CPSC 418: Parallel Computing Homework #2

Homework #2 Solution Set

All problems for this homework were programming problems. The source code for the solution is available at

http://www.ugrad.cs.ubc.ca/~cs418/2016-2/hw/2/sol/hw2.erl Here are some explanatory remarks about the solution for each question.

1. mean(WTree, DataKey) (20 points)

Solution: Since this function is just returning a single value, it can easily be implemented with a reduce. All the mean function needs to do is call wtree:reduce, passing in the worker tree, WTree, and the appropriately defined functions Leaf, Combine, and Root. It basically works just like a function for sum, except we also need to know the length of the list. So the Leaf function gets its data from the ProcState variable (in the form of a list of numbers), and puts it into a tuple of the form {Sum, Length}. The Combine function then just needs to add the corresponding values in the two tuples it gets, and Root returns the average by dividing Sum by Length for the tuple it gets.

2. vec_mean(WTree, DataKey) (20 points)

Solution: This is almost identical to question 1 if you use the vec_sum function that is given, so again, it can easily be implemented with a reduce. The Leaf function puts the (vector) sum of its data into a tuple, along with the number of vectors it received. The Combine function adds the vectors and lengths that it receives and returns them in a new tuple. And Root just needs to divide each element in the vector it receives by the total number of vectors. Note that, given Sum (the sum of all vectors) and N (the number of vectors), the expression [$X / N \parallel X <-$ Sum] returns a list of elements X / N for all X in the list Sum.

3. set_nth(N, Fun, List) (10 points)

Solution: There are many ways to solve this problem, but it can be done simply and concisely using pattern matching and recursion. The solution provided simply applies Fun to the head of List if N is 1. Otherwise it appends the head element to the result of the recursive call.

One question that came up in discussing this problem is "Do we need to use tail-recursion?". The problem did not specify tail-recursion; so we won't punish head-recursive solutions. In particular, the provided solution is head-recursive. There is a design trade-off. The head-recursive implementation is simpler, and "obviously" correct. The tail recursive version is needed if we are worried about the case where N is huge (i.e. greater than one million). In this case, the solution set chose: "Do the simple way first.", or, as Don Knuth noted "The road to hell is paved with premature optimization.".

There are some cases where tail-recursion is a must. In Erlang, server processes are implemented with recursive functions. Unless these functions are tail recursive, the server will gradually use more memory until it crashes. Tail-recursion is a must for cases like this. There are other cases where head-recursion is clearly harmless, for example, traversing a balanced binary tree. The height of the tree, and thus the depth of the recursive calls, is bounded by the log of the number of leaves of the tree. There will never be a tree so large that we can't afford the stack frames for a traversal.

4. bank_statement(WTree, SrcKey, DstKey, InitialBalance) (20 points)

Solution: Unlike in questions 1 and 2, here we want to return a list of values (i.e., the balance at

each point in time), and processes need information about values to their left in order to calculate their own final values. So we should use a scan. We need to choose the functions Leaf1, and Leaf2, Combine, and the initial value Acc0. The hardest part of this question is probably figuring out how to deal with interest transactions. The idea is for each node to calculate a balance along with a tally of accumulated compound interest. The key observation is that any sequence of transactions can be summarized as a linear function:

FinalBalance = A * InitialBalance + B

where A and B depend on the particular sequence of transactions.

Leaf1 uses the lists:fold1 function, calling the helper bs_leaf1 on each element of Source, its data. This returns a tuple of the form {Balance, CompoundInterest}, where Balance is the result of applying all the transactions in the node's data, and CompoundInterest is the (multiplicative) total of any interest transactions. The helper bs_leaf1 applies a single transaction to the running total, returning a tuple of the form {Balance, CompoundInterest}.

Leaf2 gets a list of transactions (from ProcState) and the result (i.e., balance and accumulated interest) of all transactions to its left. It then needs to update the DstList in ProcState with the appropriate values. We want to return a list of balances, given a list of transactions and an initial balance. This is a good place to use a map (actually, a mapfoldl since we are accumulating the values). The function we are passing to mapfoldl is essentially the same as bs_leaf1, but the return value needs to be of the form {Balance, {Balance, CompoundInterest}} to fit what mapfoldl is expecting. The wrapper function bs_leaf2 is just converting the result of bs_leaf1 to this form. Leaf2 then stores this result in the ProcState so we can access it after computation finishes.

Combine gets two tuples of the form {Balance, CompoundInterest}. Since the transactions of the left tuple occurred before those of the right tuple, the interest of the right tuple needs to be applied to the balance of the left tuple. The value of Balance returned is the sum of this and the right balance. The new value of CompoundInterest is just the product of the interest from the two input tuples.

Acc0 needs to be a tuple of the form {Balance, CompoundInterest}, so we provide {InitialBalance, 1.0}. We chose 1.0 for the initial value for CompoundInterest because 1.0 has the effect of applying no interest. However, if you check the calculations, you'll note that the initial value of CompboundInterest only affects the final compound interest and has no impact on the per-transaction balances. So, we could put any (numeric) value we want for the CompoundInterest field.

5. sliding_average(WTree, SrcKey, DstKey, Kernel, InitialPrefix) (25 points)

Solution: Again, a scan is appropriate for this question because processes need information about values to their left in order to calculate their own final values. In this case they will need the length(Kernel)-1 values that come immediately before their own data. This way they can calculate the sliding average of their first length(Kernel)-1 values.

Leaf1 just needs to pass up its rightmost length(Kernel)-1 values. These values will be used in the Leaf2 function to calculate the sliding average for the leftmost values coming from nodes to the right.

Combine also just needs to return the rightmost length(Kernel)-1 values from the two sublists it gets so that nodes to the right can calculate the sliding average of their leftmost values. It is possible that length(Right) < length(Kernel)-1, in which case we also need to pass up some of the values from Left.

Leaf2 gets the length(Kernel)-1 values that come immediately before its own data in AccIn. So it can just calculate the sliding average of this using the helper for the sequential algorithm.

6. Test cases (5 points)

Solution: Make sure to test edge cases such as empty lists and badly formed inputs. In some cases your code can just error out on these inputs, but others it should handle properly. You can use assertError and assertException to make sure you are handling error scenarios properly.