Discussion: BlinkDB

- First contribution: Organize sampling around query column sets - (a) A small set may "cover" all queries
- - Drawn from workloads
 - (b) MILP formulation which picks these column sets
- Second contribution: Determine on the fly, the qcs that gives the best "bang for the buck"
 - Neat idea of selectivity number of rows selected divided by the number of rows read

BlinkDB: My main takeaways

Where could the BlinkDB approach fail?

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- QCSes are not stable
- # of rare subgroups are high, dimensionality bad
 - Other approaches would also fail

- For example, if three groups have dimensionality 10000 each, the stratified sample of the cross-product could be GIGANTIC

What are the drawbacks of the BlinkDB system?

Let's talk about Optimization Techniques Query Class Repeated Queries

What are the drawbacks of the BlinkDB system: Optimization Techniques?

- Not clear
 - how to tune various parameters: K, M
 - whether the MILP is close to optimal
 - if techniques will apply to other workloads or case studies (two datasets is limited)

What are the drawbacks of the BlinkDB system: **Optimization Techniques?**

What are the drawbacks of the BlinkDB system: Query Classes?

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 Does not handle joins/nesting – Simple queries

Other Drawbacks

- A QCS either receives K or none at all. Is this wise?
 - Should depend on variance, distribution
 - If variance is small, a smaller sample may suffice
- Paper claims partial covering is OK if need be. Is that true?
 - Partial covering may lead to a biased sample – May completely miss some groups (incomplete answers)

Aqua vs. BlinkDB

- Very similar ideas for offline precomputed samples
- Aqua
 - Is query agnostic, will take full collection of group-by columns to construct stratified sample
 - May be too much
 - Broader class of queries
 - General enough to apply to joins (foreign key)

When would you use BlinkDB vs. Materialized Views?