The Development of an Imitation Model of a Multi-tenant Database Cluster

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Abstract: This paper deals with the main principles forming the foundation of an imitation model of a multi-tenant database cluster — the concept of reliable and easy to use data storage for high load cloud applications with thousands of customers, based on ordinary relational database servers. The main architectural principles of cloud applications are discussed; some statistics about real multi-tenant cloud application is given. This statistics is interpreted to detect the key characteristics of a flow of queries. Basing on this interpretation, the architecture of the imitation model and the process of its development are described.

1 INTRODUCTION

The development of a cloud application throws a lot of architectural challenges to software engineers. One of such challenges is the problem of organizing the storage of data in the cloud with the requirement of high performance, fault-tolerance and reliable tenants’ data isolation from each other. At the moment these tasks are usually solved by designing an additional layer of application logic at the level of application servers. Such a technique is discussed in a lot of specialized papers for application developers and other IT-specialists (Chong et al., 2006a/b, Candan et al., 2009). There are also some projects of providing native multi-tenancy support at the level of a single database server (Schiller et al., 2011). This paper is devoted to the alternative concept of a multi-tenant database cluster which proposes the solution of the above problems at the level of a data storage subsystem. In particular, the key characteristics of the query flow are being studied basing on the statistics of an existing multi-tenant cloud application. The correlation between various parameters of a tenant and its query flow are studied. The process of the development of cluster imitation model is described and basic principles of the model architecture and mode of operation are highlighted.

2 THE ARCHITECTURE OF THE MULTI-TENANT DATABASE CLUSTER

A multi-tenant database cluster (Boytsov and Sokolov, 2012) is a concept of data storage subsystem for cloud applications. It is an additional layer of abstraction over ordinary relational database servers with a single entry point which is used to provide the isolation of cloud application customers’ data, load-balancing, routing of queries among servers and fault-tolerance. The main idea is to provide an application interface which has most in common with the interfaces of traditional RDBMS (relational database management system). At the moment a typical scenario of interaction with the cluster from the developer point of view is seen as the following:

Connect(TenantId,ReadWrite/ReadOnly);
SQL-commands
Disconnect();

A multi-tenant cluster consists of a set of ordinary database servers and specific control and query routing servers. The query routing server is a new element in a chain of interaction between application servers and database servers. This is the component application developers will deal with. In fact, this component of the system is just a kind of a proxy server which hides the details of the cluster structure, and whose
main purpose is to find an executor for a query and route the query to him as fast as possible. It makes a decision based on the map of a cluster.

![Multi-tenant database cluster architecture](image)

Figure 1: Multi-tenant database cluster architecture.

It is important to note that a query routing server has a small choice of executors for each query. If the query implies data modification, there is no alternative than to route it to the master database of a tenant, because only their data modification is permitted. If the query is read-only, it can also be routed to a slave server, but in the general case there would be just one or two slaves for a given master, so even in this case the choice is very limited.

The data distribution and load balancing server is the most important and complicated component of the system. Its main functions are:

- initial distribution of tenants data among servers of a cluster during the system deployment or addition of new servers or tenants;
- management of tenant data distribution, based on the collected statistics, including the creation of additional data copies and moving data to another server;
- diagnosis of the system for the need of adding new computing nodes and storage devices;
- managing the replication.

This component of the system has the highest value since the performance of an application depends on the success of its work.

### 3 ANALYSIS OF EXISTING APPLICATION

Analysis of existing applications and their mode of operation is the first thing to study when designing an imitation model. In the context of the multi-tenant cluster theme the most interesting question is the characteristics of the query flow, since this component has the greatest impact on the results obtained during the modelling. As the multi-tenant cluster is a queuing system, the Poisson flow of events is a good basic model of a query flow. The key points to explore are:

1. intensity distribution of incoming query flows among clients;
2. presence or absence of dependency between an average time of query execution and characteristics of the client which this query belongs to;
3. characteristics of a customer base;
4. characteristics of customer base changes over time.

Since questions 1 and 2 have a significant impact on the distribution of queries between servers thus making a decisive contribution to the assessment of the efficiency of load balancing across the cluster as a whole, they are very important. The answer to the fourth question will allow us to adequately simulate the dynamism inherent to all cloud systems and therefore offer an effective long-term data management strategy.

There are many factors that possibly can affect parameters of a client query flow. At the initial stage of the study it was decided to take the size of the data that the client stores in the cloud as its key characteristic. The relationship between this parameter and the intensity of the query flow or an average time of query execution has been studied. The following assumptions seemed to be reasonable:

1. the most of client schemas are approximately of the same size, but there are also significant (but rare) variations in both directions;
2. client query flow intensity is directly dependent on the size of client data (the greater data the client has, the more often they are accessed);
3. the query execution time is directly dependent on the size of client data (the greater data the client has, the more data are accessed by the average query, thus its execution time increases);
4. client data size and activity smoothly change over time.

The verification of the above assumptions has been performed on the basis of statistics and logs of the existing multi-tenant cloud application. This application is the online service that provides an electronic flow of documents and accounting. The diversity of offered services leads to the diversity of possible scenarios of interaction between a client and the application, thus making a complicated query flow. The application uses Postgres SQL server as its primary data storage. All management stuff is performed by a set of specialized services and routers. Currently, the cluster consists of about
database 60 servers and continues to grow. The statistics being investigated is incomplete and rather inaccurate due to the way it is collected. There is a dedicated service that is used to collect this statistics. It “wakes up” every 24 hours and processes logs of all application servers within the cluster, making conclusions about average values of key characteristics. Since this service is only able to analyse the entire client’s remote procedure call, which can consist of several SQL queries, it can not provide accurate data about the amount of queries and their execution time. Despite this fact the statistics allows to find out some notable trends since the most of remote procedure calls consist of a single SQL query.

At first, the graph of clients’ data size distribution was built to better understand the characteristics of the application’s user base, that currently consists of about 40 000 clients. The most significant part of this graph is shown in Figure 2.

![Figure 2: Client data size distribution.](image)

The values on the horizontal axis correspond to the size of client’s database schema in megabytes. The values on the vertical axis correspond to the amount of clients with that data size.

Here we can see that most clients of the application being studied (about 37000 of 40000) have a schema with 7-8 megabytes of data. The curve of clients data size distribution basically corresponds to the curve of lognormal distribution. This means, that if we want to visualize dependency between the client data size and some other characteristics, it is better to use the logarithmic scale.

To verify the correctness of assumptions 1 and 2, the data set containing 40 000 records with the format (client data size, total amount of queries for a week) was built and analysed. The graph that visualizes this data set is given below in Figure 3. The values on the horizontal axis correspond to the size of client data. The horizontal axis has the logarithmic scale and every next interval is about 1.5 times longer than the previous one. The values on the vertical axis correspond to the average amount of queries from clients that fall into the corresponding interval.

The analysis of the above data set confirmed assumptions 1 and 2 about the relationship between the data size and the client activity: the coefficient of correlation between the size of the client’s schema and the number of client queries is about 0.7, which indicates a fairly strong correlation between these two values.

At the next stage the assumption about the dependency between the average query execution time and the client data size was studied. Another data set with the format (client data size, average query execution time for the last 24 hours) with 7500 records was built to confirm or reject this assumption. According to that data set, the coefficient of correlation between these two values equals to 0.03, that corresponds to a very weak dependency or its absence. This allows to throw away Assumption 3 as not confirmed.

The above data set also allowed to build a graph of average query execution time distribution (Figure 3).
4. The values on the horizontal axis correspond to an average query execution time in milliseconds, and the values on the vertical axis correspond to a number of clients with that average query execution time. The resulting curve basically corresponds to the curve of lognormal distribution.

Another interesting point about the cloud application is the dynamics of users base grow. Unfortunately, the application studied is currently not very mature and goes through a phase of initial accumulation of a customer base. The dynamics of the total data size within the cluster is given in Figure 5, where the values on the horizontal axis correspond to a number of weeks and the values on the vertical axis correspond to the total size of client data in megabytes.

4 THE IMITATION MODEL OF THE MULTI-TENANT CLUSTER

The imitation model of the multi-tenant database cluster has been developed according to the analysis of the collected data. The model is a GUI application that runs under Linux OS. It was developed using Qt framework and C++ programming language.

The model works in virtual discrete time from one event to another. There are the following main entities that can produce events:

1. a query generator, including queries for creating of new tenants;
2. database servers inside of the cluster – query execution finish events;
3. a generator of failures – a special entity, which is used to simulate failures of cluster components.

Models of other cluster subsystems are implemented according to an “observer” design pattern. That means that they are notified by the model core about all occurring events and are able to generate derivative events (for example, data replication queries after finishing the execution of primary data modification query at the client’s master server). A data distribution algorithm, in particular, is one of such observers that allow it to collect some statistics and analyse cluster performance for further optimization of its operation.

The Poisson flow of events is the basic model used for the query generator. The flow intensity can be set in two ways: as a static value, representing the amount of queries per unit of model time or as a dynamic value depending on the amount of clients (for example, 0.05 queries per client per unit of model time). When the generator operates in the second mode, the total intensity of the flow is equal to the product of the total amount of clients by the
value set that is it changes with a rise of the clients base. Every new query is supplied with a weight coefficient that affects the query execution time by the server. According to the above statistics these coefficients are distributed by the lognormal law and distribution parameters can be changed during the generator setup.

Besides, the model is able to simulate server’s failures and a denial of service. For that purpose there is a special entity – the failure generator. Every server inside the cluster has a non-negative probability of its failure per unit of model time. If this probability is not equal to zero, then sooner or later the server stops to serve queries for some time. The time of server’s malfunctioning is determined by the distribution with parameters set by the user. This feature allows to research the cluster for fault-tolerance.

A lot of graphical reports on the cluster operation can be generated by the model (an average response time for the last 100 queries, a queue size at each server and a total queue size, a query flow distribution between servers and so on). These reports can be used to monitor a state of the model and to analyse trends in operation of the cluster.

There is also a special mode of mass parallel simulation to collect statistics about characteristics of algorithms tested. When this mode is used, the predefined configuration of the cluster and the profile of query flow is tested in combination with the offered algorithms of query routing and data distribution. There are several special stop conditions that are used to indicate the completion of modelling (a period of time, a specified queue size at any server, a specified total queue size at the level of the entire cluster). Many identical experiments run in parallel. A complete statistics about the state of the cluster and the combination of the used algorithms are stored in resulting files by the model.

The formation of queues leads to a degradation of application performance from the point of view of the client whose data are stored by these servers. The above fact means that primitive strategies should not be used in the real production environment. Further study will be devoted to the model improvement and identification of key factors that influence the efficiency of the proposed cluster control system.

REFERENCES