Exercise 1: Depth

<u>Note:</u> This exercise was already given last week. If you have already done it, briefly discuss the solution as a group before moving to the next exercise.

Review the notion of depth seen in the video lectures. What does it represent?

Below is a formula for the depth of a *divide and conquer* algorithm working on an array segment of size L, as a function of L. The values c, d and T are constants. We assume that L>0 and T>0.

$$D(L) = \begin{cases} c \cdot L & \text{if } L \leq T \\ max(D(\lfloor \frac{L}{2} \rfloor), D(L - \lfloor \frac{L}{2} \rfloor)) + d & \text{otherwise} \end{cases}$$

Below the threshold T, the algorithm proceeds sequentially and takes time c to process each single element. Above the threshold, the algorithm is applied recursively over the two halves of the array. The results are then merged using an operation that takes d units of time.

Question 1

Is it the case that for all $1 \le L_1 \le L_2$ we have $D(L_1) \le D(L_2)$?

If it is the case, prove the property by induction on L. If it is not the case, give a counterexample showing values of L_1 , L_2 , T, c, and d for which the property does not hold.

Somewhat counterintuitively, the property doesn't hold. To show this, let's take the following values for L_1 , L_2 , T, c, and d.

$$L_1 = 10$$
, $L_2 = 12$, $T = 11$, $c = 1$, and $d = 1$.

Using those values, we get that:

$$D(L_1) = 10$$

 $D(L_2) = max(D(6), D(6)) + 1 = 7$

Prove a logarithmic upper bound on D(L). That is, prove that D(L) is in $O(\log L)$ by finding specific constants a,b such that $D(L) \le a \log_2 L + b$.

Proof sketch

Define the following function D'(L).

$$D'(L) = \begin{cases} c \cdot L & \text{if } L \leq T \\ \max(D'(\lfloor \frac{L}{2} \rfloor), D'(L - \lfloor \frac{L}{2} \rfloor)) + d + \underline{c \cdot T} & \text{otherwise} \end{cases}$$

Show that $D(L) \leq D'(L)$ for all $1 \leq L$.

Then, show that, for any $1 \le L_1 \le L_2$ we have $D'(L_1) \le D'(L_2)$. This property can be shown by induction on L_2 .

Finally, let n be such that $L \le 2^n < 2L$. We have that:

$$D(L) \leq D'(L) \qquad \qquad Proven \ earlier.$$

$$\leq D'(2^n) \qquad \qquad Also \ proven \ earlier.$$

$$\leq log_2(2^n) \ (d+cT)+cT$$

$$< log_2(2L) \ (d+cT)+cT$$

$$= log_2(L) \ (d+cT)+log_2(2) \ (d+cT)+cT$$

$$= log_2(L) \ (d+cT)+d+2cT$$

Done.

Exercise 2 : Aggregate

In the video lectures of this week, you have been introduced to the aggregate method of ParSeq[A] (and other parallel data structures...). It has the following signature:

def aggregate[B](z: B)(f: (B, A)
$$\Rightarrow$$
 B, g: (B, B) \Rightarrow B): B

Discuss, as a group, what aggregate does and what its arguments represent.

Question 1

Consider the parallel sequence xs containing the three elements x1, x2 and x3. Also consider the following call to aggregate:

The above call might potentially result in the following computation:

But it might also result in other computations. Come up with at least 2 other computations that may result from the above call to aggregate.

Some examples:

- g(f(z, x1), f(f(z, x2), x3))
- g(f(f(z, x1), x2), f(z, x3))
- g(g(f(z, x1), f(z, x2)), f(z, x3))
- g(f(z, x1), g(f(z, x2), f(z, x3)))

Below are other examples of calls to aggregate. In each case, check if the call can lead to different results depending on the strategy used by aggregate to aggregate all values contained in data down to a single value. You should assume that data is a parallel sequence of values of type BigInt.

Variant 1

This might lead to different results.

Variant 2

```
data.aggregate(0)((acc, x) => x - acc, _ + _)
```

This might lead to different results.

Variant 3

data.aggregate(0)((acc,
$$x$$
) => acc - x , _ + _)

This is always leads to the same result.

Variant 4

data.aggregate(1)((acc,
$$x$$
) => $x * x * acc, _ * _)$

This is always leads to the same result.

Under which condition(s) on z, f, and g does aggregate always lead to the same result? Come up with a formula on z, f, and g that implies the correctness of aggregate.

<u>Hint:</u> You may find useful to use calls to foldLeft(z)(f) in your formula(s).

A property that implies the correctness is:

Question 4

Implement aggregate using the methods map and/or reduce of the collection you are defining aggregate for.

A solution:

```
def aggregate(z: B)(f: (B, A) => B, g: (B, B) => B): B = if (this.isEmpty) z else this.map((x: A) => f(z, x)).reduce(g)
```

Implement aggregate using the task and/or parallel constructs seen in the first week and the Splitter[A] interface seen in this week's videos. The Splitter interface is defined as:

```
trait Splitter[A] extends Iterator[A] {
  def split: Seq[Splitter[A]]
  def remaining: Int
}
```

You can assume that the data structure you are defining aggregate for already implements a splitter method which returns an object of type Splitter[A].

Your implementation of aggregate should work in parallel when the number of remaining elements is above the constant THRESHOLD and sequentially below it.

<u>Hint:</u> Iterator, and thus Splitter, implements the foldLeft method.

A solution:

```
def aggregate(z: B)(f: (B, A) => B, g: (B, B) => B): B = {
    def go(s: Splitter[A]): B = {
        if (s.remaining <= THRESHOLD) {
            s.foldLeft(z)(f)
        }
        else {
            val splitted = s.split

            val subs = splitted.map((t: Splitter[A]) => task { go(t) })
            subs.map(_.join()).reduce(g)
        }
    }
    go(splitter)
}
```

Question 6

Discuss the implementations from questions 4 and 5. Which one do you think would be more efficient?

The version from question 4 may require 2 traversals (one for map, one for reduce) and does not benefit from the (potentially faster) sequential operator f.

Exercise 3 : Parallel Encoding

This question has been moved to the third exercise session.