

Advanced Distributed Load Balancing

Milestone 5

Praktikum: Cloud Databases

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Agenda

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1.1 Issue

- Allocating the key-ranges of KV-Servers based on a hash of the IP:Port string
 - → leads to an imbalanced distribution of key-ranges
- Even with equally distributed key-ranges, the inserted data can be distributed unevenly among all servers
 - → we waste computational power by not being able to adjust the key-ranges



1.2 What's new?

- Customizable start & end key-range for each KV-Server
- KV-Server key-range partitioning into N buckets
- Usage metrics measuring load on a KV-Server
- Offloading of buckets (based on a threshold T) to neighbouring nodes when KV-Server is experiencing higher load (GET/PUT/DELETE) than neighbours

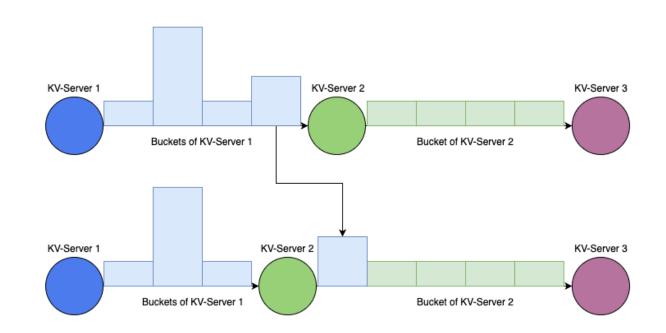
Result

Eventual keyrange distribution among servers that matches the key distribution of the underlying data & divides computational requirements among all available servers more equally



1.3 What happens under load?

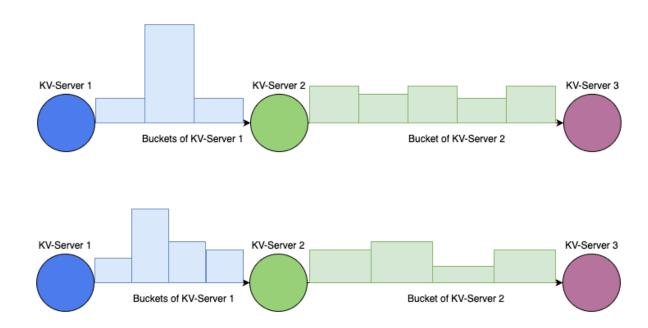
- KV-Server experiences high load → requests usage metrics form its neighbouring KV-Servers
- Neighbour with lowest load takes over responsibility of at least T% of the keys that the high-load server holds





1.4 Recalculation of Buckets

- Both KV-Servers undertake a recalculation of their buckets split the keyrange up
- They resize their bucket keyranges to accommodate a total of N buckets once again

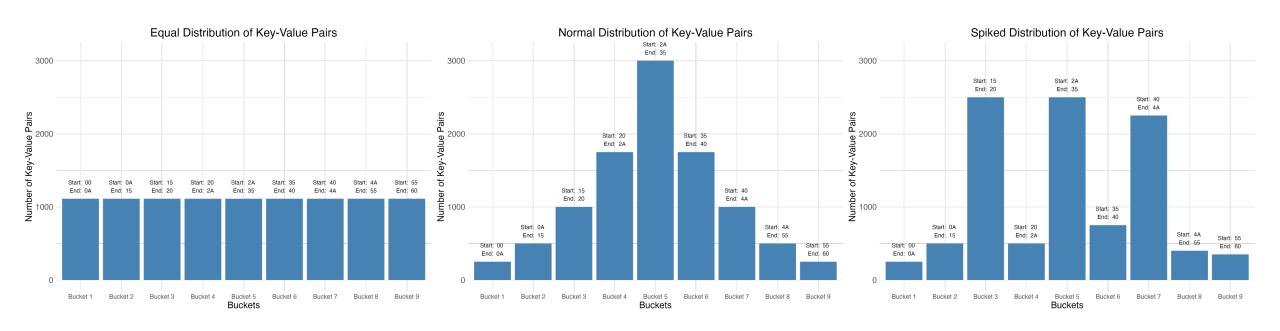






3.1 Benchmarking - Approach

- We pre-defined three different key-value pair distributions to mimic real world datasets and their intrinsic key distributions
- 10k keys in all different combinations





3.1 Benchmarking - Approach

- Based on these key distributions we conducted a sensitivity analysis to find the optimal values for the number of buckets N and offload threshold value T
- Using these optimal value for bucket count N and threshold T we compared the performance of GET/PUT/DELETE operations and a mixed case of GETs and PUTs

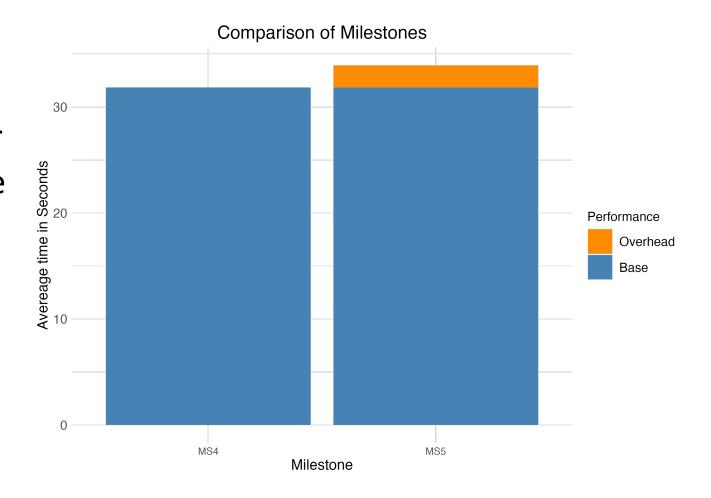
Test Environment (Local):

- Apple M1
- 8GB Memory



When forcing MS5 to act under MS4 conditions (no load-averse offloading of keys)

→ overhead of managing the buckets and frequency table of one KV server to be ~2s



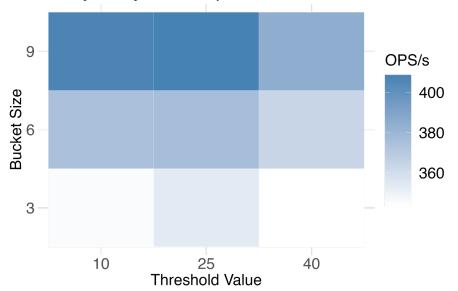


Equal Distribution (F1)

Threshold / Bucket Count	10%	25%	45%
9	24.56	24.48	25.97
6	26.67	26.57	27.44
3	28.95	28.27	29.12

OPS/s Heatmap





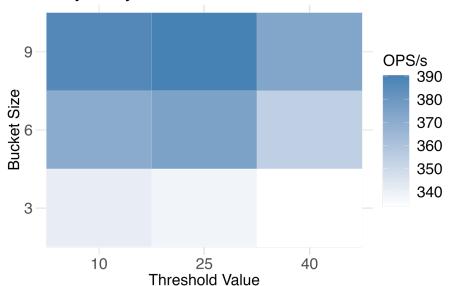


Normal Distribution (F2)

Threshold / Bucket Count	10%	25%	45%
9	25.83	25.62	26.76
6	26.96	26.66	28.25
3	29.31	29.58	29.97

OPS/s Heatmap





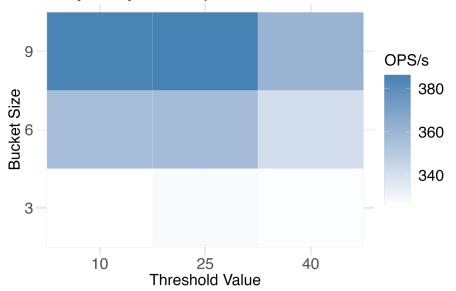


Spiked Distribution (F3)

Threshold / Bucket Count	10%	25%	45%
9	25.96	25.89	27.71
6	28.06	28.05	29.35
3	30.71	30.48	30.62

OPS/s Heatmap







- By accessing the server(s) with 3 concurrent clients we allow for concurrent GET/PUT/DELETE operations
 → performance improvement
- Due to the distribution and offloading of keys, the computational load is split among more than one single KV-Server
 - → better performance by distribution the load

