

## CAAM 454/554 Handout: Sparse Cholesky

Cholesky decomposition is the most efficient direct method to solve linear systems with symmetric, positive definite coefficient matrices. If  $A$  is symmetric, positive definite, then it can be factorized into  $A = R^T R$  where  $R$  is upper triangular. Therefore, the solution to the linear system  $Ax = b$  can be expressed, in Matlab notation, as

$$R = \text{chol}(A); \quad x = R \setminus (R' \setminus b).$$

Matlab is smart enough to solve the triangular systems efficiently. If  $A$  is sparse, then rows and columns of  $A$  usually need to be reordered to make the factor  $R$  sparse (in solving  $Ax = b$ ,  $x$  and  $b$  need to be reordered accordingly). One of the ordering functions in Matlab is `symamd`. The following script generates pictures showing how ordering affects sparsity.

```
% minimum degree ordering and sparse cholesky
load ship04s; % load a sparse matrix A
B = A*A' + speye(402); % Guarantee positive definiteness
p = symamd(B); % SYMmetric Approximate Minimum Degree ordering

% spy, spy, spy and spy
subplot(221); spy(B); title('B');
subplot(222); spy(B(p,p)); title('B(p,p)');
subplot(223); spy(chol(B)); title('chol(B)');
subplot(224); spy(chol(B(p,p))); title('chol(B(p,p))');
```

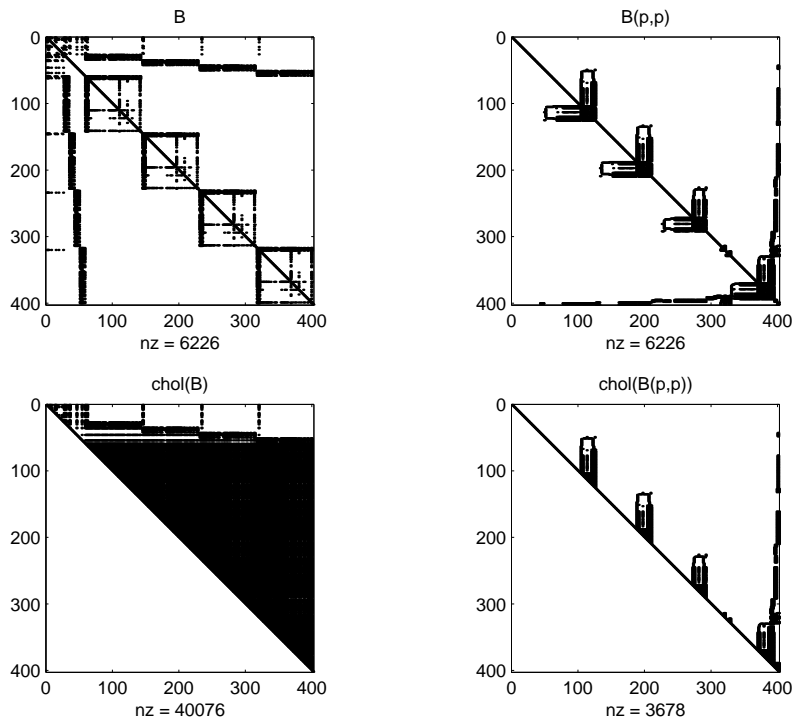


Figure 1: A Comparison of Sparsity (nz = number of nonzeros)