

From SQL to Chat:

How to Revolutionize Enterprise

Data Analysis with NVIDIA



Overview

SQL-to-Chat Debut in Complex Enterprise-level Business Analysis Scenarios

- LUI conversational interaction: just chat with data agent like an old friend!
- Low-latency in Response: let conversational interaction not a dream!

Demo



Motivation

Highly Complex Business Domain, Deep Understanding Required



Complex schema

- 100k+ fields
- Semantic ambiguity
- Complex field parse



Diverse Demand

- Multi-dimension analysis
- For sales, product, R&D users



Frequent updates

- 200 times per month
- 10k+ fields invalid

Chat-like Data Analysis

Little prior knowledge required

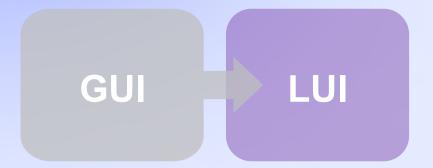
- User demand in natural language
- SQL auto generation

Instant response

- Chat-like experience
- Speed-up: minutes to seconds

Solution

Little prior knowledge required



Text-to-SQL on LLM

- Business Understanding:
 - Schema Alignment, Personalization
- Model Optimization: RAG、SFT、MoE

Instant response

Machine speed

Conversation speed

Hardware-Software Collaborative Acceleration

Software:

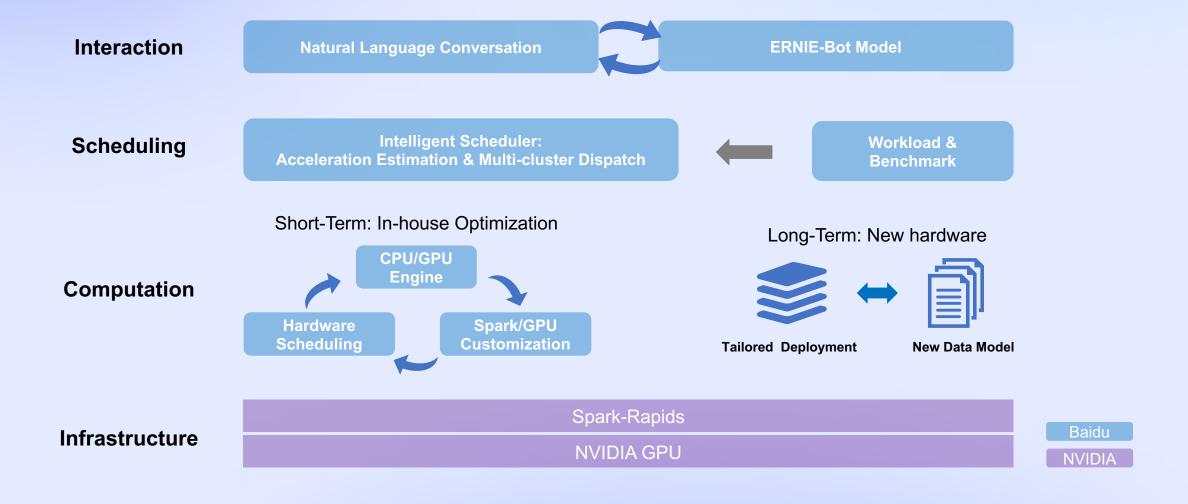
Spark-Rapids application

Hardware:

GPU/SSD customization

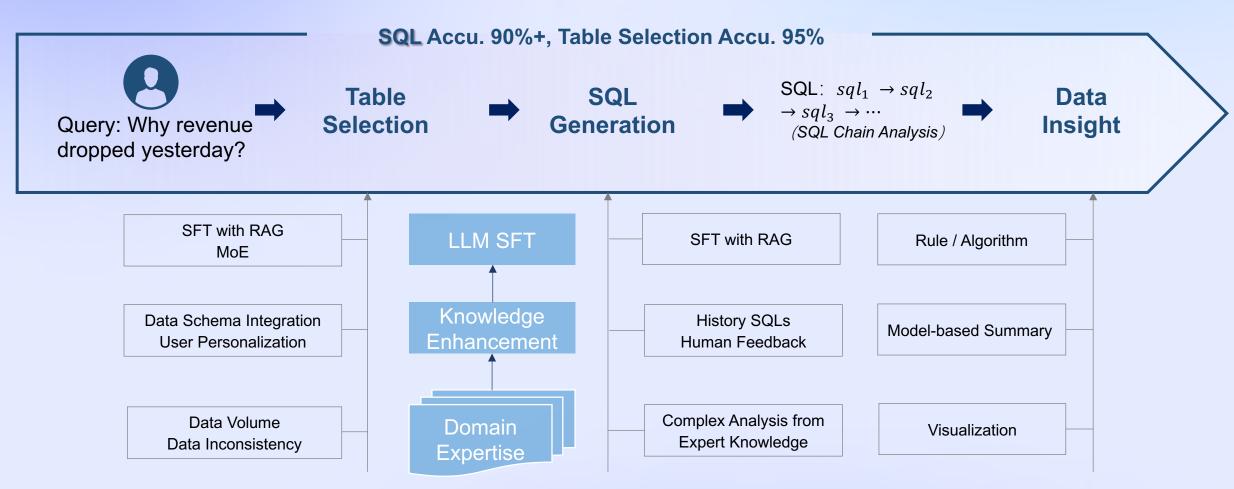
Solution | Architecture

- First Adoption of Conversational Data Analysis
- In-depth Collaboration with NVIDIA on Spark Rapids (up to 5x IMPROV.)



Solution | Revolutionize the Interaction with LLM

Build World-Class Text-to-SQL capability to fulfil complicated business demands based on



Computational speed is crucial for conversational interaction!

Spark-rapids is the ONLY choice for Conversational Interaction

Conversational Interaction O(seconds) **SQL** execution **Complicated Use Cases**

- Large Resource Consumption
- Diverse business requirement
- Complex data join scenario

Hardware-Software Co-Optimization

Software: Spark CPU

• Low ceiling, no 10x increase, cannot break the Von Neumann Architecture

SW/HW Co-op: Spark-rapids + GPU

- ~10x speedup
- Compatible with native Spark
 - Same SQL dialects
 - No Data Integration Costs
 - Suitable for Complex Scenarios

MPP: ClickHouse

- Slow Performance on Joins
- Complicated Schema Integration
- Different SQL dialects

Hardware: AEP + SSD

- Speedup only in I/O, but no 10x e2e
- High integration cost

Only Accelerate on Queries that can be accelerated!

Previous Attempts at Baidu

- Spark-rapids: not universally applicable
- Requires thousands of GPUs, low ROI

Failed reasons

- Accelerate everything and all pattern is not feasible
- Not clear about actual distribution of business and underlying data

SQL Identification & Scheduling

Our User Cases

- Diversity in Use Cases
 - 1k+ R&Ds & PMs, 1k tables
 - Execution time: o(Seconds) to o(Hours)
 - Data Tilting: CPU-bound or I/O-Bound
- Fluctuation
 - Seasonality
 - Customer changes

Success experiences

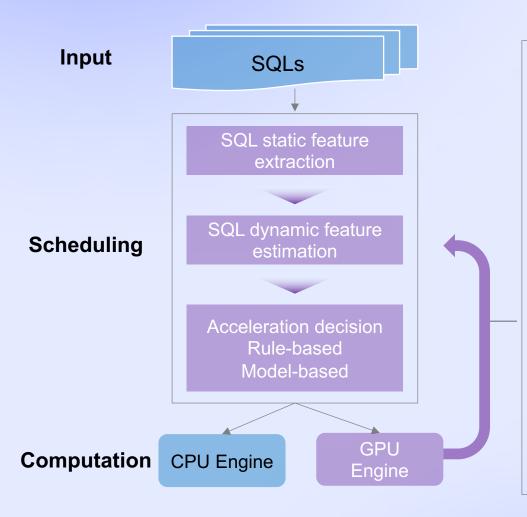
- Only accelerate SQL queries with high ROI and business importance
- The Scheduling layer

Spark-rapids: Excellent Speed up on diverse scenarios

Task Type	Reasons for Acceleration	Speedup Ratio
High Filtering Ratio	Efficient columnar processing with high parallelism on GPU	Exec-Time: 912s => 239sRatio: 3.8
Aggregation on Tilting Data	GPU compute performance far surpasses CPU on tasks with computational bottleneck	Exec-Time: 705s => 51sRatio: 13
Complex Joins	Better algorithm implementation, used GpuShuffledHashJoin to reduce overhead of sorting	Exec-Time: 237s => 46sRatio: 5.1

Heterogeneous computing

CPU/GPU Scheduling, new paradigm for Data warehouse



Scheduling on Isomeric HW

Rule-based Engine

- 1. Rule
- Static feature: SQL Ops, types
- Dynamic feature: Data volume, distribution
- 2. Characterization:
- Scan Partition Number and Data Size
- Data Discrimination, Compression Ratio

Model-based Engine

Spark-rapids Sweetspot

Data Size: 10s to 100s GBs

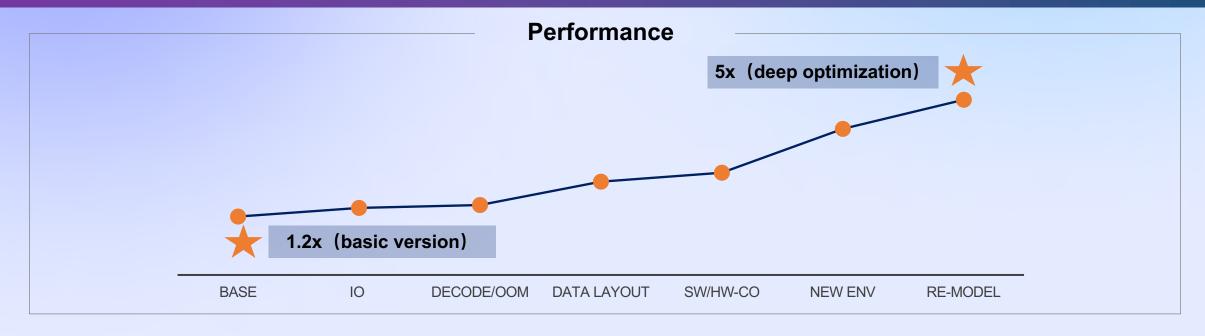
- Exec Optimization: Join/Agg/Sort/Window Op
- Typical Scenarios:
 - High-filtering Ratio
 - Data Tilting
 - Multiple Tables Join
 - Top N Sorting

Best Practices in Baidu

Deep Collaboration with NVIDIA to enhance Spark Rapids Ecosystem

Scenario / Issue	Solution	Related PR
20% data meets 80% demands	Data hot and cold tiered storage	
Data characters among difference scenarios	Suitable data layout for GPU batches	
OOM during scanning data with high compression ratio	Parquet sub-rowgroup reading	https://github.com/rapidsai/cudf/pull/ 14360
Performance bottleneck of Parquet: decompression & decode	Optimization on GPU Parquet decode More compression algorithm: Iz4_raw Hybrid utilization of CPU and GPU	
Feature & Bug-fix	Support new func: conv, parse_url Enhancing implementation: like, get_json_object	https://github.com/NVIDIA/spark- rapids/pull/8925 https://github.com/NVIDIA/spark- rapids/issues/10254

Conclusion



Business benefits

- First Adoption of Conversational Data Analysis
- Significant efficiency improvement
 - Revenue analysis: day -> seconds
 - Experiment optimization: hours -> minutes

Future Planning

Tailored Deployment on SW/HW

New Data Model for GPU Acceleration

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