

## Solution of Serie 11

**Exercise 1** 2. A solution is given in the following figure.

B	R	B	R
R	B	R	B
B	R	B	R
R	B	R	B

3. There is no dependency in each equation: red nodes are updated only with black nodes values and black nodes only with red nodes values.
4. The double loops for updating the node values and also the loops for computing the residual can be easily parallelized by the directive `#pragma omp parallel for private (i,j)`.
5. For small  $N$  (e.g.  $N = 25$ ), the sequential version is always better; the speedup of the parallel version is quite good for a number of processors  $\leq 4$ . For large  $N$ , the parallel version is better for a number of processors  $\geq 2$ ; the speedup is quite good for a processor number  $\leq 8 - 12$ . This behaviour is due to the few number of operations in the  $j$  loops. The increase of the number of iterations with  $N$  comes from the increase of the condition number of the matrix.

**Exercise 2** 1. The error  $E(\mathbf{x})$  is defined by

$$E(\mathbf{x}) = \frac{1}{2} (A(\hat{\mathbf{x}} - \mathbf{x}), (\hat{\mathbf{x}} - \mathbf{x})) = \frac{1}{2} \sum_{l,m} (\hat{x}_l - x_l) a_{lm} (\hat{x}_m - x_m).$$

where  $A\hat{\mathbf{x}} = \mathbf{b}$ . The  $i$ -th component of the gradient  $g_i$  of  $E(\mathbf{x})$  is given by

$$(\nabla E(\mathbf{x}))_i = \frac{1}{2} \frac{\partial}{\partial x_i} \left( \sum_{l,m} (\hat{x}_l - x_l) A_{lm} (\hat{x}_m - x_m) \right) = \sum_m A_{im} (x_m - \hat{x}_m) = (A\mathbf{x} - \mathbf{b})_i = -r_i.$$

One rewrites the first version of the algorithm using  $\mathbf{r}^k = -\mathbf{g}^k$  and with changing the signs of  $\alpha_k$  and  $\beta_{k+1}$ :

$\alpha_k = (\mathbf{g}^k, \mathbf{p}^k) / (A\mathbf{p}^k, \mathbf{p}^k)$
$\mathbf{x}^{k+1} = \mathbf{x}^k - \alpha_k \mathbf{p}^k$
$\mathbf{g}^{k+1} = A\mathbf{x}^{k+1} - \mathbf{b}$
$\beta_k = -(A\mathbf{p}^k, \mathbf{g}^{k+1}) / (A\mathbf{p}^k, \mathbf{p}^k)$
$\mathbf{p}^{k+1} = \mathbf{g}^{k+1} + \beta_k \mathbf{p}^k$

2. From the second and third equations, one finds that:

$$\mathbf{g}^{k+1} = A\mathbf{x}^{k+1} - \mathbf{b} = A(\mathbf{x}^k - \alpha_k \mathbf{p}^k) - \mathbf{b} = \mathbf{g}^k - \alpha_k A\mathbf{p}^k, \quad (1)$$

that is the relationship of point 3.

3. From the first and last equations, one obtains:

$$\alpha_k = \frac{(\mathbf{g}^k, \mathbf{p}^k)}{(A\mathbf{p}^k, \mathbf{p}^k)} = \frac{\|\mathbf{g}^k\|^2}{(A\mathbf{p}^k, \mathbf{p}^k)} + \beta_{k+1} \frac{(\mathbf{g}^k, \mathbf{p}^{k-1})}{(A\mathbf{p}^k, \mathbf{p}^k)}.$$

But

$$(\mathbf{g}^k, \mathbf{p}^{k-1}) \stackrel{(1)}{=} (\mathbf{g}^{k-1}, \mathbf{p}^{k-1}) - \alpha_{k-1} (A\mathbf{p}^{k-1}, \mathbf{p}^{k-1}) = 0, \quad (2)$$

and then

$$\alpha_k = \frac{\|\mathbf{g}^k\|^2}{(A\mathbf{p}^k, \mathbf{p}^k)}. \quad (3)$$

4. The fourth equation

$$\beta_k = -\frac{(A\mathbf{p}^k, \mathbf{g}^{k+1})}{(A\mathbf{p}^k, \mathbf{p}^k)} \stackrel{(1)}{=} \frac{1}{\alpha_k} \frac{(\mathbf{g}^{k+1} - \mathbf{g}^k, \mathbf{g}^{k+1})}{(A\mathbf{p}^k, \mathbf{p}^k)} \stackrel{(3)}{=} \frac{\|\mathbf{g}^{k+1}\|^2}{\|\mathbf{g}^k\|^2} - \frac{(\mathbf{g}^k, \mathbf{g}^{k+1})}{\|\mathbf{g}^k\|^2}, \quad (4)$$

implies, using the fifth one:

$$\begin{aligned} (\mathbf{g}^k, \mathbf{g}^{k+1}) &\stackrel{(1)}{=} \|\mathbf{g}^k\|^2 - \alpha_k(\mathbf{g}^k, A\mathbf{p}^k) = \|\mathbf{g}^k\|^2 - \alpha_k(\mathbf{p}^k, A\mathbf{p}^k) + \alpha_k\beta_k(\mathbf{p}^{k-1}, A\mathbf{p}^k) \\ &\stackrel{(3)}{=} \alpha_k\beta_k(\mathbf{p}^{k-1}, A\mathbf{p}^k) \end{aligned}$$

and, again with the help of the fifth equation:

$$(\mathbf{p}^{k-1}, A\mathbf{p}^k) = (A\mathbf{p}^{k-1}, \mathbf{p}^k) = \beta_k(A\mathbf{p}^{k-1}, \mathbf{p}^{k-1}) + (A\mathbf{p}^{k-1}, \mathbf{g}^k) = 0.$$

Hence

$$\beta_k = \frac{\|\mathbf{g}^{k+1}\|^2}{\|\mathbf{g}^k\|^2}.$$

5. The second version requires only the operation  $(A\mathbf{p}^k, \mathbf{p}^k)$  while in the first version one needs to compute  $(A\mathbf{p}^k, \mathbf{p}^k)$  and  $(A\mathbf{p}^k, \mathbf{r}^{k+1})$ , that is one more matrix-vector multiplication.