

A Way Forward in Parallelising Dynamic Languages

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Abstract

Dynamic languages became very popular in recent years. At some point, the need for concurrency arose, and many of them made the choice to use a single global interpreter lock (GIL) to synchronise the interpreter in a multithreading scenario. This choice, however, makes it impossible to actually run code in parallel.

Here we want to compare different approaches to replacing the GIL with a technology that allows parallel execution. We look at fine-grained locking, shared-nothing, and transactional memory (TM) approaches. We argue that software-based TM systems are the most promising, especially since they also enable the introduction of large, parallelisable atomic blocks as a better synchronisation mechanism in the language.

Keywords transactional memory, dynamic languages, parallelism, global interpreter lock

1. Introduction

In a world where computers get more and more cores and single-thread performance increases less and less every year, many dynamic languages have a problem. While there is certainly a lot of popularity around languages like Python and Ruby, their ability to make use of multiple cores is somewhat limited. For ease of implementation, they chose to use a single, global interpreter lock (GIL) to synchronise the execution of code in multiple threads. While this is a straight-forward way to eliminate synchronisation issues in the interpreter, it prevents parallel execution. Code executed in multiple threads will be serialised over this GIL so that only one thread can execute at a time.

There exist several solutions and workarounds to remove or avoid the GIL in order to benefit from multiple cores. We are going to discuss several of them and try to find the best way forward. The first approach uses fine-grained locking to replace the single GIL. Then there are shared-nothing models that use for example multiple processes with multiple interpreters and explicit message passing. Finally, one can also directly replace the GIL with transactional memory (TM), either software-based (STM) or hardware-based (HTM).

The approach that wins in the end should perform similarly for single-threaded execution as compared to the GIL and be able to

execute code in parallel on multiple cores. Furthermore, we will also take into account the compatibility with existing code that may already use threads for concurrency, as well as the changes that are required to the interpreter itself.

These requirements are not easy to meet. The author's position is that STM provides the best way forward. While STM currently has a big performance problem, it gets more points in the other categories. We think that it is the only solution that also provides a better synchronisation mechanism to the application in the form of parallelisable atomic blocks. In the following section, we try to present a balanced view of the compared approaches.

2. Discussion

In this section we first explain the motivation for using a GIL and then examine different approaches to remove or avoid it – highlighting their advantages and disadvantages.

2.1 Why is there a GIL?

The GIL is a very simple synchronisation mechanism for supporting multithreading in an interpreter. The basic guarantee is that the GIL may only be released in between bytecode instructions¹. The interpreter can thus rely on complete isolation and atomicity for the instructions' execution. Also, accesses to data structures like dictionaries and lists happen atomically and do not need additional protection from data races when shared between threads.

The GIL also provides the application with a sequential consistency model [13]. This can be very valuable as it means less surprises for the programmer. For example in Table 1, the programmer can expect the critical section to only be entered by one thread. On the other hand, if the model allowed to buffer the writes, both threads may enter the critical section at the same time.

Thread 1	Thread 2
A = B = 0	
A = 1	B = 1
if B == 0:	if A == 0:
only one thread enters here (e.g. critical section)	

Table 1. Critical section with a sequential consistency model.

As a consequence, applications can rely on certain operations to be atomic and that they will always be executed in the order in which they appear in the code. While depending on this may not always be a good idea, it is done in practice. A GIL-replacement

¹ This also applies to Abstract Syntax Tree (AST) interpreters, where the GIL may only be released between interpreting two AST nodes. We talk about “bytecode instructions” in a general way as a basic step in the interpreter.

should therefore uphold these guarantees, while preferably also be as easily implementable as a GIL for the interpreter. The latter can be especially important since many of these languages are developed and maintained by very large open-source communities, which are not easy to coordinate.

The GIL also allows for easy integration with external C libraries that may not be thread-safe. For the duration of the calls, we simply do not release the GIL. External libraries that are explicitly thread-safe can voluntarily release the GIL themselves in order to still provide some parallelism. This is done for example for potentially long I/O operations. Consequently, I/O-bound, multithreaded applications can actually parallelise to some degree. Again, a potential solution should be able to integrate with external libraries with similar ease. We will however focus our argumentation more on running code in the interpreted language in parallel, not the external C code.

Since the GIL is mostly an implementation detail of the interpreter, it is not exposed to the application running on top of it. To synchronise memory accesses in applications using threads, the state-of-the-art still means explicit locking everywhere. It is known that using locks for synchronisation can be hard at times [14, 15, 18]. They are non-composable, have overhead, may deadlock, limit scalability, and add to the overall complexity of the program logic. For a better parallel programming model for dynamic languages, we propose another, well-known synchronisation mechanism called *atomic blocks* [16, 17]. This is also suggested by [14, 15] as an easier mechanism than locks.

Atomic blocks are composable, deadlock-free, higher-level and expose useful atomicity and isolation guarantees to the application for a series of instructions. Interpreters using a GIL can simply guarantee that the GIL is not released during the execution of the atomic block. Of course, this still means that no two atomic blocks can execute in parallel or even concurrently. Potential solutions are preferable if they provide a good way to implement atomic blocks (or another, comparable synchronisation mechanism) that are also able to be executed in parallel.

2.2 Potential Solutions

For the discussion, we define a set of criteria to evaluate the potential solutions for removing or avoiding the GIL and its limitations:

Performance: How much does the approach impact performance on a single thread and how much on multiple threads? Can it make use of parallelism?

Existing applications: How big are the changes required to integrate with and parallelise existing applications?

Better synchronisation: Does the approach enable better, parallelisable synchronisation mechanisms for applications (e.g. atomic blocks)? Many synchronisation mechanisms can be built on top of all solutions (e.g. message passing, monitors). We look for mechanisms that are directly enabled by the contending approaches.

Implementation: How difficult is it to implement the approach in the interpreter?

External libraries: Does the approach allow for easy integration of external libraries?

2.2.1 Fine-Grained Locking

The first obvious candidate to replace the GIL is to use multiple locks instead of a single global lock. By refining the granularity of the locking approach, we gain the ability to run code that does not access the same objects in parallel. What we lose instead is the simplicity of the GIL approach. With every additional lock, the likelihood of deadlocks grows, as well as the overhead that acquiring and

releasing locks produces. The former means that sometimes it is necessary to fall back to less fine-grained locking, preventing some potential parallelism in order to keep the complexity manageable. The latter means that we lose a bit of performance compared to the GIL, which requires much less acquire-release operations.

Jython [2] is one project that implements an interpreter for Python on the Java Virtual Machine (JVM) and that uses fine-grained locking² to correctly synchronise the interpreter. For a language like Python, one needs quite a few, carefully placed locks – every dictionary, list, instance, or mutable object in general needs a lock. Compared to e.g. Java, object attributes are backed by a dictionary. Accesses to it must be synchronised because the interpreter could crash otherwise. Since there is no central location for all these locks, the complexity of the implementation is quite a bit larger compared to using a GIL. Integrating external, non-thread-safe libraries should however be very simple too. One can simply use one lock per library to avoid this issue.

In the end, fine-grained locking can transparently replace the GIL and therefore parallelise existing applications, generally without any changes. An implementation has to follow the GIL semantics very closely, otherwise it may expose some latent data races in existing applications which are just not exposed with a GIL³. This approach does however not provide a better parallelising synchronisation mechanism to the application and still requires explicit locking in the application.

2.2.2 Shared-Nothing

There are also approaches that work around the GIL instead of trying to replace it. If an application can be split into completely independent parts that only very rarely need to share something, or only do so via an external program like a database, then it is sensible to have one GIL per independent part. At the extreme, there are applications that parallelise perfectly simply by running independent processes; some web servers and some numeric computations do.

We will consider here a slightly more general approach: the *multiprocessing* [3] module of Python. In essence, it uses process-forking to provide the application with multiple interpreters that can run in parallel. Communication is then done explicitly through pipes.⁴

The model of explicit communication is sometimes seen as a superior way to synchronise concurrent applications because of its explicitness. However, not every application fits well into this model and its applicability is therefore limited. Performance is good as long as the application does not need to communicate a lot, because inter-process communication is relatively expensive. Also the implementation of this approach is very cheap since one can actually take an unmodified GIL-supported interpreter and run several of them in parallel. That way, we also inherit the easy integration of external libraries without any changes.

2.2.3 Transactional Memory

Transactional memory (TM) can be used as a direct replacement for a single global lock. Transactions provide the same atomicity and isolation guarantees as the GIL provides for the execution of bytecode instructions. So instead of acquiring and releasing the

² The performance impact of fine-grained locking is milder on the JVM than it would be in a typical piece of C code; see e.g. [20].

³ There are rare cases where not having atomic bytecodes actually changes the semantics. E.g. in Jython, `dict1.update(dict2)` is not atomic: it first reads data from `dict2` with `dict2`'s lock, and then puts it into `dict1` with `dict1`'s lock. A lot can happen in-between.

⁴ There are multiple alternative designs like e.g. actors or tuple spaces. Since they are similar and do not replace the GIL directly, we focus on the example of *multiprocessing*.

	GIL	Fine-grained locking	Shared-nothing	HTM	STM
Performance (single threaded)	++	+	++	++	--
Performance (multithreaded)	--	+	+	+	+
Existing applications	++	++	--	++	++
Better synchronisation	o	o	+	o	++
Implementation	++	-	++	++	++
External libraries	++	++	++	++	++

Table 2. Comparison between the approaches (--/o/+/++)

GIL between these instructions, this approach runs the protected instructions inside transactions.

TM can be implemented in software (STM) or in hardware (HTM). There are also hybrid approaches, which combine the two. We count these hybrid approaches as STM, since they usually provide the same capabilities as software-only approaches but with different performance characteristics. We will now first look at HTM, which recently gained a lot of popularity by its introduction in common desktop CPUs from Intel (Haswell generation) [4, 19].

HTM provides us with transactions like any TM system does. It can be used as a direct replacement for the GIL [4, 8, 9]. However, as is common with hardware-only solutions, there are quite a few limitations that can not be lifted easily. For this comparison, we look at the implementation of Intel in recent Haswell generation CPUs.

HTM in these CPUs works on the level of caches. This has a few consequences like false-sharing on the cache-line level, and most importantly it limits the amount of memory that can be accessed within a transaction. This transaction-length limitation makes it necessary to have a fallback in place in case this limit is reached. In recent attempts, the usual fallback is the GIL [4, 9]. In our experiments, the current generation of HTM proved to be very fragile and thus needing the fallback very often. Consequently, scalability suffered a lot from this.

The performance of HTM is pretty good as it does not introduce much overhead (< 40% [4]). And it can transparently parallelise existing applications to some degree. The implementation is very straight-forward because it directly replaces the GIL in a central place. HTM is also directly compatible with any external library that needs to be integrated and synchronised for use in multiple threads. The one thing that is missing is support for a better synchronisation mechanism for the application. It is not reasonable in general to expose the hardware-transactions to the application in the form of atomic blocks, because doing so would require the system to support much longer transactions.

STM provides all the same benefits as HTM except for its performance. It is not unusual for the overhead introduced by STM to be between 100% to even 1000% [6, 7]. While STM systems often scale very well to a big number of threads and eventually overtake the single-threaded execution, they often provide no benefits at all for low numbers of threads (1-8). There are some attempts [5, 12] that can reduce the overhead a lot, but scale badly or only for certain workloads. Often the benefits on more than one thread are too small in real world applications.

However, STM compared to HTM does not suffer from the same restricting limitations. Transactions can in principle be arbitrarily long. This makes it possible to expose transactions to the application in the form of atomic blocks – thereby attacking the issues of parallelisation and synchronisation in a unified way. While many synchronisation mechanisms can be bolted on top of any GIL replacement, this is the only approach that directly enables a better, parallelising synchronisation mechanism than locks. We think this is a very important point because it not only gives dynamic

languages the ability to parallelise (already commonplace in most other languages), but also pushes parallel programming forward in a way that other approaches cannot. Together with sequential consistency, it provides an environment for parallel applications that has much less surprises than e.g. Java or C#.

On the implementation level, while one can argue that STM requires the insertion of read and write barriers in the whole interpreter, this can be done automatically and locally by a program transformation [10]. There are attempts to do the same for fine-grained locking [11] but they require a whole program analysis since locks are inherently non-composable — and their effectiveness is doubtful in our use case, since we execute bytecode instructions in any order defined by a script only known at runtime. This makes it close to impossible to order locks consistently or to know in advance which locks a transaction will need.

3. The Way Forward

Following the above argumentation for each approach, we assembled a general overview in Table 2. The points were assigned according to the criteria described in 2.2. Since the criteria are defined intuitively, there are no formal justifications for the number of points. The reader is thus advised to take the result with a grain of salt and form their own opinion.

The general picture is everything else than clear. It looks like HTM may be a good solution to replace the GIL in the near future. Current implementations are however far too limiting, not widely available, and do not provide good scaling.

Allowing for parallel execution just means that dynamic languages catch up to all other languages that already provide real parallelism. This is why we think that only the STM approach is a viable solution in the long-term. It unifies both, the simple memory model (sequential consistency) and the synchronisation of memory accesses using composable atomic blocks. It is not *just* a simple GIL replacement.

Unfortunately, STM has a big performance problem, which currently makes it lose this comparison. Particularly, for our use case there is not much static information available since we are executing a program only known at runtime. Additionally, replacing the GIL means running every part of the application in transactions, so there is not much code that can run outside and that can be optimised better. The performance of the TM system is vital.

One way to get more performance is to develop STM systems that make better use of low-level features in existing OS kernels. We are currently working on an STM system that makes use of several such features like virtual memory and memory segmentation. We further tailor the system to the discussed use case, which gives us an advantage over other STM systems that try to be more general or simply focus on other use cases. With this approach, initial results suggest that we can keep the overhead of STM well below 50%. A hybrid TM system, which also uses HTM to accelerate certain tasks, looks like a very promising direction of research too.

We think that further work to reduce the overhead of STM is very worthwhile. In fact, considering some analogies that have been

drawn between garbage collection and transactional memory [1], we think that it is worthwhile to focus the STM research more specifically onto the context shown in this paper – for use in implementations of high-level languages, rather than as a tool directly used by the programmer.

Acknowledgments

We would like to thank Maciej Fijałkowski and Carl Friedrich Bolz for their valuable inputs and the many fruitful discussions.

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