# Sparse vs. Dense Matrices

# Lecture 24 Sparse Matrix Algorithms

MIT 18.335J / 6.337J Introduction to Numerical Methods

> Per-Olof Persson November 28, 2006

- A sparse matrix is a matrix with enough zeros that it is worth taking advantage of them [Wilkinson]
- A structured matrix has enough structure that it is worthwhile to use it (e.g. Toeplitz)
- A dense matrix is neither sparse nor structured

## **MATLAB Sparse Matrices: Design Principles**

- Most operations should give the same results for sparse and full matrices
- Sparse matrices are never created automatically, but once created they propagate
- Performance is important but usability, simplicity, completeness, and robustness are more important
- Storage for a sparse matrix should be O(nonzeros)
- Time for a sparse operation should be close to O(flops)

#### **Data Structures for Matrices**

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#### Full:

- Storage: Array of real (or complex) numbers
- Memory: nrows\*ncols

31	0	53
0	59	0
41	26	0
double *A		

b c a b

#### Sparse:

- Compressed column storage
- Memory: About
- 1.5\*nnz+.5\*ncols



## **Compressed Column Format - Observations**

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- Element look-up:  $O(\log \text{#elements in column})$  time
- Insertion of new nonzero very expensive
- Sparse vector = Column vector (not Row vector)



# Permutations of the 2-D Model Problem

- 2-D Model Problem: Poisson's Equation on  $n \times n$  finite difference grid
- Total number of unknowns  $n^2 = N$
- Theoretical results for the fill-in:
  - With natural permutation:  $O(N^{3/2})$  fill
  - With any permutation:  $\Omega(N \log N)$  fill
  - With a nested dissection permutation:  $O(N \log N)$  fill

# **Nested Dissection Ordering**

- A separator in a graph G is a set S of vertices whose removal leaves at least two connected components
- A nested dissection ordering for an  $N\mbox{-vertex graph}\ G$  numbers its vertices from 1 to N as follows:
  - Find a separator S, whose removal leaves connected components  $T_1,T_2,\ldots,T_k$
  - Number the vertices of S from  $N-\left|S\right|+1$  to N
  - Recursively, number the vertices of each component:  $T_1$  from 1 to  $|T_1|,\,T_2$  from  $|T_1|+1$  to  $|T_1|+|T_2|,\,{\rm etc}$
  - If a component is small enough, number it arbitrarily
- It all boils down to finding good separators!

# **Heuristic Fill-Reducing Matrix Permutations**

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- Banded orderings (Reverse Cuthill-McKee, Sloan, etc):
  - Try to keep all nonzeros close to the diagonal
  - Theory, practice: Often wins for "long, thin" problems
- Minimum degree:
  - Eliminate row/col with fewest nonzeros, add fill, repeat
  - Hard to implement efficiently current champion is "Approximate Minimum Degree" [Amestoy, Davis, Duff]
  - Theory: Can be suboptimal even on 2-D model problem
  - Practice: Often wins for medium-sized problems

# **Heuristic Fill-Reducing Matrix Permutations**

- Nested dissection:
  - Find a separator, number it last, proceed recursively
  - Theory: Approximately optimal separators => approximately optimal fill and flop count
  - Practice: Often wins for very large problems
- The best modern general-purpose orderings are ND/MD hybrids

## **Fill-Reducing Permutations in Matlab**

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- Reverse Cuthill-McKee:
  - p=symrcm(A);
  - Symmetric permutation: A(p,p) often has smaller bandwidth than A
- Symmetric approximate minimum degree:
  - p=symamd(A);
  - Symmetric permutation: chol(A(p,p)) sparser than chol(A)
- Nonsymmetric approximate minimum degree:
  - p=colamd(A);
  - Column permutation: lu(A(:,p)) sparser than lu(A)
- Symmetric nested dissection:
  - Not built into MATLAB, several versions in the MESHPART toolbox

# Complexity of Direct Methods Time and space to solve any problem on any well-shaped finite element

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