Outline of Lecture 15

- Parallel programming in Haskell
- Lazy evaluation and parallel evaluation (the Eval monad)
- Dataflow programming (the Par monad)
- Functional programming and its applications outside Haskell
- Exam information. The topics for exam exercises

Parallel Haskell

- Parallel Haskell is aimed at providing access to multiple processors in a natural and robust way
- Advantage: parallel programming in Haskell is deterministic (thus, adding parallelism leads to no race conditions or deadlocks)
- Advantage: Haskell program are high level and declarative
 - more likely to work on wide range of parallel hardware
 - by default, take advantages of the current highly tuned technologies (like parallel garbage collection) and their future advancements
- **Disadvantage**: A lot of execution details are hidden ⇒ performance problems can be sometimes hard to understand

Parallel Haskell (cont.)

- The main thing that the parallel Haskell programmer has to think about is partitioning the problem into parallel computations
- Two sub-issues
 - Granularity: not too high, not too low to keep the processors busy.
 Depend on static or dynamic partitioning
 - Data dependencies: when one task depends on another, they must be performed sequentially. Two approaches in Haskell: entirely implicit (using the Eval monad) or explicit (using the Par monad).

The latter (explicit) one is less concise but it can be easier to understand and fix problems

Lazy evaluation and basic paralellism

- Haskell is lazy ⇒ expressions are not evaluated until they are required
- Example:

```
Prelude> x = 1 + 2 :: Int
```

- Semantically, it looks equivalent to x = 3
- Calculationally, the computation of 1 + 2 has not happened yet, so we might compute it in parallel with something else
- We say that at this point x is unevaluated
- Printing the actual value of x without causing it to be evaluated:

```
Prelude> :sprint x
x = _
```

- Unevaluated, the expression 1+2 is just an object in memory (called thunk) and x is a pointer to it
- Forcing evaluation:

```
Prelude> x
3
Prelude> :sprint x
x = 3
```

• Dependent definitions:

```
Prelude> x = 1 + 2 :: Int
Prelude> y = x + 1
Prelude> :sprint x
x = _
Prelude> :sprint y
y = _
```

 One step of evaluation with the built-in Haskell function seq (evaluates the first argument and returns the second one)

```
Prelude> seq y ()
()
Prelude> :sprint x
x = 3
Prelude> :sprint y
y = 4
```

 seq forces one-step evaluation, e.g., for the first constructor, and stops there:

```
Prelude> xs = map (+1) [1..10] :: [Int]
Prelude> :sprint xs
xs = _
Prelude> seq xs ()
()
()
Prelude> :sprint xs
_ : _
```

Applying length to xs (no need to evaluate the actual elements):

```
Prelude> length xs
10
Prelude> :sprint xs
xs = [_,_,_,_,_,_]
```

Forcing full evaluation of xs:

```
Prelude> sum xs
65
Prelude> :sprint xs
xs = [2,3,4,5,6,7,8,9,10,11]
```

 By using Haskell pre-defined features from the Control module, we can control the evaluation process: which evaluations to do in parallel or in sequence

The Eval monad

 The basic functionality for creating parallelism is provided by the Eval monad (the module Control.Parallel.Strategies):

```
data Eval a
instance Monad Eval

runEval :: Eval a -> a

rpar :: a -> Eval a
rseq :: a -> Eval a
```

- The rpar combinator creates parallelism, i.e., instructs the unevaluated argument to be evaluated in parallel. Does not wait for the result before passing the control
- The rseq combinator forces sequential evaluation: evaluates the arguments and waits for the result
- runEval performs the Eval computation and returns its result

The Eval monad (cont.)

• Example:

```
runEval $ do
  a <- rpar (f x)
  b <- rpar (f y)
  return (a,b)</pre>
```

- f x and f y begin to evaluate in parallel, while return happens immediately
- The rest of the program will continue to execute while f x and f y are being evaluated in parallel
- By default, only a single evaluation step (for the first encountered constructor) is done

The Eval monad (cont.)

• Example:

```
runEval $ do
  a <- rpar (f x)
  b <- rpar (f y)
  rseq a
  rseq b
  return (a,b)</pre>
```

- f x and f y begin to evaluate in parallel
- rseq a and rseq b force to wait until the evaluation is over before returning
- Many other evaluation strategies are possible

• The given Sudoku module contains a function

```
solve :: String -> Maybe Grid
```

that solves a single Sudoku problem (represented as String). The result is Nothing if no solution exists

 Sequential code (sudoku1.hs) to solve a set of Sudoku problems from a file:

```
import Sudoku
...
main :: IO ()
main = do
  [f] <- getArgs
  file <- readFile f
  let puzzles = lines file
        solutions = map solve puzzles
  print (length (filter isJust solutions))</pre>
```

• Compiling sudoku1.hs with full optimisation:

```
$ ghc -02 sudoku1.hs -rtsopts
[1 of 2] Compiling Sudoku ( Sudoku.hs, Sudoku.o )
[2 of 2] Compiling Main ( sudoku1.hs, sudoku1.o )
Linking sudoku1 ...
```

 Running sudoku1.hs on 1000 sample problems (with extra options to get statistics):

```
$ ./sudoku1 sudoku17.1000.txt +RTS -s
1000
...
Total time 0.890s ( 0.899s elapsed)
```

- Let us split the list of 1000 problems in two and run the solver in parallel on two cores
- Parallel code (sudoku2.hs):

```
main :: IO ()
main = do
  [f] <- getArgs</pre>
  file <- readFile f
 let puzzles = lines file
      (as,bs) = splitAt (length puzzles 'div' 2) puzzles
      solutions = runEval $ do
         as' <- rpar (force (map solve as))
         bs' <- rpar (force (map solve bs))
         rseq as'
         rseq bs'
         return (as' ++ bs')
 print (length (filter isJust solutions))
```

- force full evaluation instead of one evaluation step
- Compiling sudoku2.hs with full optimisation (threaded version):

```
$ ghc -02 sudoku2.hs -rtsopts
[2 of 2] Compiling Main ( sudoku2.hs, sudoku2.o )
Linking sudoku2 ...
```

Running sudoku2.hs on 1000 sample problems on two cores (option –N2):

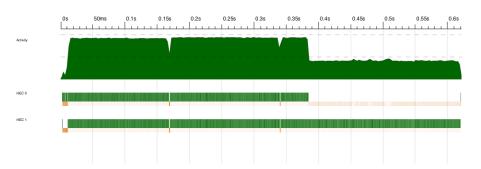
```
$ ./sudoku2 sudoku17.1000.txt +RTS -N2 -s
1000
...
Total time 0.975s (0.608s elapsed)
```

• Speedup on the elapsed time (comparing to the sequential version): 0.899 / 0.608 = 1.48

- The speedup 1.48 is not bad, but could be better. Where are the losses?
- Creating an event log for analysis:

```
$ rm sudoku2; ghc -02 sudoku2.hs -rtsopts
-threaded - eventlog
$ ./sudoku2 sudoku17.1000.txt +RTS -N2 -1
```

• Using the ThreadScope tool to analyse the log



- It looks like the work load between two lists was unevenly distributed and one processor finished executing earlier and stayed idle for quite some time
- How to improve on that problem? We have to make work "chunks" (the tasks to be handled in parallel) smaller

- Haskell supports dynamic partitioning of such work tasks given to rpar to the available idle processors
- We just have to create a pool with enough tasks by calling rpar often enough so it can do its job and balance the work evenly
- First, let's define abstraction that will let us apply a function to a list in parallel:

```
parMap :: (a -> b) -> [a] -> Eval [b]
parMap f (a:as) = do
  b <- rpar (f a)
  bs <- parMap f as
  return (b:bs)</pre>
```

- Let us apply dynamic partitioning and run the solver in parallel on two cores:
- Parallel code (sudoku3.hs):

```
main :: IO ()
main = do
  [f] <- getArgs
file <- readFile f
  let puzzles = lines file
      solutions = runEval (parMap solve puzzles)
  print (length (filter isJust solutions))</pre>
```

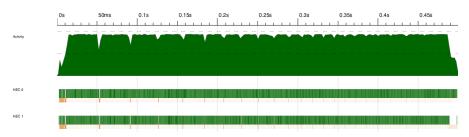
• Compiling sudoku3.hs with full optimisation (threaded version):

```
$ ghc -02 sudoku3.hs -rtsopts
[1 of 2] Compiling Sudoku ( Sudoku.hs, Sudoku.o )
[2 of 2] Compiling Main ( sudoku2.hs, sudoku2.o )
Linking sudoku3 ...
```

Running sudoku3.hs on 1000 sample problems on two cores:

```
$ ./sudoku3 sudoku17.1000.txt +RTS -N2 -s
1000
...
Total time 0.949s (0.490s elapsed)
```

• Speedup on the elapsed time (comparing to the sequential version): 0.899 / 0.490 = 1.83



- Much better work load distribution
- The last run: the same sudoku3.hs, but on 4 cores

```
$ rm sudoku3; ghc -02 sudoku3.hs -rtsopts
$ ./sudoku3 sudoku17.1000.txt +RTS -N4 -s
...
Total time 0.996s (0.265s elapsed)
```

• Speedup on the elapsed time (comparing to the sequential version): 0.899 / 0.265 = 3.39

Dataflow parallelism: the Par monad

- The Eval monad heavily relies on lazy evaluation to express parallelism, allows decoupling of the algorithm from parallelism as well as building different parallel strategies compositionally
- Another parallel programming model, the Par monad, is more explicit about granularity and data dependencies. We have to give more details, but in return gain more control
- Implements dataflow parallelism (dataflow networks)

Dataflow parallelism: the Par monad (cont.)

• The interface:

```
newtype Par a
instance Monad Par

runPar :: Par a -> a
fork :: Par a -> Par a
```

- A computation in the Par monad can be run using runPar to return a pure result
- Parallelism is introduced by fork, the purpose of which is to create parallel tasks
- The Par computation passed as as the fork argument ("the child") is executed in parallel with fork caller

Dataflow parallelism: the Par monad

- fork does not return anything to the caller. How can we get a result back from a created parallel computation?
- Data can be passed between Par computations using the IVar type and its operations:

```
new :: Par (IVar a)
put :: NFData a => IVar a -> a -> Par a
get :: IVar a -> Par a
```

• Here, NFData a (normal-form data) – type class of fully evaluated data, i.e, values with no unevaluated expressions

Dataflow parallelism: the Par monad

- We can think of IVar value as a box that starts empty (after the operation new)
- The put operation stores a value in the box, while the get operation reads the value
- If get finds the box empty, it waits until its is filled by put. The box can be written only once
- So IVar lets us communicate values between parallel Par computations, because we can put a value in the box in one place and get in another

Dataflow parallelism: simple example

 Two independent computations in parallel (calculating Fibonacci values for the distinct arguments m and n)

```
runPar $ do
  i <- new
  j <- new
  fork (put i (fib n))
  fork (put j (fib m))
  a <- get i
  b <- get j
  return (a+b)</pre>
```

Functional programming in other languages

- Increasing number of functional programming features in the current imperative languages (Python, Java, JavaScript, C#): making functions into "first class citizens", constructing functions on-the-fly with lambda notation, functional composition, immutability
- Synergy of FP and OOP in Scala
- Java: a lot of new functional features in Java 8
- Extra material: FP in Java 8, FP and Scala

Applications of functional programming in real world

- Frameworks for Big Data and data analytics (Hadoop Map Reduce, Cascalog, Apache Spark, R Studio, etc.): chaining functional computations, combining and transforming data sets
- Building interactive fault-tolerant concurrent systems and NoSql database systems with Erlang: using immutable data, pattern matching, functional programming for interacting processes exchanging data messages
- Extra material: FP in R, FP and Erlang

Exam information

- Exam date: January 12th, 12.00 14.00, Didlaukio 102 auditorium
- 4-6 small Haskell tasks (slightly easier than homework exercises)
- Must-know exam task topics:
 - Pattern matching; Guards; Local definitions
 - Primitive recursion:
 - Working with lists, including list comprehensions;
 - Higher order functions (such as map, fold, filter), function composition and partial function application;
 - User-defined datatypes;
 - Haskell type classes (no functors or monads!).