



How *t*-digest works and why

Ted Dunning June 1, 2015

T-digest

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A New Look at Anomaly Detection

by Ted Dunning and Ellen Friedman © June 2014 (published by O'Reilly)



e-book available courtesy of MapR

http://bit.ly/1jQ9QuL



Last October: *Time Series Databases*

by Ted Dunning and Ellen Friedman © Oct 2014 (published by O'Reilly)



Available Now: Real World Hadoop

by Ted Dunning and Ellen Friedman © Feb 2015 (published by O'Reilly)



Practical Machine Learning series (O'Reilly)

- Machine learning is becoming mainstream
- Need pragmatic approaches that take into account real world business settings:
 - Time to value
 - Limited resources
 - Availability of data
 - Expertise and cost of team to develop and to maintain system
- Look for approaches with big benefits for the effort expended

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Agenda

- Why should we estimate quantiles?
- How t-digest works
- How can you get it?
- Questions



Why on-line algorithms?





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Why Quantiles (percentiles)





Suppose You Have ...

- 100 M users, 1K sites touched each day
- What is 99.9% latency for each user/site combination?

for each user?

for each site?

for users in Kansas?

for users who complained?

for users who complained, but before they complained?



Or Suppose ...

- 1000 nodes, each with 24 disks, 100 unique RPC calls
- Want latencies for all disks, all RPC calls between all nodes
 50 %-ile, 99%-ile, 99.9%-ile
- <100ns overhead per measurement
- <10MB overhead per node
- No logs except for exceptionally slow cases
- Summary at any time



What about accuracy?

What Accuracy Required?

- 50%-ile ± 0.5%
- 99.99%-ile ± 0.5%
- 99.99%-ile ± 0.001%
- 50%-ile ± 0.001%



What Accuracy Required?

- 50%-ile ± 0.5% Often just fine
- 99.99%-ile ± 0.5% *Nonsense*
- 99.99%-ile ± 0.001% By definition
- 50%-ile ± 0.001% *Over-kill*



The internals

Variable Cluster Size for Constant Relative Accuracy





Second-Order Accuracy via Interpolation





Translation Between Quantile and Cluster





The Algorithm

- Static Buffers
 - n1 new points
 - n2 existing centroids
 - n2 merge space
- Algorithm
 - Collect new points until full
 - Sort new points
 - Merge with existing centroids
 k2 k1 < 1 criterion for merging
 - Swap centroids and merge space

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- Can be implemented with inplace merge
- Can use approximate q-k mapping for speed
- Completely static memory

Using t-digest



Available As

- An aggregator in Elastic Search
- In stream-lib
- As a UDF for Apache Drill (soon!)
- In Apache Mahout
- From Maven Central

```
<dependency>
  <groupId>com.tdunning</groupId>
  <artifactId>t-digest</artifactId>
  <version>3.1</version>
</dependency>
```



The Upshot

- Streaming approximations are important
- Accurate quantiles are important
- The *t*-digest algorithm is simple and very accurate
- You can use it almost anywhere



Special Thanks To

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- Adrien Grand (best tree implementation)
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Who I am

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