

Interactive Querying with Spark SQL



Extracting information from data using SQL.

What about querying metadata of data?



Scheduled Query: Query NOT executed by a user.

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Interactive Query: Query executed by a user.

Scheduled vs Interactive Querying

	Scheduled	Interactive
Predictability	High	None
Results	Based on SLAs	Immediate
Syntax/Analysis errors	Low	High
ROI on Tuning	High	Low
Resource Efficiency Importance	High	Low

Data Scale at Pinterest

- 700+ PB in AWS S3
- 150K+ Data Compute Jobs per day (not including metadata only operations)
- ~20K Hadoop Nodes
- 1000+ Presto Workers
- 300K+ Hive Tables
- Everything in Cloud (AWS)

Query Engines at Pinterest

- Spark SQL
- Presto
- Flink SQL

Learn more about our Hive to Spark SQL migration

Advanced Migrations: From Hive to Spark SQL

Wednesday, June 29 @4:45 PM

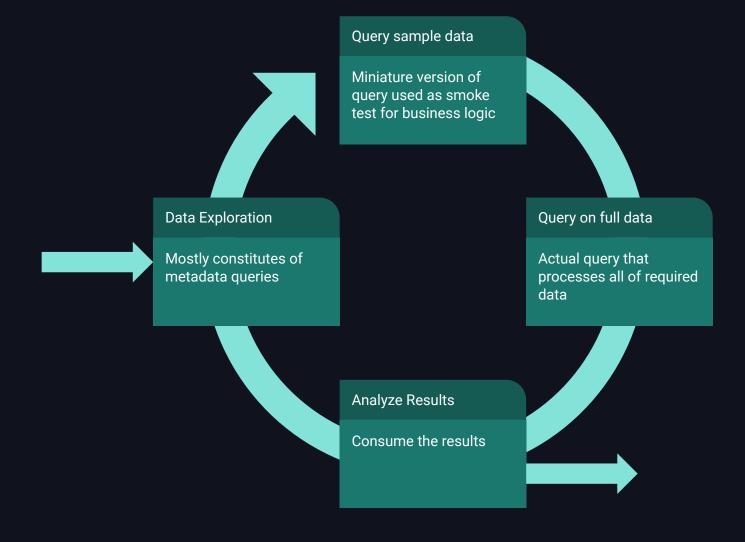
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By Zaheen Aziz



Interactive Querying

Usage pattern

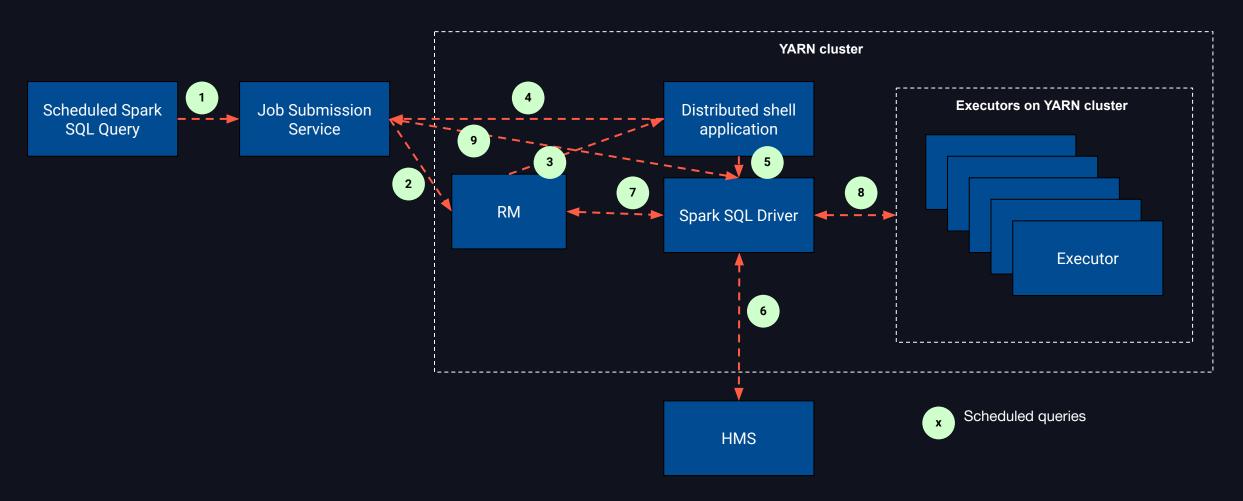


Interactive Querying

Requirements

- Seamless query submission
- Fast metadata queries
- Fail fast
- Easy access to results
- Error handling and tuning suggestions
- Auto error handling and retries

Scheduled Querying With Spark SQL



Interactive Querying

Requirements Check

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Architectural Choices

Distributed Shell Application

- Spark-sql CLI as dist shell application on YARN
- Waiting for container allocations
- Retrieving results is tricky
- Tracking statement-level progress is hard

Apache Spark Thrift Server

- Similar to HiveServer2
 - JDBC/ ODBC protocols
- No isolation between queries

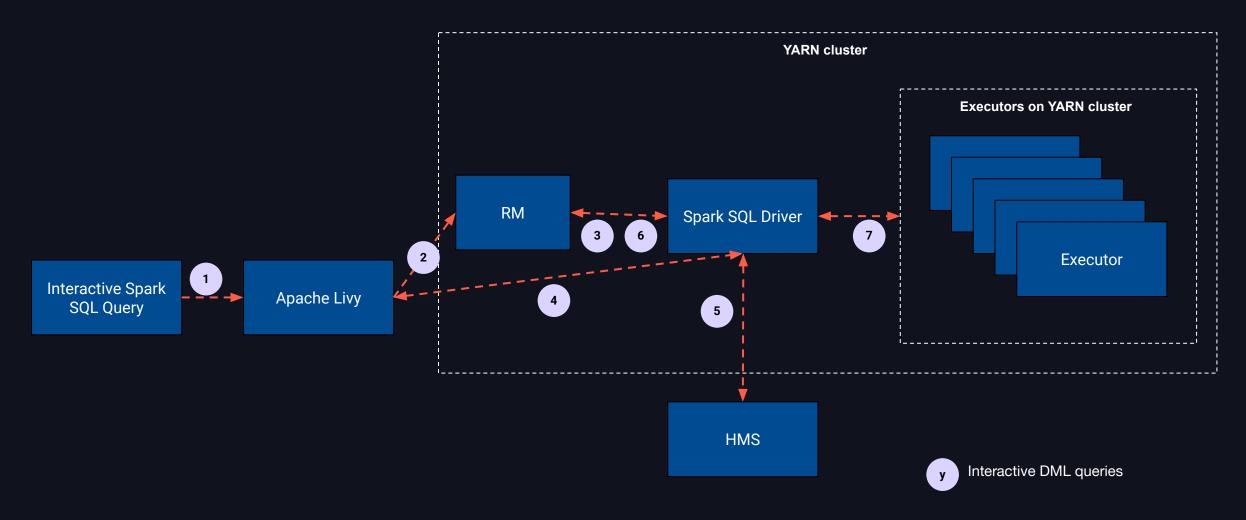
Apache Livy – Batch Sessions

- REST interface to Spark cluster
- Similar to spark-submit
- Multi-tenancy, high availability, and failure isolation

Apache Livy – Interactive Sessions

- Enables reusing sessions
- Multi-tenancy, high availability, and failure isolation

Querying With Spark SQL



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Seamless Query Submission

From existing query clients like querybook, jupyter, etc

 Generic DB-API 2.0 compliant Python client on top of Livy's REST API

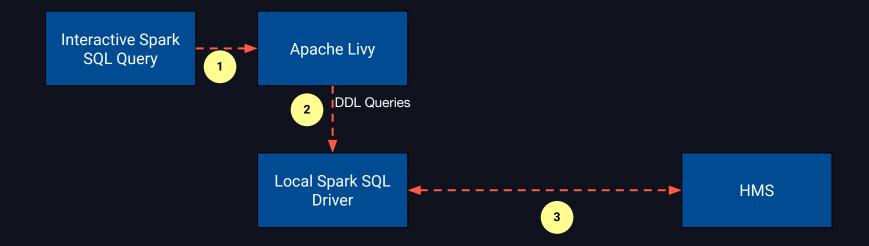
```
from bigpy.hive.livy.livy client import Connection
from bigpy.hive.livy.livy client import Cursor
cursor = Connection(livy_connection_urls='host1:port1').cursor()
cursor.execute('SELECT * FROM my awesome data LIMIT 10', async=True)
status = cursor.poll().operationState
while status == Cursor. STATE RUNNING):
  logs = cursor.fetch logs()
  for message in logs:
    print message
  status = cursor.poll().operationState
print cursor.fetchall()
```

Fast Metadata Queries

The problem..

- Waiting for container allocations
- Waiting for Spark Context to start up

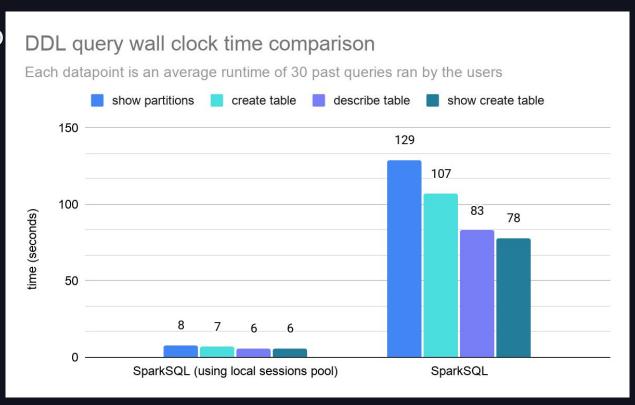
Fast Metadata Queries



Interactive DDL queries

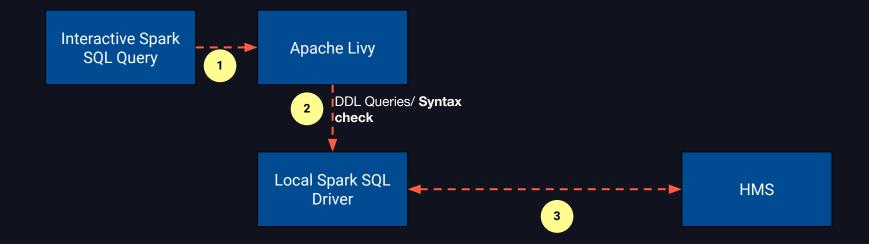
Fast Metadata Queries

- Pool is initialized on server startup
- Self-reliant, some of the features include:
 - Automatic garbage collection
 - Health monitoring
 - Asynchronous loading



With this design, we reduced query latency from 70 seconds to an average of 10 seconds (~6.3x improvement)

Fail Fast



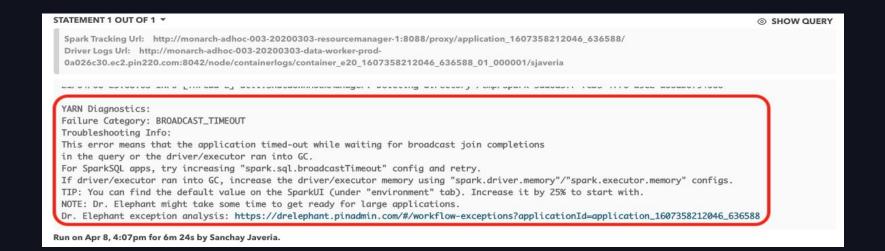
Interactive DDL queries

Easy Access to Results

- Livy limits the number of rows a query can return
- Needed to avoid Livy from choking on memory
- For large results, we redirect results to AWS S3
- Clients read results from S3 on query completion

Error Handling Recommendations

- Utilize YARN diagnostic and Spark error messages
- Fail YARN application based on last query's execution status
- Added an error classifier/ troubleshooting information to Dr. Elephant
- Clients retrieve troubleshooting information from Dr. Elephant APIs

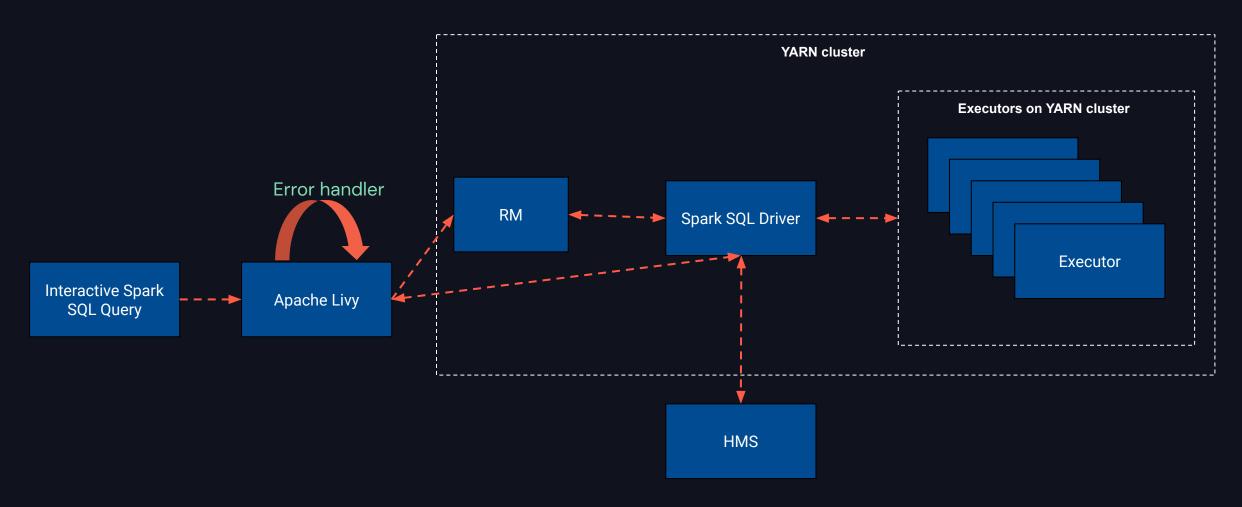


Resource Tuning Suggestions

- User applications are often too generous with memory
- Provide visibility into over/ under memory consumption



Auto Error Handling and Retries



Interactive Querying

Requirements Check

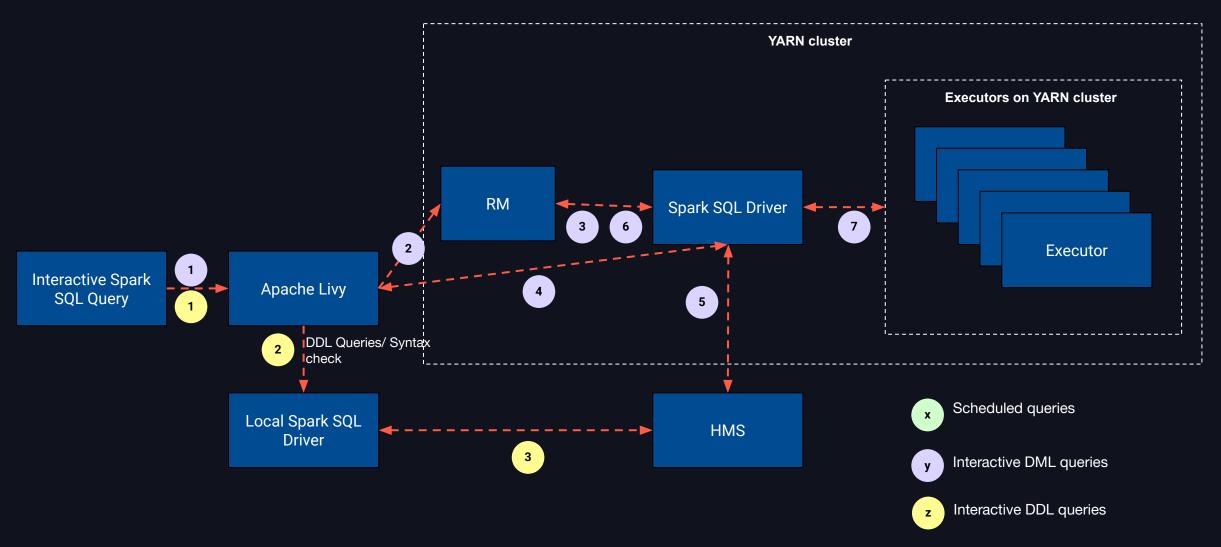
- Seamless query submission
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Interactive UDFs

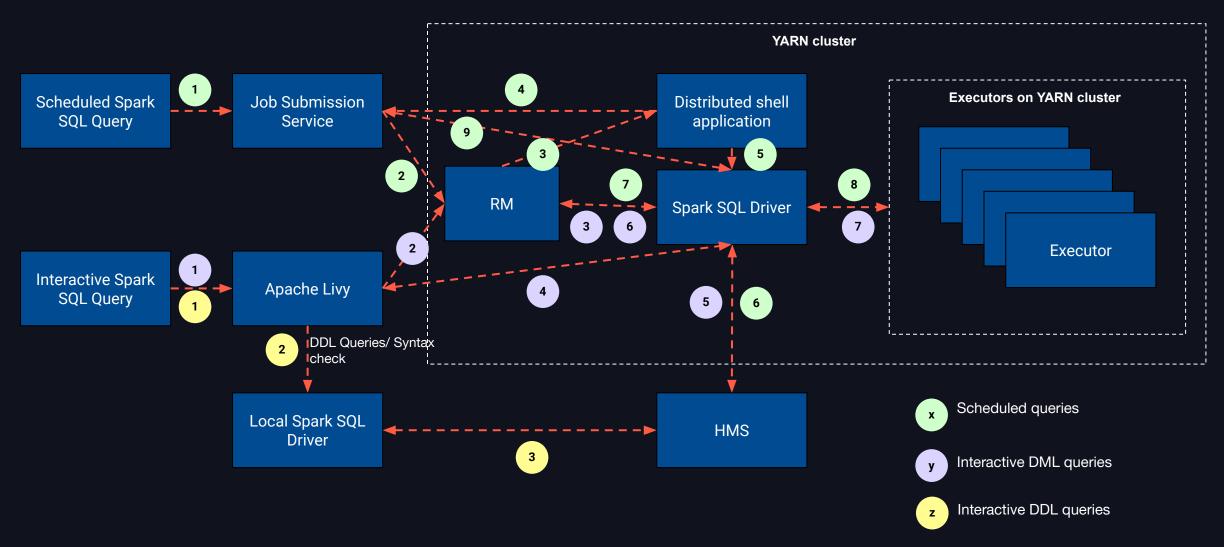
Allow users to author UDFs along with queries.



Uniteractive Querying With Spark SQL



Querying with Spark SOL Querying With Spark SQL



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Thank you

