# **Cost of LU Factorization**

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# Notation: Big-O and Asymptotic

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

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

$$\frac{f(n)}{g(n)} \leqslant C$$
, 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)  $x^Ty$  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 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 1

x1 = A \ b1;

x2 = A \ b2;
```

```
%% method 2

[L, U, P] = lu(A);

x1 = U \ (L \ (P*b1));

x2 = U \ (L \ (P*b2));
```