

2.4 Solving Linear Systems $Ax = b$

transform into a linear system that is easy to solve.
e.g., triangular system:

$$\begin{pmatrix} x & x & x & x \\ & x & x & x \\ & & x & x \\ & & & x \end{pmatrix} \begin{pmatrix} x \\ x \\ x \\ b \end{pmatrix} = \begin{pmatrix} x \\ x \\ b \\ x \end{pmatrix}$$

Ⓚ How would you solve it?

If M is non-singular, then solution is not affected by

$$MAz = Mb$$

because

$$z = (MA)^{-1}Mb = A^{-1}M^{-1}Mb = A^{-1}b \checkmark = x \checkmark$$

[Ex. 2.9] Permutation Matrix

- identity with rows + cols permuted

(row permutation)

$$PAx = Pb$$

reorders the equation
 x unchanged

(col permutation)

$$APz = b \Rightarrow z = (AP)^{-1}b = P^T A^{-1}b$$

$$AP(P^T x) = b$$

$$= P^T x \text{ row perm}$$

[Ex. 2.10] Diagonal Scaling

$$DAx = Db \quad D.Az = Db$$

$$ADz = b$$

$$\Rightarrow z = (AD)^{-1}b = D^{-1}A^{-1}b = D^{-1}x$$

$$ADD^{-1}z = b$$

row scaling — doesn't change exact

column scaling

soln, but does change

numerical accuracy.

§ 2.4.2 Triangular Linear Systems.

lower triangle L

$$l_{ij} = 0 \text{ when } i < j$$

upper triangular U

$$u_{ij} = 0 \text{ when } i > j$$

$$Lx = b$$

$$\sum_{k=1}^n l_{ik} x_k = b_i$$

$$x_i = \frac{1}{l_{ii}} \left(b_i - \sum_{k=1}^{i-1} l_{ik} x_k \right)$$

$$Ux = b$$

$$\sum_{k=i}^n u_{ik} x_k = b_i$$

$$x_i = \frac{1}{u_{ii}} \left(b_i - \sum_{k=i+1}^n u_{ik} x_k \right)$$

Ex. 2.11 Triangular Linear System
3x3

§ 2.4.3 Elementary Elimination Matrices

$$\begin{pmatrix} a \\ b \\ c \end{pmatrix} \rightarrow \begin{pmatrix} a \\ 0 \\ c \end{pmatrix}$$

for $j = 1 \dots n$ (*)
if $l_{jj} = 0$ stop
 $x_j = b_j / l_{jj}$
for $i = j+1 \dots n$
 $b_i = b_i - l_{ij} x_j$
end
end.

$$\begin{pmatrix} 1 & 0 \\ & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \text{ leave first row unchanged.}$$

$$\begin{pmatrix} -\frac{y}{x} & 1 \\ & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} -\frac{y}{x}x + y \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ y \end{pmatrix}$$

$$\begin{pmatrix} 1 & 0 \\ -\frac{y}{x} & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x \\ 0 \end{pmatrix}$$

$M_k a =$
"elementary elimination"
matrix or
"Gauss Transformation"

$$\begin{pmatrix} 1 & & & & \\ & 1 & & & \\ & & \dots & & \\ & & -m_{k1} & & \\ & & -m_{k2} & & 1 \\ & & -m_{kn} & & & 1 \end{pmatrix} \begin{pmatrix} a_1 \\ a_2 \\ a_3 \\ \vdots \\ a_n \end{pmatrix} = \begin{pmatrix} a_1 \\ a_2 \\ a_3 \\ \vdots \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

$$m_i = \frac{a_i}{a_k}, \quad i = k+1, \dots, n$$

pivot

Forward Substitution

for $j = 1 \dots n$
 if $l_{jj} == 0$ stop

- ① $x_j = b_j / l_{jj}$
- ② for $i = j+1 \dots n$
 $b_i = b_i - l_{ij} x_j$
 end
- end

$$\begin{matrix} \textcircled{1} \\ \textcircled{2} \end{matrix} \begin{pmatrix} \textcircled{x} & & & \\ x & x & & \\ & x & x & \\ & & x & x \end{pmatrix} \begin{pmatrix} \textcircled{x} \\ x \\ x \\ x \end{pmatrix} = \begin{pmatrix} \textcircled{x} \\ x \\ x \\ x \end{pmatrix}$$

$$\begin{pmatrix} - & & & \\ - & \textcircled{x} & & \\ - & \textcircled{x} & x & \\ - & \textcircled{x} & x & x \end{pmatrix} \begin{pmatrix} - \\ \textcircled{x} \\ x \\ x \end{pmatrix} = \begin{pmatrix} - \\ \textcircled{x} \\ x \\ x \end{pmatrix}$$

$$\begin{pmatrix} - & & & \\ - & - & & \\ - & - & \textcircled{x} & \\ - & - & \textcircled{x} & x \end{pmatrix} \begin{pmatrix} - \\ - \\ \textcircled{x} \\ x \end{pmatrix} = \begin{pmatrix} - \\ - \\ \textcircled{x} \\ \textcircled{x} \end{pmatrix}$$

$$\begin{pmatrix} - & & & \\ - & - & & \\ - & - & - & \\ - & - & - & \textcircled{x} \end{pmatrix} \begin{pmatrix} - \\ - \\ - \\ \textcircled{x} \end{pmatrix} = \begin{pmatrix} - \\ - \\ - \\ \textcircled{x} \end{pmatrix}$$

$$\begin{pmatrix} x & x & x & \textcircled{x} \\ & x & x & \textcircled{x} \\ & & x & \textcircled{x} \\ & & & \textcircled{x} \end{pmatrix} \begin{pmatrix} x \\ x \\ x \\ \textcircled{x} \end{pmatrix} = \begin{pmatrix} x \\ x \\ x \\ \textcircled{x} \end{pmatrix}$$

Backward Substitution

for $j = n \dots 1$
 if $u_{jj} == 0$ then stop
 $x_j = b_j / u_{jj}$
 for $i = j-1 \dots 1$
 $b_i = b_i - u_{ij} x_j$
 end

end

$$\begin{pmatrix} 1 & 0 \\ -\frac{a_2}{a_1} & 1 \end{pmatrix} \begin{pmatrix} a_1 \\ a_2 \end{pmatrix} = \begin{pmatrix} a_1 \\ -\frac{a_2}{a_1} \cdot a_1 + a_2 \end{pmatrix} = \begin{pmatrix} a_1 \\ 0 \end{pmatrix}$$

Elementary Elimination Matrix or Gauss Transform

$$M_k =$$

eliminate element 2 using multiple of element 1

$$M_k \begin{pmatrix} 1 & & & & & \\ & 1 & & & & \\ & & \ddots & & & \\ & & & 1 & & \\ -m_{k1} & & & & 1 & \\ & & & & & \ddots \\ & & & & & & 1 \end{pmatrix} \begin{pmatrix} a_1 \\ a_2 \\ a_3 \\ \vdots \\ a_n \end{pmatrix} = \begin{pmatrix} a_1 \\ \vdots \\ a_k \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

$$m_i = \frac{a_i}{a_k}, i = k+1 \dots n$$

leave first k elements unchanged pivot

$j > k$

$$\begin{pmatrix} 1 & & & & & \\ & 1 & & & & \\ & & \ddots & & & \\ & & & 1 & & \\ -m_{j1} & & & & 1 & \\ & & & & & \ddots \\ & & & & & & 1 \end{pmatrix} \begin{pmatrix} 1 & & & & & \\ & 1 & & & & \\ & & \ddots & & & \\ & & & 1 & & \\ -m_{kj} & & & & 1 & \\ & & & & & \ddots \\ & & & & & & 1 \end{pmatrix}$$

$$\vec{e}_j^T M_k = (0 \dots 0 \dots 0) \begin{pmatrix} 0 \\ \vdots \\ m_{k+1} \\ \vdots \\ m_n \end{pmatrix} = m_j$$

$$\times M_j M_k = (I - \vec{m}_j \vec{e}_j^T) (I - \vec{m}_k \vec{e}_k^T) = I - m_j \vec{e}_j^T - m_k \vec{e}_k^T + \vec{m}_j (\vec{e}_j^T m_k) \vec{e}_k^T = I - \vec{m}_j \vec{e}_j^T - \vec{m}_k \vec{e}_k^T + m_j \vec{m}_k \vec{e}_k^T$$

$$\checkmark M_k M_j = (I - \vec{m}_k \vec{e}_k^T) (I - \vec{m}_j \vec{e}_j^T) = I - \vec{m}_k \vec{e}_k^T - \vec{m}_j \vec{e}_j^T + \vec{m}_k (\vec{e}_k^T \vec{m}_j) \vec{e}_j^T = I - \vec{m}_k \vec{e}_k^T - \vec{m}_j \vec{e}_j^T$$

$$\checkmark M_k^{-1} M_j^{-1} = (I + m_k \vec{e}_k^T) (I + m_j \vec{e}_j^T) = I + m_k \vec{e}_k^T + m_j \vec{e}_j^T$$

① M_k - lower triangular - unit diagonal - non-singular.

② $M_k = I - \vec{m}_k \vec{e}_k^T$, $m_{ik} = \begin{pmatrix} 0 \\ \vdots \\ m_{k+1} \\ \vdots \\ m_n \end{pmatrix}$

③ $M_k^{-1} = I + \vec{m}_k \vec{e}_k^T$

check

$$\begin{aligned} (I - m_k e_k^T)(I + m_k e_k^T) &= I + m_k e_k^T - m_k e_k^T - m_k (e_k^T m_k) e_k^T \\ &= I \end{aligned}$$

④ $j > k$

$$\begin{aligned} M_j M_k &= (I - m_j e_j^T)(I - m_k e_k^T) \\ &= I - m_j e_j^T - m_k e_k^T + m_j e_j^T m_k e_k^T \end{aligned} \quad X$$

$$\begin{aligned} M_k M_j &= (I - m_k e_k^T)(I - m_j e_j^T) \\ &= I - m_k e_k^T - m_j e_j^T + m_k e_k^T m_j e_j^T \\ &= I - m_k e_k^T - m_j e_j^T \end{aligned}$$

$$\begin{aligned} M_k^{-1} M_j^{-1} &= (I + m_k e_k^T)(I + m_j e_j^T) \\ &= I + m_k e_k^T + m_j e_j^T + m_k (e_k^T m_j) e_j^T \end{aligned}$$

⊛ and true $M_i^{-1} M_k^{-1} M_j^{-1} = I + m_k e_k^T + m_j e_j^T + m_j e_j^T m_k e_k^T$ for $k < j$

§2.4.4 Gaussian Elimination & LU Factorization

$$\begin{aligned} M_1 A x &= M_1 b && a_{11} \text{ pivot} \\ M_2 M_1 A x &= M_2 M_1 b \\ &\vdots \\ M_n \dots M_2 M_1 A x &= M_n \dots M_2 M_1 b \end{aligned}$$

this makes easy to write down way

$$\underbrace{M}_M A x = M b$$

upper triangular

$$\begin{aligned} M^{-1} (M A) x &= b && \Rightarrow A = LU \\ \underbrace{(M_1^{-1} M_2^{-1} \dots M_n^{-1}) M A}_L U x &= b \end{aligned}$$

A

$$L U x = b \quad \text{Ⓚ How to solve now?}$$

LU factors can be reused for different right hand sides

4x4

LU Factorization

$$\begin{pmatrix} \boxed{x} & x & x & x \\ \hline x & x & x & x \\ x & x & x & x \\ x & x & x & x \end{pmatrix}$$

$$\begin{pmatrix} 1 & & & \\ m_{21} & 1 & & \\ m_{31} & & 1 & \\ m_{41} & & & 1 \end{pmatrix} \begin{pmatrix} 1 & & & \\ -m_{21} & 1 & & \\ -m_{31} & & 1 & \\ -m_{41} & & & 1 \end{pmatrix} \begin{pmatrix} x & | & x & x & x \\ \hline x & | & x & x & x \\ x & | & x & x & x \\ x & | & x & x & x \end{pmatrix}$$

$$\begin{pmatrix} 1 & & & \\ m_{21} & 1 & & \\ m_{31} & & 1 & \\ m_{41} & & & 1 \end{pmatrix} \begin{pmatrix} x & | & x & x & x \\ \hline 0 & | & x & x & x \\ 0 & | & x & x & x \\ 0 & | & x & x & x \end{pmatrix} = M_1^{-1} M_1 A$$

$$\begin{pmatrix} 1 & & & \\ m_{21} & 1 & & \\ m_{31} & & 1 & \\ m_{41} & & & 1 \end{pmatrix} \begin{pmatrix} 1 & & & \\ & 1 & & \\ & m_{32} & 1 & \\ & m_{42} & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & & & \\ & 1 & & \\ -m_{32} & 1 & & \\ -m_{42} & & 1 & \end{pmatrix} \begin{pmatrix} x & x & | & x & x \\ \hline 0 & x & | & x & x \\ 0 & x & | & x & x \\ 0 & x & | & x & x \end{pmatrix} = M_1^{-1} M_2^{-1} M_2 M_1 A$$

$$\begin{pmatrix} 1 & & & \\ m_{21} & 1 & & \\ m_{31} & m_{32} & 1 & \\ m_{41} & m_{42} & 0 & 1 \end{pmatrix} \begin{pmatrix} x & x & | & x & x \\ \hline 0 & x & | & x & x \\ 0 & 0 & | & x & x \\ 0 & 0 & | & x & x \end{pmatrix}$$

$$\begin{pmatrix} 1 & & & \\ m_{31} & 1 & & \\ m_{31} & m_{32} & 1 & \\ m_{41} & m_{42} & m_{43} & 1 \end{pmatrix} \begin{pmatrix} x & x & x & x \\ 0 & x & x & x \\ 0 & 0 & x & x \\ 0 & 0 & 0 & x \end{pmatrix} = (M_1^{-1} M_2^{-1} M_3^{-1}) (M_3 M_2 M_1 A) = LU$$