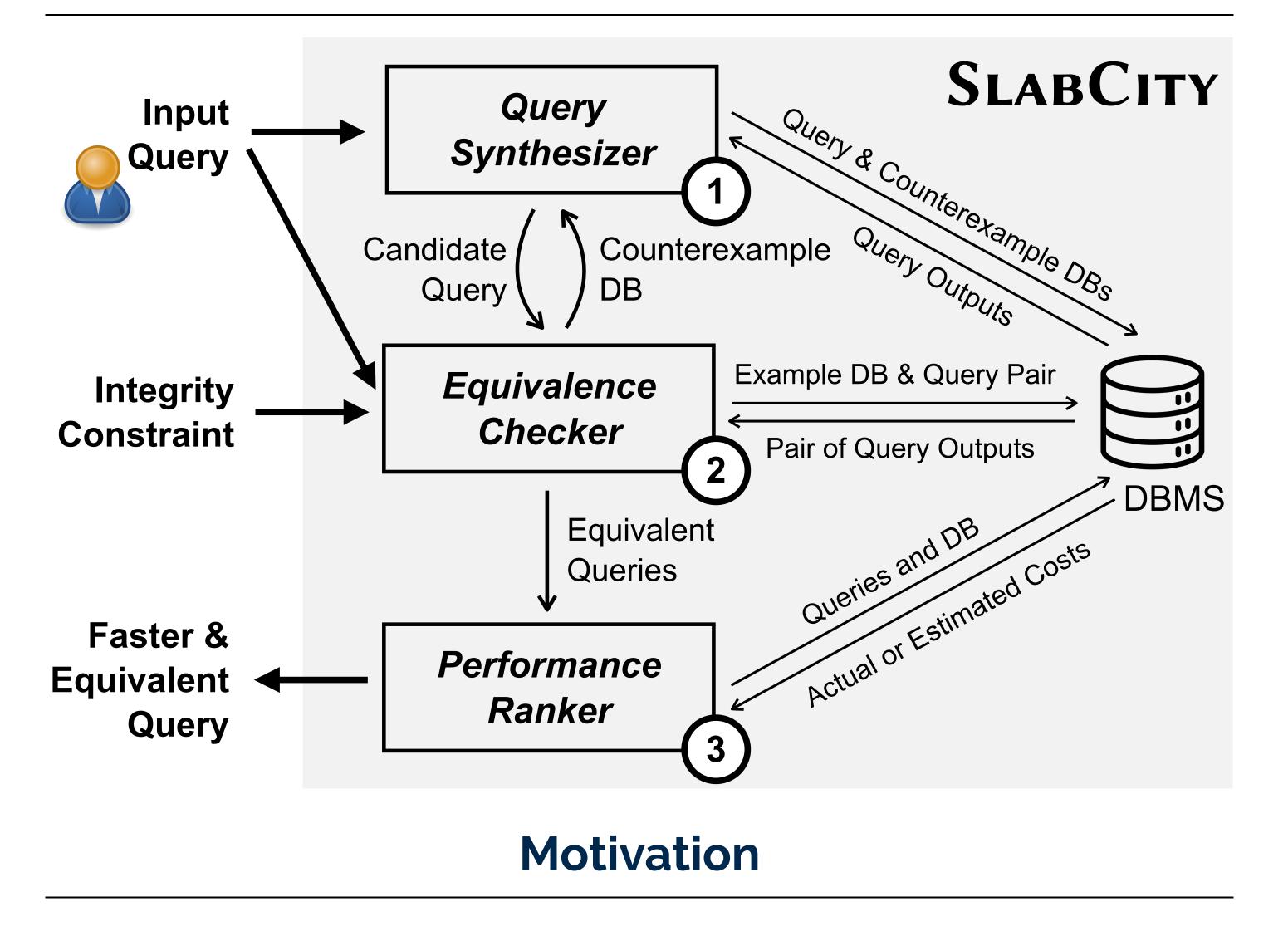


SlabCity: Whole-Query **Optimization using Program** Synthesis



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Workflow of SlabCity



Approach

Dataflows capture how data is computed and used

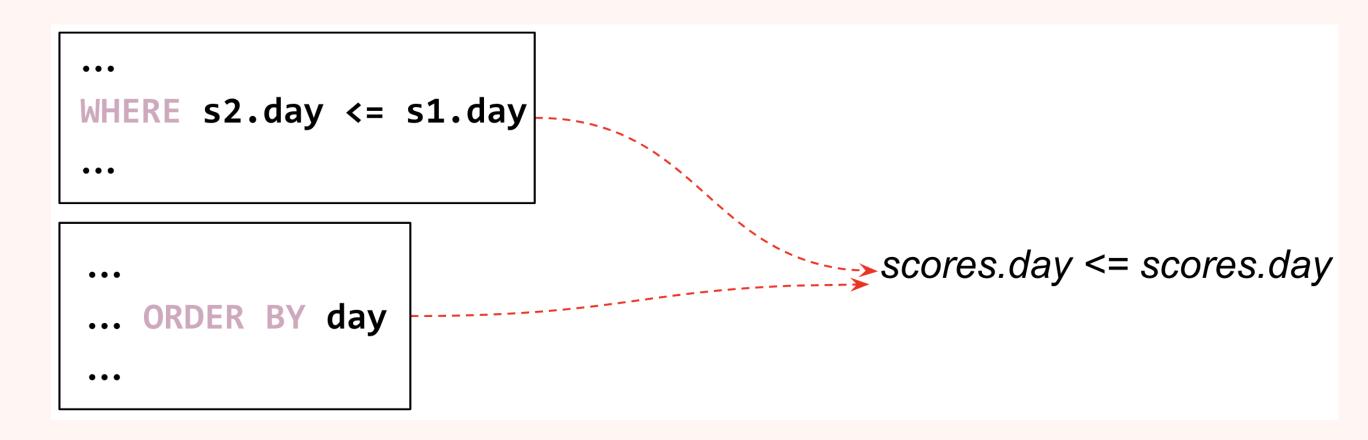
scores.gender

scores.gender = scores.gender

- Poorly-written queries are a major problem in the industry.
- Query rewriting transforms a query into another that is semantically equivalent but faster.
- Current query rewriting solutions usually rely on rewrite rules.
- However, it is impossible to enumerate every slow query pattern and design a rewrite rule for each of them.

scores.day scores.day <= scores.day scores.score -> SUM GroupBy(scores.gender, scores.day)

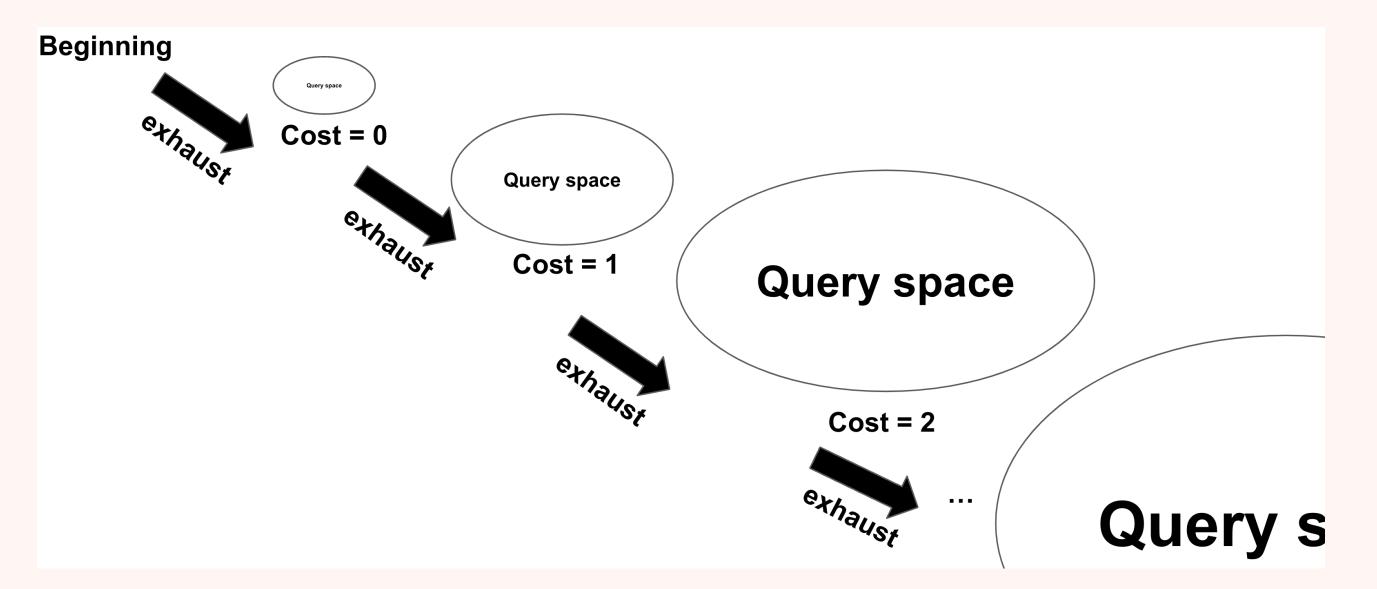
Equivalent queries often shares dataflows



Derive query cost using dataflows

Given a input query \mathbf{Q} , for any query component \mathbf{C} in search space, $cost(\mathbf{C}) = #$ of dataflows in **C** but not in **Q**

Search queries in ascending order based on cost



If there is a slow query pattern not covered by existing rules, how can we optimize it?

Let us look at two queries both written by real LeetCode users for calculating running total.

SELECT DISTINCT s1.gender, s1.day, SUM(s2.score) **FROM** scores **AS** s1 **JOIN** scores **AS** s2 **ON** s1.gender = s2.gender WHERE s2.day <= s1.day **GROUP BY** s1.gender, s1.day

SELECT gender,

day,

Q2

Q1

SUM(score) OVER PARTITION BY gender ORDER BY day **FROM** scores

- Q2 is much faster than Q1.
- People who wrote Q1 may not realize Q1 is very inefficient.
- People who wrote Q2 may not understand why someone would write a query like **Q1**.

Evaluation

Four **workloads**: LeetCode user submissions, Calcite rule testing pairs, TPC-H, TPC-DS

Two **baselines**: WeTune¹, LearnedRewrite²

- Optimization Coverage: SlabCity can optimize more queries.
- Latency Reduction: SlabCity can find faster queries.

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It's hard to design a rule rewriting Q1 to Q2 before you see Q1.

Key Contributions



- Propose the first synthesis-based query rewriting technique capable of whole-query optimization without requiring predefined rewrite rules.
- Define dataflows for SQL queries and exploit them for efficient query synthesis.
- Contribute a new benchmark by curating more than 1000 real-life queries from LeetCode participants.
- [1] Zhaoguo Wang, Zhou Zhou, Yicun Yang, Haoran Ding, Gansen Hu, Ding Ding, Chuzhe Tang, Haibo Chen, and Jinyang Li. Wetune: Automatic discovery and verification of query rewrite rules. In Proceedings of the 2022 International Conference on Management of Data, pages 94–107, 2022.
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