## Solutions Exercise 1.2 from PSC2

(a) Minimum finding: determine the index $j$ of the component with the minimum value and subtract this value from every component: $y_{i}=$ $x_{i}-x_{j}$, for all $i$.

The basic idea is to find the local minimum first, send it to all processors, determine the global minimum redundantly, and then subtract it from all local values.

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input: \(\quad \mathbf{x}\) vector of length \(n\), \(\operatorname{distr}(\mathbf{x})=\) block.
output: \(\mathbf{y}\) vector of length \(n, \operatorname{distr}(\mathbf{y})=\) block,
    \(j=\operatorname{argmin}\left\{x_{i}: 0 \leq i<n\right\}, y_{i}=x_{i}-x_{j}\), for all \(i\).
    \(b=\lceil n / p\rceil ; \quad \triangleright \operatorname{Superstep}(0)\)
    minval \(_{s}:=\infty\);
    for \(i:=s b\) to \(\min ((s+1) b, n)-1\) do
        if \(x_{i}<\) minval \(_{s}\) then
            \(j_{s}:=i\);
            minval \(_{s}:=x_{i} ;\)
    for \(t:=0\) to \(p-1\) do \(\quad \triangleright\) Superstep (1)
        put \(j_{s}\), minval \(_{s}\) in \(P(t)\);
    minval \(:=\infty\); \(\quad \triangleright\) Superstep (2)
    for \(t=0\) to \(p-1\) do
        if minval \(_{t}<\) minval then
            \(j:=j_{t} ;\)
            minval \(:=\) minval \(_{t} ;\)
    for \(i:=s b\) to \(\min ((s+1) b, n)-1\) do
        \(y_{i}:=x_{i}-\) minval \(;\)
```

The BSP cost of the algorithm is

$$
2\lceil n / p\rceil+p+2(p-1) g+3 l,
$$

where we counted 1 flop for a comparison, and no flops for an assignment.
(b) Rotating to the right: assign $y_{(i+k) \bmod n}=x_{i}$.

The basic idea is to put the data into the correct destination processor in a single communication superstep. We may assume without loss of generality that $k \leq n / 2$, because otherwise we have a rotation of $n-k$ to the left, which is similar.
We compute the BSP cost as follows. For $k<n / p$, every processor $P(s)$ puts its $k$ rightmost data into processor $P((s+1)) \bmod p)$, at a cost of $k g+l$. The remaining $n / p-k$ local data are shifted to the right in a local memory copy. For $k \geq n / p$, the cost is $(n / p) g+l$.
(c) Smoothing: replace each component by a moving average $y_{i}=1 /(k+1)$ $\sum_{j=i-k / 2}^{i+k / 2} x_{j}$, where $k$ is even. Assume here that $x_{j}=0$ for $j<0$ or $j \geq n$.
Assume for simplicity that $k / 2 \leq n / p$ (a commonly used value these days is $k=6$ ). The remaining case is similar, but a bit more elaborate. The rightmost vector component of a processor $P(s)$ has to obtain at most $k / 2$ values from $P(s+1)$. The leftmost vector component has to obtain at most $k / 2$ values from $P(s-1)$. With this information, all local components $y_{i}$ can be computed.
The BSP cost of the algorithm is

$$
3\lceil n / p\rceil+k+k g+2 l .
$$

Here, we first obtain all locally needed data in $k g$ time. After that, we need $k$ flops to add the first $k$ values, and then we produce a moving average by subtracting the left value and adding the right value of the moving window in $2\lceil n / p\rceil$ time. Finally, we divide all local data by $k+1$ in $\lceil n / p\rceil$ time.
Note that if you perform a cost analysis, this forces you to be precise about what happens in the algorithm. Still, you do not always have to give the full details as in (a).
(d) Partial summing: compute $y_{i}=\sum_{j=0}^{i} x_{j}$, for all $i$.

The basic idea is that every processor $P(s)$ computes its local partial sums, without regard for the others. $P(0)$ will then already have the correct result, but the others will have to add a correction, which consists of the total sums for all processors $P(t)$ with $t<s$. This can be done by processor $P(t)$ sending its rightmost value to all highernumbered processors. The received sums are then added locally. In a final pass through the local data, the correction is added.
The BSP cost of the algorithm is (please check)

$$
2\lceil n / p\rceil+p-1+(p-1) g+3 l .
$$

In an implementation, it would be useful to write a separate sequential function to compute partial sums, and use this function in the parallel program. Reuse of code!

