Lec 15: Square Linear Systems – Analysis

Efficiency

Notation: Big-O and Asymptotic

Let f, g be positive functions defined on $\mathbb{N}$.

- $f(n) = O\big(g(n)\big)$ ("f is *big-O* of g") as $n \to \infty$ if

$$\frac{f(n)}{g(n)} \leqslant C, \quad \text{for all sufficiently large } n.$$

- $f(n) \sim g(n)$ ("f is *asymptotic* to g") as $n \to \infty$ if

$$\lim_{n \to \infty} \frac{f(n)}{g(n)} = 1.$$

Timing Vector/Matrix Operations – FLOPS

- One way to measure the "efficiency" of a numerical algorithm is to count the number of floating-point arithmetic operations (FLOPS) necessary for its execution.

- The number is usually represented by $\sim cn^p$ where c and p are given explicitly.

- We are interested in this formula when n is large.

FLOPS for Major Operations

Vector/Matrix Operations

Let $x, y \in \mathbb{R}^n$ and $A, B \in \mathbb{R}^{n \times n}$. Then

- (vector-vector) $x^{\mathrm{T}} y$ requires $\sim 2n$ *flops*.
- (matrix-vector) Ax requires $\sim 2n^2$ *flops*.
- (matrix-matrix) AB requires $\sim 2n^3$ *flops*.

Cost of PLU Factorization

Note that we only need to count the number of *flops* required to zero out elements below the diagonal of each column.

- For each $i > j$, we replace R_i by $R_i + cR_j$ where $c = -a_{i,j}/a_{j,j}$. This requires approximately $2(n - j + 1)$ *flops*:
    - 1 division to form c
    - $n - j + 1$ multiplications to form cR_j
    - $n - j + 1$ additions to form $R_i + cR_j$
- Since $i \in \mathbb{N}[j + 1, n]$, the total number of *flops* needed to zero out all elements below the diagonal in the jth column is approximately $2(n - j + 1)(n - j)$.
- Summing up over $j \in \mathbb{N}[1, n - 1]$, we need about $(2/3)n^3$ *flops*:

$$\sum_{j=1}^{n-1} 2(n - j + 1)(n - j) \sim 2 \sum_{j=1}^{n-1} (n - j)^2 = 2 \sum_{j=1}^{n-1} j^2 \sim \frac{2}{3} n^3$$

Cost of Forward Elimination and Backward Substitution

Forward Elimination

- The calculation of $y_i = \beta_i - \sum_{j=1}^{i-1} \ell_{ij} y_j$ for $i > 1$ requires approximately $2i$ *flops*:

  - 1 subtraction

  - $i - 1$ multiplications

  - $i - 2$ additions

- Summing over all $i \in \mathbb{N}[2, n]$, we need about n^2 *flops*:

$$\sum_{i=2}^{n} 2i \sim 2\frac{n^2}{2} = n^2.$$

Backward Substitution

- The cost of backward substitution is also approximately n^2 *flops*, which can be shown in the same manner.

Cost of G.E. with Partial Pivoting

Gaussian elimination with partial pivoting involves three steps:

- PLU factorization: $\sim (2/3)n^3$ *flops*
- Forward elimination: $\sim n^2$ *flops*
- Backward substitution: $\sim n^2$ *flops*

Summary

The total cost of Gaussian elimination with partial pivoting is approximately

$$\frac{2}{3}n^3 + n^2 + n^2 \sim \frac{2}{3}n^3$$

flops for large n.

Application: Solving Multiple Square Systems Simultaneously

To solve two systems $A\mathbf{x}_1 = \mathbf{b}_1$ and $A\mathbf{x}_2 = \mathbf{b}_2$.

Method 1.

- Use G.E. for both.
- It takes $\sim (4/3)n^3$ *flops*.

```
%% method 1
x1 = A \ b1;
x2 = A \ b2;
```

Method 2.

- Do it in two steps:

  ① Do PLU factorization $PA = LU$.

  ② Then solve $LU\mathbf{x}_1 = P\mathbf{b}_1$ and $LU\mathbf{x}_2 = P\mathbf{b}_2$.

- It takes $\sim (2/3)n^3$ *flops*.

```
%% method 2
[L, U, P] = lu(A);
x1 = U \ (L \ (P*b1));
x2 = U \ (L \ (P*b2));
```

```
%% compact implementation
X = A \ [b1, b2];
x1 = X(:, 1);
x2 = X(:, 2);
```