CPN based Data Integrity Evaluation for Cloud Transactions

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Abstract: Cloud computing environments, especially the PaaS environments, are one of the most promising platforms for high capacity and low cost transaction processing. However, today’s PaaS environments provide us with limited capability for data integrity in database and transaction processing, compared with traditional transaction management systems. Therefore, we need to evaluate whether the cloud environment currently considered provides enough capability for our data integrity requirements. This paper presents a Colored Petri Net (CPN) based approach to modeling and evaluating generalized transaction systems including cloud environments. The evaluation is performed based on the given constraints on database records expressed in the form of predicate logic formulae, and examined by CPN/ML codes implemented as the guard functions of the transitions for integrity verification.

1 INTRODUCTION

Transactional integrity is one of the most critical and challenging issues in cloud computing, since today’s cloud environments, especially the PaaS (Platform as a Service) environments, e.g. Google App Engine (GAE) (Sanderson, 2009) or Amazon Elastic Compute Cloud (EC2) (Vliet and Paganelli, 2011), degrade the capability of data integrity in exchange for implementing highly scalable, distributed, and virtualized platforms. The data integrity in traditional transaction processing relies on the ACID property of each transaction, which is implemented as a major functionality of every transaction management system (Gray and Reuter, 1993).

Unlike these management systems, cloud environments provide us with more simplified mechanism for data integrity, which is often referred to as BASE property in contrast with ACID (Pritchett, 2008). One of the most distinctive characteristics of BASE transactions is that they are not totally isolated for preserving data integrity, and therefore the “C” (Consistency) of the ACID property cannot be assured in BASE environments, and consequently the data integrity might be broken. Due to this incomplete integrity assurance, mission critical applications have been thought not to be suitable to cloud computing environments.

However, for the evolution of cloud computing, these applications are one of the critical success factors, since there are huge number of them are running in the form of transactions. In order to deploy these applications over cloud environments, we first have to reveal their transactional behavior in the environments, along with the impact analysis to data integrity.

In order to deal with a variety of cloud environments including emerging ones, it seems a good practice to build a platform independent framework for evaluating data integrity. In this framework, transaction systems must be formalized for generalization. Among numerous formalization tools, Colored Petri Net (CPN) (Jensen and Kristensen, 2009) is one of the most appropriate tools for this evaluation, since it can express dynamic systems from both behavioral and functional viewpoints simultaneously, along with the simulation capability (Martin et al., 2004).

This paper presents a CPN based formal approach to modeling and evaluating the transactional behavior of database applications in cloud environments. The paper is organized as follows. In section 2, we introduce the basic concepts of “data integrity” from several viewpoints. Section 3 presents CPN based modeling techniques for the transactional behavior from platform independent viewpoints. Section 4 discusses how the CPN based models are evaluated for the consistency.


2 DATA INTEGRITY FROM APPLICATION VIEWPOINTS

Data integrity is one of the most important issues in software and system design relating to database and transaction processing. However, the concept of “integrity” has many facets, and we need to discuss it from different several viewpoints. The typical aspects that the concept shows are “physical”, “logical”, “application”, “system”, and so on. Various database or transaction management systems provide us with sophisticated mechanism for the data integrity, in cooperation with the operating systems and hardware controllers, under which they are running.

In spite of these mechanisms, each application is still responsible for the data integrity from its own application viewpoints. Several relational database management systems provide us with a few functionality for this kind of data integrity, which includes “foreign key”, “referential integrity”, “domain”, “trigger”, and “stored procedure”.

In order to utilize this functionality properly, we first have to define the application data integrity rigorously. However, the data integrity from application viewpoints depends too much on individual applications, and consequently it seems difficult to generalize and formalize it. In this section, we introduce a predicate logic based generalization and formalization of the data integrity. For the generalization, we define data integrity as a set of constraints on the databases to be used. These constraints can be categorized into the following three classes.

1. Constraints on data values of the attributes in the database records.
2. Constraints on the properties of a collection of database records.
3. Constraints on the existence of particular database records.

The above three constraints are referred to as “value oriented”, “set oriented”, and “existence oriented” constraints respectively in this paper.

One of the most rigorous ways to express these constraints is to describe them in the form of predicate logic formulae (Shinkawa and Matsumoto, 2001). For this description, we first have to define the “language” \( \mathcal{L} \) which includes constants, functions, predicates, and logical symbols, along with their semantics in the form of the “structure” \( S \) that is composed of the domain and interpretation. Since the language deals with databases, it consists of the followings.

1. **Constants**: all the possible values that each attribute can take.
2. **Functions**: all the arithmetic and SQL functions, along with the functions for control structures, e.g. “if-then-else”, “while”, and “break” functions.
3. **Predicates**: All the arithmetic and SQL predicates such as “\( \leq \)”, “\( \geq \)”, “BETWEEN”, “IN”, and so on.
4. **Logic Symbols**: follow the traditional predicate logic notation, e.g. “\( \forall \)”, “\( \wedge \)”, “\( \neg \)”, “\( \vee \)”, or “\( \exists \)”.

In addition to the above, we can use variables to represent the objects in the system. In order to express various objects in the databases effectively, indexed variables are used, e.g. \( db_i \) to express the \( i \)th database in the system, \( \alpha_{ij}^{(i)} \) to represent the \( j \)th database record in \( db_i \), and \( \alpha_{ij}^{(k)} \) to represent \( k \)th attribute value in \( \alpha_{ij}^{(i)} \). Other symbols may be used in the form of indexed or non-indexed variables.

The structure \( S = (U,I) \) that gives the semantics to the language \( \mathcal{L} \) is defined as follows.

1. Let the schema of \( i \)th database be \( DB_i = A_1^{(i)} \times \ldots \times A_n^{(i)} \) and a database instance from it be \( DB_i^t \).
2. The domain \( U \) of the language \( \mathcal{L} \) is a set of objects that are referred to as arguments of the functions and predicates, and are expressed as

\[
U = \bigcup_{i} DB_i \cup \bigcup_{i,j} \alpha_{ij}^{(i)} \cup \bigcup_{i} \mathcal{R}_i
\]

where \( \mathcal{R}_i = \{ \langle \alpha_1^{(i)}, \ldots, \alpha_n^{(i)} \rangle \} \) and \( \alpha_\alpha^{(i)} \in A_j^{(i)} \), that is, the set of all the database records.

3. The interpretation \( I \) follows the rules defined by arithmetic, logic, and SQL.

Using the above language \( \mathcal{L} \) and the structure \( S \), the generalized form of the constraints are as follows.

1. For a value oriented constraint,

\[
Q_1 \cdot \ldots \cdot Q_n \bigvee_{j} \bigwedge_{i} P_{ij}(t_{ij}^{(i)} \ldots t_{mij}^{(i)})
\]

where \( Q_i \) is a variable with the quantifier “\( \forall \)”, e.g. \( \forall x_i \), \( P_{ij} \) is a predicate, and \( t_{ij}^{(i)} \) is a term composed of variables, constants, and functions.

2. For a set oriented constraint, it can be expressed in the same form as a value oriented constraint, however each term represent a set of objects.

3. For an existence oriented constraint,

\[
Q_1 \cdot \ldots \cdot Q_n \bigvee_{j} \bigwedge_{i} P_{ij}(t_{ij}^{(i)} \ldots t_{mij}^{(i)})
\]

where \( Q_i \) represents a variable with the quantifier “\( \exists \)”, e.g. “\( \exists y_i \). The rest of the formula is the same as that of the value oriented constraints.
The above forms of constraints can be intermixed.

For example, if the \( db_i \) represents an employee database, and is imposed the constraints “the salary of the boss of an employee must be higher than him”, it is expressed as

\[
\forall j \left( \text{salary}(\text{boss}(r_j^{(i)})) > \text{salary}(r_j^{(i)}) \right)
\]

where “salary” and “boss” are the functions that return the salary and the boss of the employee that corresponds to the database record \( r_j^{(i)} \). The other example is that “the number of the members in each section must be less than 100” is expressed as

\[
\forall a_k^{(j)} \left( \text{count}(db_i, a_k^{(j)}) < 100 \right)
\]

where “count” is the function to return the number of records, which takes two arguments, one is a database defined as a set of tuples, and the other is a grouping key.

3 CPN MODELING FOR TRANSACTION PROCESSING

Since the values in databases are confirmed at each “commit” or “aborted” time, above discussed constraints must hold at that time. In order to examine these constraints at appropriate times, we need an environment to simulate the behavior of transactions, unless we have the real environment for the transactions. The benefit to have a simulation environment is that we can efficiently examine the constraints in early phases of system development, before a real environment becomes available. In this paper, we use Color Petri Net (CPN) as a simulation tool.

CPN is an extension of a regular Petri net, which introduces functionality and data types for more flexible control of the regular Petri net. CPN is formally defined as a nine-tuple \( CPN=(P, T, A, \Sigma, V, C, G, E, I) \), where

- \( P \): a finite set of places.
- \( T \): a finite set of transitions.
  - (a transition represents an event)
- \( A \): a finite set of arcs \( P \cap T = P \cap A = T \cap A = \emptyset \).
- \( \Sigma \): a finite set of non-empty color sets.
  - (a color represents a data type)
- \( V \): a finite set of typed variables.
- \( C \): a color function \( P \rightarrow \Sigma \).
- \( G \): a guard function \( T \rightarrow \text{expression} \).
  - (a guard controls the execution of a transition)
- \( E \): an arc expression function \( A \rightarrow \text{expression} \).
  - (an arc expression function \( P \rightarrow \text{closed expression} \).

All the above functions, like arc and guard functions, are expressed using a functional programming language CPN ML (Jensen et al., 2007), which is an extension of the standard ML (Harper et al., 1997).

In order to determine the behavior of transactions, we need to express such three aspects of transaction processing as “control structure”, “data structure”, and “application structure”.

[Control Structure]
This aspect represents how the system deals with each transaction, and can be composed of the following CPN model components. Figure 1 represents an example of the structure, where some arc and guard functions are omitted for readability.

1. Queuing and Dequeuing.
   These components put transactions into the waiting line outside the system, and take them out of it for scheduling. In a CPN model, they can be regarded as external entities of the model, and therefore are expressed as a single place to hold the transactions to be processed. In Figure 1, this place is shown at the top of the figure, which is labeled “TRQ”.

2. Scheduling.
   This component executes each dequeued transaction according to a predefined scheduling algorithm. The component is expressed in a CPN model as a transition “Sch” to obtain a transaction from the place “TRQ”. A place “SEQ” to give a unique transaction id, a place “TQ” that represents an internal queue, and succeeding \( n \) parallel transitions for concurrent execution. It is shown in Figure 1, below the place “TRQ”.

3. Database Access.
   This component manipulates databases by “insert”, “update”, “delete”, or “select” operations. The component is expressed in a CPN model as a place to hold a database operation such as “select” or “update”, transitions for processing each database, and a place to hold all the database record instances. In Figure 1, the place labeled “SQLR”, “DB”, “DBF” and the transition labeled “DB1” and “DB2” belong to this component. In order to represent the “buffering” operation, the database is denoted by the two places “DB” and “DBF”, which correspond to the buffer and finalized database respectively.

4. Exclusive Control.
   This component performs one of the most important functionality to preserve data integrity. The simplest implementation by CPN is to use a transition for “lock” operation and a place to hold the tokens representing lock acquisition and release.
Each token is expressed as a list of the locks for database records. In Figure 1, the transition “LC” and the place “Lock” compose the component.

5. Logging.
This component stores the operation logs for database related activities. In a CPN model, it is expressed as a transition for log writing, and a place for a log file, which are represented as the transition “Log” and the place “LF” respectively in Figure 1.

6. Commit and Abort.
These components respectively confirm or discard the database update. Each of them is expressed as a transition connected to the places representing the databases and the log file. In Figure 1, the two transitions “Commit” and “Abort” represent this component, where the place “CMPL” is a place to control the simulation (see below).

7. Other Complementary Components.
These components include “restart”, “backup”, “reorganization”, and so on. They are used under particular situations, e.g. database recovery. Since this paper deals with data integrity in steady operation, they are not included in the CPN models.

In addition to the above components for transaction processing, several components to control the CPN model are needed, which are used to repeat or terminate the simulation. In Figure 1, the transition “End”, “Post”, and the place “CMPL” compose such components.

[Data Structure]
In Transaction processing, data integrity is defined as that the database records within a system satisfy the given constraints. In a CPN model, each database record can be expressed as a token associated with a color set. This color set represents the data type which corresponds to the database schema for the database record. However, the data types used in a CPN model and database schema are mutually different, therefore we need to define the transformation rules between them. One of the simplest ways for this transformation is to map the schema data types into “int” for the numeric types, and into “string” for others in CPN/ML. For example, the following color set definition represents a database schema with “integer” key and “varchar” data.

```
colset DB = int;
colset Key = int;
colset Data = string;
colset KD = product Key * Data;
colset KDList = list KD;
colset DBRecord = product DB * KDList;
```

[Application Structure]
The purpose of modeling the application structure of a transaction system is to reveal how each transaction accesses the databases to simulate the interactions between them for consistency evaluation. A transaction is an entity in a transaction system, and can be expressed as a token in a CPN model. This token is required to express the logic included in the transaction. The required information from the application logic is the order of database accesses, in order to evaluate data integrity. Therefore, this paper expresses it as a list of data access information, or more concretely, we represent each transaction as a product of “transaction-id” and “database access list”, each element of which is composed of “database id”, “access type”, and “access key”. As color sets, they are expressed as follows.

```
colset Action = int;
colset SEQ = int;
colset State = int;
colset TID = int;
colset SQL = product DB * Action * Key * Data;
colset SQLList = list SQL;
colset TranReq = product TID * SQLList;
colset Tran = product TID * SEQ * State * SQLList;
```

where the color set “Action” represents the type of database operation, e.g. “1” represents “select”, “2” represents “update” and so on. The color set “SEQ” represents a unique transaction id.

4 INTEGRITY EVALUATION
The CPN models discussed in the previous section are to be evaluated whether they satisfy the constraints expressed in the form of predicate logic formulae. Unlike the model checkers, e.g. SPIN, SMV, or LTSA, CPN is not equipped with the verification capability. Therefore, we need to add constraint evaluation mechanism to the models. This mechanism works every time a transaction issues “commit” or “abort”, which makes the database update confirmed. Therefore, the mechanism must be synchronized with the “commit” and “abort” components of the CPN model. One of the implementation of this mechanism is shown in Figure 2 (This figure is denoted by a simplified notation of CPN for readability). This mechanism consists of

1. A place for synchronization with the “commit” and “abort” components (The place “C” in Figure...
2. A transition for assuring data integrity of which guard function is boolean and logically equivalent to the constraints for the integrity (Transition “A” in Figure 2).

3. A transition for detecting the breaking of data integrity, of which guard function is boolean and logically equivalent to the negation of the above constraints (Transaction “D” in Figure 2).

4. Two output-only places, one of which is connected to the above transition “A” to hold the “OK” messages, and one of which is connected to the above transition “D” to hold the error messages.

This mechanism works as follows.

1. When the “commit” or “abort” transition is fired, a token to activate one of the evaluation transitions “A” or “D” is provided into the place “C”. At this moment, all the databases are confirmed.
2. By marking the place “C” in Figure 2, the evaluation transitions “A” and “D” become eligible to fire. The transition “A” is associated with the guard which becomes true if the constraints are satisfied. On the contrary, the transition “D” is associated with that which becomes true if the constraints are not satisfied.

3. Under the control of the above guards, the transition “A” is activated if the integrity constraints are satisfied. In such a case, a token is sent back to the place “C”, and the succeeding transitions of the “commit” or “abort” transitions become eligible to fire. Otherwise, the transition “D” is activated, and no token is sent back. Consequently, the succeeding processes of the corresponding transaction are halted.

By adding the above mechanism to a CPN model for transaction processing, erroneous transactions to disturb the data integrity are detected based on the constraints expressed in the form of predicate logic formulae. The situation of the data integrity problems is reported as a token in the place “D”.

5 CONCLUSIONS

Cloud computing environments, especially the PaaS environments, provide us with the platforms for highly productive development, flexible operation, and easy maintenance of transaction systems. One of the bottlenecks of them is the low capability of preserving data integrity, which is often referred to as “CAP theorem”. This paper presented a CPN based modeling and evaluation techniques for transaction systems from data integrity viewpoints.

Firstly, the definition of data integrity, which is one of the most vague concepts in database systems and applications, is introduced from three different viewpoints, and then it is formalized using predicate logic. Although there are a variety of transaction management systems based on different technologies and mechanism, the essential functionality is common and can be formally modeled.

The paper used CPN as a formalization and modeling tool to express the behavior of transactions, and the integrity rules are expressed within CPN models using CPN/ML codes. The common functional components among the different transaction management systems, e.g. transaction queueing, scheduling, commit, or abort, are represented as CPN structures with appropriate CPN/ML codes for color, guard, and arc functions. Integrity evaluation mechanism is also implemented as a CPN structure, and is added to the above CPN models. It evaluates the original CPN models examining whether they satisfy the given integrity criteria.

The paper deals with the common functionality among different transaction systems, however for practical use, more platform dependent models are needed, e.g. those for Google App Engine (GAE) or Amazon E2C.

REFERENCES


