

## Scientific Computing II

### Incomplete Cholesky Decomposition

#### Programming Exercise 7: Incomplete Cholesky Decomposition

In the following, we review the incomplete Cholesky factorisation which computes an approximation  $\tilde{A} \in \mathbb{R}^{N \times N}$  to a matrix  $A \approx \tilde{A}$ . We decompose  $\tilde{A}$  into a lower left triangular matrix  $L$  and—analogue to the slides from the lecture—a diagonal matrix  $D$  such that  $\tilde{A} = LD^{-1}L^\top$ :

```

for i=1 to N do
  for j=1 to i-1 do
     $L_{ij} = A_{ij} - \sum_{k=1; (i,k), (j,k) \in S}^{j-1} L_{ik} D_{kk}^{-1} L_{jk}$  if  $(i, j) \in S$ 
  end
   $L_{ii} = D_{ii} = A_{ii} - \sum_{k=1; (i,k) \in S}^{i-1} L_{ik}^2 D_{kk}^{-1}$ 
end

```

where  $S$  denotes the set of all indices  $(i, j)$  with  $A_{ij} \neq 0$ .

- For which matrices can we apply the (incomplete) Cholesky decomposition?
- Implement the algorithm in a matlab function `incompleteLDL(A)` which returns the matrices  $L, D^{-1}$ .

Test your algorithm using a 2D Poisson problem with Dirichlet conditions as input. The respective stencil reads

$$S := \begin{bmatrix} 0 & -1 & 0 \\ -1 & 4 & -1 \\ 0 & -1 & 0 \end{bmatrix}$$

for all inner points. Write a function `generate2DPoisson(x,y)` which for a given number of grid points  $x \times y$  generates the (full) matrix  $A$ . Include both inner and Dirichlet boundary points into the matrix description. You may use a lexicographic ordering of the grid points.

Validate your implementation by comparison with Matlab's internal routine `ichol(A)`. Hint: `ichol(A)` works on matrices in sparse format only. You may use Matlab's routines `full(sparseMatrix)` to convert a sparse matrix into a full matrix and `sparse(matrix)` to convert a full matrix into a sparse matrix.

- (c) Optimise your implementation `incompleteLDL(A)` such that you obtain a matrix-free (with respect to the input data) incomplete LDL-factorisation. The corresponding function `incompleteLDLPoisson2D(x,y)` only takes the number of grid points  $x \times y$  as arguments.

Validate your implementations by comparing the resulting matrices  $L$  and  $D^{-1}$  with the results from `incompleteLDL(A)`.

What is the complexity of your optimised implementation?

**Solution:**

- (a) The decomposition works for symmetric positive definite matrices.
- (b) See `generate2DPoisson.m`, `incompleteLDL.m` and `exercise7.m`. We compare Matlab's matrix  $L_{matlab}$  and our matrices  $L, D^{-1}$  by computing  $\tilde{L} := L \cdot D_{\sqrt{\cdot}}^{-1}$  where  $D_{\sqrt{\cdot}}^{-1}$  is a diagonal matrix with entries  $(D_{\sqrt{\cdot}}^{-1})_{ii} = \sqrt{D_{ii}^{-1}}$ . Both matrices  $\tilde{L}$  and  $L_{matlab}$  should be identical.
- (c) See `incompleteLDLPoisson2D.m` and `exercise7.m`. Considering the nested for-loops (xx,yy) in the function, we observe that we have a  $O(N)$ -algorithm.