

# Overview

- CH: cost
- CH: apps (CA)
- CH for interpolation

Q: - ill vs. well  
- P mat

border line → "well"

$$10^{-10} \leq 10^5 \cdot 10^{-15}$$

rel. out err      cond #      rel. inp. err

$$\text{ill} \rightarrow 10^{\cdot} \leq 10^5 \quad 10^{-4}$$

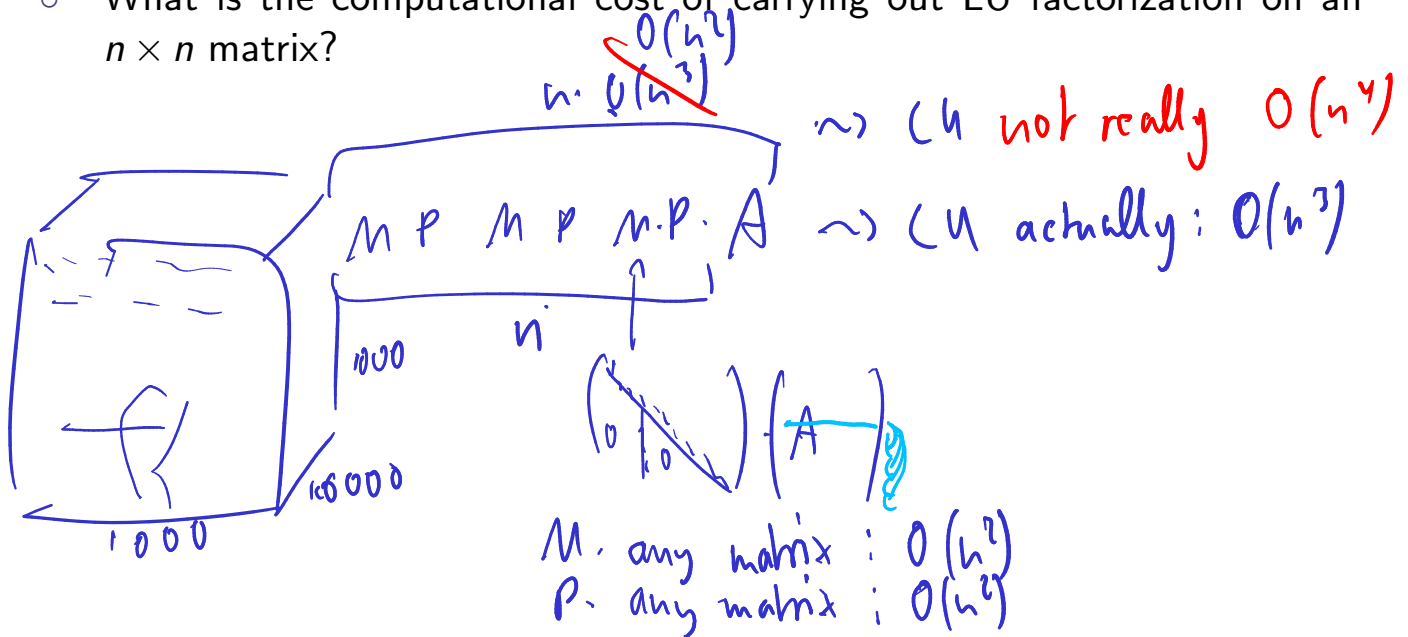
$$\begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \\ 2 \end{pmatrix} = \begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix}$$

## Computational Cost

- What is the computational cost of multiplying two  $n \times n$  matrices?

$$O(n^3)$$

- What is the computational cost of carrying out LU factorization on an  $n \times n$  matrix?

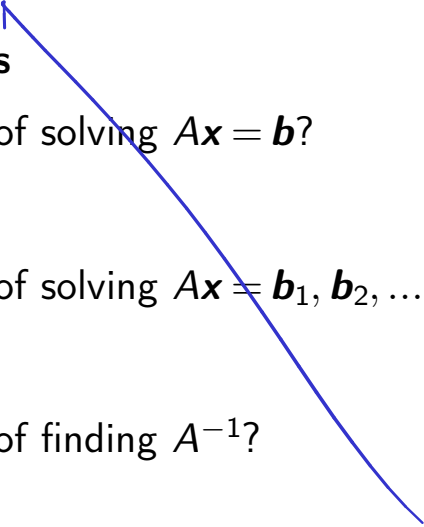


## Demo: Complexity of Mat-Mat multiplication and LU

n goes up  
by  $10^6$

$$\left( \begin{array}{l} 1000 \times 1000 \sim 1s \\ 10^9 \times 10^9 \sim (10^6)^3, 1s \end{array} \right.$$

## More cost concerns

- What's the cost of solving  $A\mathbf{x} = \mathbf{b}$ ?
  - What's the cost of solving  $A\mathbf{x} = \mathbf{b}_1, \mathbf{b}_2, \dots, \mathbf{b}_n$ ?
  - What's the cost of finding  $A^{-1}$ ?
- 

## Cost: Worrying about the Constant, BLAS

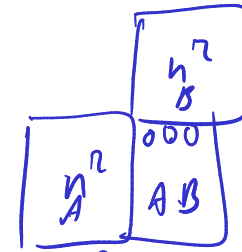
$O(n^3)$  really means

$$\alpha \cdot n^3 + \beta \cdot n^2 + \gamma \cdot n + \delta.$$

All the non-leading and constants terms swept under the rug. But: at least the leading constant ultimately matters.

Getting that constant to be small is surprisingly hard, even for something deceptively simple such as matrix-matrix multiplication.

**Idea:** Rely on library implementation: **BLAS** (Fortran)



$n^2$  data  $\rightarrow$   $n^3$  computation

lots of opportunity for data reuse

opportunity for cache mgmt.

Level 1  $\mathbf{z} = \alpha \mathbf{x} + \mathbf{y}$

vector-vector operations

$O(n)$

?axpy

Level 2  $\mathbf{z} = \mathbf{A}\mathbf{x} + \mathbf{y}$

matrix-vector operations

$O(n^2)$

?gemv

Level 3  $\mathbf{C} = \mathbf{A}\mathbf{B} + \beta \mathbf{C}$

matrix-matrix operations

$O(n^3)$

?gemm

BLAS

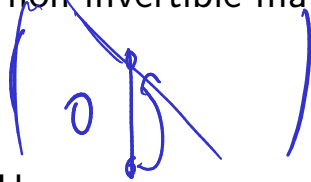
LAPACK: Implements 'higher-end' things (such as LU) using BLAS

Special matrix formats can also help save const significantly, e.g.

- banded
- sparse

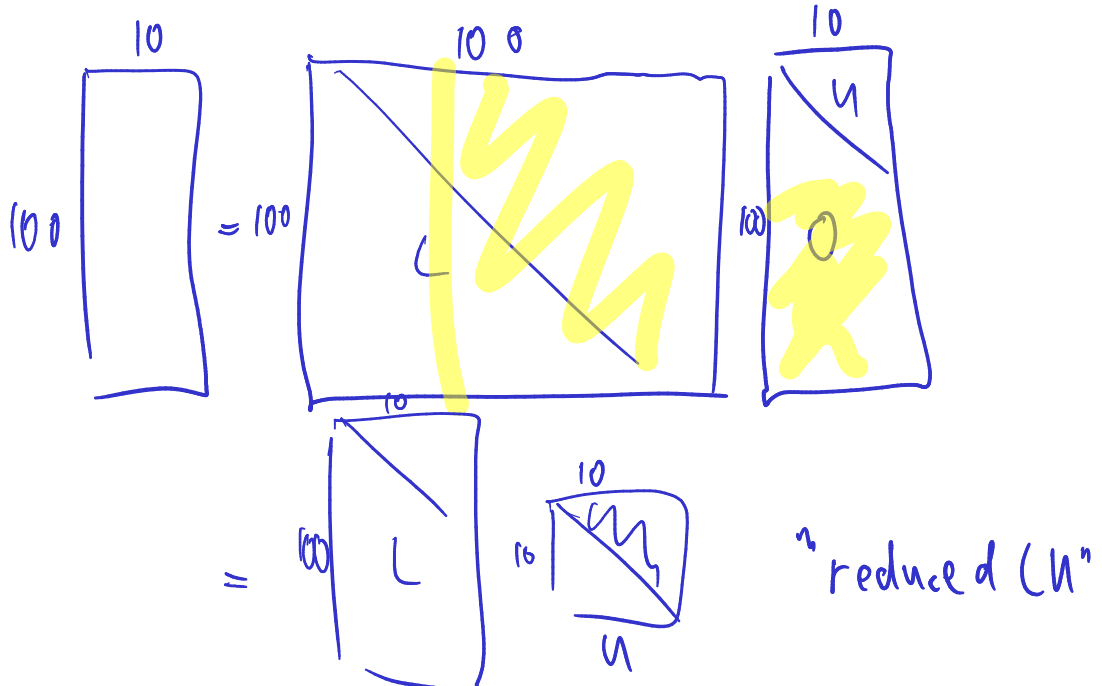
## LU: Special cases

- What happens if we feed a non-invertible matrix to LU?



$$PA = LU$$

- What happens if we feed LU an  $m \times n$  non-square matrices?





## In-class activity: LU factorization 2

# 9 LU: Applications

## 9.1 Linear Algebra Applications

## Solve a Linear System

- LU factorization gives us

$$PA = LU, \quad \rightarrow \quad A = P^r LU$$

so that  $P$  is a permutation matrix,  $L$  is lower triangular,  $U$  is upper triangular. How does that help solve a linear system  $Ax = b$ ?

$$Ax = b$$

$$P^r LUx = b$$

$$L(Ux) = Pb$$

$$Ly = Pb \quad \text{fw subst}$$

$$Ux = y \quad \text{bw subst.}$$

## Solve a Matrix Equation

- Suppose we want to solve  $AX = B$ .

$A$  and  $B$  are given,  $X$  is unknown.

(Assume: square and have same size) How can we do that using LU?

$$\left( \begin{array}{c} A \\ B \end{array} \right) \left( \begin{array}{c} x_1 \\ \vdots \\ x_n \end{array} \right) = \left( \begin{array}{c} b_1 \\ \vdots \\ b_n \end{array} \right)$$

$\swarrow$   $x_1$   
 $\nwarrow$   $b_1$

$$A \vec{x}_1 = \vec{b}_1$$

$$A \vec{x}_2 = \vec{b}_2$$

↳ solve column-by-column

no: solve  $n$  linear systems:  $O(n^4)$ ?

yes: calculate  $L$  &  $U$  :  $O(n^3)$

↳  $n \times$  fw/lw subst:  $n \cdot O(n^2) = O(n^3)$

## Compute an Inverse

- Suppose we want to compute the inverse  $A^{-1}$  of a matrix  $A$ .  
How do we do that using LU?

$$\text{solve } AX = I$$

- What's the computational cost of doing so?

$$O(n^3)$$

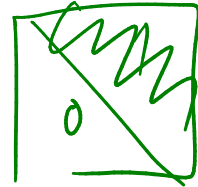
## Find the Determinant of a Matrix

- How can we find the determinant of a matrix using LU?

$$PA = LU \quad \leadsto \quad A = P^T LU$$

$$\det(A) = \det(P^T) \cdot \det(L) \cdot \det(U)$$

$\pm 1$



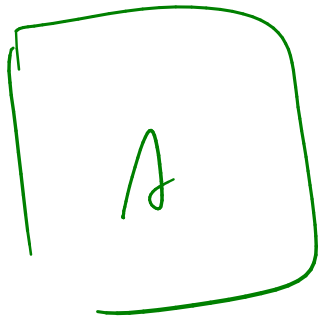
$u_{11} \cdot u_{22} \cdot \dots \cdot u_{nn}$

$O(n^3)$

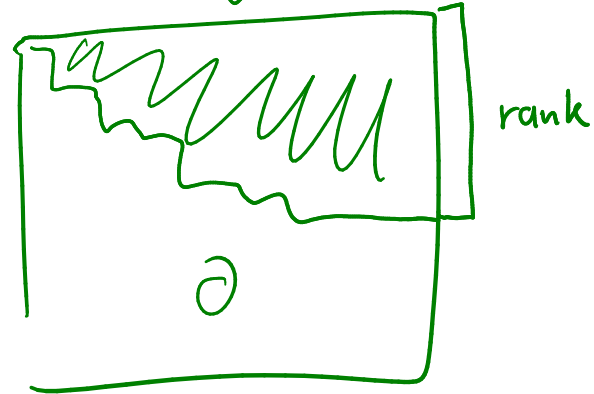
## Find Row Echelon Form... if we can?

1st hat  $U$ ? **No!**

- The factor  $U$  in pivoted LU looks like it is in upper echelon form. Is it?



$\rightsquigarrow$   
RREF



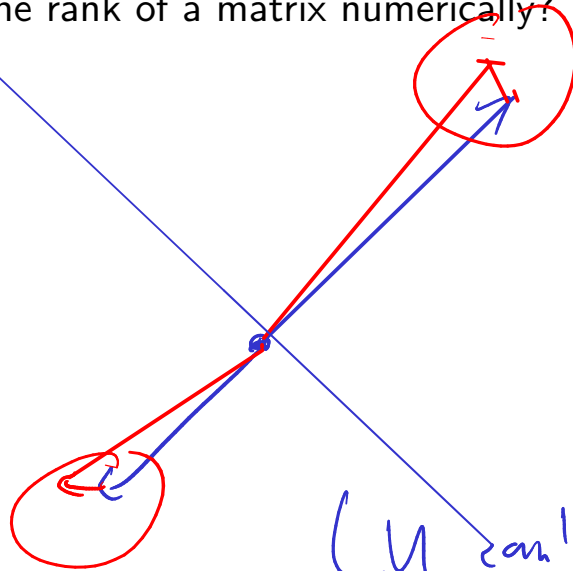
$$\text{rank}(A) \approx \dim(\text{colspace}) \approx \dim(\text{rowspace})$$



# Finding the Rank of a Matrix Numerically... if we can?

- Can we find the rank of a matrix numerically?

$$\begin{pmatrix} - & - \\ - & - \end{pmatrix}$$



To ask about rank, we need a tolerance

LU can't do tolerances  
can't do RREF } → postpone

## 9.2 Interpolation

$$\begin{matrix} \text{points} \downarrow & \text{functions} \rightarrow & \begin{pmatrix} \phi_0(x_0) & \dots & \phi_n(x_0) \\ \vdots & & \vdots \\ \phi_0(x_n) & \dots & \phi_n(x_n) \end{pmatrix} & \begin{pmatrix} \alpha_0 \\ \vdots \\ \alpha_n \end{pmatrix} & = & \begin{pmatrix} y_0 \\ \vdots \\ y_n \end{pmatrix} \end{matrix}$$

## Recap: Interpolation

Starting point: Looking for a linear combination of functions  $\varphi_i$  to hit given data points  $(x_i, y_i)$ .

Interpolation becomes solving the linear system:

$$y_i = f(x_i) = \sum_{j=0}^{N_{\text{func}}} \underbrace{\alpha_j \varphi_j(x_i)}_{V_{ij}} \quad \Leftrightarrow \quad V\alpha = \mathbf{y}.$$

Want unique answer: Pick  $N_{\text{func}} = N \rightarrow V$  square.

$V$  is called the (generalized) Vandermonde matrix.

Main lesson:

$V(\text{coefficients}) = (\text{values at nodes}).$

$$V = \begin{pmatrix} \varphi_0(x_0) & \dots \\ \vdots & \\ \varphi_0(x_n) & \dots \end{pmatrix} \quad \varphi_n(x_n)$$

$$V\vec{\alpha} = \vec{y}$$

## Rethinking Interpolation

We have so far always used monomials  $(1, x, x^2, x^3, \dots)$  and equispaced points for interpolation. It turns out that this has *significant problems*.

**Demo:** Monomial interpolation

## Demo: Choice of Nodes for Polynomial Interpolation

## Interpolation: Choosing Basis Function and Nodes

Both function basis and point set are under our control. What do we pick?

Ideas for basis functions:

- Monomials  $1, x, x^2, x^3, x^4, \dots$
- Functions that make  $V = I \rightarrow$  'Lagrange basis'  $\leftarrow$
- Functions that make  $V$  triangular  $\rightarrow$  'Newton basis'  $\leftarrow$
- Splines (piecewise polynomials)  $\leftarrow$
- Orthogonal polynomials
- Sines and cosines
- 'Bumps' ('Radial Basis Functions')



Ideas for nodes:

- Equispaced
- 'Edge-Clustered' (so-called Chebyshev/Gauss/... nodes)

## Better Conditioning: Orthogonal Polynomials

- What caused monomials to have a terribly conditioned Vandermonde?
  - What's a way to make sure two vectors are *not* like that?
  - But polynomials are functions!
- 
- But how can I practically compute the Legendre polynomials?