

The State of the Histogram

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[SLO Conf](#) - May 2021



Heinrich Hartmann - SRE @  zalando ()

Why Histograms?



Data



Histogram

- small
- accurate percentiles
- mergeable

Application: Latency measurements & SLOs.

⇒ Latency SLOs
done right.

- @FOSDEM 2019
- Circus/blog.

History of

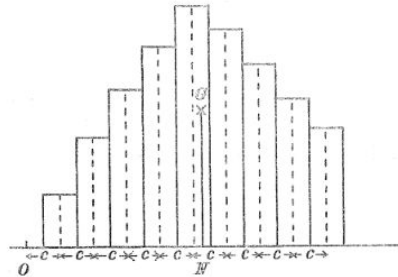
Histograms

boundary of the first rectangle, on the line of common bases, and let y_r be the height of the r^{th} rectangle, or

$$y_r = \frac{\alpha}{c} \frac{n(n-1)\dots(n-r+2)}{r-1} p^{n-r+1} q^{r-1},$$

while

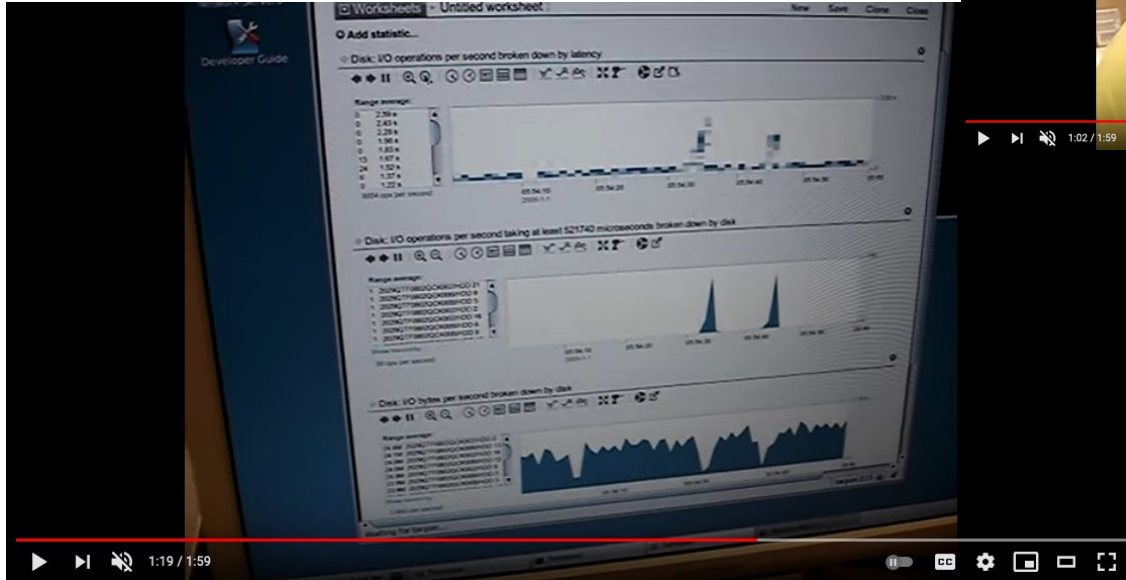
$$y_1 = \alpha p^n / c.$$



Let us find the values of

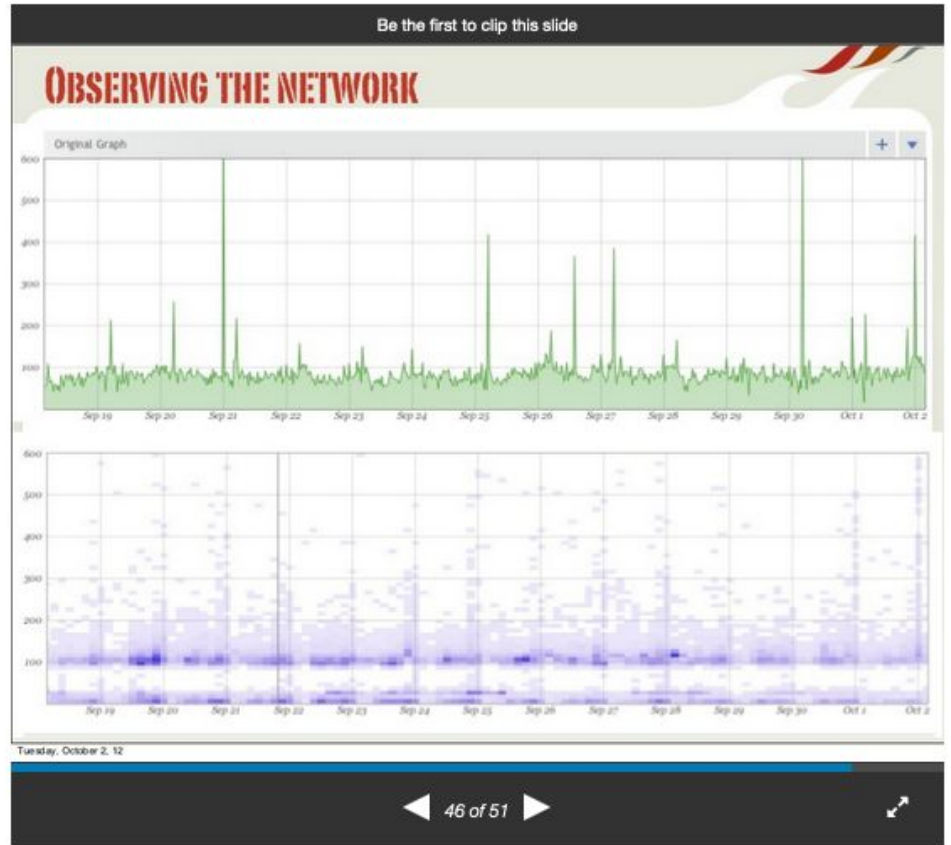
$$\Sigma \{y_r c \times (rc)^r\},$$





Brendan Gregg @ Sun -- Shouting at the DataCenter

2009



Technology • 7,763 views • Oct. 02, 2012

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Theo Schlossnagle @ Circonus -- Monitoring & Observability

2012




HdrHistogram

- Provides tools for iteration
 - Linear, Logarithmic, Percentile
- Supports percentile iterators
 - Practical due to high dynamic range
- Convenient percentile output
 - 10% intervals between 0 and 50%
 - 5% intervals between 50% and 75%
 - 2.5% intervals between 75% and 87.5%...
- Very useful for feeding percentile distribution graphs...

Value	Percentile	TotalCountIncludingThisValue
0.003	0.000000000000	7
0.057	0.100000000000	807222
0.056	0.200000000000	2235747
0.059	0.300000000000	1694413
0.060	0.400000000000	1994719
0.062	0.500000000000	2373326
0.064	0.550000000000	2620309
0.066	0.600000000000	2795011
0.070	0.650000000000	3010126
1.280	0.700000000000	3228296
5.552	0.750000000000	3456162
7.712	0.775000000000	3574401
9.856	0.800000000000	3649655
12.016	0.825000000000	3805210
14.176	0.850000000000	3920746
16.320	0.875000000000	4036566
17.408	0.887500000000	4096671
18.464	0.900000000000	4150910
19.584	0.912500000000	4209006
20.832	0.925000000000	4267165
22.208	0.937500000000	4324157
22.976	0.947500000000	4352952
23.608	0.950000000000	4381652
24.736	0.958750000000	4410732
25.760	0.962500000000	4439554
26.880	0.968750000000	4467918
27.488	0.971875000000	4482272
28.160	0.975000000000	4496605
28.896	0.978125000000	4511359
29.696	0.981250000000	4525423
30.056	0.984375000000	4539989
31.200	0.989375000000	4547261
31.776	0.987500000000	4554465
32.384	0.989062500000	4561828
33.038	0.990625000000	4569070

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Computing Extremely Accurate Quantiles Using t -Digests

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Summary. We present on-line algorithms for computing approximations of rank-based statistics that give high accuracy, particularly near the tails of a distribution, with very small sketches. Notably, the method allows a quantile q to be computed with an accuracy relative to $\max(q, 1 - q)$ rather than absolute accuracy as with most other methods. This new algorithm is robust with respect to skewed distributions or ordered datasets and allows separately computed summaries to be combined with no loss in accuracy.

An open-source Java implementation of this algorithm is available from the author. Independent implementations in Go and Python are also available.

Keywords: quantiles, median, rank statistics, t -digest

DDSketch: A Fast and Fully-Mergeable Quantile Sketch with Relative-Error Guarantees

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ABSTRACT

Summary statistics such as the mean and variance are easily maintained for large, distributed data streams, but order statistics (i.e., sample quantiles) can only be approximately summarized. There is extensive literature on maintaining quantile sketches where the emphasis has been on bounding the rank error of the sketch while using little memory. Unfortunately, rank error guarantees do not preclude arbitrarily large relative errors, and this often occurs in practice when the data is heavily skewed.

Given the distributed nature of contemporary large-scale systems, another crucial property for quantile sketches is mergeability, i.e., several combined sketches must be as accurate as a single sketch of the same data. We present the first fully-mergeable, relative-error quantile sketching algorithm with formal guarantees. The sketch is extremely fast and accurate, and is currently being used by Datadog at a wide-scale.

PVLDB Reference Format:

Charles Masson and Jee E Rim and Homin K. Lee. DDSketch: A fast and fully-mergeable quantile sketch with relative-error guarantees. *PVLDB*, 12(12): 2195-2205, 2019.
DOI: <https://doi.org/10.14778/3352063.3352135>

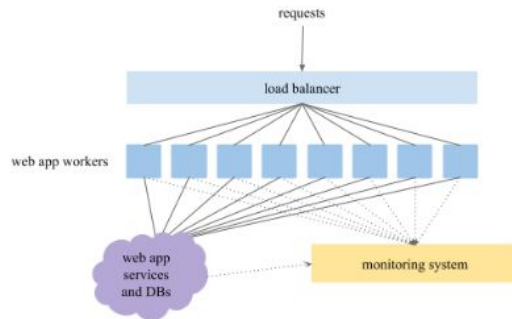


Figure 1: A distributed web application, with each container sending metrics to the monitoring system.

an event storage system) can be high enough that simply forwarding all this information can strain the capacities (network, memory,

CIRCLLHIST

A LOG-LINEAR HISTOGRAM DATA STRUCTURE FOR IT INFRASTRUCTURE MONITORING

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January 22, 2020

ABSTRACT

The circllhist histogram is a fast and memory efficient data structure for summarizing large numbers of latency measurements. It is particularly suited for applications in IT infrastructure monitoring, and provides nano-second data insertion, full mergeability, accurate approximation of quantiles with a-priori bounds on the relative error.

Open-source implementations are available for C/lua/python/Go/Java/JavaScript.

1 Introduction

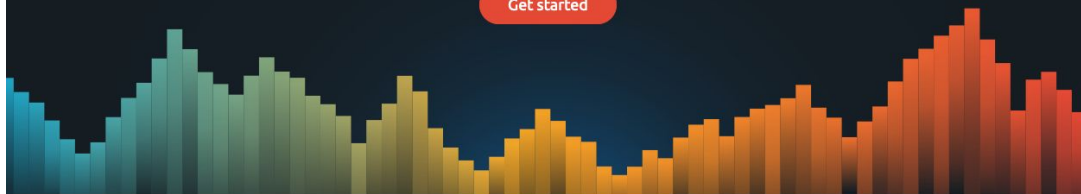
Latency measurements have become an important part of IT infrastructure and application monitoring. The latencies of a wide variety of events like requests, function calls, garbage collection, disk IO, system-call, CPU scheduling, etc. are of great interest of engineers operating and developing IT systems.

There are a number of technical challenges associated with managing and analyzing latency data. The volume emitted by a single data source can easily become very large. Furthermore, data has to be collected and aggregated from a large number of different sources. The data has to be stored over long time periods (months, years), in order to allow historic comparisons and long-term service quality estimations (SLOs).

Vendor-neutral log-linear histograms for the compression, mergeability, and analysis of telemetry data

OpenHistogram is an open source data structure for quickly and efficiently summarizing large numbers of measurements.

[Get started](#)



Organizations using OpenHistogram



SmugMug 

SPARKPOST 

2021

So many Histograms!

HdR

Circ Histogram

open histogram

t-Digest

DDsketch

Which should you use?

Desirable Properties

- (1) small, bounded size
- (2) fast insertion
- (3) accurate quantiles
- (4) mergeability.
- (5) zero configuration / compatibility

Evaluation - Contenders

- exact Exact quantile computation based on numpy arrays [6].
- prom Quantile estimation based on Prometheus Histograms [4].
- hdr The HDR Histogram data-structure introduced in [5].
- dd The DDSketch data-structure introduced in [1].
- t-digest The t-digest data-structure introduced in [3].
- circllhist The circllhist data-structure described in this document.

Summary

	Array	Prun. "Histogram"	t-Digest	Log-Histograms		
				HdR	DD	Cirellist
size	X	✓	✓	✓	✓	✓
accuracy	✓	X	✓	✓	✓	✓
performance	✓	✓	✓	✓	✓	✓
mergeability	✓	✓	?	✓	✓	✓
zero-conf.	✓	X	○	X	○	✓

Adoption in

Open Source
Technology



Sparse high-resolution histograms for Prometheus

- Author: Björn Rabenstein
- 2020-06-08: Started work on draft.
- 2021-02-10: First published version.
- 2021-02-25: Prometheus dev-summit approves doc as guideline for experiments.
- 2021-03-15: All review comments so far addressed.

Terminology

In this document, the capitalized word *Histogram* refers to the Prometheus metric type, while the lower-case word *histogram* refers to the statistical concept in general.

Further terms explained:



Björn Beorn Rabenstein

Principal Software Engineer at Grafana Labs



Add exponential bucketing to histogram protobuf #149

Open yzhuge wants to merge 5 commits into [open-telemetry:main](#) from [yzhuge:exponential-histogram](#)

Conversation 22 Commits 5 Checks 1 Files changed 1

YZ

yzhuge commented on 8 Mar · edited

Follow up of [open-telemetry/opentelemetry-protobuf#226](#)

👍 1