# Augmented OLAP for Big Data

- Apache Kylin PMC Chair
- Microsoft Reginal Director & MVP



#### Luke Han | luke.han@Kyligence.io

Co-founder & CEO of Kyligence

#### **Strata Global Sponsor**

BOOTH #410



# About Luke Han



- Luke Han
- Co-founder & CEO at Kyligence
- **Co-creator and PMC Chair of Apache Kylin**
- **Apache Software Foundation Member**
- Microsoft Regional Director & MVP
- Former eBay Big Data Product Manager Lead





# About Apache Kylin

- Leading Open Source OLAP for Big Data
- Rank 1 from googling "big data OLAP"
- Rank 1 from googling "hadoop OLAP"
- Open sourced by eBay in 2014
- Graduated to Apache Top Project in 2015
- 1000+ Adoptions world wild
- 2015 InfoWorld Bossie Awards
- 2016 InfoWorld Bossie Awards







Inc. 2019.



# About Kyligence

- Pains in Big Data Analysis
- Kyligence's solution: Augmented OLAP
  - Video Demo
  - Benchmark
- Use Cases



# Kyligence = Kylin + Intelligence



- Founded in 2016 by the original creators of Apache Kylin
- CRN Top 10 Big Data Startups 2018
- Backing by leading VCs:
  - Redpoint Ventures
  - Cisco
  - CBC Capital
  - Shunwei Capital
  - Eight Roads Ventures (Fidelity International Arm)
  - Coatue



- Global Offices:
  - Shanghai
  - Beijing
  - Shenzhen
  - San Jose
  - New York
  - Seattle

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# Trusted by Global Leaders

#### Most of them are Global Fortune 500















# VIVO OPO Lenovo



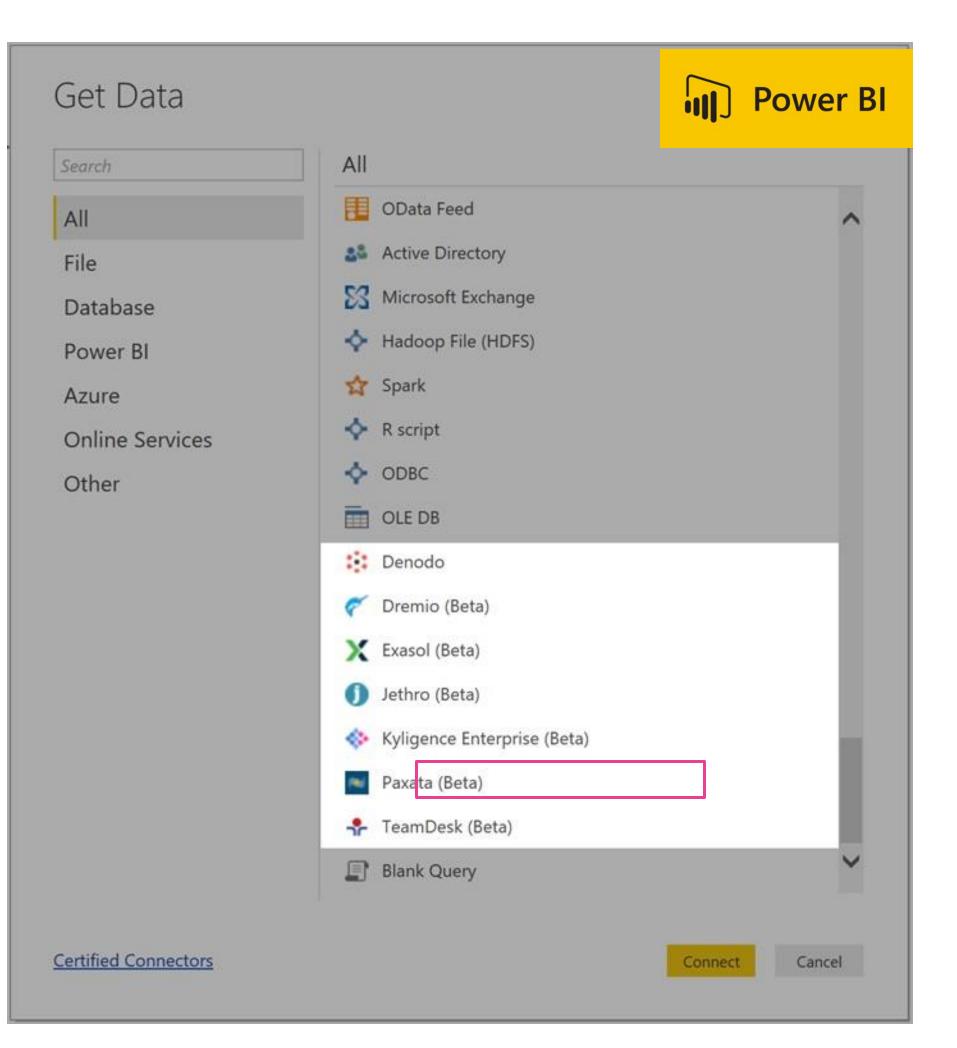












Inc. 2019.



 About Kyligence Pains in Big Data Analysis Kyligence's solution: Augmented OLAP • Use Cases



# Let's talk about photography story...



https://technave.com/data/files/mall/article/201812271418327393.jpg



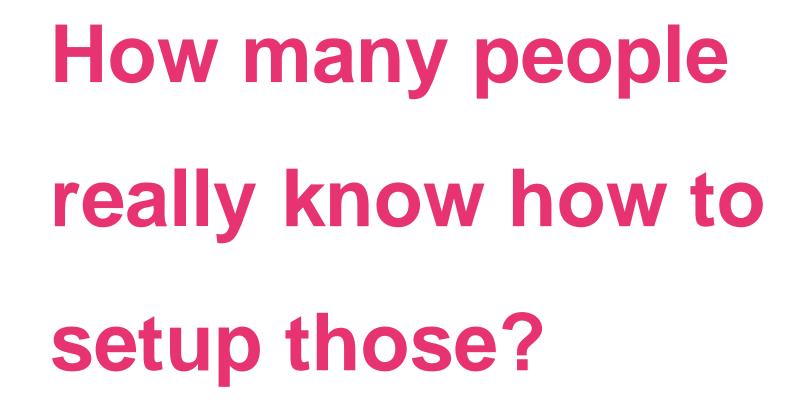
Inc. 2019.

#### Let's talk about photography story... $\blacklozenge$



Shooting Mode Exposure Control







# Let's talk about photography story...

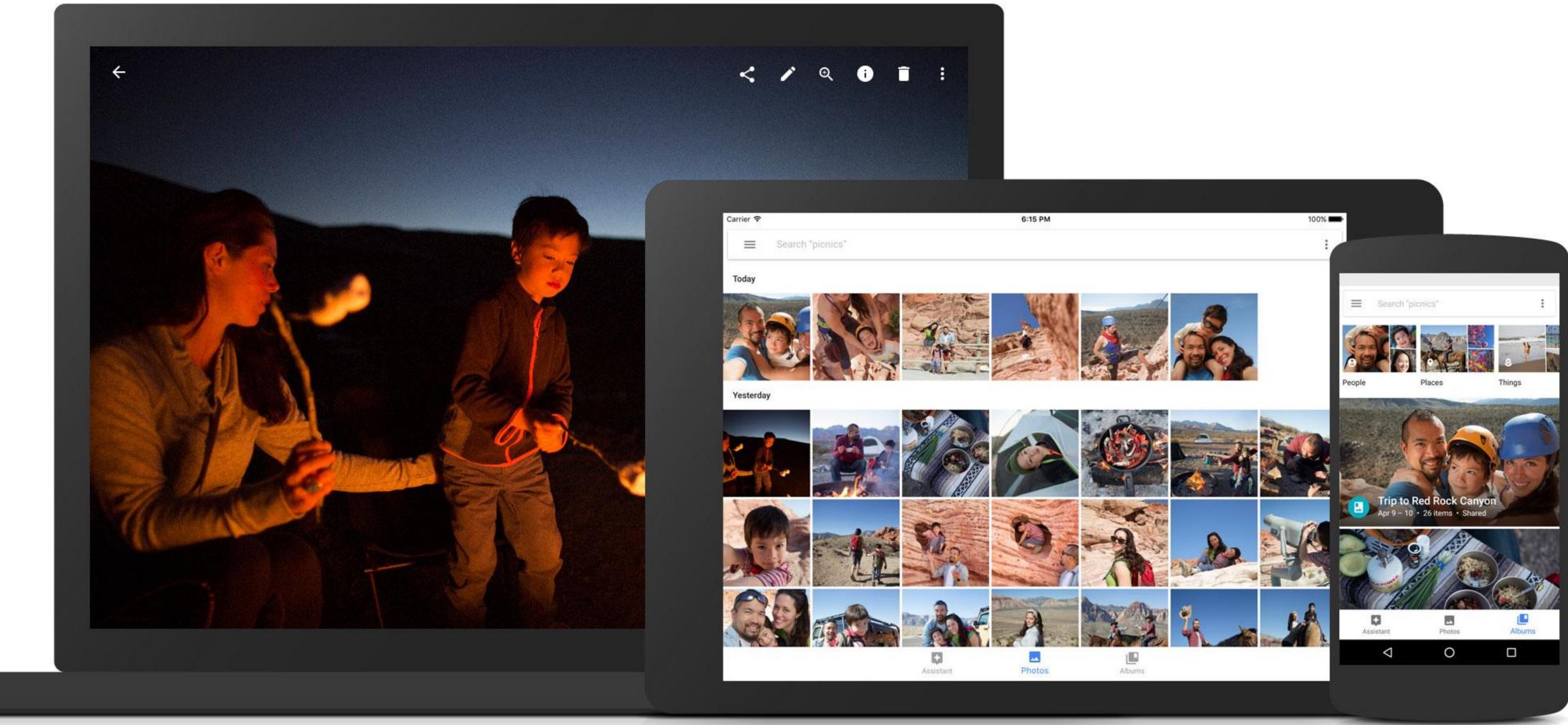


https://technave.com/data/files/mall/article/201812271437395703.jpg

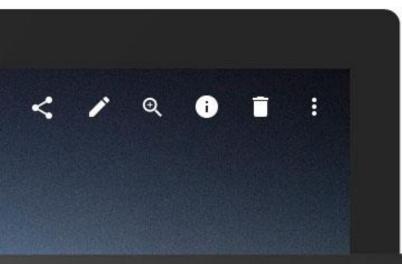


Inc. 2019.

# Let's talk about photography story...

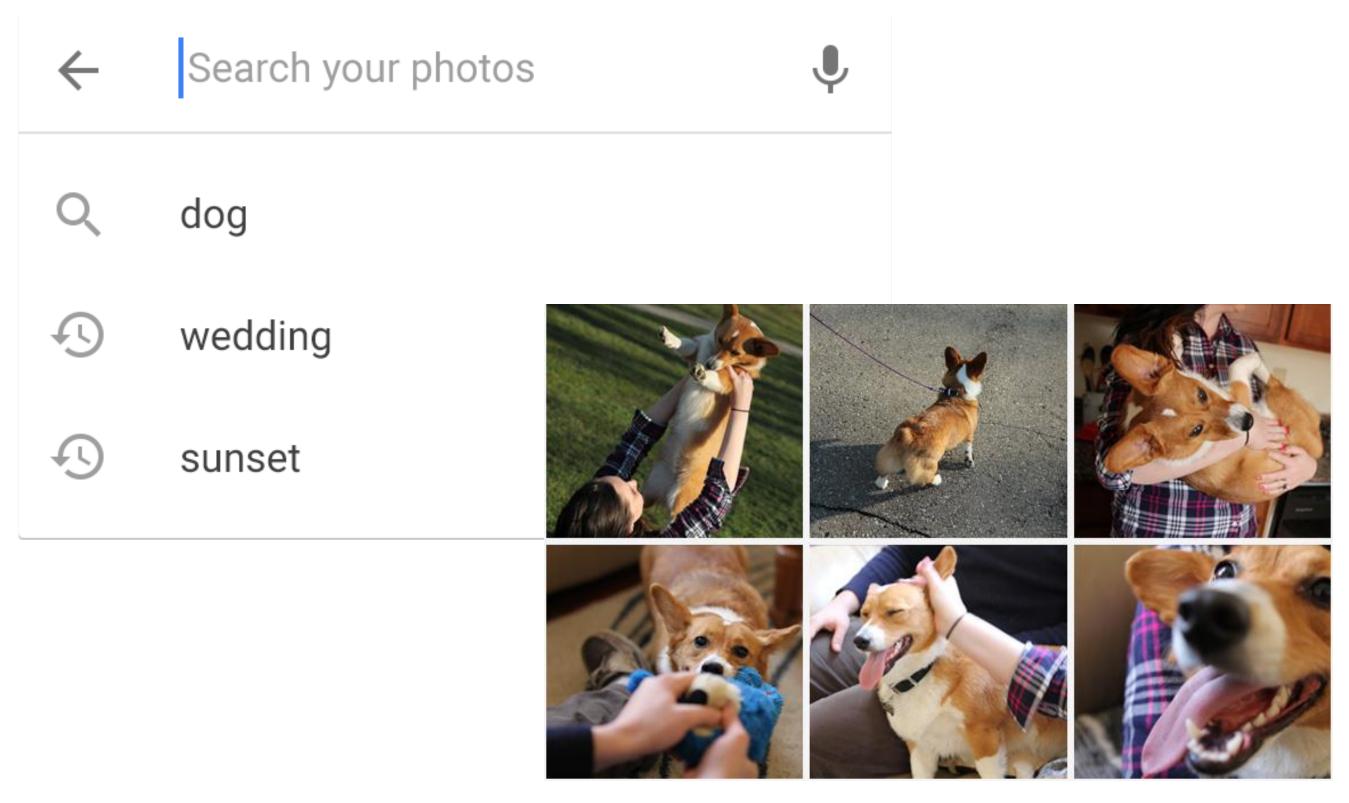






Google Photos

#### Let's talk about photography story... $\blacklozenge$







# How do you manage your 100,000+ photos?

Google Photos



# how about your enterprise



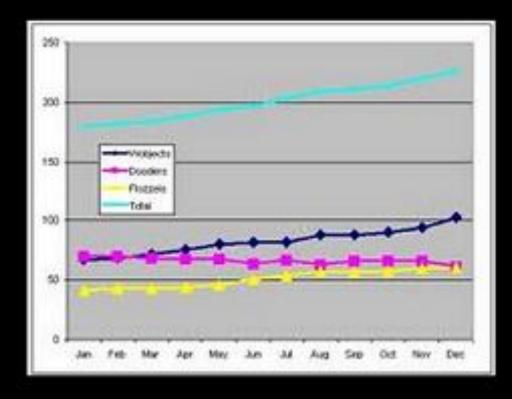
# Then...

# clata?

# Data Scientist

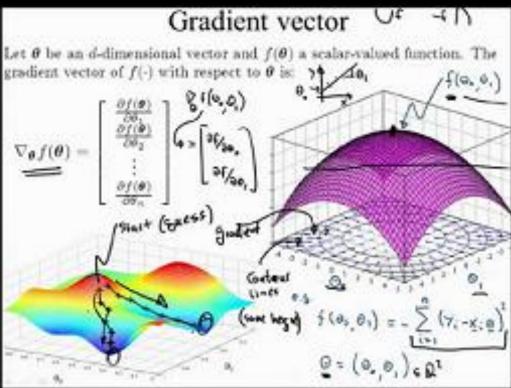


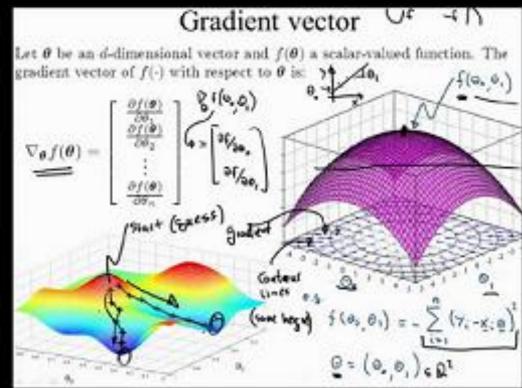
#### What my friends think I do



What my boss thinks I do







https://www.sintetia.com/wp-content/uploads/2014/05/Data-Scientist-What-I-really-do.png



What my mom thinks I do



What society thinks I do

What I think I do



What I actually do

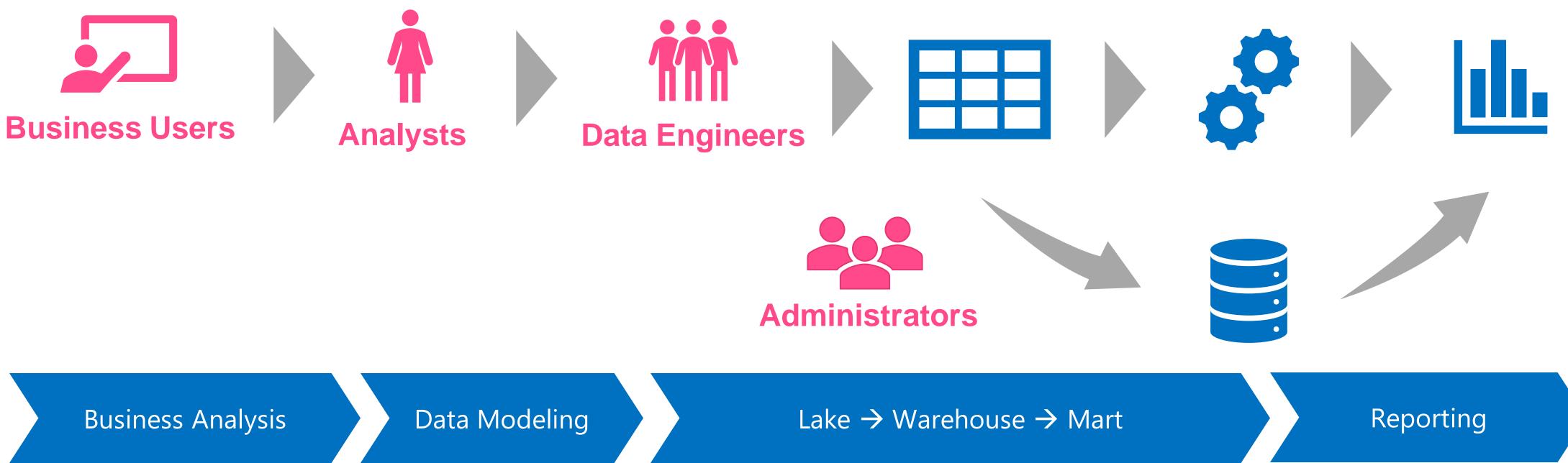
# Fast and Changing **Analysis Demand**

VS



# **Slow and Heavy Big Data Operations**





•

# **Slow Time-to-Insight:**

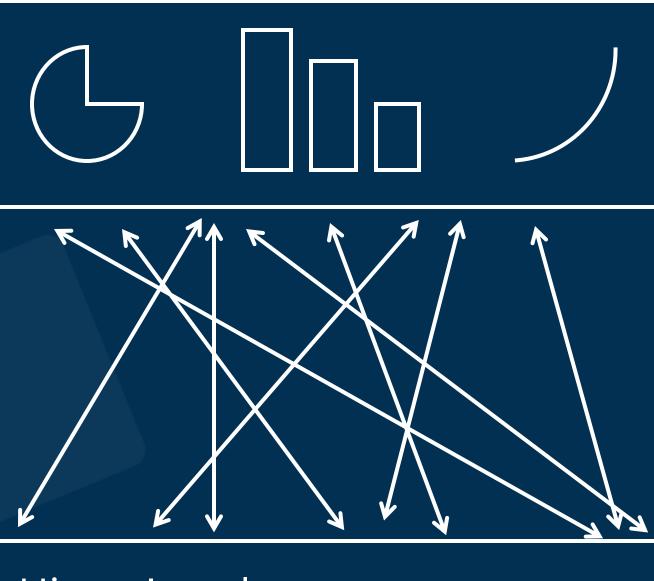


High Cost :  $\mathbf{\Psi} \mathbf{\Psi} \mathbf{\Psi}$ 

## $\blacklozenge$

# Pains in the "Throw in some People" Approach

#### Presentation Visualization



Data Lake

Hive Impala Spark SQL Drill MapReduce Spark .....



#### Time-to-value Pain

Weeks of waiting breaks the "online" promise.

## **Collaboration Pain**

Hard to reuse asset across teams. Each team fights their own path.

#### **Resource Pain**

Hard to scale. Where to find so many skilled big data engineers?



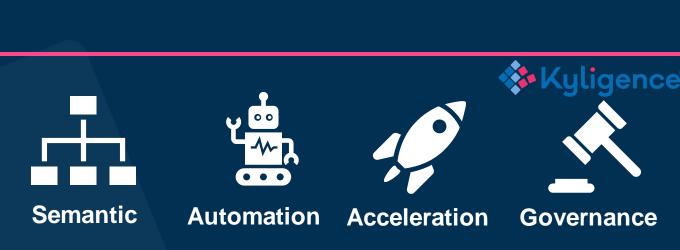
# About Kyligence Pains in Big Data Analysis Kyligence's solution: Augmented OLAP Use Cases



# Throw in some Intelligence!

Presentation Visualization

Augmented OLAP Data Mart



Data Lake

Impala Spark SQL Drill Hive MapReduce Spark . . . . . . .



### Let a system replace the people.

- Transparent SQL Acceleration
- On-demand Data Preparation
- Interactive Query Performance
- High Concurrency
- Centralized Semantic Layer

Faster time to market. Stay "online".

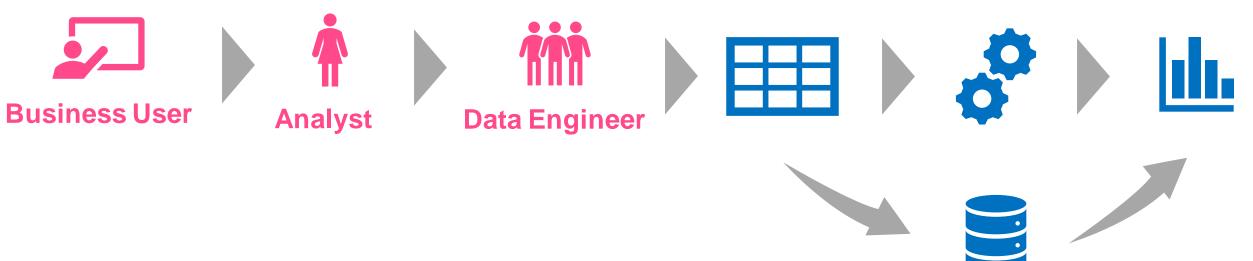
# A Learning OLAP System



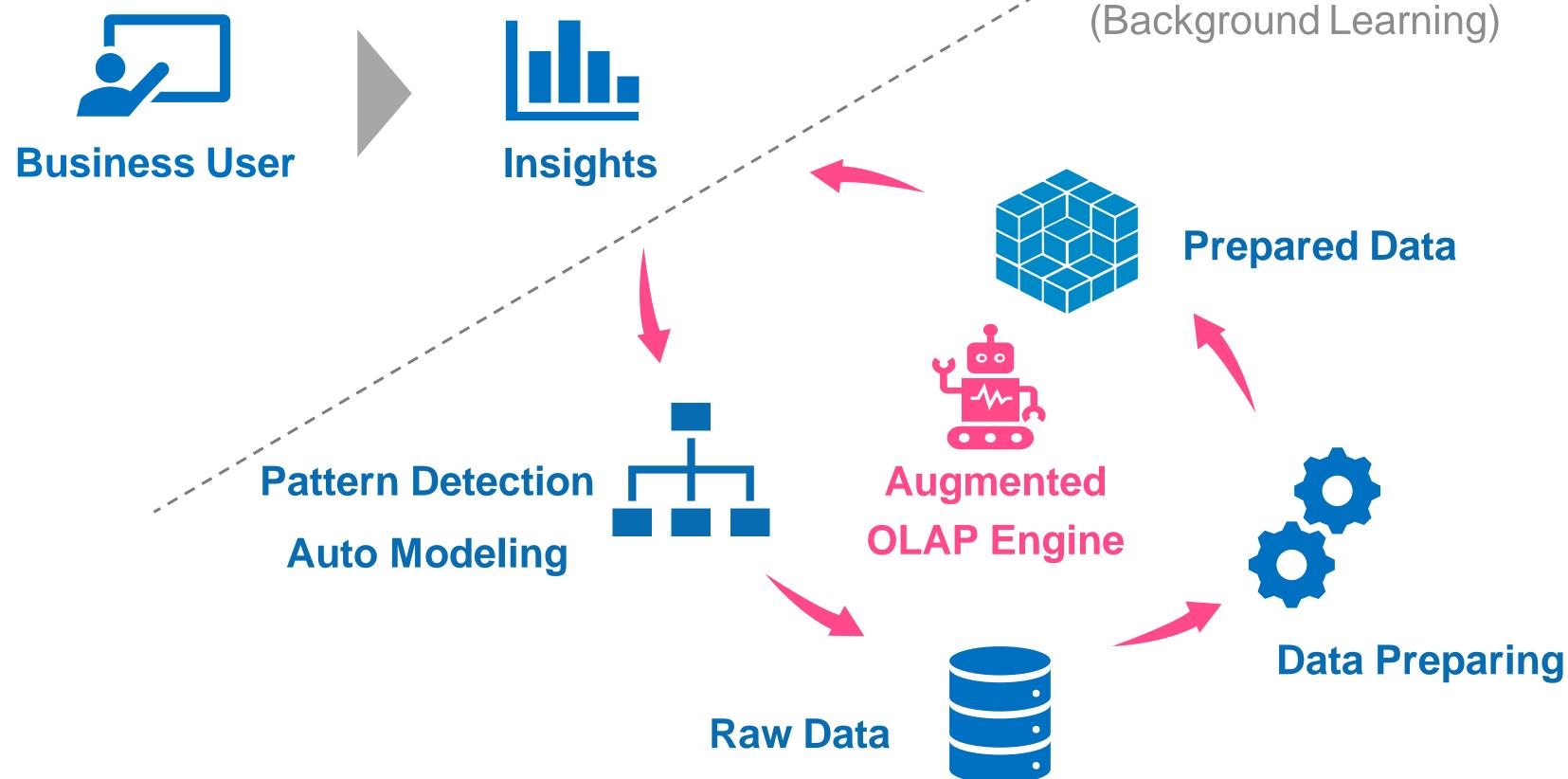
VS







# A Learning OLAP System

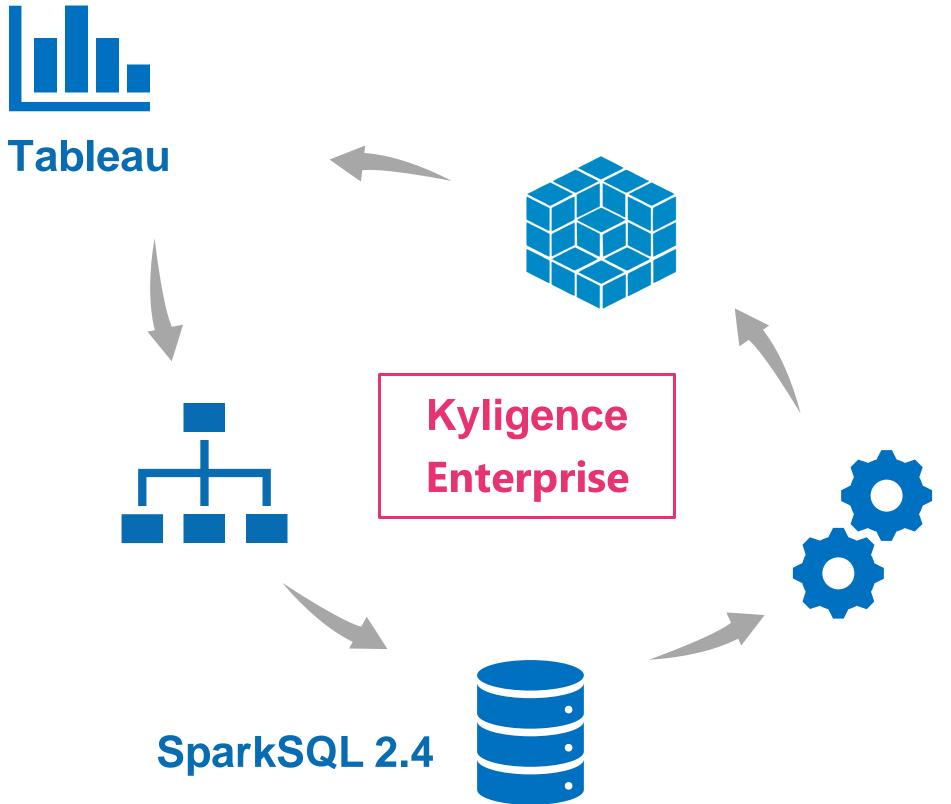




#### (Background Learning)



















How to improve the first slow exploration?

What if the analyst operates differently the second time?

More comprehensive performance benchmark?







Inc. 2019



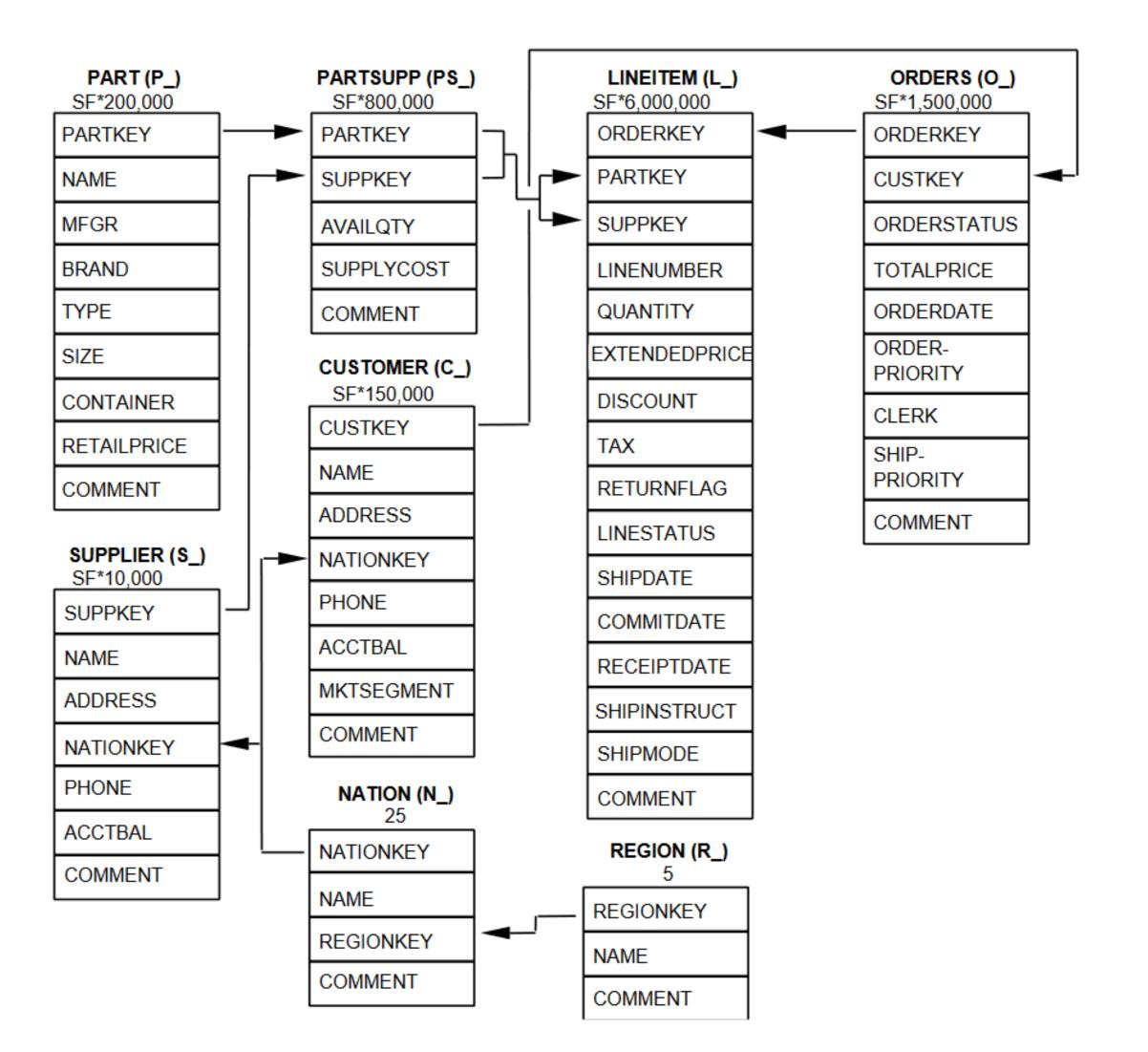
## **TPC-H Benchmark**

- Examine large volumes of data
- High complexity queries
- Answers critical business questions
- 22 decision making queries

#### E.g. The Shipping Priority Query

retrieves the shipping priority and potential revenue of the orders having the largest revenue among those that had not been shipped as of a given date. Top 10 orders are listed in decreasing order of revenue.



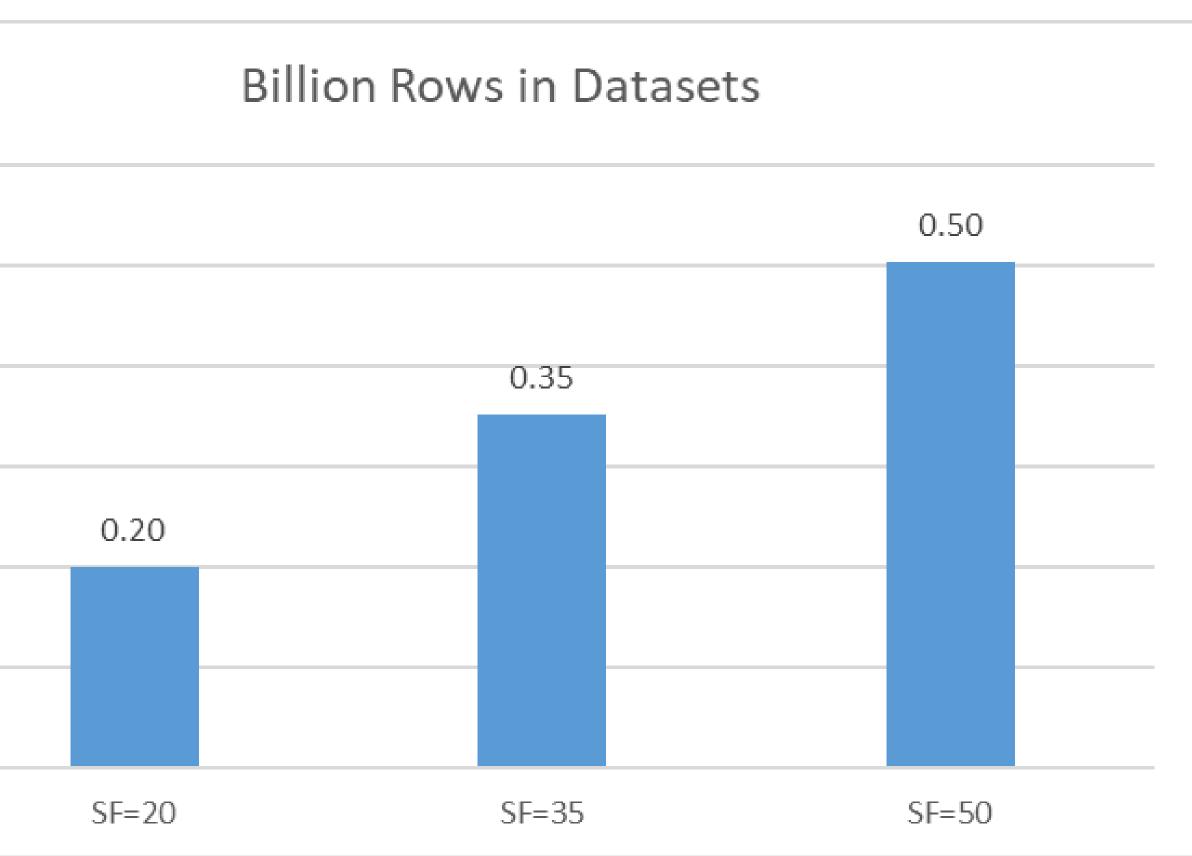


# Kyligence Enterprise 4 Beta vs SparkSQL 2.4

| To see the trend as data grows  | Billion |
|---|---------|
|   | 0.60 —— |
| <ul> <li>3 datasets</li> <li>Scale Factor = 20, 35, 50</li> </ul>                       | 0.50 —— |
|   | 0.40 —— |
|   | 0.30 —— |
|   | 0.20 —  |
| <ul> <li>TPCH_SF1: Consists of the base row size (several million elements).</li> </ul> | 0.10    |
| • TPCH_SF20: Consists of the base row size x 20.  | 0.00 —— |
| • TPCH_SF35: Consists of the base row size x 35.  |         |
|   |         |

• TPCH\_SF50: Consists of the base row size x 50 (several hundred million elements).





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#### Hardware Configurations

## Same 4 physical nodes

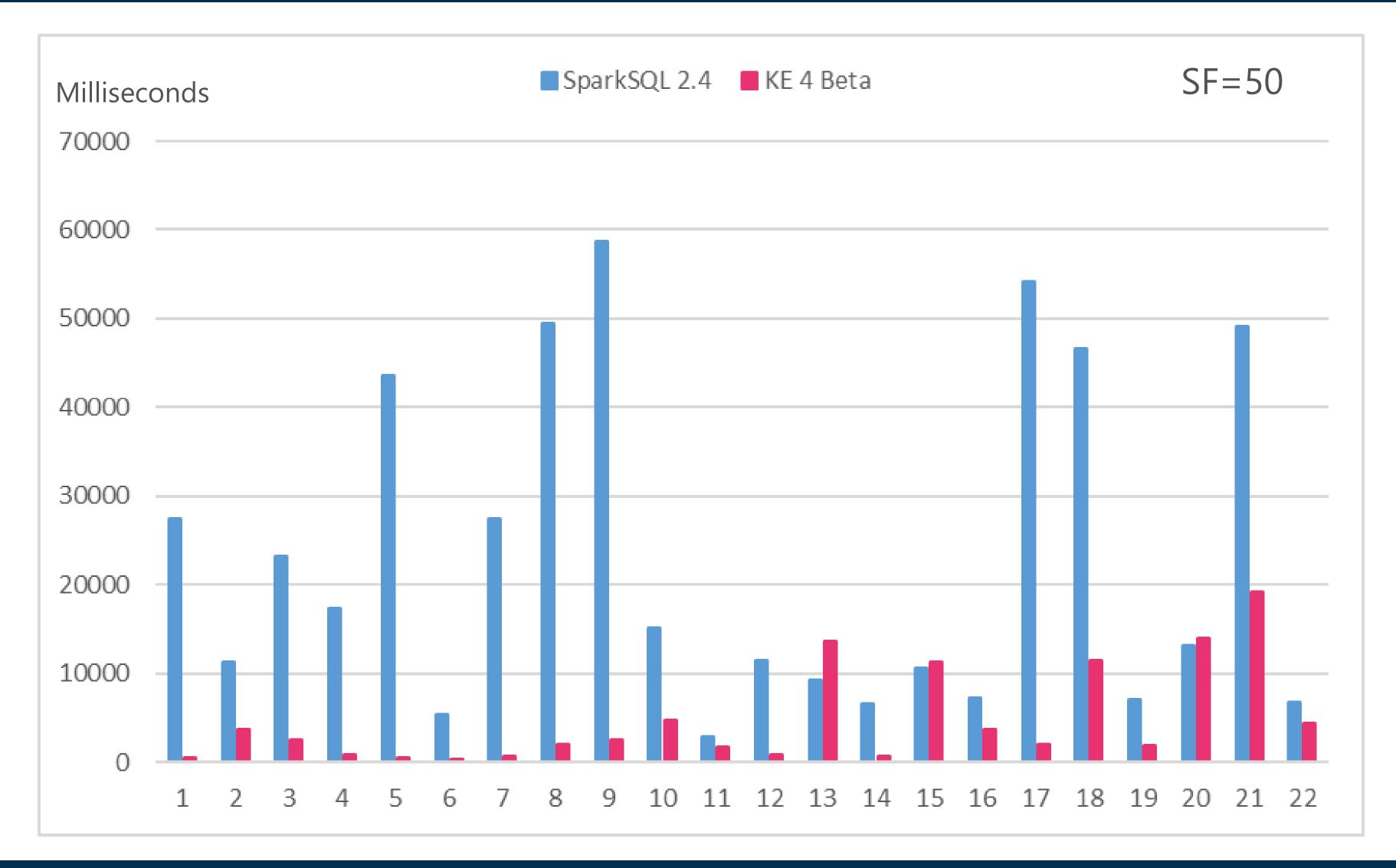
- Intel(R) Xeon(R) CPU E5-2630 v4 @ 2.20GHz \* 2 -
- Totally 86 vCores, 188 GB mem —

Same Spark configuration for both KE 4 Beta and SparkSQL 2.4

- spark.driver.memory=16g
- spark.executor.memory=8g -
- spark.yarn.executor.memoryOverhead=2g \_
- spark.yarn.am.memory=1024m \_
- spark.executor.cores=5 -
- spark.executor.instances=17 -







**Wyligence** 

# Query Response Time | KE 4 Beta vs. SparkSQL 2.4

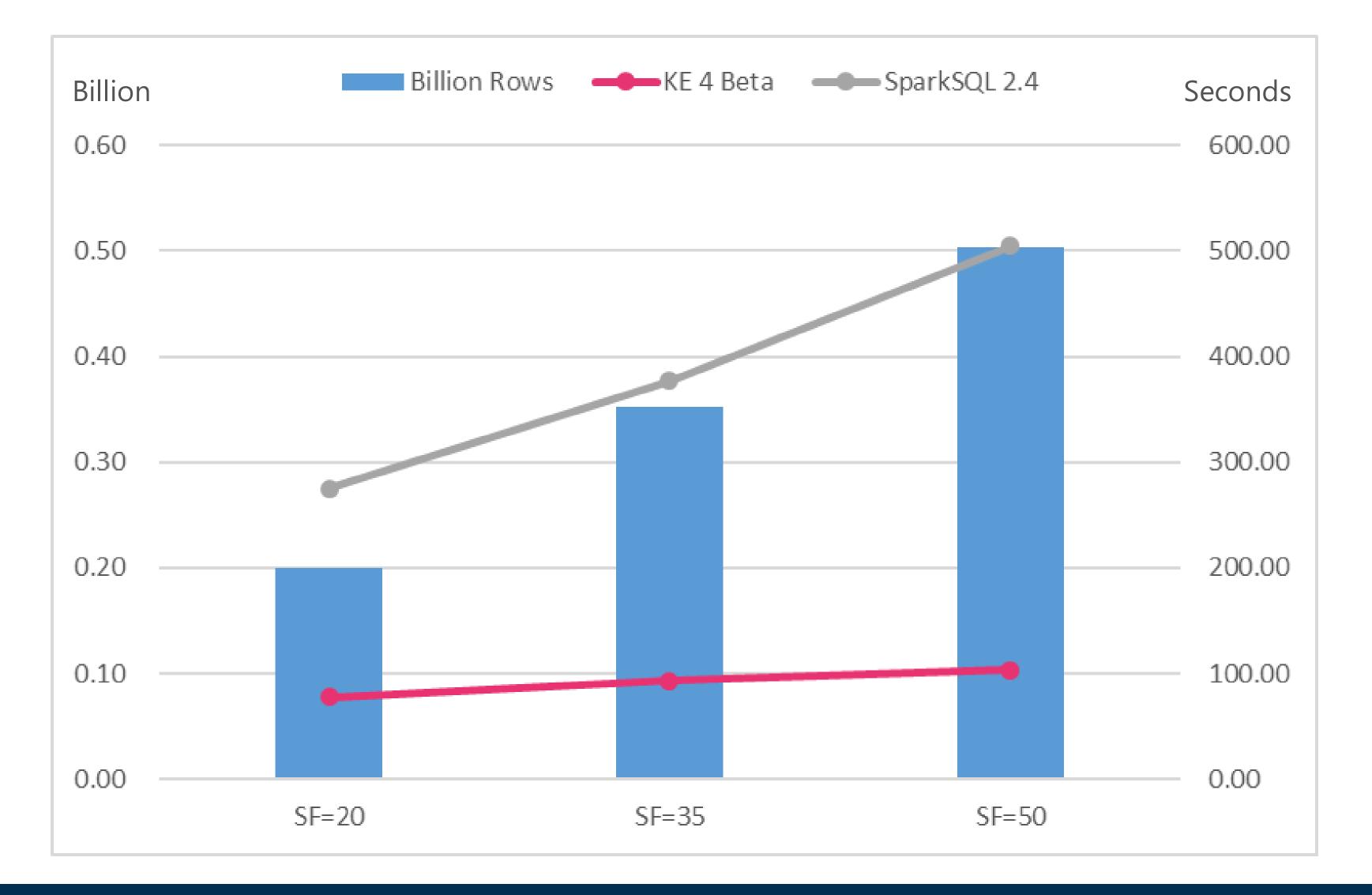
TPC-H 22 queries

For each dataset

- Run each query 3 times
- Record the average time
- No warm up

Lower is better.

# Total Response Time | KE 4 Beta vs. SparkSQL 2.4



**Wyligence** 

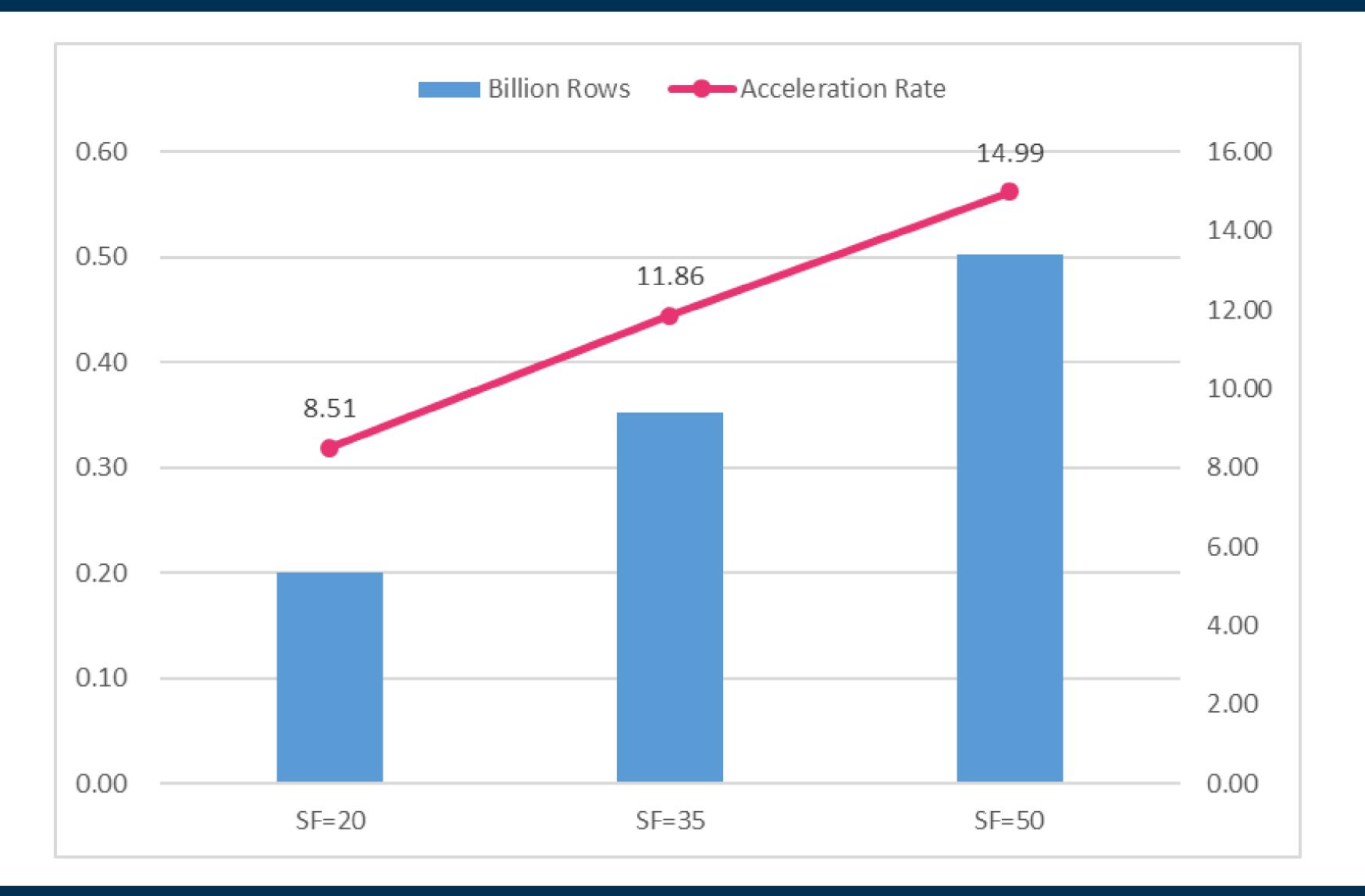
Total response time is the sum of 22 queries' response time.

Compare over the size of datasets and feel the trend.

Scale out for the future.



# 



**Wyligence** 

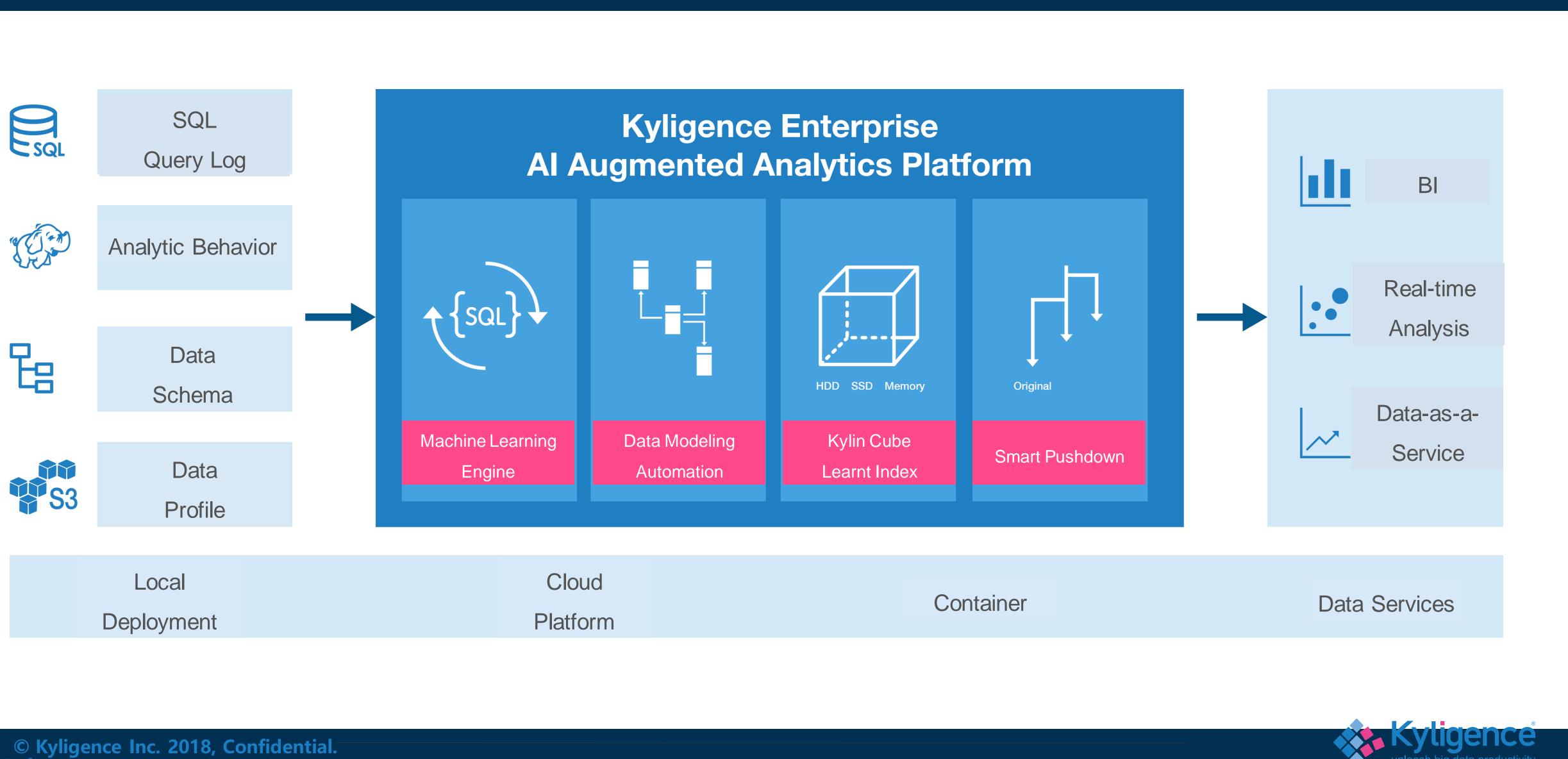
# Avg. Acceleration Rate | KE 4 Beta vs. SparkSQL 2.4

Acceleration Rate

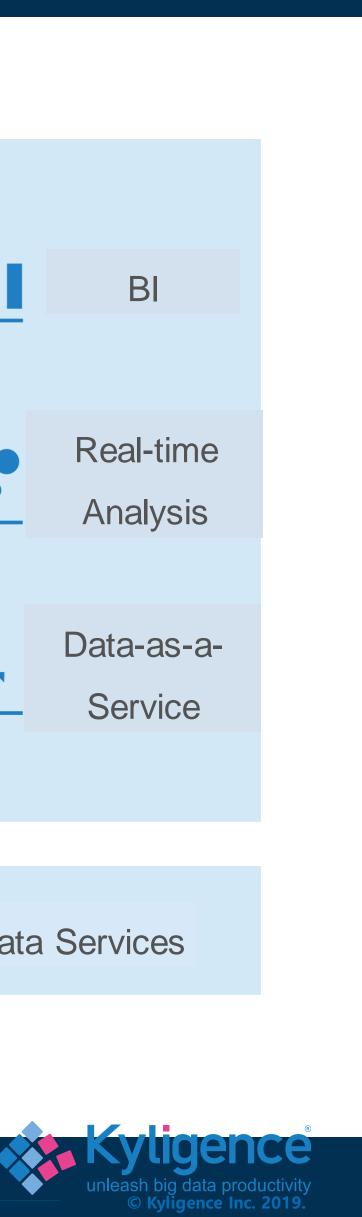
= SparkSQL time / KE time

Take average of the 22 and compare over size of datasets.

# **Al-Augmented Analytics Platform**









# About Kyligence Pains in Big Data Analysis Kyligence's solution: Augmented OLAP Use Cases



# Use Case: IBM Cognos Replacement

# One Kyligence Cube for 800+ Cognos Cubes





Org. Daily Cube

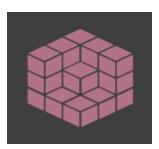


Org. Monthly Cube

Geo Scene



Shanghai Merchants





Merch. Daily





Channel Merch. Monthly Cube Monthly Cube



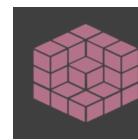
Zhejiang **Merchants** 



**Channel Daily** 

Cube

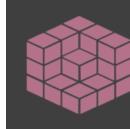
Anhui **Merchants** 



**Region Daily** Cube



Region Monthly Cube



Guangdong **Merchants** 

800+ Cognos Cube, 1000+ ETL jobs



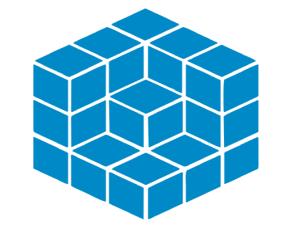












Data: 300+ B Records Merchants: 10+m Cards: 10+B

Card Transaction

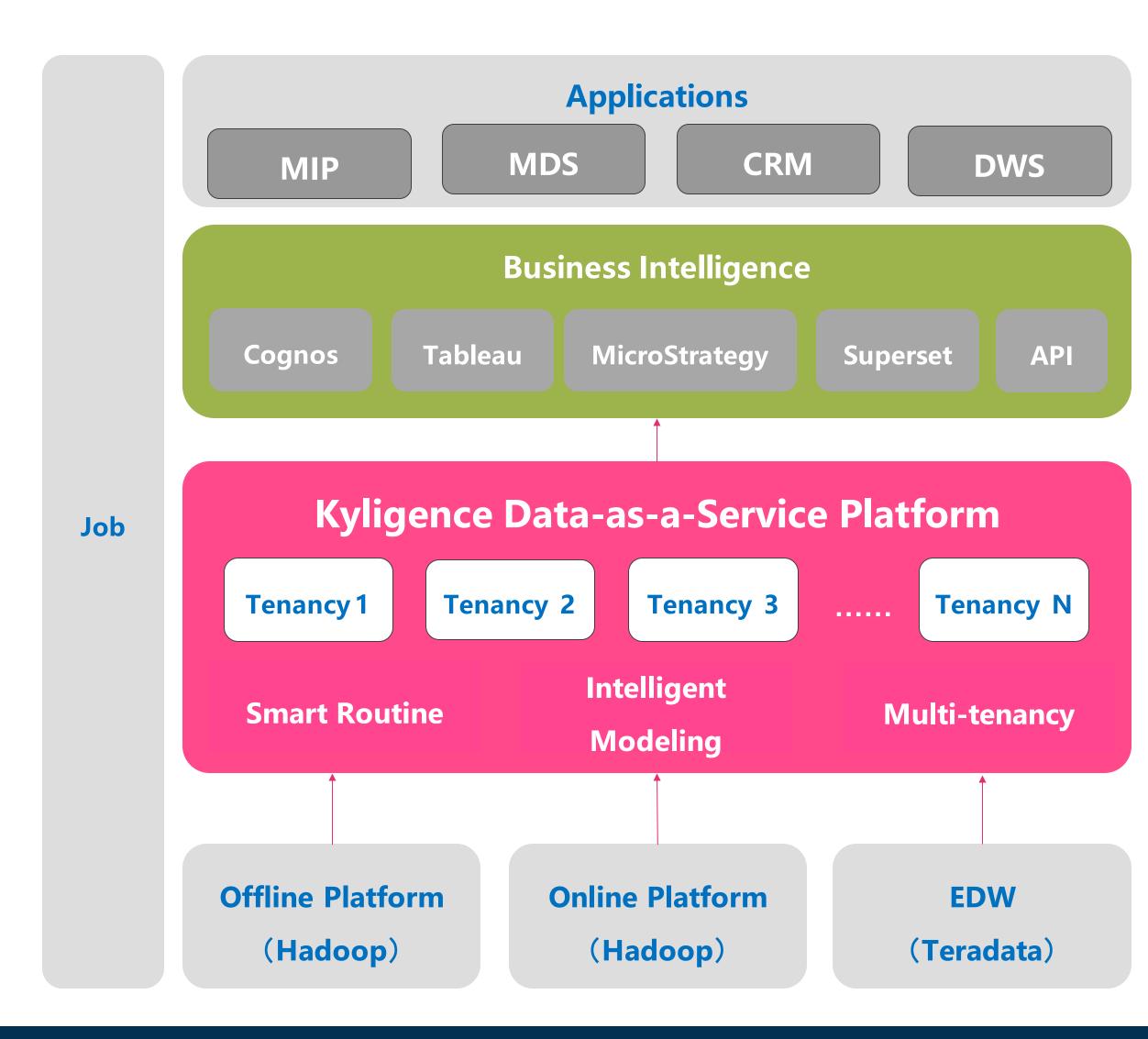
Dimensions: 167

Measures: 20



# Use Case: Data as a Services Platform

**Security** 



**Kyligence** 



In the past, due to the limitations of our previous multi-dimensional analytic tool, we faced challenges of constrained time range in queries......We are considering leveraging multi-dimensional data cubes to replace a number of fragmented legacy tabular reports in more business units, so that we can provide better analytic services to our business users."

-- Laments Wu Ying, VP of CMBs Development Center,









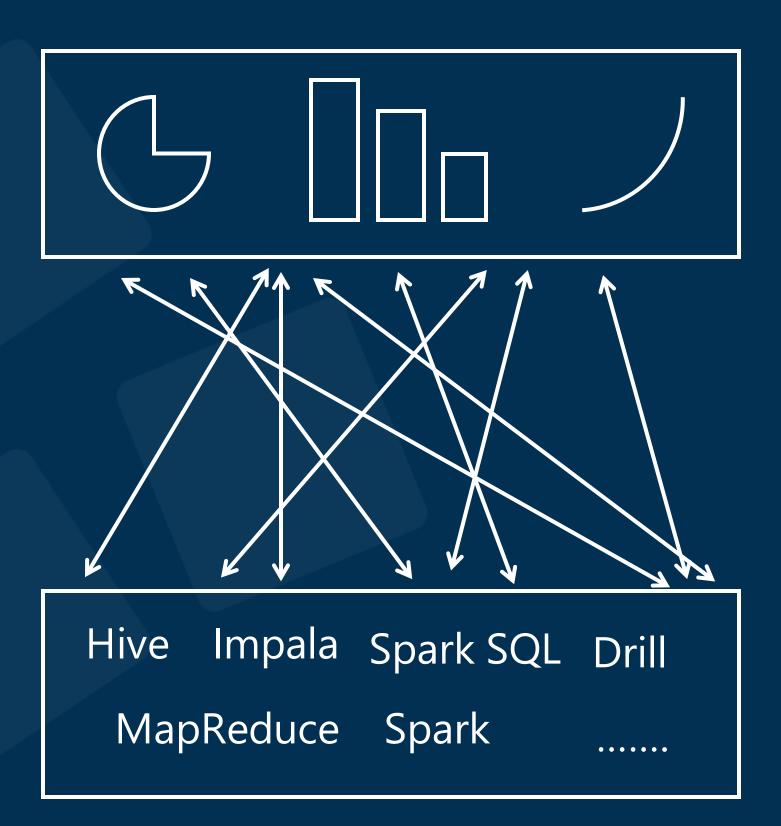




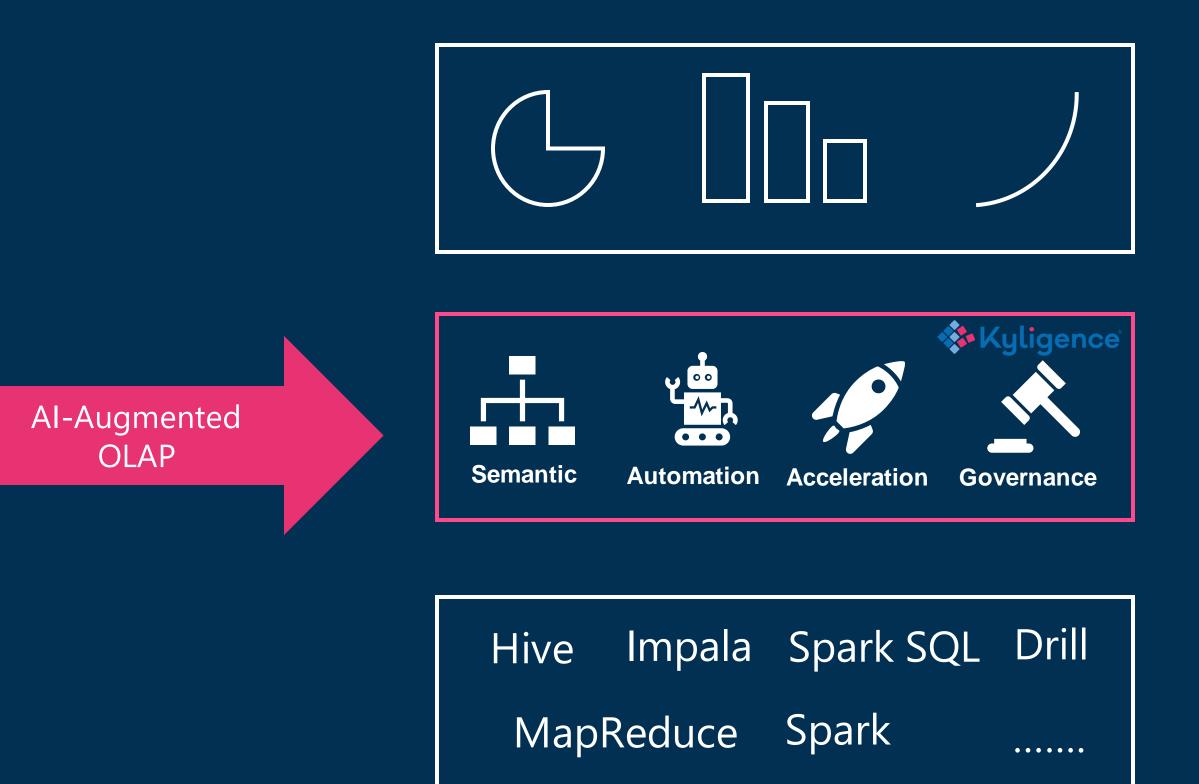




# Take away: Augmented OLAP, the future for analytics







# Thanks

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- Homepage: http://kyligence.io
  - @kyligence
  - #410

