Return to java.util.concurrent

CSCI 5828: Foundations of Software Engineering Lecture 24 — 11/13/2014

Goals

- Explore the services of java.util.concurrent
 - ExecutorService
 - Callable/Future
 - ForkJoinPool and ForkJoinTask

ExecutorService (I)

- ExecutorService is a Java interface that defines a common set of services for an abstract "thread pool";
 - this interface has a variety of concrete implementations that provide a choice of concurrent behavior to developers
- What's a thread pool? (A review)
 - Thread creation is a slow process
 - Thread pools create a bunch of threads all at once (typically at init time)
 - When a new thread is needed, one is taken from the pool and it starts executing immediately
 - very helpful in situations where, e.g., a server is responding to incoming network requests

ExecutorService (II)

- Static factory methods on the Executors class are used to create instances of the ExecutorService; for instance
 - CachedThreadPool: creates threads as needed but will reuse previous ones if they are available
 - FixedThreadPool: creates a fixed set of threads
 - ScheduledThreadPool: creates a thread pool that can execute tasks after a delay or periodically
 - SingleThreadExecutor: creates a thread pool with only a single thread
- You can write code that only depends on the interface ExecutorService and then be free to change the actual threading behavior you get at run-time based on external factors (you can even switch threading behaviors on the fly)

ExecutorService (III)

- The basic API of the ExecutorService allows you to
 - submit a single task for execution
 - submit a collection of tasks for execution
 - where you want all of the tasks results (invokeAll)
 - or where you want just one of the results (invokeAny)
 - shutdown the thread pool when you are done with it

Callable/Future: Making this all work

- In order to give tasks to the thread pool and receive results back, you make use of two additional interfaces
 - Callable<T> and Future<T>
- Both make use of Java generics to give flexibility in the return types of the computation
 - For instance, I can promise that my task returns a string
 - Callable<String> callMe = new Callable<String>() {
 - public String call() throws Exception { ...; return result; }
 - }
 - callMe is now a Task that I can hand to an ExecutorService

Callable/Future (II)

- When I give callMe to an ExecutorService, it is going to hand the task to a thread and ask the thread to execute it
 - At the time, we have no idea how long it will take for the task to complete
 - Thus, the ExecutorService gives me an instance of Future<String> back so I can get the value once the task is complete
 - Future<String> myString = service.submit(callMe);
 - This call does NOT block, I get a reference to myString almost immediately
 - I can then decide to retrieve the string whenever I need it by calling get()
 - String result = myString.get();
 - This call MAY block, if the task is still being executed; otherwise, I get the result right away. I can also call a version of get() that accepts a timeout.

Portfolio Calculator (I)

- A simple program to retrieve stock quotes from Yahoo
 - Designed as one abstract superclass AbstractNAV (Net Asset Value)
 - · methods for
 - reading in stock symbols and number of shares
 - timing how long it takes to calculate the portfolio
 - Two subclasses
 - one sequential => loops over list of symbols and calls Yahoo to get current price
 - one concurrent => creates futures for each stock to retrieve prices;
 hands them all over to an executor service to execute in parallel

Portfolio Calculator (II)

- Since these tasks are IO Bound, the program needs to decide how many threads it will need
 - Each task is going to be blocked for most of its life and then it will do a single calculation (numberOfShares * retrievedPrice)
 - We estimate that waiting for Yahoo to give us the current price is going to take about 90% of the task's life cycle
 - So to estimate the number of threads, we use the following formula
 - Number of Threads = Number of Cores / (1 Blocking Coefficient)
 - My machine has 8 cores, so
 - 8/(1-0.9) = 8/.1 = 80 threads

Portfolio Calculator (III)

- It then creates an array to store all of our Callable objects
 - final List<Callable<Double>> partitions = new ArrayList<Callable<Double>>();
- It populates that List by creating one Callable<Double> for each stock symbol
 - It then hands the list over to the executor service which hands back a list of Future<Double> objects
 - final List<Future<Double>> valueOfStocks =
 executorPool.invokeAll(partitions, 10000, TimeUnit.SECONDS);
 - Finally, it loops over each Future object and totals up the final value
 - for (final Future<Double> valueOfAStock : valueOfStocks) {netAssetValue += valueOfAStock.get();}
 - It is this call to get() that can finally block, waiting for the task to complete

Portfolio Calculator (IV)

- Let's see this in action
 - As you will see, the concurrent version of the program is significantly faster
 - Why? => Latency!
 - Each request to Yahoo takes a certain amount of time to create the connection, wait for Yahoo to retrieve the data, and stream the result back
- With the sequential version of the program, we take that latency and add it for each stock request; say the latency was 2 seconds
 - For 80 stocks, we would expect to wait 160 seconds for the whole sequential program to complete
 - In the concurrent program, all tasks contact Yahoo at the same time, the 2 second latency for each task overlaps. As a result, the program takes ~2-3 seconds

Finding Primes (again)

- The reason we saw such an amazing speed-up with the previous program was due to the fact that its tasks were IO-bound.
 - With compute-bound tasks, we have to limit the number of threads to the number of cores
- I won't spend much time on this example since we've seen it many times
 - Let's just look at the code briefly to see how the code creates a bunch of Callable<Integer> tasks that count primes for a given partition
 - The executor service then gives us back a list of Future<Integers> and we call those to total up the number of primes in a given range

Coordinating Threads (I)

- A key challenge in the design of concurrent systems is the coordination of threads
 - We may want to
 - start them
 - wait for them to finish
 - assign tasks to them
 - retrieve results from them
 - allow threads to exchange data
 - etc.

Coordinating Threads (II)

- With the ExecutorService, the most typical case now involves
 - submitting a task to a thread pool of type Callable
 - receiving a Future in response
 - when ready, calling get() on the Future to retrieve the result
- Let's see this in action with an example called File Size Calculator
 - First, let's take a look at the sequential version of this program
 - The examples in this lecture come from the Programming Concurrency on the JVM book from Pragmatic Programmers
 - Design is straightforward; recursive function that returns either the size for a single file or for directories, the combined size of all of its children

Disk Cache

- With programs that target the disk, performance will vary
 - The first time through a particular section of the disk, the time will be slower than subsequent runs on the same section of the disk
- The reason for this is the disk cache
 - The operating system will
 - take the most recently read sections of disk
 - and cache them in memory
 - under the assumption that they will be read again fairly soon
- The difference may not be major but it will be there
 - First run sequential on /usr: 34.1 seconds; Second run: 30.9 seconds

First Stab at Concurrency

- Creates a thread pool of 100 threads
- Makes use of recursive function to calculate size of files and directories
 - If its handed a file, return the file size
 - If its handed a directory
 - loop through children
 - submit() a task to the thread pool to calculate the size of the child
 - Each task is a Callable<Long>
 - Thread pool returns a Future<Long> that gets added to an array
 - loop through array calling get() on each Future to add up subtotals
 - return the result

Result? DEADLOCK!

- This approach to the program has a flaw that appears on "deep directories"
 - Each task adds new tasks to the thread pool and then waits for those tasks to return
 - That means that the calling task is STILL ON THE POOL
 - blocked waiting for its subtasks to complete
 - If your directory has lots of subdirectories (more than 100 in this case)
 - You can get into the situation where each of the 100 threads in the thread pool are blocked waiting for subdirectory calculations to complete
 - when this happens, the program deadlocks
 - or thanks to the timeout that we set, eventually the timeout fires and the program terminates

Discussion

- This problem is unfortunate because
 - the approach is straightforward and understandable
 - you'd likely come up with it on a first pass design
- But, a machine's resources are finite
 - you might be able to make this code work on more directories by upping the number of threads
 - but that approach is not generic
 - eventually you'll run into the limit concerning the number of threads the operating system will allow a single process to create
 - and you'll be stuck

New Approach: Find Directories, Total Later

- To make progress, we need an approach that
 - submits tasks for sub-directories
 - but doesn't require the submitting task to hang around for the results
- New Approach
 - Create a data structure that holds the total size of a directory's files and a list of all of that directory's sub-directories
 - Tasks now calculate the size of files in their assigned directory and create a list of all subdirectories; allowing them to complete and not stick around
 - The main thread takes care of submitting new tasks and totaling results
 - Demo

Terrific Results But...

- increased complexity!
 - We got great results but the approach we used is not intuitive
 - Creating a class to store partial (immutable) results
 - Creating the function executed by tasks such that it completes quickly
 - Adopting a while loop strategy in main to iterate while there were directories to process
 - and then ensuring that the while loop would not terminate until all directories had been processed
- Let's look at features that java.util.concurrent has that might reduce the complexity of the code

CountDownLatch (I)

- The next approach examines the use of a CountDownLatch
 - plus it relaxes our constraint to avoid shared mutability
- but it achieves the same results with simpler code
 - Simplicity is not to be discounted
 - it has significant impacts on the ability to maintain software systems

CountDownLatch (II)

- What's a CountDownLatch?
 - It is a synchronization aid to help coordinate threads
 - It maintains a count and has three primary methods
 - CountDownLatch(n) creates the latch with a specific count
 - await() block the calling thread until the latch's count == 0
 - countDown() -- decrement the count of the latch
- Typical scenario:
 - · create a bunch of threads and start() them
 - but don't let them run() until some point in the future
 - i.e. have their first line in run() call await()

New Approach

- Instead of returning subdirectories, we let each task update two shared variables
 - each an instance of AtomicLong (like AtomicInteger but stores long value)
- One AtomicLong stores the total file size
- The second AtomicLong stores the number of "pending file visits"
 - This value gets incremented each time we find a subdirectory to visit
 - It gets decremented each time we are done processing a subdirectory
 - When this value equals zero, we call countDown() on the latch
- The main thread initializes the latch to a value of 1, starts the directory search, and calls await()

Performance

- Comparable performance to previous approach
 - but with simpler code
- We actually anticipate that this approach would be slightly slower than the previous approach due to the extra thread synchronization
 - Each call to AtomicLong involves thread synchronization
 - threads do not necessarily block
 - (only happens when there is contention)
 - but a monitor of some sort will be checked and that slows things down

Third Approach: Queue (I)

- We have seen two approaches for exchanging data between threads
 - Callable/Future and Atomic<Type>
- both techniques ensured that we could pass information between threads
- A third approach is to use a data structure such as a queue to pass information between threads
 - as long as there is space in the queue, producers will not block
 - · as long as there are items on the queue, consumers will not block
 - contention will occur only when the queue is full (producers) or when it is empty (consumers)

Third Approach: Queue (II)

- This version of the program creates a blocking queue with 500 slots
- An atomic long is used to keep track of pending file visits
- Tasks traverse the directories as normal, adding file sizes to the queue and updating the atomic long as they submit more tasks to the thread pool
- The main program kicks off the traversal and then sits in a loop
 - that reads items off the queue until there are no more file visits pending and the queue is empty
- Performance:
 - First Run: 24.6 seconds; Second Run: 10.9 seconds
 - Same performance, just slightly different abstractions, perhaps simpler
 - not by much

Java 7: Fork-Join API

- Java 7 introduced a new type of thread pool and task
 - ForkJoinPool and ForkJoinTask
- The key benefit of this new thread pool is that threads can steal tasks generated by other active tasks
 - This solves the problem we encountered with the first approach to the concurrent file size calculator
 - When a task generates a bunch of other tasks and blocks, its thread can let it go and work on the other tasks
- With this approach, we get a program very similar to our "naive" approach
 - without the danger for deadlock like we saw before

Summary

- We learned the ins and outs of using the ExecutorService in various ways
 - Saw how Callable and Future work to allow us to pass information between threads
 - Explored various problems that can still occur when using ExecutorService
 - Saw a number of different ways to design the same program
 - Performance was usually the same
 - What was different was the complexity of each design
 - Certain designs provided more simplicity than others
 - If two designs perform the same, prefer the one that is less complex to make it easier to maintain that solution

Coming Up Next

- Lecture 25: Refactoring a poorly designed concurrent program written in Java
- Lecture 26: Design of Design, Part 2