Evaluation of geo-distributed databases using the DeathStar Hotel Reservation benchmark

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1. Introduction

Geo-distributed database: A large database composed of multiple databases spread across large physical distances.

Database benchmark: A method of comparing database performance using a simulated workload.

DeathStar[1] Hotel Reservation: A benchmark which aims to represent a modern travel booking application.

2. Research Question

Research Question: How do geo-distributed database systems perform when benchmarked using the DeathStarBench hotel reservation scenario?

This evaluation will be done under several scenarios. namely:

- Network delays - Baseline
 - Packet loss
- Skewed - Sunflower
- Scalability

Performance will be measured using the following metrics:

- Bytes transferred - Throughput
- Latency
- Cost

2. Background

Database benchmarking is important to determine the strengths and weaknesses of a system before deciding on it's use.

Traditional benchmarks may no longer be representative of modern workloads, creating a mismatch between expected and actual performance, especially for geodistributed databases.

This project and the others in the group aim to implement several benchmarks in an appropriate environment for a comparison of some modern database benchmarks.

Which geo-distributed databases will be measured: - SLOG[4]

Detock	[2]

- Calvin[3]

- Janus[5]

4. Experimental Setup



5. Results





The skew scenario. Performance is not significantly affected.





---- Calvin --- SLOG --- Detock Median Latency Bytes 300 0.8 ള് 0.2 0.5 0.6 0.7 0.8 0.9 1.0 0.5 0.6 0.7 0.8 0.9 sunflower skev sunflower skew

The network latency scenario. Performance drops quickly. Calvin outperforms the others.





1500

1000 -

The scalability scenario. Throughput increases until fully saturated. At this point, it stagnates and latency, bytes, and cost increase.





The sunflower scenario.

Performance is not

significantly affected.





has results similar to the network latency scenario. Performance drops quickly and Calvin outperforms the others.





Detock -

Latency breakdown for the baseline scenario with large transactions.



Latency breakdown with small transactions. More time is spent coordinating, rather than idle.

6. Conclusions

We can draw three main conclusions from this data:

- Detock and SLOG perform better when there are few multi-home transactions

- Calvin and Janus perform consistently, and are best when there are many multi-home transactions.

- Calvin is the most cost-effective and performs best in poor network conditions.

Future work:

- Add more databases.

- Evaluate the effect of multi-partition transactions on performance.

7. Limitations

The main limitations of this work are the following:

- Client number may be too high and impacting latency.
- Latency breakdown is not detailed enough.

- DeathStar hotel benchmark does not have enough contention for skew or sunflower.

- Cross region latency is too low (not accurate simulation of real database configurations).

8. References

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