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September 17, 2024

Announcements

- HWs

Goals

- LU
- Linear systems
- maybe LSQ?

Review

$$A = LU$$

$$Ax = b$$

$$\leadsto \underbrace{LU}_{y} x = b$$

Saving the LU Factorization

What can be done to get something *like* an LU factorization?

$$PA = LU \quad \leftarrow \text{partial}$$

vs complete

Demo: LU Factorization with Partial Pivoting [cleared]

Saving the LU Factorization

What can be done to get something *like* an LU factorization?

Idea from linear algebra class: In Gaussian elimination, simply swap rows, equivalent linear system.

- ▶ Good idea: Swap rows if there's a zero in the way
- ▶ Even better idea: Find the largest entry (by absolute value), swap it to the top row.

The entry we divide by is called the *pivot*.

- ▶ Swapping rows to get a bigger pivot is called **partial pivoting**.
- ▶ Swapping rows *and columns* to get an even bigger pivot is called **complete pivoting**. (downside: additional $O(n^2)$ cost *per step* to find the pivot!)

Demo: LU Factorization with Partial Pivoting [\[cleared\]](#)

Cholesky: LU for Symmetric Positive Definite

LU can be used for SPD matrices. But can we do better?

$$A = LL^T \quad \begin{pmatrix} l_{11} & \vec{l}_{12}^T \\ & L_{22} \end{pmatrix} \quad A = A^T$$
$$\begin{matrix} 1 \\ \vdots \\ n-1 \end{matrix} \left\{ \begin{matrix} l_{11} \\ \vec{l}_{21} \\ \vdots \\ l_{n1} \end{matrix} \right\} \begin{matrix} l_{11} \\ \underbrace{\vec{l}_{21}}_{1 \times (n-1)} \\ \vdots \\ \underbrace{l_{n1}}_{n-1} \end{matrix} \begin{matrix} \vec{a}_{11} \\ \vec{a}_{21} \\ \vdots \\ A_{22} \end{matrix}$$
$$l_{11}^2 = a_{11} \Leftrightarrow l_{11} = \sqrt{a_{11}}$$
$$l_{21} l_{11} = \vec{a}_{21} \Leftrightarrow \vec{l}_{21} = \vec{a}_{21} / l_{11}$$
$$l_{21}^T l_{21} + \underbrace{L_{22} L_{22}^T}_{= A_{22}}$$

More cost concerns

What's the cost of solving $Ax = b$?

$$O(n^3) LU + fw/bw \quad O(n^2)$$

What's the cost of solving $Ax = b_1, b_2, \dots, b_n$?

$$O(n^3) LU + n \times (fw/bw) \quad O(n \cdot n^2)$$

What's the cost of finding A^{-1} ?

$$AX = I$$

$$O(n^3)$$


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.

Shrinking the constant: surprisingly hard (even for 'just' matmul)

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

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, ?trsm

Demo: BLAS Level 2 vs Level 3 [cleared]

LAPACK

LAPACK: Implements 'higher-end' things (such as LU) using BLAS
Special matrix formats can also help save const significantly, e.g.

- ▶ banded
- ▶ sparse
- ▶ symmetric
- ▶ triangular

Sample routine names:

- ▶ dgesvd, zgesdd
- ▶ dgetrf, dgetrs

LU on Blocks: The Schur Complement

Given a linear system

$$\begin{matrix}
 & & n & \nearrow & A \text{ square} \\
 m & \begin{bmatrix} A & B & | & \mathbf{b}_1 \\ C & D & | & \mathbf{b}_2 \end{bmatrix}, & & & \\
 & & & \searrow &
 \end{matrix}
 \quad \left[\begin{array}{c} -CA^{-1} \\ \mathbb{0} \end{array} \right]
 \quad \begin{array}{|c|c} A & B \\ \hline C & D \end{array} \xrightarrow{x} \mathbf{b}$$

can we do 'block Gaussian elimination' to get a *block triangular matrix*?

$$\left(\begin{array}{cc|c} A & B & \mathbf{b}_1 \\ \mathbb{0} & D - CA^{-1}B & \mathbf{b}_2 - CA^{-1}\mathbf{b}_1 \end{array} \right)$$

$$A: k \times k$$

$$C: (m-k) \times k$$

Schur complement.

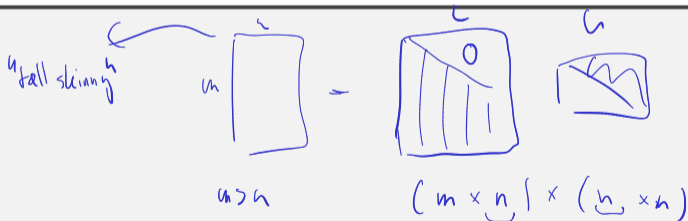
LU: Special cases

What happens if we feed a ^{square} non-invertible matrix to LU?

$$RA \rightarrow LU \quad \text{in v.}$$

(Note: In the original image, 'square' is written above 'non-invertible', and 'in v.' is written to the right. The 'A' in 'RA' has a red 'x' above it, and the 'U' in 'LU' has a red 'x' above it. The 'R' and 'L' have green checkmarks above them.)

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



Round-off Error in LU without Pivoting

Consider factorization of $\begin{bmatrix} \epsilon & 1 \\ 1 & 1 \end{bmatrix}$ where $\epsilon < \epsilon_{\text{mach}}$:

$$L = \begin{pmatrix} 1 & \\ \frac{1}{\epsilon} & 1 \end{pmatrix} \quad U = \begin{pmatrix} \epsilon & 1 \\ 0 & 1 - \frac{1}{\epsilon} \end{pmatrix}$$

$$\text{fl}(U) = \begin{pmatrix} \epsilon & 1 \\ 0 & 1 - \frac{1}{\epsilon} \end{pmatrix}$$

$$\text{fl}(L) = \begin{pmatrix} \epsilon & 1 \\ 1 & 0 \end{pmatrix}$$

$$\text{bu. err.} = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$$

Round-off Error in LU with Pivoting

Permuting the rows of A in partial pivoting gives $PA = \begin{bmatrix} 1 & 1 \\ \epsilon & 1 \end{bmatrix}$



Changing matrices

Seen: LU cheap to re-solve if RHS changes. (Able to keep the expensive bit, the LU factorization) What if the *matrix* changes?

$A \rightarrow A + \vec{u}\vec{v}^T$

$A = LU$

compute $A^{-1}\vec{b}$
in $O(n^2)$

Sherman-Morrison $(A + \vec{u}\vec{v}^T)^{-1} = A^{-1} - \frac{A^{-1}\vec{u}\vec{v}^T A^{-1}}{1 + \vec{v}^T A^{-1}\vec{u}}$

$\tilde{A}\vec{x} = \vec{b} \Leftrightarrow \vec{x} = \tilde{A}^{-1}\vec{b}$

$A^{-1}\vec{b} - \frac{(A^{-1}\vec{u}\vec{v}^T A^{-1})(A^{-1}\vec{b})}{1 + \vec{v}^T A^{-1}\vec{u}}$
 $O(n^2)$

Demo: Sherman-Morrison [cleared]

if parenthesized right.

In-Class Activity: LU

In-class activity: LU and Cost