Efficient SQL Query Compilation with JIT Compilers

Saumya Verma, Sudip Chatterjee, Suprio Ray

Faculty of Computer Science, University of New Brunswick

{nivan, sverma3, sudip.chatterjee, sray}@unb.ca

Mark Stoodley, Calisto Zuzarte, Ian Finlay

IBM Canada {mstoodle, calisto, finlay}@ca.ibm.com

ECLIPSE OMR JITBUILDER

Eclipse OMR JitBuilder, developed using Eclipse OMR project, simplifies the task for a runtime system to incorporate a JIT compiler. It is easy to implement and with just a few hundred lines of code, it can target more than one platform.

MULTI-LEVEL INTERMEDIATE REPRESENTATION (MLIR)

MLIR is a modular and extensible compiler infrastructure. It provides a common intermediate representation that can optimize code generation across different hardware targets. It supports custom optimizations, transformations and analyses.

CONTRIBUTIONS

- Development of CasaDB based on in-memory database query compilation
- Conference Paper Published on query compilation
- Presented multiple posters
- Presented Efficient Compilation Of SQL Queries at CASCON x Evoke 2022
- Received best video award at CASCON x EVOKE 2022

GOALS

 Light-weight integration of Eclipse OMR JitBuilder into PostgresSQL-12.5 to JIT compiler operations

PROBLEM STATEMENT

- Evaluation of SQL Expressions and tuple materialization consumes a significant portion of query execution time.
- With relational database systems following an interpretation based tuple-at-a-time model of volcano style query execution, this process becomes more inefficient and slow.
- Recent research emphasizes the need to compile the query plans, but the integration of these techniques within a database system requires re-architecting the query engine.

80

60

40

20

SF4 SF8 SF16

Q1

SF4

SF8

SF16

Interpreted JIT compiled

Execution time for interpreted vs. JIT compiled

SF4

SF8

Q14



 A fast and multi-target backend generation technique using JIT compilers (Eclipse OMR and MLIR) supporting code generation for single node and distributed nodes.

APPROACHES

- Data-centric, run-time code generation model
- Use of pipelines and Intermediate Representation



std::vector<MaterializedTupleRef_V7_0_1_2_3_4_5_6_7_8> V7; arrow::Result<std::vector<MaterializedTupleRef_V7_0_1_2_3_4_5_6 _7_8>> res_V7

Code generation, compilation, and execution of TPCH Q6 using MLIR Vs PostgreSQL





*Contributors : Stephen A. MacKay, DeVerne Jones, Nithin Ivan, Shubh Sharma and Debajyoti Datta

cdb_arrow::arrow_read_MaterializedTupleRef_V7_0_1_2_3_4_5_6_7_8
("orders");

if(res_V7.ok()){V7 = res_V7.ValueOrDie();}
else {std::cerr << res_V7.status();}
std::cerr << "RETURNED from : orders" << std::endl;
std::cout << "RECORDS:V7 = " << V7.size() << std::endl;
tuplesFromAscii<MaterializedTupleRef_V7_0_1_2_3_4_5_6_7_8>("ord
ers");

auto end_2 = walltime(); auto runtime_2 = end_2 - start_2; std::cout << "pipeline 2: " << runtime_2 << " s" << std::endl; std::cout << "timestamp 2 end " << std::setprecision(15) << end_2 << std::endl; auto start_4 = walltime();

std::cout << "timestamp 4 start " << std::setprecision(15) <<
start_4 << std::endl;</pre>

std::vector<MaterializedTupleRef_V8_0_1_2_3_4_5_6_7> V8;

SQL query to Query Execution Plan (QEP) using pipeline

