

Declarative Data-Centric Consistency

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Abstract

The utilization of relaxed data consistency models has increased in recent years, especially in distributed applications, where maintaining strong consistency comes on the cost of availability and performance. Developing applications with correct consistency properties is difficult, because current development tools do not provide abstractions that enable programmers to write their code in a replicated, asynchronous environment intuitively. In this research prospectus, we survey various techniques and tools proposed to control data consistency, and describe our vision of a data-centric, asynchronous framework that would facilitate developing distributed applications with replicated data.

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1 Introduction and Problem Domain

In an ideal concurrent environment, it is reasonable for programmers to expect that the latest update issued to a shared data abstraction will be the one visible to the next read issued by any client. However, a system will almost always include multiple copies of the same data abstraction through caching, buffering and replication, and since updates do not propagate among the copies in the speed of light, it is possible to observe multiple versions of data at the same time. Furthermore, updates are often reordered, either to allow optimizations, which is the case in many CPU architectures such as ARM [10], or because of communication medium delays, familiar to network application developers. Consequently, defining what ‘latest update’ means is no longer straight forward. Programming in such environments requires defining a consistency model [1] which serves as a contract between the data abstraction and its clients. Strong consistency models such as sequential consistency [23] and linearizability [20] come close to fulfilling the above mentioned expectation, but they are expensive in terms of time for certain applications, because they disallow many execution interleavings through some form of synchronization, ultimately forcing clients to wait. They also forbid some reordering-based optimizations. In reality, most frameworks provide more relaxed consistency models, such as causal consistency [33] and eventual consistency [38].

Even when built on top of such frameworks, the vast majority of multicore applications attempt to exhibit a strongly-consistent behavior. The Exceptions such as [24] and thread schedulers in Fortran (@Nate Doug told me about these but I could not find a source, any idea?) tend to be probabilistic in nature or tolerate nondeterminism, which are not common properties of multicore applications. On the contrary, Internet has brought many more use-cases where relaxed consistency is employed on application-level with little compromise. As a motivating example, consider an online shopping application that replicates product data. When a client requests a list of all products that have 5-star ratings, it is not necessary to query all replicas in order to get the most up-to-date rating data. In fact, requiring acknowledgments from all replicas would cause unnecessary system load for a task that is inherently subjective. Some real-world use-cases include metrics collection and product recommendations at eBay [32], and inbox search at facebook [22]. Both employ Cassandra [22], an eventually-consistent [40] storage backend.

Relaxed consistency is rarely sufficient on its own for the requirements of a complete application. Typically, components requiring different consistency policies co-exist and operate. Great care must be taken to ensure correct program semantics in this case. Continuing with the previous example, consider when the client adds an item from the list of items with 5-star ratings to his shopping cart. Because the list was not constructed by querying all replicas, it is possible that it includes some items that are no longer in stock. If the programmer forgets to query all replicas to ensure that the selected item is in stock prior to checkout. The client will end up buying a non-existing item.

Employing weak and strong consistency side to side, on application level in a distributed environment is a relatively recent shift from the traditional approach of developing strongly consistent applications on top of weakly consistent frameworks. We believe that many programming pitfalls, similar to the above mentioned can be avoided by providing programmers with the right development environment and tools. Ongoing research on concurrency, programming languages and computer systems has produced many methodologies and tools that facilitate consistency management, and we are investigating how to combine and extend existing ideas in order to construct a framework that allows programmers to safely employ both strong and relaxed consistency in their distributed applications.

The core idea we are studying is based on a declarative, data-centric approach to consistency management, originally proposed by Vaziri et al. [39]. This approach allows programmers to declaratively express atomicity as a property of data, and we are investigating how to allow expressing and inferring other consistency properties and constraints as well, so that it is known at compile time which parts of the program are strongly consistent and which parts are not. This way, programs become clearer, and it becomes possible to delegate many parts of consistency management logic to the compiler and runtime. In this research prospectus, I start by describing the core ideas of this approach, then demonstrate how it can be adapted to support multiple consistency levels of replicated data. Then I will list related work on data abstractions, consistency constraints and program properties that we believe are useful for programmers to express declaratively. After that I will overview the research directions and challenges, and how we plan to conduct our work.

2 Data-Centric Constraints

When programming in a shared-memory environment, programmers must ensure that concurrent accesses to the same memory location do not lead to data races and inconsistent results. This is traditionally achieved by ensuring that instructions accessing shared data acquire locks, or are within a transaction. Such approaches have an operational, control centric flavor. The problem is that data races may occur if the programmer forgets to synchronize even a single control flow path.

2.1 Atomic Sets

Vaziri et al. [39] outlined a solution in 2006, that associates synchronization constraints with data, and based on these constraints, the compiler generates synchronization code for all operations that access this data. Figure 1(counter example from the paper) shows an example taken from that paper written in Java. It involves incrementing a counter concurrently. An atomic set is declared, which is a memory region to which all accesses must be synchronized, then class fields that need to be protected are added to this set. The compiler then synchronizes all public class methods that interact with the atomic set. The system also allows atomic sets that span multiple objects, and provides other features as well.

2.2 Adding Replicated Data Consistency Constraints

In many replicated key-value storage systems, such as Cassandra [22], Dynamo [18], and Riak [34], clients can tune consistency by specifying the number of replicas on which an operation must succeed before returning an acknowledgment to the client application. In Cassandra, consistency is configured on the level of cluster, data center, or individual I/O operations. Going back to our online shopping example, instead of relying solely on the programmer to choose the correct consistency level every time he writes code that interacts with the contents of the shopping cart, we can attach a policy to shopping cart data region, stating that all operations must have the consistency level ALL, meaning all replicas must acknowledge the successful completion of the operation. We gain two benefits following this data-centric approach: it is now possible to infer the correct consistency level for all operations dealing with the shopping cart data, and if the programmer specifies a lower consistency level, such as ONE, which only requires an acknowledgment from one replica, it is possible now to show an error/warning stating that this is probably not what is meant to be done. Furthermore, Since the list or 5-star products was read with consistency level ONE, it represents a memory region that is not strongly consistent. It will not be possible to assign an item form this region to the shopping cart, as it represents a forbidden information flow path.

3 Research Prospectives

Inspired by the above mentioned work, we aim to present a programming environment that enables developing distributed applications with better data abstractions. There are two main aspects we need to investigate: defining data region abstractions, and identifying data and operation properties and constraints that would be useful to infer/express. In the following sections, we describe these two research problems, and show related work.

3.1 Data Region Abstractions

We need to determine which data region abstraction(s) to employ, because choosing the right abstractions eases defining constraints on data. Many data abstraction in various systems come with baked-in consistency constraints. Having good default definitions of such constraints often relieves the programmer from having to define them himself. We also believe that having replication defined as an explicit property of data makes reasoning about it easier for the programmer. The following list contains various data abstractions and built-in consistency guarantees.

1. The Java memory model [31] ensures atomic semantics for one-word primitive types, such as `int`. `long` and `double` are not required to be atomic by the language specification.
2. In many database systems, row updates are atomic. For instance, In MySQL [29], the MYISAM storage engine locks tables when an update is issued, while InnoDB engine locks the row being updated. In Cassandra, rows are atomic units.
3. Objects in object oriented languages span over multiple memory locations, and group together data and operations [35]. Objects can be arbitrarily nested and passed around to form complex graphs, and accesses to object data are typically unordered and synchronous. Objects present a difficult case for constraint attachment, because many threads can have references to different parts of an object and update them concurrently, possibly making temporary invariant violation visible among threads. Even if each object in the graph is properly synchronized, the collection of objects might still have incorrect semantics [5]. Furthermore, it is hard to reason about nesting of objects bearing conflicting constraints. There is a large body of ongoing research on ownership types and alias management that aim to restrict arbitrary referencing of objects [28, 11, 16].
4. In the current implementation of atomic sets [12], Vaziri et al. employ a simple ownership type system that ensures objects are only accessed via their owners. Without this constraint, it would be hard to perform static analysis. Atomic sets presented in this work are static data containers that may span the data of multiple

objects. data within an atomic set share the same lock. However atomic sets only express “shallow-locking”, if nested objects must be locked, they need to be explicitly added to the containing object’s atomic set. It is also not possible to dynamically grow or shrink atomic sets.

5. DPJ [7] regions are hierarchal heap partitions used to disambiguate accesses to distinct objects, as well as distinct parts of the same object. A type-checker then ensures that there are no conflicting accesses to overlapping memory regions between concurrent tasks.
6. X10 [9] supports places [13], which are a distribution-friendly data abstraction that encapsulate the binding of activities and globally addressable memory. An activity may synchronously access data items only at the place in which it is running. It may atomically update one or more data items, but only in the current place. If it becomes necessary to read or modify an object at some other place, a place-shifting operation can be used. This approach ensures that expensive operations (e.g., those which require communication) are readily visible in the code. The implicit constraint in this system is the locality of data.
7. The actor model [2] differs in two fundamental ways from the object model: all accesses to actor state are serial and asynchronous. This model is a better fit for a networked environment, and is gaining popularity with languages such as Erlang [4] and Scala [30] employing it successfully. Typically, actors must not expose their internal data by any means other than message passing. In practice, systems such as Akka [8] do not enforce that, and rely on the programmer to maintain actor encapsulation. Because of their properties, actors are easily distributable, however it is hard to introduce concurrency within an actor. Habanero Java and Habanero Scala [21] enable controlled concurrency within an actor.
8. Coqa [25] is a programming language that supports different kinds of objects with properties similar to traditional objects or actors.

3.2 Consistency-Related Properties and Constraints

We now list some other properties and constraints that enable better control of consistency by making some program properties explicit to the programmer or the development environment.

1. Monotonicity: this property means that the final order or content of input will never cause any earlier output to be “revoked” once it has been generated. This means that any order of operation generates the same consistent result. Bloom is a distributed language that exploits this property by providing programmers with order-independent abstractions, which encourages a programming style that minimizes coordination requirements and guarantees consistency. [3]
2. Latency and failure: X10, Scala and other languages supports futures [17]. These constructs materialize latency and failure as datatypes and operation effects. Their advantage is that they make these code properties explicit. In distributed environment, failure of operations should always be considered. Multicore programmers need not be concerned with a CPU core going down or instructions getting lost between system buses, but network developers always face node failure and message loss. Futures make failure handling a first-class action in a distributed language.
3. Locality: Loci [41] is a pluggable type system that enables programmers to make thread locality of data explicit and statically checkable. This has many advantages: programmers need not to worry about the consistency of data proved to be local, and compilers can perform more optimizations with such data.
4. Commutativity and Inverses: DPJ and the transactional boosting method [19] allow programmers to specify commutative and inverse methods. Commutativity permits re-ordering of method calls while ensuring deterministic semantics, and inverses allow creating more compact rollback logs for transactional memory.
5. Isolation Levels: In databases, isolation levels can be attached to transactions [6]. The consistency of dataset resulting from a transaction would depend on the specified isolation level.
6. Effects: A problem with object-oriented systems is that operations on data are grouped as functions. A function binding in many systems is determined at runtime. Functions can perform arbitrary reads and writes, and because they are dynamically bound, it is hard to tell at compile-time which pieces of data are affected by a function. Effect systems allow the programmer to declaratively attach effects to methods, and some of them can infer the effects partially [27]. Read/write effect systems exist [26]. Cormac and Shaz [14] propose an effect system that declares atomicity as a property of methods.
7. Transactional Memory: Software transactional memory [36] borrows the ideas of database transactions and apply them to generic programs. Such systems are known for they high overhead because of the delay of creating operation logs and performing roll-backs. Not all software systems can tolerate such overheads. However transactions are easier to reason about than locking, and they are composable.

8. Other replicated-data consistency guarantees: There are many other consistency level between ONE and ALL mentioned earlier. Cassandra also support QUORUM, which employs a quorum-based technique [15]. Session guarantees [37] provide an application with a view of the replicated database that is consistent with its own reads and writes performed in a session even though these operations may be directed at different servers. Guarantees are provided within the context of a session. a single application may create several sessions that it uses to exercise fine-grain control over the guarantees it desires.

4 Current State of Work and Research Directions

As mentioned, we aim to facilitate programming with multiple consistency levels in the same application. This can be done by providing programmers with programming constructs to express data-centric consistency properties, and building a compiler and runtime toolchain that enforces these constraints. I have implemented atomic sets [12] in Scala programming language, and studied other program properties that contribute to building applications with known consistency properties. I also examined the concurrent skip list map [24] from the Java collections library as a possible use-case, because it contains both strongly and weakly consistent data regions. It did not seem at that point that multicore applications with similar properties are plentiful, so I started examining some eventually consistent, distributed key-value stores, focusing on Cassandra, and also studied the actor model and reactive programming to get to know distributed programming patterns and to find other use cases. I am currently designing language constructs to allow appending Cassandra consistency policies to data regions in Scala. I am also building a sample application to demonstrate their use and benefits.

I believe that the approach we propose would be most appreciated in distributed applications. Consequently, we need to ensure that it integrates well with a distributed programming model, which includes, but is not limited to the following: programmers should be able to intuitively program assuming asynchronous-by-default operations, similar to that proposed by the actor model, which is now popular because it truly reflects how the network operates. Furthermore, development frameworks nowadays also make other properties of a distributed application first-class citizens in code. These include failure, as in Scala's reactive streams and try types, latency represented by futures, as implemented in Java, Scala and X10, and data residing at different places as implemented in X10. I would argue that a major problem in distributed programming is the current data abstractions. We cannot tell, by merely looking at a data type, whether it represents replicated data, inconsistent or strongly consistent data or some other essential data properties. As we have shown earlier, objects as data containers are challenging to reason about, so we need to investigate how to tame this model to be used in our approach, which most likely means that we need to implement an ownership and alias management type system, as well as various effect systems. We may also need to introduce further data abstractions, similar to DPG regions or X10 places.

We aim to develop our framework as a declarative, domain specific language. The declarative approach of programming now trending, and programming languages such as Scala facilitate it, and encourage programmers to create embedded languages tailored for the needs of a class of applications.

Finally, to summarize our vision: we plan to build declarative programming language constructs that enable expressing, inferring and enforcing various consistency policies and distributed application properties in a global, data-centric manner.

5 Achievements

So far I have achieved the following:

- Attended a workshop on software correctness and reliability at ETH Zurich.
- Attended the following courses: Bugs 2013 (4 ECTS) and Introduction to Ph.D. studies (2 ECTS)
- Attended ECOOP 2014 with co-located workshops and summer schools.
- Served as a TA for the following courses: Programming Fundamentals 3 (4 ECTS) and Automata and Formal Languages (2 ECTS). I am currently the TA of Massively Parallel Programming (3 ECTS)

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