

Overview

In this problem set, you will implement a simplified version of the MapReduce framework. The main part of this assignment will require you to implement a coordinator that will distribute work to independent workers and collect the results of the computation. You will implement various applications that make use of your framework.

Objectives

- [TODO — fix]

Recommended Reading

- We have provided documentation for a subset of the Async library that should be sufficient for this assignment: [TODO — link].
- Real World OCaml, chapter 18 [TODO — link]
- Lectures 17 and 18 [TODO — link]
- Recitations [TODO — fix]

Part One: Logic

This part of the assignment asks you to work with formal proof trees. Although groups will submit a single writeup, we strongly recommend that both partners be able to do both problems.

Exercise 1:

For each of the following propositions, construct a tree to serve as a proof. In the 3110 LaTeX package, we have provided tools useful for typing up these proofs.

- (a) $((P \Rightarrow Q) \Rightarrow P) \Rightarrow P$
- (b) $P \Rightarrow (P \Rightarrow (Q \Rightarrow (Q \Rightarrow (P \Rightarrow P))))$
- (c) $((P \wedge Q) \Rightarrow R) \iff (P \Rightarrow (Q \Rightarrow R))$ To prove a bidirectional statement, you must prove it for both directions (i.e. if you are proving $A \iff B$, you must prove both $A \Rightarrow B$ and $B \Rightarrow A$).

Exercise 2:

[TODO — review] The following proof contains an error. Find and correct the error so that the proof is correct.

Let

- $S(d)$: There is a blizzard in Ithaca on day d
- C : Cornell's classes are cancelled
- H : Cornell Students are happy

Given:

- $\forall d S(d) \Rightarrow C$
- $C \Rightarrow H$
- $\neg H$

$$\begin{array}{c}
 \frac{\frac{\frac{\forall d S(d) \rightarrow C \wedge (C \rightarrow H) \quad ((p \rightarrow q) \wedge (q \rightarrow r)) \rightarrow (p \rightarrow r)}{\forall d S(d) \rightarrow H}}{\neg H} \quad (p \rightarrow q) \rightarrow (\neg q \rightarrow \neg p)}{\neg H \rightarrow \forall d \neg S(d)} \\
 \hline
 \forall d \neg S(d)
 \end{array}$$

Part Two: Async warmup

These exercises are intended to get you used to asynchronous computation and the Async programming environment.

Exercise 3:

Implement the asynchronous queue interface defined in `aQueue.mli`. We recommend that you use the `Async.Std.Pipe` module.

Exercise 4:

Write a function `fork` that takes a deferred and two blocking functions, and runs the functions when the deferred becomes determined:

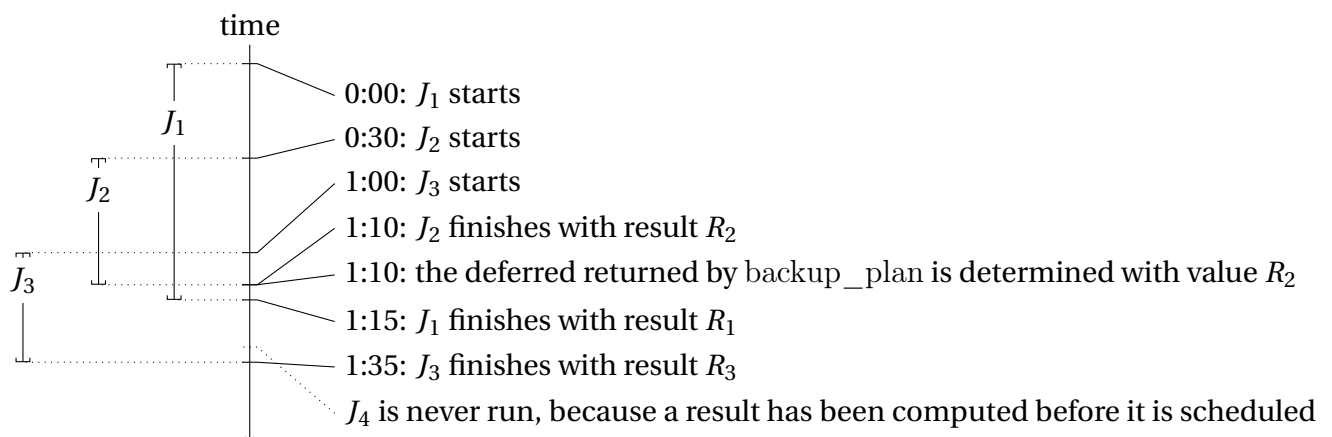
```
val fork : 'a Deferred.t -> ('a -> 'b Deferred.t)
        -> ('a -> 'c Deferred.t) -> unit
```

The output of the two blocking functions should be ignored.

Exercise 5:

Write a function `backup_plan` that takes a list of blocking functions and tries them one at a time, with a 30 second delay in between. It should return a `Deferred.t` that becomes determined when the first of the jobs finishes. Once any of the jobs have been successfully completed, no new jobs should be started.

For example, suppose that we have jobs `j1`, `j2`, `j3`, and `j4`. Suppose that `j1` takes 1:15. `j2` takes 40 seconds, `j3` takes 35 seconds, and `j4` takes 5 seconds. Calling `backup_plan [j1; j2; j3; j4]` would initiate the following sequence of events:



Exercise 6:

Implement the `FilterReader` module. An `'a FilterReader.t` wraps an `'a Pipe.Reader.t` and allows programs to wait for data that matches a predicate to be read from the pipe.

For example:

```
# let (r,w) = Pipe.create ();;
# let filterer = FilterReader.create r;;

# FilterReader.read filterer ((>) 5)
>>| fun x -> printf "reader 1: %i\n" x;;
# FilterReader.read filterer ((>) 3)
>>| fun x -> printf "reader 2: %i\n" x;;
# FilterReader.read filterer ((>) 8)
>>| fun x -> printf "reader 3: %i\n" x;;

# Pipe.write w 3;; (* no readers *)
- : TODO = TODO
# Pipe.write w 4;; (* reader 1 doesn't accept, so reader 2 gets it *)
reader 2: 4
- : TODO = TODO
# Pipe.write w 4;; (* no readers --- reader 2 is done *)
- : TODO = TODO
# Pipe.write w 10;; (* reader 1 and reader 3 could handle, but reader 1 came first. *)
reader 1: 10
- : TODO = TODO
# Pipe.write w 10;; (* reader 1 is gone so reader 3 gets it *)
reader 3: 10
- : TODO = TODO
```

Exercise 7:

Part Three: MapReduce Framework

Modern applications rely on the ability to manipulate massive data sets in an efficient manner. One technique for handling large data sets is to distribute storage and computation across many computers. Google's MapReduce is a computational framework that applies functional programming techniques to parallelize applications.

MapReduce Overview

MapReduce was spawned from the observation that a wide variety of applications can be structured into a **map phase**, which transforms independent data points, and a **reduce phase** which combines the transformed data in a meaningful way. This is a very natural generalization of the `List.fold` function with which you are intimately familiar.

MapReduce jobs provide the code to run in these two phases by implementing the `MapReduce.Job` interface. A MapReduce Job is executed as follows:

- The `map` function takes a single input and transforms the value into a collection of intermediate key, value pairs. The `map` function is called once for each element of the input list.

```
val map : input -> (key * inter) list Deferred.t
```

- The `map` results are then aggregated in a `combine` phase. Values associated with the same key are merged into a single list.
- Finally, the `reduce` function takes a key and an `inter list` to compute the output corresponding to that key. `reduce` is called once for each independent key.

```
val reduce : key * inter list -> output Deferred.t
```

The advantage of this design is that each call to `map` or `reduce` is independent, so the work can be distributed across a large number of machines. This allows MapReduce applications to process very large data sets quickly while using limited resources on each individual machine.

An example: Word Count

Figure 1 shows a distributed execution of a word counting application. The input to the application is a list of filenames. During the map phase, the controller sends a `MapRequest` message for each input filename to a mapper. The mapper invokes the `map` function in the `WordCount.Job` module, which reads in the corresponding file and produces a list of `(word, count)` pairs. The mapper then sends these pairs back to the controller.

Once the controller has collected all of the intermediate `(word, count)` pairs, it groups all of the pairs having the same word into a single list, and sends a `ReduceRequest` to a reducer. The reducer invokes the `reduce` function of the `WordCount.Job` module, which simply returns the sum of all of the counts. The reducer sends this output back to the controller, which collects all of the reduced outputs and returns them to the `WordCount` application.

Communication protocol

In our implementation of MapReduce, the Controller communicates with Workers by sending and receiving the messages defined in the `Protocol` module.

Messages are strongly typed; a message from the Controller to the Worker will have type `WorkerRequest.t`; responses have the type `WorkerResponse.t`. Messages can be sent and received by using the `send` and `receive` functions of the corresponding module. For example, the Controller should call `WorkerRequest.send` to send a request to a worker; the worker will call `WorkerRequest.receive` to receive it.

The `WorkerRequest` and `WorkerResponse` modules are parameterized on the `Job`, so that the messages can contain data of the types defined by the `Job`. This means that before the worker can call `receive`, it needs to know which `Job` it is running. As soon as the controller establishes a connection to a worker, it should send a single line containing the name of the job. After the job name is sent, the controller should only send `WorkerRequest.ts` and receive `WorkerResponse.ts`.

We have provided code in the `Worker` module's `init` function that receives the job name and calls `Worker.Make` with the corresponding module.

Once a connection is established and the job name is sent, the Controller will send some number of `WorkerRequest.MapRequest` and `WorkerRequest.ReduceRequest` messages to the worker. The worker will process these messages and send `WorkerRequest.MapResult` and `WorkerRequest.ReduceResult` messages respectively. When the job is complete, the controller should close the connection.

Error handling

There are a variety of errors that you will have to consider.

infrastructure failure If the controller is unable to connect to a worker, or if a connection is broken while it is in use, or if the worker misbehaves by sending an inappropriate message, the controller should simply close the connection to the worker and continue processing the job using the remaining workers.

If all of the workers die, the `map_reduce` function should raise an exception [**TODO — what exception?**].

If a worker encounters an error when communicating with the controller, it should simply close the connection.

application failure If the application raises an exception while executing the `map` or `reduce` functions, then the worker should return a `JobFailed` message. Upon receiving this message, the controller should cause `map_reduce` to raise an exception [**TODO — what exception?**].

The name and stack trace of the original exception can be found using the `Printexc` module from the OCaml standard library; these should be returned to the controller in the `JobFailed` message, and used to construct the [**TODO — fix**]exception.

slow workers It is possible that some workers may simply be slow. If a worker takes more than [**TODO — time**] processing a single `map` or `reduce` request, then the controller should start a second worker to process the same job. This retry process should be repeated: if

the second worker takes more than [TODO — fix] a third worker should be started and so on. The old workers should not be terminated; the first of these workers to respond should determine the output of the job.

Exercise 8: Implement RemoteController

Implement the `RemoteController` module. The `init` function should simply record the provided list of addresses for future invocations of `Make.run`.

The `Make.map_reduce` function is responsible for executing the MapReduce job. It should use `Tcp.connect` to connect to each of the workers that were provided during `init`. It should then follow the protocol described above to complete the given `Job`.

You can use the controller to run a given app by running the `controllerMain.ml`:

```
% cs3110 compile controllerMain.ml
% cs3110 run controllerMain.ml
```

Exercise 9: Implement Worker

Implement the `Worker.Make` module in the `map_reduce` directory. The `Make.run` function should receive messages on the provided `Reader.t` and respond to them on the provided `Writer.t` according to the protocol described above.

You can run the worker on a given port by invoking `workerMain.ml`:

```
% cs3110 compile workerMain.ml
% cs3110 run workerMain.ml 31100
```

The list of addresses and ports that the controller will try to connect to is given in the file `addresses.txt`.

Part Four: MapReduce Apps

In this part of the assignment you will implement various MapReduce applications.

A MapReduce application implements the `MapReduce.App` interface, which provides a `main` function. This function will typically read some input, and then invoke the provided controller with one or more `Jobs`. We have provided you with the `WordCount` example application described above to get you started.

We have also provided you with a local controller that you can use to test your apps without a completed implementation of the MapReduce framework. To run an app locally, simply pass the `-local` option to `controllerMain.ml`. You should be able to run `WordCount` out of the box:

```
% cs3110 compile controllerMain.ml
% cs3110 run controllerMain.ml -local SteamWordCount apps/word_count/filenames.txt
```

Exercise 10:

[TODO — fix]

Exercise 11:

[TODO — fix]

Exercise 12:

[TODO — fix]