Stats on intermediate tables

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The role of statistics in a query optimizer

- Cost estimation
 - What statistics to collect
 - □ When to collect them
 - □ Under which circumstances to re-optimize the query
- These statistics are then used to optimize the execution of the query

What we have seen

Runtime collection of statistics

- Dynamic collectors of Tukwila
 - Gather statistics about each operation dynamically
- Mid-Query Re-Optimization of Sub-Optimal Query Execution Plans
 - Filter and statistics collector
- Static collection from the start
 - SIT and MNSA
 - The strategies are for cardinality estimation and not for plan generation

Traditional optimizer:

- Quality of the query execution plan depends on the accuracy of cost estimates.
- Cost estimation depends on cardinality estimation of various intermediate results.
- Optimizer uses statistics built over base tables

Problem:

□ Propagation statistics lead to large estimation errors

Core Idea

SIT

- Histograms
- Materialized view
 - Do some intensive work of the results in advance
- Use existing SITs to model the distribution of tuples on intermediate nodes

MNSA

Select a small subset of SITs that are sufficient to increase the quality of the query plans by the optimizer

Traditional query optimizer

- Cardinality estimation using histograms
- Selection
- Join



Problem

propagation of statistics through predicates



Materialized view

What is a materialized view?

- Like a normal view yet in that it contains ACTUAL data
- Data for the view is assembled when the view is created or refreshed

Motivation

Saving the overhead of performing the work already done by the materialized view

SIT: Statistics on Query Expressions

Applies

- Histograms
- □ Materialized view
- SITs are created by the system and we assume that they are dependable sources
 Real computation
 - Approximate query processing

Implementation

By implementing a wrapper on top of the original cardinality estimation module of the RDBMS



Cardinality estimation using SITs

- Analyze the input
- Identify and apply relevant SITs
 - □ SIT-Sets should be applied to the query
 - If predicates of the query are not covered by the SIT, apply the auxiliary SITs
- Estimation and return the cardinality of the transformed query plan

MNSA: Magic Number Sensitivity Analysis

Goal:

- To select the most influenced subset of SITs that are sufficient to increase the quality of the query plans
- Approach: Consider workload information
 - Given a query workload and a space constraint
 - Find the set of SITs that fits in the available space
 - So that actual cost is minimized or substantially reduced

MNSA algorithm

- Incrementally identifies and builds new statistics over the base tables until no additional statistic is needed
 - Magic selectivity number (extreme predicted selectivities) to estimate the absence of statistics
 - Verifies whether the optimized query plans are toptimizer-cost equivalent
- Problem:
 - □ MNSA can not apply directly to the optimizer system

Independence strategy



Alternative to get the cardinality estimation : Extreme cardinality estimation

- Max and Min strategy
 - Get the max(min) number of tuples in the join results
 S.A < 10 (30 tuples)</p>



Selecting SITs

- Which generating query to use for some SIT
 - Max and Min cost differences should be significant
 - Selectivity estimation for the Min and Max strategies
 - Score of SIT
- Discard non-essential statistics



Conclusions

Benefits:

- □ Better performance
- Do not need to store and maintain materialized views but only build statistics over those views

Weakness:

- □ Where does the materialized view come from?
- Histograms
- □ The quality of SITs

Future Works:

Extending and evaluate the methodology ofr more complex queries and more complex statistics