

There is no quiz this week.

Exercise Let $A = \begin{bmatrix} 2 & 2 & -4 \\ 1 & 1 & 5 \\ 1 & 3 & 6 \end{bmatrix}$ and $b = \begin{bmatrix} 10 \\ -2 \\ -5 \end{bmatrix}$

Use Gaussian elimination with partial pivoting

find L and U , where U is upper triangular.

and L is unit lower triangular with $|l_{ij}| \leq 1$ for all

$i > j$ and $LU = \hat{A}$, where \hat{A} can be obtained

by making row interchanges. Use the LU decomposition.

to solve the system $Ax = b$.

$$L = \begin{bmatrix} 1 & 0 & 0 \\ * & 1 & 0 \\ * & * & 1 \end{bmatrix} \quad U = \begin{bmatrix} 2 & 2 & -4 \\ 1 & 1 & 5 \\ 1 & 3 & 6 \end{bmatrix}$$

$$\text{Row 2} - \frac{1}{2} \text{Row 1} \Rightarrow \begin{bmatrix} 2 & 2 & -4 \\ 0 & 0 & 3 \\ 1 & 3 & 6 \end{bmatrix} \quad L = \begin{bmatrix} 1 & 0 & 0 \\ \frac{1}{2} & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\text{Row 3} - \frac{1}{2} \text{Row 1} \Rightarrow \begin{bmatrix} 2 & 2 & -4 \\ 0 & 0 & 3 \\ 0 & 2 & 4 \end{bmatrix} \quad L = \begin{bmatrix} 1 & 0 & 0 \\ \frac{1}{2} & 1 & 0 \\ \frac{1}{2} & 0 & 1 \end{bmatrix}$$

$$\text{Row 2} \leftrightarrow \text{Row 3} \Rightarrow \begin{bmatrix} 2 & 2 & -4 \\ 0 & 2 & 4 \\ 0 & 0 & 3 \end{bmatrix} \quad L = \begin{bmatrix} 1 & 0 & 0 \\ \frac{1}{2} & 1 & 0 \\ \frac{1}{2} & 0 & 1 \end{bmatrix}$$

$$\hat{A} = \begin{bmatrix} 2 & 2 & -4 \\ 1 & 3 & 6 \\ 1 & 1 & 5 \end{bmatrix}$$

(matrix A with rows 2 and 3 interchanged)

$$\hat{b} = \begin{bmatrix} 10 \\ -5 \\ -2 \end{bmatrix}$$

$$\hat{A}x = \hat{b}$$

$$\hat{A} = LU = \begin{bmatrix} 2 & 2 & -4 \\ 0 & 2 & 4 \\ 0 & 0 & 3 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ \frac{1}{2} & 1 & 0 \\ \frac{1}{2} & 0 & 1 \end{bmatrix}$$

$$\text{Let } Ux = y$$

$$\text{then } Ly = \hat{b}$$

1.1.7

$$Ux = y \Rightarrow \begin{bmatrix} 2 & 2 & -4 \\ 0 & 2 & 4 \\ 0 & 0 & 3 \end{bmatrix} x = \begin{bmatrix} 10 \\ -10 \\ -7 \end{bmatrix}$$

$$\begin{bmatrix} 2 & 2 & -4 \\ 0 & 2 & 4 \\ 0 & 0 & 3 \end{bmatrix} x = \begin{bmatrix} 10 \\ -10 \\ -7 \end{bmatrix}$$

$$3x_3 = -7$$

$$x_3 = -\frac{7}{3}$$

$$2x_2 + 4x_3 = -10$$

$$2x_2 = -10 + \frac{28}{3}$$

$$2x_2 = -\frac{2}{3}$$

$$x_2 = -\frac{1}{3}$$

$$2x_1 + 2x_2 - 4x_3 = 10$$

$$2x_1 + 2(-\frac{1}{3}) - 4(-\frac{7}{3}) = 10$$

$$2x_1 - \frac{2}{3} + \frac{28}{3} = 10$$

$$2x_1 = \frac{30}{3} - \frac{26}{3}$$

$$2x_1 = \frac{4}{3}$$

$$x_1 = \frac{2}{3}$$

$$\text{then } Ly = b$$

$$\begin{bmatrix} 1 & 0 & 0 \\ \frac{1}{2} & 1 & 0 \\ \frac{1}{2} & 0 & 1 \end{bmatrix} y = \begin{bmatrix} 10 \\ -10 \\ -7 \end{bmatrix}$$

$$y = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix}$$

$$y_1 = 10$$

$$\frac{1}{2}y_1 + y_2 = -10$$

$$5 + y_2 = -10$$

$$y_2 = -15$$

$$5 + y_3 = -7$$

$$y_3 = -12$$

$$y = \begin{bmatrix} 10 \\ -10 \\ -7 \end{bmatrix}$$

$$x = \begin{bmatrix} \frac{2}{3} \\ -\frac{1}{3} \\ -\frac{7}{3} \end{bmatrix}$$

Cholesky Decomposition - works if A is symmetric and positive definite

Gaussian Elimination (without pivots) - works to find solution under certain conditions

Gaussian Elimination (with pivots) - works well to find solutions to $Ax = b$

if we do not have an ill-conditioned system.

The most extreme case of an ill-conditioned system is where A is singular, $\det(A) = 0$

$Ax = b$ does not have one unique solution,

if A is a singular matrix in the process of ...

uniqueness,
If A is a singular matrix in the process of Gaussian elimination there will be a step where all possible pivots are zero.

But with each step of the algorithm the entries of the matrix are subject to round off.

And we may arrive at a "solution".

But it is not in fact a solution.

The additional cost of row interchanges is not great, we make $n^2/2$ comparisons is relatively ^{small} compared, $O(n^3)$ of the algorithm,

Roughly the cost of Gaussian elimination with partial pivoting is $\frac{2}{3}n^3$ flops.

Back to example from last Thursday

$$A = \begin{bmatrix} 0 & 4 & 1 \\ 1 & 1 & 3 \\ 2 & -2 & 1 \end{bmatrix} \quad \text{and we find } \hat{A} = \begin{bmatrix} 2 & -2 & 1 \\ 0 & 4 & 1 \\ 1 & 1 & 3 \end{bmatrix}$$

Goal Find P such that $\hat{A} = PA$

$$\begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 0 & 4 & 1 \\ 1 & 1 & 3 \\ 2 & -2 & 1 \end{bmatrix} = \begin{bmatrix} 2 & -2 & 1 \\ 0 & 4 & 1 \\ 1 & 1 & 3 \end{bmatrix}$$

Definition A permutation matrix is a matrix that has exactly one 1 in each row and in each column, all other entries are zero.

Exercise Show that if P is a permutation matrix then $P^T P = P P^T = I$, thus P is nonsingular.

Exercise

then $P^T P = P P^T = I$, thus P is nonsingular
and $P^{-1} = P^T$.

$$P = \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \quad P^T = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix}$$

$$P P^T = \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Let P be a permutation matrix that is $n \times n$

Let $Q = P^T$
 $P Q_{ij} = (\text{row } i \text{ of } P) \times (\text{column } j \text{ of } Q)$

This is nonzero if and only if

Since 1 one in each row P and 1 one in each column of Q .

Since $Q = P^T$ it would be nonzero if $i = j$.

$$P Q_{ij} = \sum_{k=1}^n P_{ik} Q_{kj}$$

Let's assume $i \neq j$ and $P Q_{ij} = 1$

then there exists a k $1 \leq k \leq n$

such that $P_{ik} = Q_{kj} = 1$

Since $Q_{kj} = 1$ then $P_{jk} = 1$ since $Q = P^T$

But we said $P_{ik} = 1$ and $i \neq j$.

So P has two 1's in column k . This contradicts P being a permutation

matrix.

And $i = j$ must be the case for

$P Q_{ij}$ to be nonzero.

Theorem Gaussian elimination with partial pivoting
matrix A produces

Theorem Gaussian elimination with partial pivoting applied to an $n \times n$ matrix A produces a unit lower triangular matrix L such that $|l_{ij}| \leq 1$, an upper triangular matrix U , and a permutation matrix P such that

$$A = P^T L U$$

Exercise

$$A = \begin{bmatrix} 2 & 2 & -4 \\ 1 & 1 & 5 \\ 1 & 3 & 6 \end{bmatrix}$$

Find P , L , and U
such that $A = P^T L U$

we already know $L = \begin{bmatrix} 1 & 0 & 0 \\ \frac{1}{2} & 1 & 0 \\ \frac{1}{2} & 0 & 1 \end{bmatrix}$

$LU = \hat{A} = \begin{bmatrix} 2 & 2 & -4 \\ 1 & 3 & 6 \\ 1 & 1 & 5 \end{bmatrix}$ and $U = \begin{bmatrix} 2 & 2 & -4 \\ 0 & 2 & 4 \\ 0 & 0 & 3 \end{bmatrix}$

$P = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}$ ↖ 1,3
↗ 3,2

Thus $P^T = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}$

$A = P^T L U$ where $P^T = P = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}$ $L = \begin{bmatrix} 1 & 0 & 0 \\ \frac{1}{2} & 1 & 0 \\ \frac{1}{2} & 0 & 1 \end{bmatrix}$
and $U = \begin{bmatrix} 2 & 2 & -4 \\ 0 & 2 & 4 \\ 0 & 0 & 3 \end{bmatrix}$

Computing A^{-1}

We know $A^{-1}A = I$

$$I x = x$$

$$(A^{-1}A)x = x$$

$$A x = I$$

where x is the columns of
 $A [x_1 \ x_2 \ x_3 \ x_4 \dots x_n] = [e_1 \dots e_n]$ some matrix X
 and e gives the columns of I .

$$A x_i = e_i \quad i = 1, \dots, n$$

If we solve those n -systems we obtain A^{-1} .

Complete Pivoting

Here both row and column interchanges are allowed,

We start by finding the largest $|a_{ij}|$ and

move to 0_{11} .

after zeroing out the first column
 we now focus on an $(n-1) \times (n-1)$ matrix
 and repeat the process

- Provides extra protection from roundoff error
- The disadvantage is seeking out the largest entry.
step 1 You must compare n^2 entries
step 2 You must compare $(n-1)^2$ entries.

$$\sum_{k=0}^{n-2} (n-k)^2 \approx \frac{1}{3} n^3 \text{ comparisons}$$

Since partial pivoting works fairly this is the more often used algorithm,

Discussion in Gaussian Elimination

Error Discussion in Gaussian Elimination

- Propagation of roundoff error in Gaussian Elimination.

Example Consider the ill-conditioned matrix

$$A = \begin{bmatrix} 1000 & 999 \\ 999 & 998 \end{bmatrix}$$

How would we get L and U ?

$$\text{Row 2} - \frac{999}{1000} \text{Row 1}$$

$$\begin{bmatrix} 1000 & 999 \\ 0 & -0.001 \end{bmatrix}$$

$$\begin{aligned} 998 - \frac{999}{1000} (999) \\ 998 - 998.001 \\ = -0.001 \end{aligned}$$

Example The ill-conditioned Hilbert matrices, defined by $h_{ij} = \frac{1}{(i+j-1)}$

for example

$$H_4 = \begin{bmatrix} 1 & \frac{1}{2} & \frac{1}{3} & \frac{1}{4} \\ \frac{1}{2} & \frac{1}{3} & \frac{1}{4} & \frac{1}{5} \\ \frac{1}{3} & \frac{1}{4} & \frac{1}{5} & \frac{1}{6} \\ \frac{1}{4} & \frac{1}{5} & \frac{1}{6} & \frac{1}{7} \end{bmatrix}$$

See if you can find L and U .

$$U = \begin{bmatrix} 1 & \frac{1}{2} & \frac{1}{3} & \frac{1}{4} \\ 0 & \frac{1}{12} & \frac{1}{12} & \frac{3}{40} \\ 0 & 0 & 0 & 0 \end{bmatrix}$$