Scriber Slides

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Summary

- Motivation
 - Disk reads are expensive, so achieve low latency by maximizing the amount of data in memory

Key Contribution

- Compressed representation of the input data
- New algorithm for querying compressed data
- Evaluation
 - Able to fit 242 GB of data on a cluster with a total memory of 150 GB, whereas Cassandra and Mongo can fit a maximum of 23 GB

Pros

- 1. Low storage does not use secondary indexes
- Low latency in-memory data access are much faster
- 3. Higher throughput than other systems if indexes do not fit in memory
- 4. Their query algorithm is 2.3x faster on average in comparison to strawman

Cons

- 1. Expensive preprocessing step (4GB/ hour/core)
- 2. Updates cannot be in-place
- 3. Poor throughput for queries involving sequential scans
- 4. Does not provide fault tolerance
- Experiments designed arounds strengths of Succinct rather than representative of real world scenarios

Discussion

- 1. How would Succinct perform if the input size became so large that it cannot be stored just in memory?
- 2. Can we create a similar system for structured data?
- 3. Is there a way to bring together BlinkDB and Succinct?
- 4. What kind of workloads would Succinct not be useful for, given lack of support for in-place updates?
- 5. What if we have binary data instead of text files?