

# Welcome to MA221! Lecture 3, Aug 28

math problem  
structure  
accuracy

- 1) Error Bounds: even if  $A$  close to your input, if  $A$  close to singular, could still get wrong answer  
We will derive "cheap" error bounds  
"condition number"  $\approx$  machine epsilon  
eg  $10^{-16}$  double

$$\text{condition number} = \frac{1}{\text{distance to nearest "ill posed problem"}}$$

- 2) "Guaranteed correct" except in "rare cases": take a few steps of Newton to make error smaller  
cheap for  $Ax=b$ , LS

popular again because of 16-bit floating point accelerators, eg Google TPU (conference now)

- 3) "Accuracy" = getting bit wise identical answers every time you run code.  
because floating point addition not associative.

## Axis 4 - Efficiency

- (1) Fewest keystrokes

$A \setminus b$

lots of existing software

- (2) What does "fewest flops" mean?  
for  $n \times n$  matmul?

Classical:  $2n^3$  flops

Strassen (1969)  $O(n^{\log_2 7}) = O(n^{2.81})$

Coppersmith-Winograd (1987)  $O(n^{2.376})$   
"galactic algorithm"

Williams (2013)  $O(n^{2.3728642})$

Le Gall (2014)  $O(n^{2.3728639})$

Williams + Ataman (2020)  $O(n^{2.3728596})$

Williams + Xu + Xu + Zhou (2023)  
 $O(n^{2.371552})$

D., Dmitry, Hottz (2008):

all standard problems ( $Ax=b$ , LS, eig) can run as fast as matmul  
 $O(n^w)$  matmul  $\rightarrow O(n^w)$  for  $Ax=b$ , etc

numerically stable,  
sometimes practical

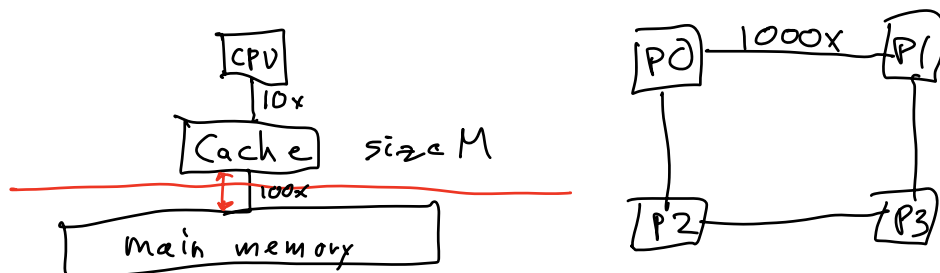
will show how to reorganize  $Ax=b$ , LS, eig, to do almost all work in matmul

(3) Counting flops not only important metric

(3.1) "End of Moore's Law" (2004)

$\Rightarrow$  parallelism necessary  
sometimes need new methods

(3.2) Arithmetic cheap compared to moving data



naive matmul (3 nested loops) can be

100x slower than optimized version  
Lots of recent algorithms attaining  
Lower bounds, sometimes same  
ops in different order, sometimes  
different math alg.

### (3.3) Minimizing Energy

Turns out minimizing data movement  
minimizes energy

Communication lower bounds

$$\text{for } O(n^3) \text{ mat mul is } \Omega\left(\frac{n^3}{\sqrt{M}}\right)$$

$$\text{for } O(n^w) \text{ mat mul is } \Omega\left(\frac{n^w}{M^{w/2-1}}\right)$$

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## Syllabus

Direct Methods for  $Ax=b$ , LS

$$LU: A = P \cdot L \cdot U$$

↑ permutation      ↓      ↓

$$LS: A = Q \cdot R$$

↑ orthogonal      ↓

Eigenproblems (not Jordan Form)

$$\text{Schur Form } A = Q R Q^T$$

$$\text{SVD } A = U \Sigma V^T$$

orthogonals      ↘ eigenvalues on diagonal  
 ↗  
 orth    diag    orth

Iterative:

Jacobi, Gauss-Seidel,

Conjugate gradients, Multigrid...

Common idea: Do many matrix-vector multiples, find "best solution" in subspace spanned by vectors

Randomized

Lecture 2: Floating Point Arith.

Error Analysis

Beyond bases

Exceptions

Variable precision

Reproducibility

Interval Arithmetic

Ex: Polynomial evaluation + zero finding

Recall bisection

want to solve  $f(x) = 0$

start with interval  $[x_1, x_2]$

where  $f(x_1) \cdot f(x_2) < 0$

evaluate  $f\left(\frac{x_1+x_2}{2}\right)$ , keep  
bisection until interval narrow  
enough.

(Matlab demo of bisection)