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# Big Data Case Study: Zhihu's 1.3 Trillion Rows of Data, Milliseconds of Response Time

1.3 trillion rows of data, milliseconds of response time.

by Xiaoguang Sun · Sep. 04, 19 · Big Data Zone · Analysis



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Trillions of rows, so little time

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Zhihu, which means "Do you know?" in classical Chinese, is the Quora of China: a question-and-answer website where all kinds of respectively. It is community of its users. As China's biggest knowledge sharing plateries, the interrupt we currently have 220 million registered users, and 30 million questions with more than 130 million answers on the site.

As our userbase grew, the data size of our applications grew out of hand. About 1.3 trillion rows of data were stored in our Monetanks! application (which stores posts users have already read). With approximately 100 billion rows of data accruing each month and SUPPOF growing, this number will reach 3 trillion in two years. We faced severe challenges in scaling our backend while maintaining good user experience.

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In this post, I'll dive deep into how we managed to keep milliseconds of query response time over such a large amount of data and how TiDB, an open source MySQL-compatible NewSQL Hybrid Transactional/Analytical Processing (HTAP) database, empowered us to get real-time insights into our data. I'll introduce why we chose TiDB, how we are using it, what we learned, best practice, and some thoughts about the future.

## **Our Pain Points**

This section covers the architecture for our Moneta application, the ideal architecture we tried to build, and **database scalability** as our major pain point.

# **System Architecture Requirements**

Zhihu's Post Feed service is a crucial system through which users are exposed to content posted on the site. The Moneta application in the backend stores the posts users have read and filters out these posts in the post stream of Zhihu's Recommendations page.

The Moneta application has the following characteristics:

- **Requires high availability for data**: Post Feed is the first screen to appear, and it plays an important role in driving user traffic SUPPC to Zhihu.
- **Handles immense write data**: For example, more than 40 thousand records are written per second at peak time, and theadverts  $\epsilon$  number of records grows by nearly three billion records every day.

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- **Stores historical data for a long time**: Currently, about 1.3 trillion records are stored in the system. With approximately Pole Thanks! billion records accruing each month and growing, historical data will reach three trillion records in about two years.
- **Tackles high-throughput queries**: At peak time, the system processes queries that are performed on 12 million posts on average per second.
- **Limits the response time for queries to 90 ms or less**: This occurs even for a long-tail query, which takes the longest time to execute.
- **Tolerates false positives**: This means that the system can recall many interesting posts for users, even when some posts have been filtered out by mistake.

Considering the facts above, we need an application architecture with the following features:

- **High availability**: It's a bad user experience to find lots of already-read posts when a user opens Zhihu's Recommendations page.
- Excellent system performance: Our application has high throughput and strict requirements for response time.
- Easy to scale out: As our business develops and our application evolves, we hope our system can easily scale-out.

# **Exploration**

To build an ideal architecture with the features above we incorporated three key components in our previous architecture.

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- **Proxy**: This forwards users' requests to available nodes and ensures the high availability of the system.
- **Cache**: This temporarily deals with requests in memory, so we don't always need to process requests in the database. This improves system performance.
- **Storage**: Before using TiDB, we managed our business data on standalone MySQL. As data volume surged, the standalone SUPPC MySQL system wasn't enough. Then we adopted the solution of MySQL sharding and Master High Availability Manager (MWA) espetit but this solution was undesirable when 100 billion new records flooded into our database each month.

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# **Shortcomings of MySQL Sharding and MHA**

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MySQL sharding and MHA is not a good solution because both MySQL sharding and MHA have their shortcomings.

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## **Shortcomings of MySQL sharding**

- The application code becomes complicated and difficult to maintain.
- It is troublesome to change the existing sharding key.
- Upgrading the application logic affects application usability.

## **Shortcomings of MHA**

- We need to implement the virtual IP (VIP) configuration by writing a script or utilizing a third-party tool.
- MHA only monitors the master database.
- To configure MHA, we need to configure passwordless Secure Shell (SSH). This may lead to potential security risks.
- MHA doesn't provide the read load balancing feature for the slave server.
- MHA can only monitor whether the master server (instead of the slave master) is available.

Database scalability was still the weak point of the overall system until we found TiDB and migrated data from MySQL to TiDB.

## What is IIDB?

The TiDB platform is a collection of components that when used together become a NewSQL database with HTAP capabilities.



## TiDB platform architecture

TiDB platform architecture

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Inside the TiDB platform, the main components are as follows:

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- **TiDB server** is a stateless SQL layer that processes users' SQL queries, accesses data in the storage layer, and returns the corresponding results to the application. It is MySQL-compatible and sits on top of TiKV.
- **TiKV server** is the distributed transactional key-value storage layer where the data persists. It uses the Raft consensus protocol for replication to ensure strong data consistency and high availability.
- **TiSpark** cluster also sits on top of TiKV. It is an Apache Spark plugin that works with the TiDB platform to support complex Online Analytical Processing (OLAP) queries for business intelligence (BI) analysts and data scientists.
- Placement Driver (PD) server is a metadata cluster powered by etcd that manages and schedules TiKV.

Beyond these main components, TiDB also has an ecosystem of tools, such as Ansible scripts for quick deployment, Syncer and TiDB Data Migration for migration from MySQL, and TiDB Binlog, for collecting the logical changes made to a TiDB cluster and providing incremental backup and replication to the downstream (TiDB, Kafka, or MySQL).

## TiDB's key features

TiDB's key features include the following:

- Horizontal scalability.
- MySQL-compatible syntax.
- Distributed transactions with strong consistency.
- Cloud-native architecture

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- Minimal extract, transform, load (ETL) with HTAP.
- Fault tolerance and recovery with Raft.
- Online schema changes.

## **How We Use TiDB**

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In this section, I'll show you how TiDB functions in Moneta's architecture and its performance metrics for the Moneta applications in the Interrest of the Moneta applications in Moneta's architecture and its performance metrics for the Moneta applications in the Interrest of the

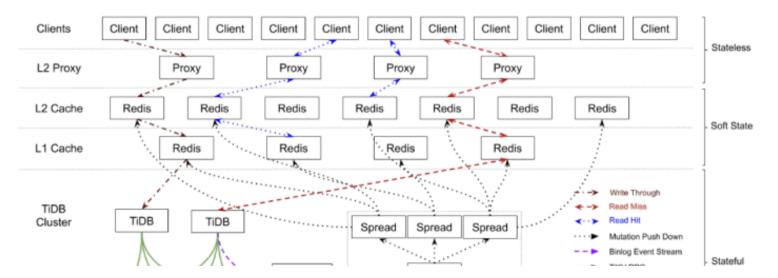
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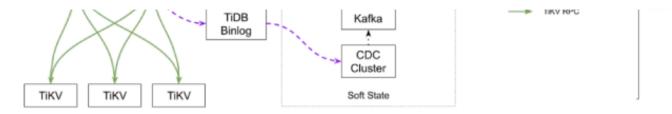
## **TiDB** in our Architecture

We deployed TiDB in our system, and the overall architecture of the Moneta application became:

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- The top layer: Stateless and scalable client APIs and proxies. These components are easy to scale out.
- **The middle layer**: Soft-state components and layered Redis caches as the main part. When services break down, these components can self-recover services via restoring data saved in the TiDB cluster.
- **The bottom layer**: The TiDB cluster stores all the stateful data. Its components are highly available, and if a node crashes, it can self-recover its service.





TiDB architecture in Zhihu's Moneta application

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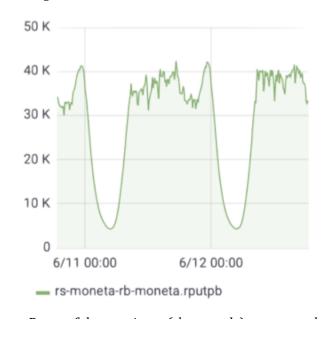
In this system, all the components are self-recoverable, and the entire system has a global failure monitoring mechanism. We then the interror view a sr can leave

## **TiDB's Performance Metrics**

Since we've applied TiDB in the production environment, our system is highly-available and easy to scale, and system performance has remarkably improved.

As an example, take a set of performance metrics for the Moneta application in June 2019:

## 40,000 rows of data were written per second at peak time.



## 30,000 queries and 12 million posts were examined per second at peak time.



Rows of data written (thousands) per second

The 99th percentile response time was about 25 ms, and the 999th percentile response time was about 50 ms. In fact, the average response time is far less than these figures, even for long-tail queries that demand a stable response time.

The 99th percentile response time

99th percentile response time



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999th percentile response time

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## What We Learned

Our migration to TiDB wasn't without a hitch. Here we'd like to share some lessons learned.

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# **Importing Data Faster**

We used TiDB Data Migration (DM) to collect MySQL incremental binlog files and then used TiDB Lightning to fast-import the data to the TiDB cluster.

To our surprise, it only took four days to import these 1.1 trillion records to TiDB. If we logically wrote the data into the system, it might have taken a month or more. We could have imported the data even faster if we had more hardware resources.

# **Reducing Query Latency**

After we finished the migration, we tested a small amount of read traffic. When the Moneta application first went online, we found that the query latency didn't meet our requirements. To solve the latency issue, we worked with PingCap engineers to tune system performance.

During this process, we accumulated precious insights into our data and data processing:

• Some queries are sensitive to query latency and some are not. We deployed a separate TiDB database to process latency-sensitive queries. (Other queries that are not latency-sensitive are processed in a different database, and the execution of the former won't affect the latter.

iatency-sensitive queries are processed in unicrem databases, and the execution of the former won carrect the fatter.

- For queries that didn't have an ideal execution plan, we wrote SQL hints to help the execution engine select an optimal execution plan.
- We used low-precision timestamp Oracle (TSO) and prepared statements to reduce network round-trips.

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## **Evaluating Resources**

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Before we tried TiDB, we didn't analyze how many hardware resources we would need to support the same amount of data we hard to the interror on the MySQL side. To reduce maintenance costs, we deployed MySQL in the single master — single slave topology. In contrast, the interror the interror on the MySQL side. To reduce maintenance costs, we deployed MySQL in the single master — single slave topology. In contrast, the interror of the interror of the master in the interror of the interror of the master — single slave topology. In contrast, the interror of the interror of the interror of the master — single slave topology. In contrast, the interror of the interror

Once our data center was set up properly, we could quickly complete our evaluation of TiDB.

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# **Expectations for TiDB 3.0**

At Zhihu, the anti-spam and Moneta applications are architected the same way. We tried Titan and Table Partition in the release candidate versions of TiDB 3.0 (TiDB 3.0.0-rc.1 and TiDB 3.0.0-rc.2) in our anti-spam application for in-production data.

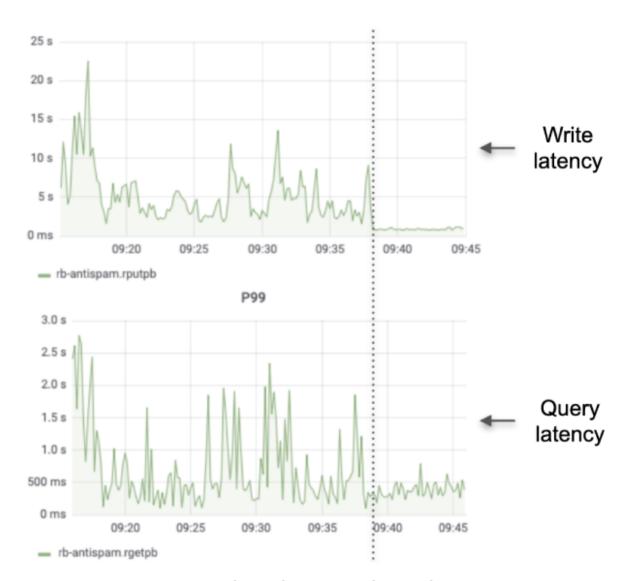
## **Titan Reduced Latency**

The anti-spam application has always been tormented with high latency of both queries and writes.

We heard that TiDB 3.0 would introduce Titan, a key-value storage engine, to reduce write amplification in RocksDB (the underlying storage engine in TiKV) when using large values.

To give this feature a try, we enabled Titan after TiDB 3.0.0-rc.2 was released. The following figure shows the latency for writes and queries respectively, compared to that of RocksDB and Titan:





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Write and query latency in RocksDB and Titan

The statistics show that after we enabled Titan, both write and query latencies dropped sharply. It was really astonishing! When we saw the statistics, we couldn't believe our eyes.

# **Table Partition Improved Query Performance**

We have also used TiDR 2 N's Table Partition feature in the anti-snam annication. Using this feature we can divide a table into

multiple partitions by time. When a query comes, it is performed on a partition that covers the target time range. This has drastically improved our query performance.

Let's consider what may happen if we implement TiDB 3.0 in both Moneta and anti-spam applications in the future.

# **TiDB 3.0 in the Moneta Application**

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TiDB 3.0 has features, such as Batch Message in gRPC, multi-thread Raftstore, SQL plan management, and TiFlash. We believe the Werts  $\epsilon$  will add lustre to the Moneta application.

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## Batch Message in gRPC and Multi-Thread Raftstore

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Moneta's write throughput is more than 40 thousand transactions per second (TPS). TiDB 3.0 can send and receive Raft messages in batch, and it can process Region Raft logic in multiple threads. We believe these features will noticeably improve our system's SUPPOF concurrency capacity.

## **SQL Plan Management**

As mentioned above, we had written quite a few SQL hints to make the query optimizer select an optimal execution plan. TiDB 3.0 adds an SQL plan management feature that binds queries to specific execution plans directly within the TiDB server. With this feature, we don't need to modify query text to inject a hint.

#### **TiFlash**

At TiDB DevCon 2019, I first heard about TiFlash, an extended analytical engine for TiDB. It uses the column-oriented storage technique to achieve a high data compression rate, and it applies the extended Raft consensus algorithm in data replication to ensure data security.

Because we have massive data with high write throughput, we can't afford to use ETL every day to replicate data to Hadoop for analysis. But with TiFlash, we're optimistic that we can easily analyze our tremendous data volume.

# **TiDB 3.0 in the Anti-Spam Application**

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