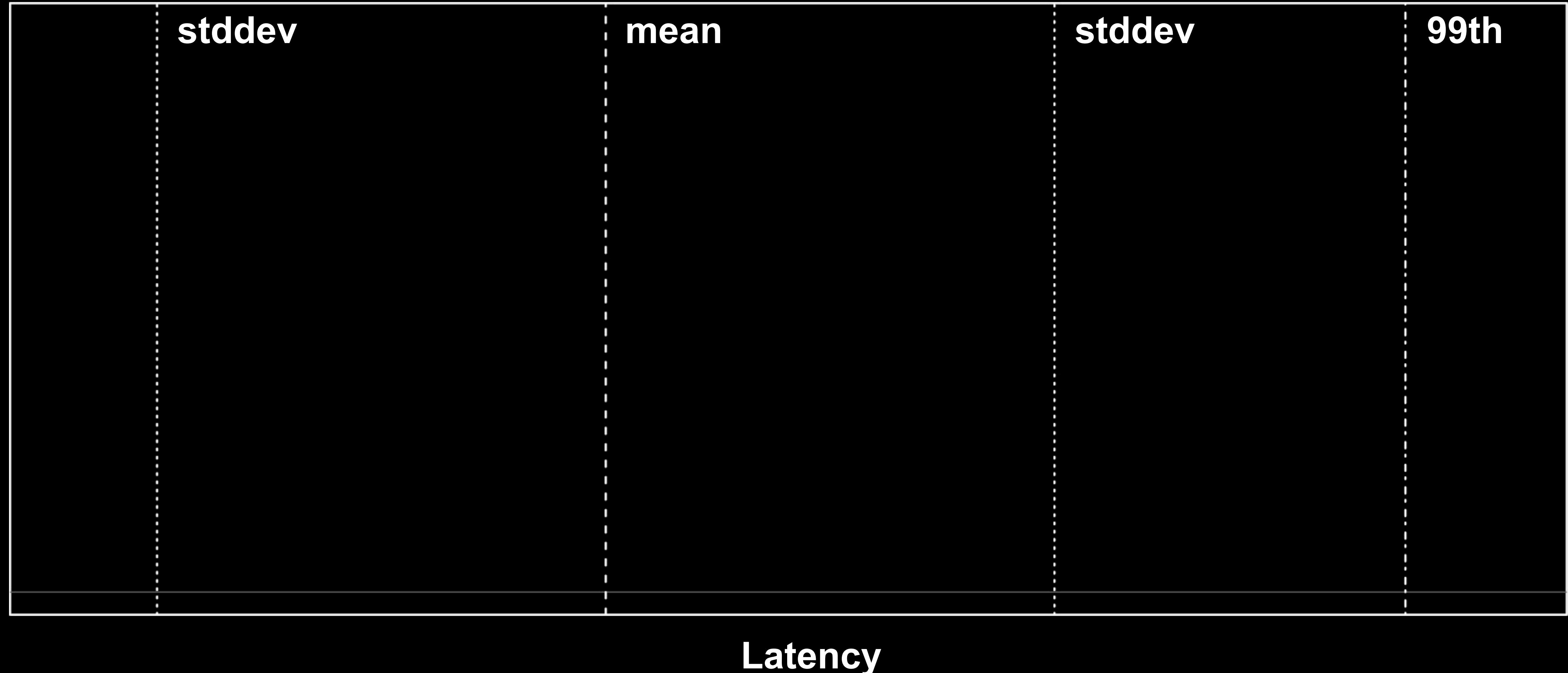
- Type I error (red not green): time wasted
- Type II error (green not red): more time wasted & undiagnosed errors
- Misleading for subjective metrics: green might not mean what you think it means - depends on tests
- Over-simplification

# Average Anti-Method

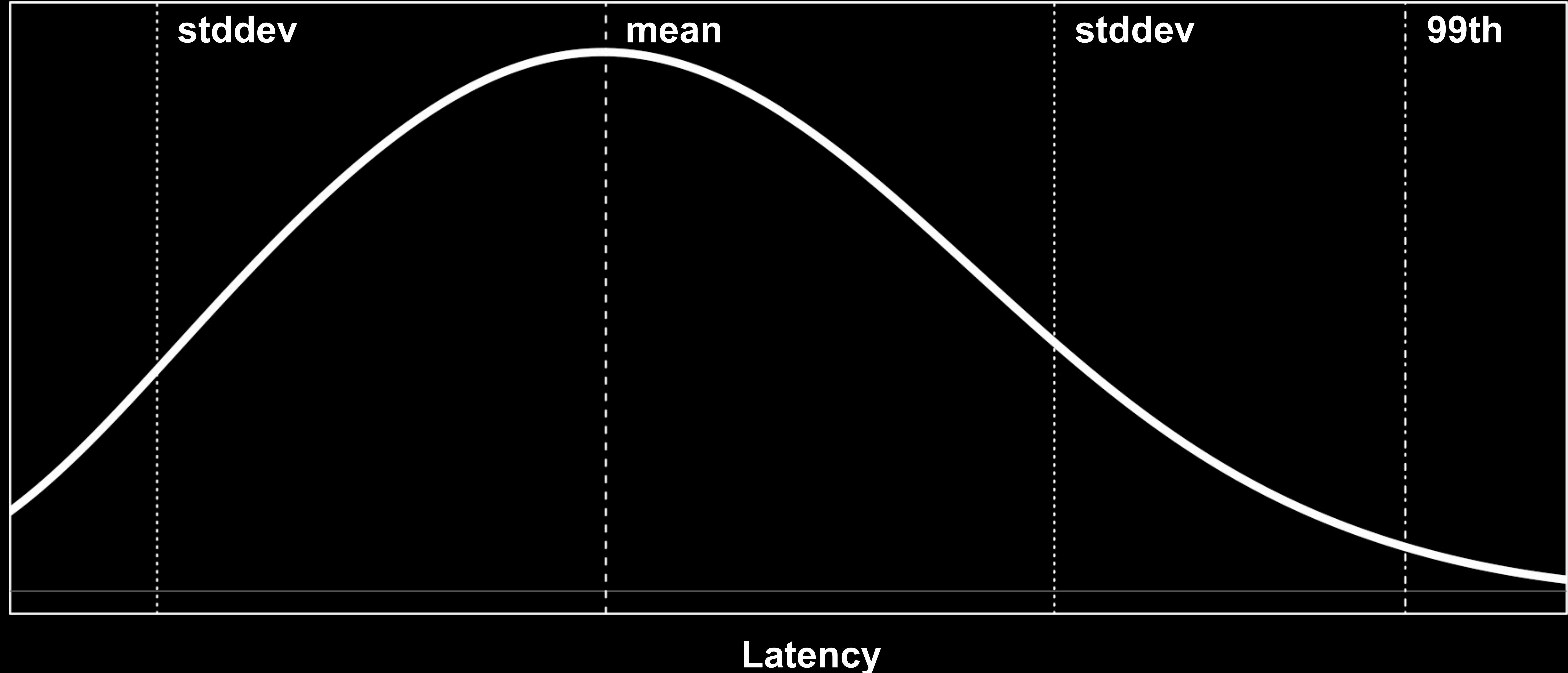
# Average Anti-Method

- 1. Measure the average (mean)
- 2. Assume a normal-like distribution (unimodal)
- 3. Focus investigation on explaining the average

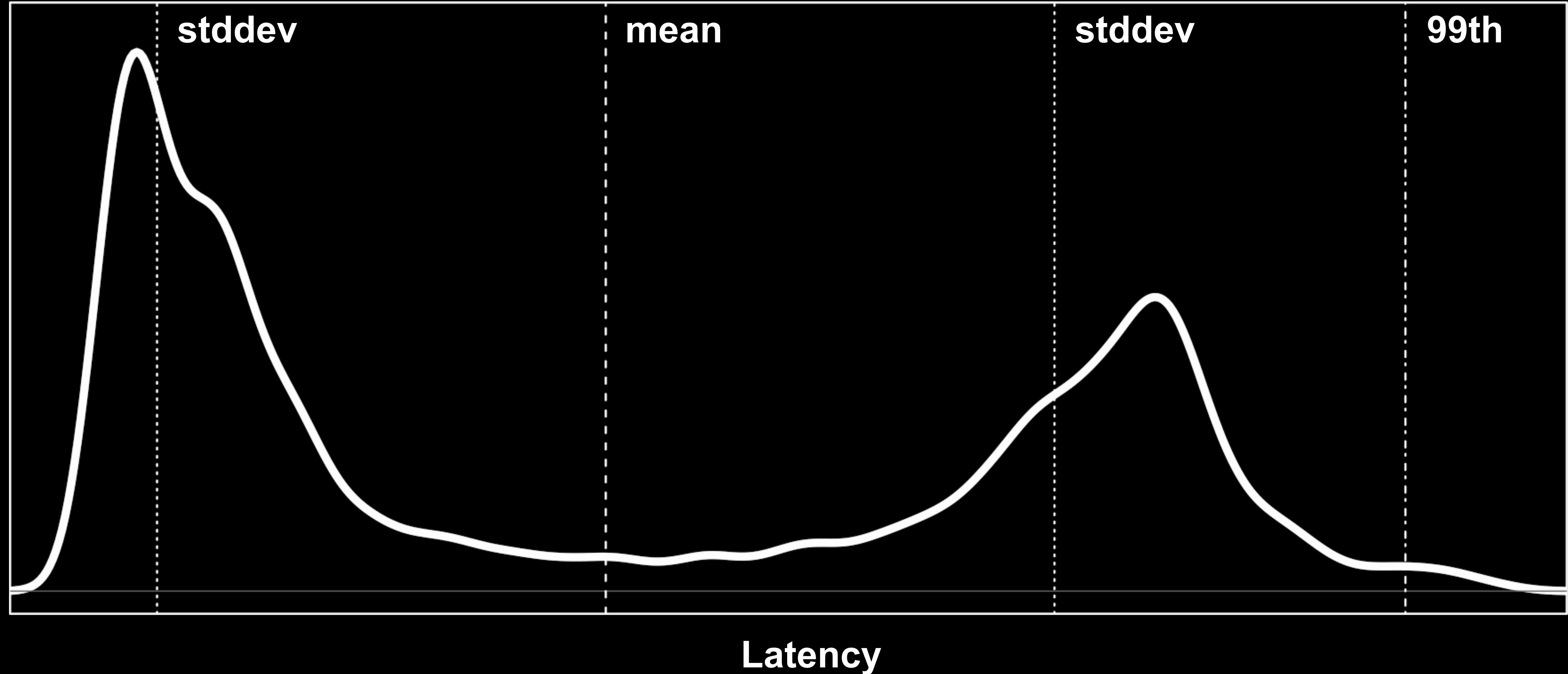
# Average Anti-Method: You Have



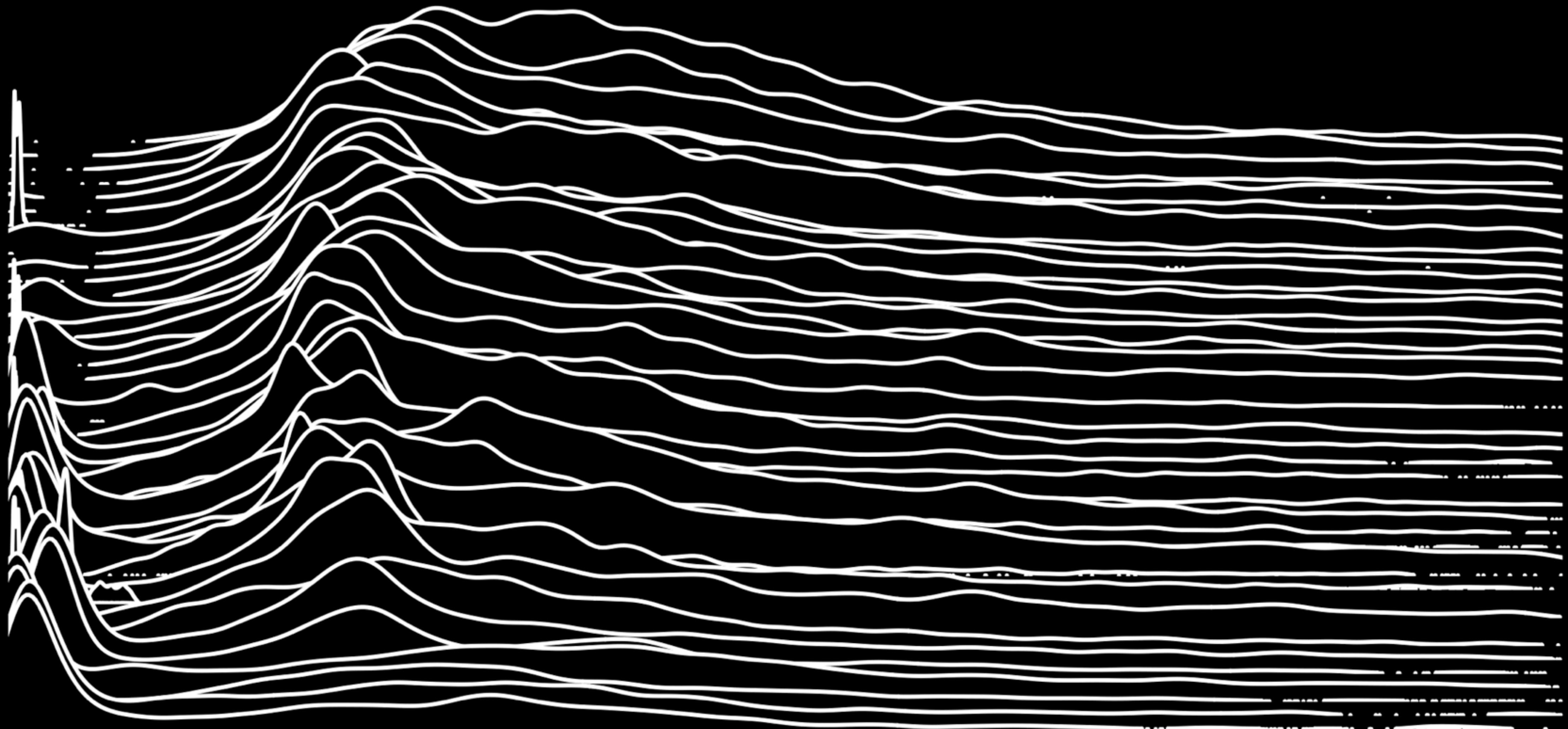
# Average Anti-Method: You Guess



# Average Anti-Method: Reality



# Average Anti-Method: Reality x50

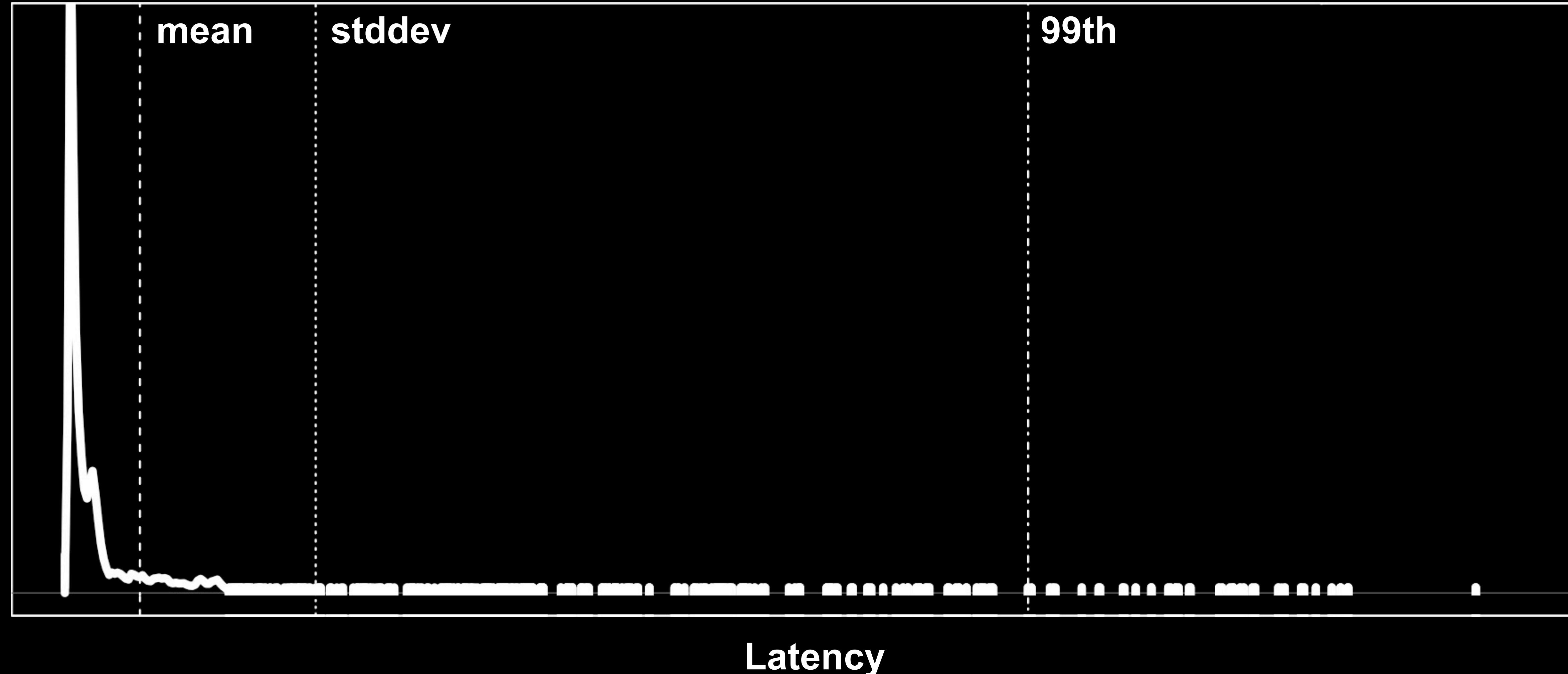


<http://dtrace.org/blogs/brendan/2013/06/19/frequency-trails>

# Average Anti-Method: Examine the Distribution

- Many distributions aren't normal, gaussian, or unimodal
- Many distributions have outliers
  - seen by the max; may not be visible in the 99...th percentiles
  - influence mean and stddev

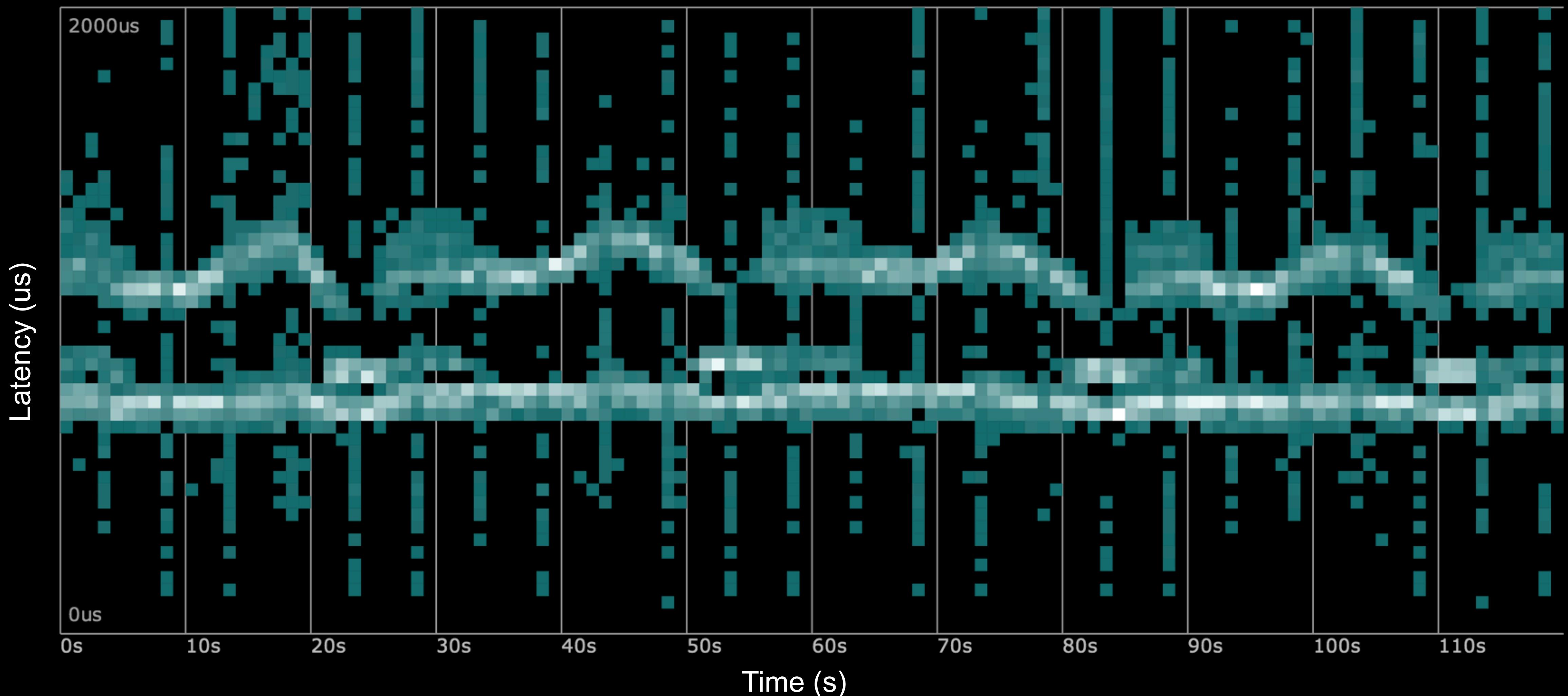
# Average Anti-Method: Outliers



# Average Anti-Method: Visualizations

- Distribution is best understood by examining it
  - Histogram summary
  - Density Plot detailed summary (shown earlier)
  - Frequency Trail detailed summary, highlights outliers (previous slides)
  - Scatter Plot show distribution over time
  - Heat Map show distribution over time, and is scaleable

# Average Anti-Method: Heat Map



<http://dtrace.org/blogs/brendan/2013/05/19/revealing-hidden-latency-patterns>  
<http://queue.acm.org/detail.cfm?id=1809426>

# Average Anti-Method

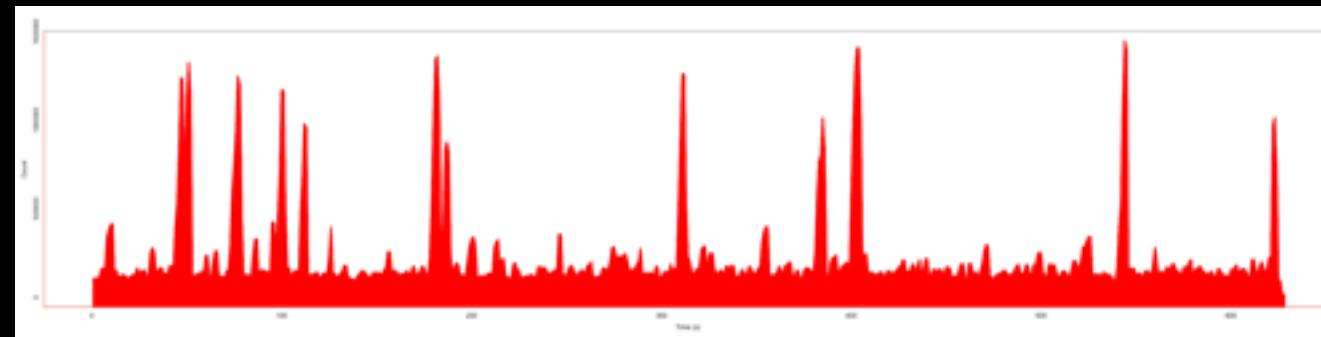
- Pros:
  - Averages are versatile: time series line graphs, Little's Law
- Cons:
  - Misleading for multimodal distributions
  - Misleading when outliers are present
  - Averages are average

# Concentration Game Anti-Method

# Concentration Game Anti-Method

- 1. Pick one metric
- 2. Pick another metric
- 3. Do their time series look the same?
  - If so, investigate correlation!
- 4. Problem not solved? goto 1

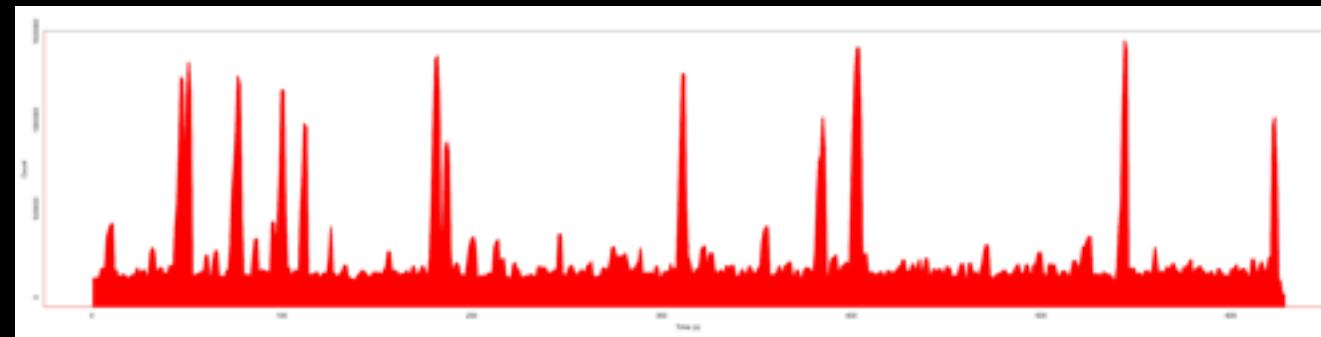
# Concentration Game Anti-Method, cont.



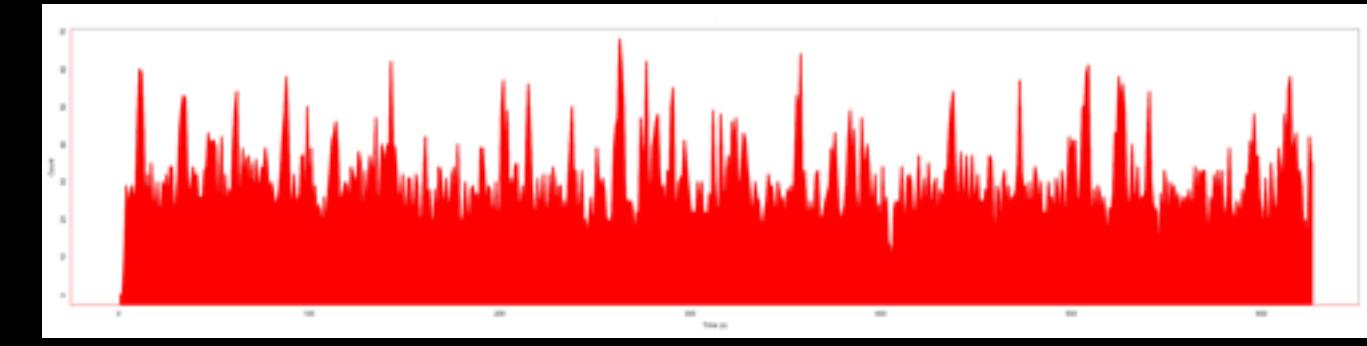
App Latency



# Concentration Game Anti-Method, cont.



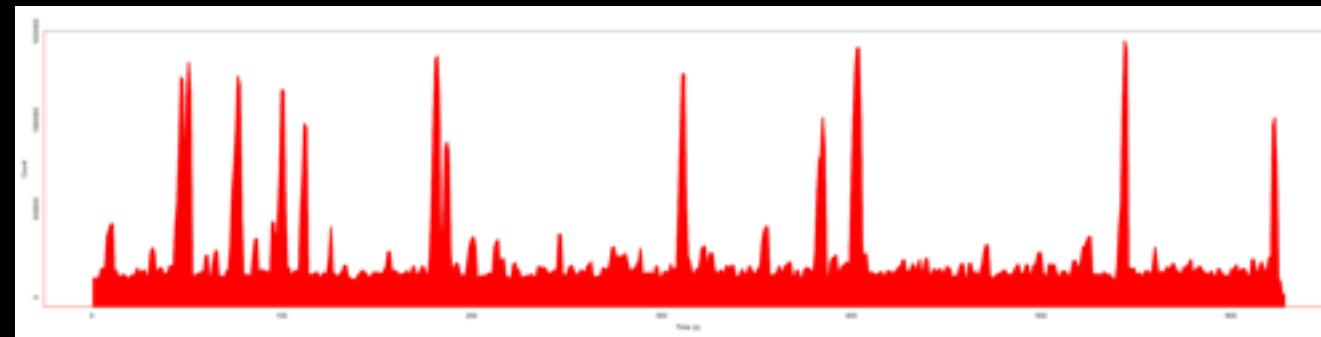
App Latency



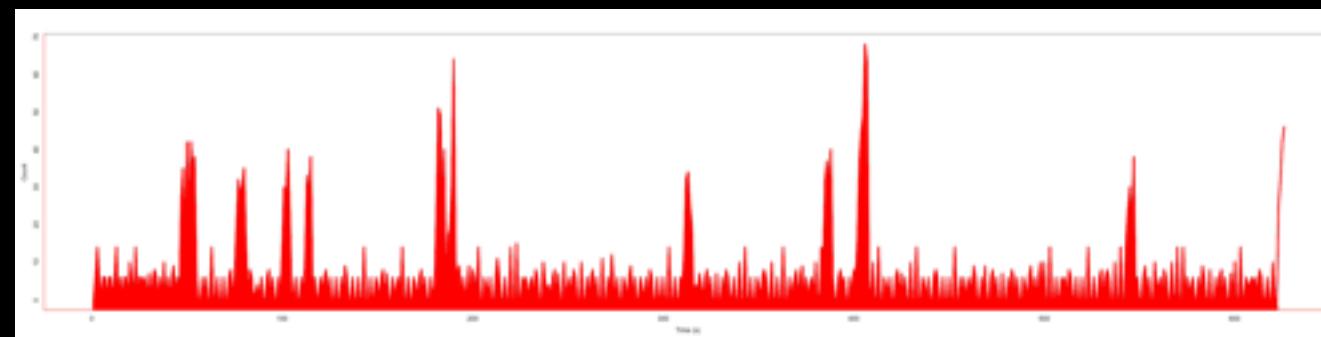
NO



# Concentration Game Anti-Method, cont.



App Latency



YES!



# Concentration Game Anti-Method, cont.

- Pros:
  - Ages 3 and up
  - Can discover important correlations between distant systems
- Cons:
  - Time consuming: can discover many symptoms before the cause
  - Incomplete: missing metrics

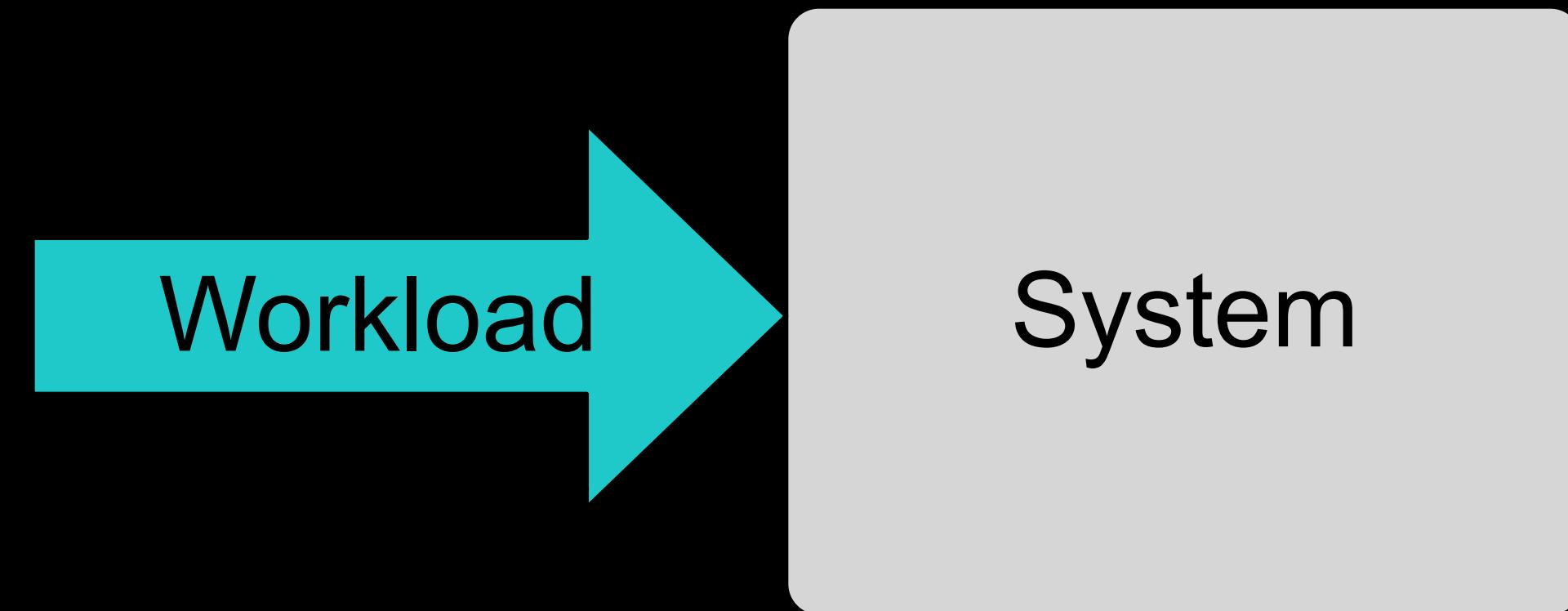
# Workload Characterization Method

# Workload Characterization Method

- 1. Who is causing the load?
- 2. Why is the load called?
- 3. What is the load?
- 4. How is the load changing over time?

# Workload Characterization Method, cont.

- 1. Who: PID, user, IP addr, country, browser
- 2. Why: code path, logic
- 3. What: targets, URLs, I/O types, request rate (IOPS)
- 4. How: minute, hour, day
- The target is the system input (the workload)  
not the resulting performance



# Workload Characterization Method, cont.

- Pros:
  - Potentially largest wins: eliminating unnecessary work
- Cons:
  - Only solves a class of issues – load
  - Can be time consuming and discouraging – most attributes examined will not be a problem

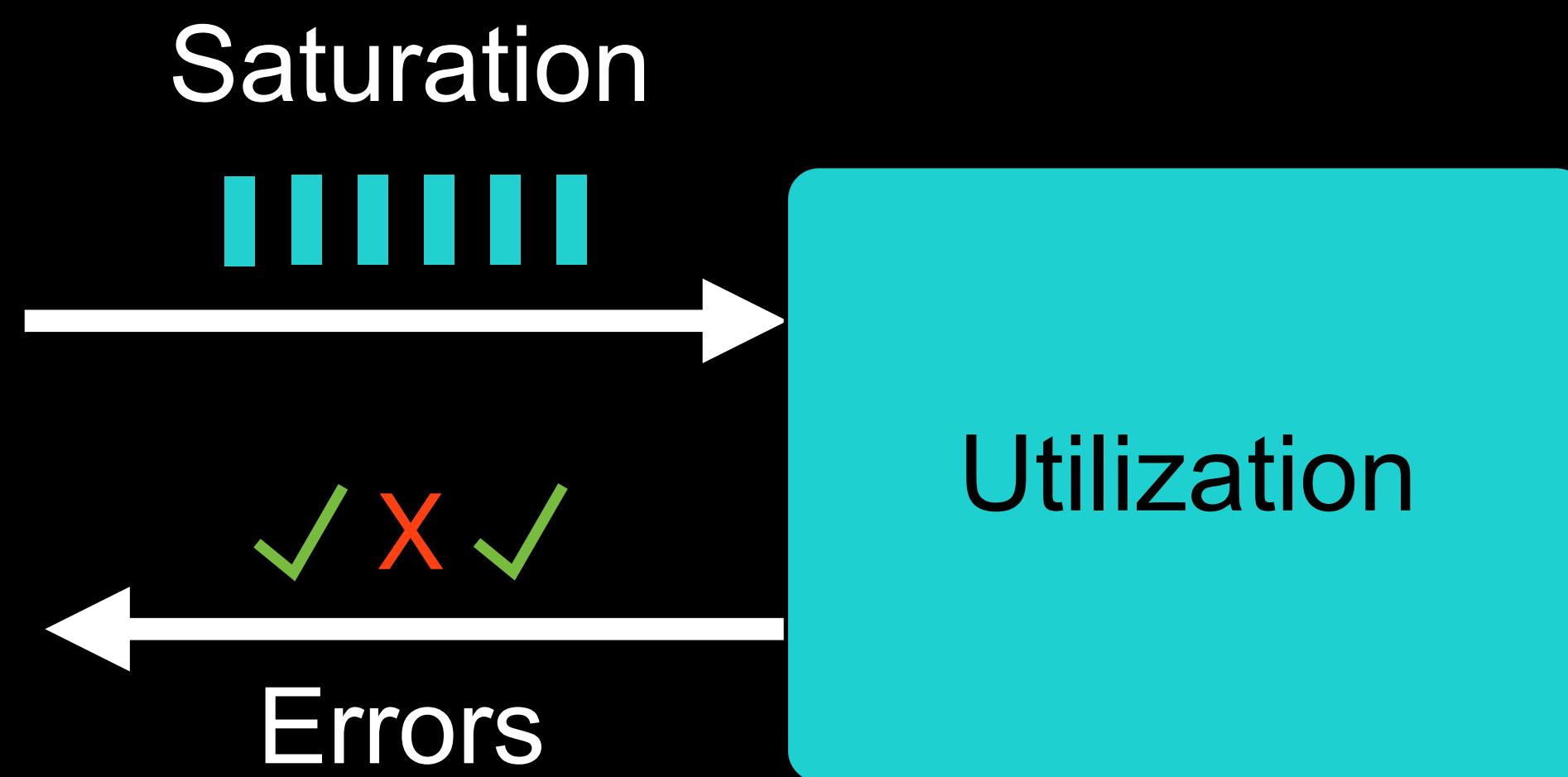
# USE Method

# USE Method

- For every resource, check:
  - 1. Utilization
  - 2. Saturation
  - 3. Errors

# USE Method, cont.

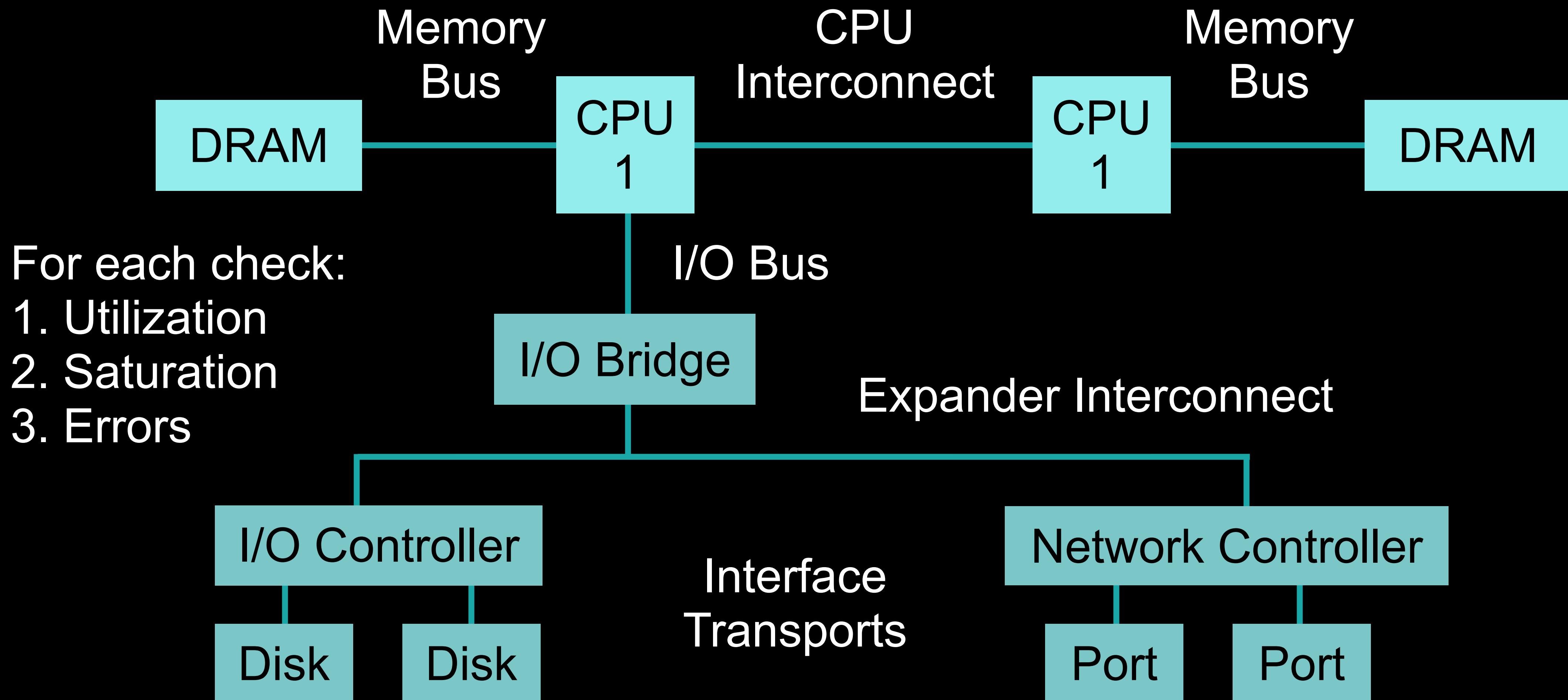
- For every resource, check:
  - 1. Utilization: time resource was busy, or degree used
  - 2. Saturation: degree of queued extra work
  - 3. Errors: any errors
- Identifies resource bottlenecks quickly



# USE Method, cont.

- Hardware Resources:
  - CPUs
  - Main Memory
  - Network Interfaces
  - Storage Devices
  - Controllers
  - Interconnects
- Find the *functional diagram* and examine every item in the *data path*...

# USE Method, cont.: System Functional Diagram

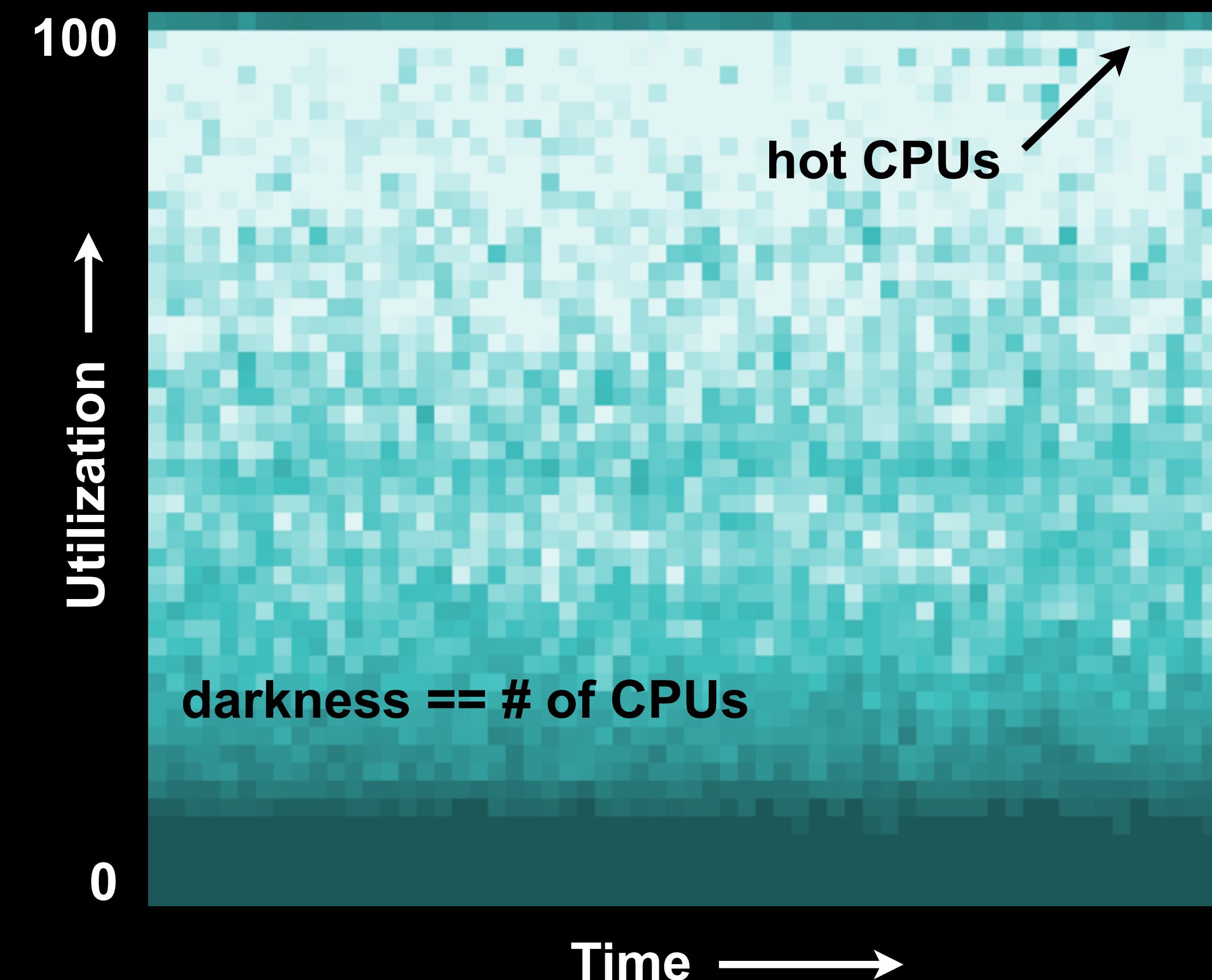


# USE Method, cont.: Linux System Checklist

Resource	Type	Metric
CPU	Utilization	per-cpu: mpstat -P ALL 1, "%idle"; sar -P ALL, "%idle"; system-wide: vmstat 1, "id"; sar -u, "%idle"; dstat -c, "idl"; per-process: top, "%CPU"; htop, "CPU%"; ps -o pcpu; pidstat 1, "%CPU"; per-kernel-thread: top/htop ("K" to toggle), where VIRT == 0 (heuristic).
CPU	Saturation	system-wide: vmstat 1, "r" > CPU count [2]; sar -q, "runq-sz" > CPU count; dstat -p, "run" > CPU count; per-process: /proc/PID/schedstat 2nd field (sched_info.run_delay); perf sched latency (shows "Average" and "Maximum" delay per-schedule); dynamic tracing, eg, SystemTap schedtimes.stp "queued(us)"
CPU	Errors	perf (LPE) if processor specific error events (CPC) are available; eg, AMD64's "04Ah Single-bit ECC Errors Recorded by Scrubber"
...	...	...

# USE Method, cont.: Monitoring Tools

- Average metrics don't work: individual components can become bottlenecks
- Eg, CPU utilization
- Utilization heat map on the right shows 5,312 CPUs for 60 secs; can still identify "hot CPUs"



<http://dtrace.org/blogs/brendan/2011/12/18/visualizing-device-utilization>

# USE Method, cont.: Other Targets

- For cloud computing, must study any resource limits as well as physical; eg:
  - physical network interface U.S.E.
  - AND instance network cap U.S.E.
- Other software resources can also be studied with USE metrics:
  - Mutex Locks
  - Thread Pools
- The application environment can also be studied
  - Find or draw a functional diagram
  - Decompose into queueing systems

# USE Method, cont.: Homework

- Your ToDo:
  - 1. find a system functional diagram
  - 2. based on it, create a USE checklist on your internal wiki
  - 3. fill out metrics based on your available toolset
  - 4. repeat for your application environment
- You get:
  - A checklist for all staff for quickly finding bottlenecks
  - Awareness of what you cannot measure:
    - *unknown unknowns* become *known unknowns*
    - ... and known unknowns can become feature requests!

# USE Method, cont.

- Pros:
  - Complete: all resource bottlenecks and errors
  - Not limited in scope by available metrics
  - No unknown unknowns – at least known unknowns
  - Efficient: picks three metrics for each resource – from what may be hundreds available
- Cons:
  - Limited to a class of issues: resource bottlenecks

# Thread State Analysis Method

# Thread State Analysis Method

- 1. Divide thread time into operating system states
- 2. Measure states for each application thread
- 3. Investigate largest non-idle state

# Thread State Analysis Method, cont.: 2 State

- A minimum of two states:

On-CPU	
Off-CPU	

# Thread State Analysis Method, cont.: 2 State

- A minimum of two states:

On-CPU	executing spinning on a lock
Off-CPU	waiting for a turn on-CPU waiting for storage or network I/O waiting for swap ins or page ins blocked on a lock idle waiting for work

- Simple, but off-CPU state ambiguous without further division

# Thread State Analysis Method, cont.: 6 State

- Six states, based on Unix process states:

Executing	
Runnable	
Anonymous Paging	
Sleeping	
Lock	
Idle	

# Thread State Analysis Method, cont.: 6 State

- Six states, based on Unix process states:

Executing	on-CPU
Runnable	and waiting for a turn on CPU
Anonymous Paging	runnable, but blocked waiting for page ins
Sleeping	waiting for I/O: storage, network, and data/text page ins
Lock	waiting to acquire a synchronization lock
Idle	waiting for work

- Generic: works for all applications

# Thread State Analysis Method, cont.

- As with other methodologies, these pose questions to answer
  - Even if they are hard to answer
- Measuring states isn't currently easy, but can be done
  - Linux: /proc, schedstats, delay accounting, I/O accounting, DTrace
  - SmartOS: /proc, microstate accounting, DTrace
- Idle state may be the most difficult: applications use different techniques to wait for work

# Thread State Analysis Method, cont.

- States lead to further investigation and actionable items:

Executing	Profile stacks; split into usr/sys; sys = analyze syscalls
Runnable	Examine CPU load for entire system, and caps
Anonymous Paging	Check main memory free, and process memory usage
Sleeping	Identify resource thread is blocked on; syscall analysis
Lock	Lock analysis

# Thread State Analysis Method, cont.

- Compare to database query time. This alone can be misleading, including:
  - swap time (anonymous paging) due to a memory misconfig
  - CPU scheduler latency due to another application
- Same for any “time spent in ...” metric
  - is it really *in* ...?

# Thread State Analysis Method, cont.

- Pros:
  - Identifies common problem sources, including from other applications
  - Quantifies application effects: compare times numerically
  - Directs further analysis and actions
- Cons:
  - Currently difficult to measure all states

# More Methodologies

- Include:
  - Drill Down Analysis
  - Latency Analysis
  - Event Tracing
  - Scientific Method
  - Micro Benchmarking
  - Baseline Statistics
  - Modelling
- For when performance *is* your day job

# Stop the Guessing

- The anti-methodologies involved:
  - guesswork
  - beginning with the tools or metrics (answers)
- The actual methodologies posed questions, then sought metrics to answer them
- You don't need to guess – post-DTrace, practically everything can be known
- Stop guessing and start asking questions!

# Thank You!

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- github: <https://github.com/brendangregg>
- blog: <http://dtrace.org/blogs/brendan>
- blog resources:
  - <http://dtrace.org/blogs/brendan/2008/11/10/status-dashboard>
  - <http://dtrace.org/blogs/brendan/2013/06/19/frequency-trails>
  - <http://dtrace.org/blogs/brendan/2013/05/19/revealing-hidden-latency-patterns>
  - <http://dtrace.org/blogs/brendan/2012/03/07/the-use-method-linux-performance-checklist>
  - <http://dtrace.org/blogs/brendan/2011/12/18/visualizing-device-utilization>