

## Sparse Cholesky Factorization

Cholesky decomposition is the most efficient way 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

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

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))');
```

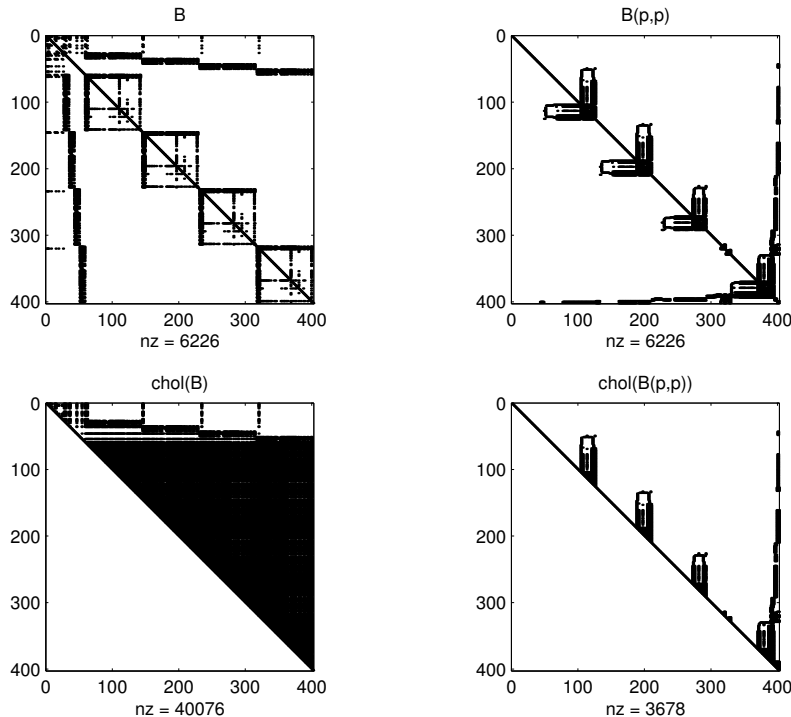


Figure 1: A Comparison of Sparsity (nz = number of nonzeros)