# Do You Really Know Your Response Times? 

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## Sky Over The Top Delivery

- Web Services
- Over The Top Asset Delivery
- NowTV/Sky Go
- Always up
- High traffic
- Highly concurrent


## OTT Endpoints



- How much traffic is hitting that endpoint?
- How quickly are we responding to a typical customer?
- One customer complained we respond slowly. How slow do we get?
- What's the difference between the fastest and slowest responses?
- I don't care about anomalies but how slow are the slowest $1 \%$ ?


## Collecting Response Times

My App Response Time Collection System

- Large volumes of network traffic
- Risk of losing data (network may fail)
- Affects application performance
- Needs measuring itself!


## Our Setup

Application instance 1


Application instance 2

## Dropwizard Metrics Library: Types of Metric

- Counter
- Gauge - 'instantaneous measurement of a value'
- Meter (counts, rates)
- Histogram - min, max, mean, stddev, percentiles
- Timer - Meter + Histogram


## Example Dashboard



## Dropwizard Metrics

- Use Dropwizard and you get
- Metrics infrastructure for free
- Metrics from Cassandra and Dropwizard bundles for free
- You can easily add timers to metrics just by adding annotations
- Ports exist for other languages
- Developers, architects, managers everybody loves graphs
- We trust and depend on them
- We rarely understand them
- We lie to ourselves and to our managers with them


## Goals of this talk

- Understand how we can measure service time latencies
- Ensure meaningful statistics are given back to managers
- Learn how to use appropriate dashboards for monitoring and alerting


## What is the $99^{\text {th }}$ Percentile Response Time?

$?$

## What is the $99^{\text {th }}$ Percentile?



## Our Setup

Reservoir Application instance 1


## Reservoirs



## Types of Reservoir

- Sliding window
- Time-base sliding window
- Exponentially decaying


## Forward Decay



Time

$$
w_{i}=e^{\alpha x_{i}}
$$

| $w_{1}$ | $w_{2}$ | $w_{3}$ | $w_{4}$ | $w_{5}$ | $w_{6}$ | $w_{7}$ | $w_{8}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| $v_{1}$ | $v_{2}$ | $v_{3}$ | $v_{4}$ | $v_{5}$ | $v_{6}$ | $v_{7}$ | $v_{8}$ |
| Sorted by value |  |  |  |  |  |  |  |

## Getting at the percentiles

- Normalise weights: $\sum_{i} w_{i}=1$
- Lookup by normalised weight


## Data retention

- Sorted Map indexed by w.random_number
- Smaller indices removed first


## Response Time Jumps for 4 Minutes



## One Percent Rise from 20 ms to 500 ms




## One Percent Rise from 20 ms to 500 ms



## Trade-off

- Autonomous teams
- Know one app well
- Feel responsible for app performance
- But...
- Can't know everything
- Will make mistakes with numbers
- We might even ignore mistakes


## One Long Request Blocks New Requests



## One Long Request Blocks New Requests



Spikes and Tower Blocks


## Splitting Things Up



## Metric Imbalance Visualised



## Metric Imbalance

- One pool gives more accurate results
- Multiple pools allow drilling down, but...
- Some pools may have inaccurate performance measurements
- Only those with sufficient rates should be analysed
- How can we narrow down on just those?
- Simpson's Paradox


## Simpson's Paradox

## Explanation

- Two variables have a positive correlation
- Grouped data shows a negative correlation
- There's a lurking third variable


## Simpson's Paradox



- Increasing traffic $\Longrightarrow X$ gets slower
- Increasing traffic $\Longrightarrow Y$ gets faster
- We move \% traffic to System Y
- We wait for prime time peak
- System gets slower???
- $100 \%$ of brand B traffic still goes to $X$
- Results are pooled by client and brand
- Classic example: UC Berkeley gender bias


## Lessons Learnt

- Want fast alerting?
- Use max
- If you don't graph the max you are hiding the bad
- Don't just look at fixed percentiles.
- Understand the distribution of the data (HdrHistogram)
- A few fixed percentiles tells you very little as a test
- Monitor one metric per endpoint
- When aggregating response times
- Use maxSeries


## So We're Living a Lie, Does it Matter?



## Conclusions and Thoughts

- Don't immediately assume numbers on dashboards are meaningful
- Understand what you are graphing
- Test assumptions
- Provide these tools and developers will confidently use them
- Although maybe not correctly!
- Most developers are not mathematicians
- Keep it simple!
- Know which numbers are real and which are lies!

Thank you

