

# Direct methods to solve linear systems

$$A \in \mathbb{R}^{n \times n} \quad A \text{ nonsingular}$$

$$A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix} \quad a_{ij} \in \mathbb{R}$$

$$b = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix} \in \mathbb{R}^n \quad A \text{ and } b \text{ given}$$

$$x = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \in \mathbb{R}^n \quad \text{unknown} \quad \text{goal: Find } x$$

$$Ax = b$$

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1$$

$$a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2$$

$$\vdots$$

$$a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nn}x_n = b_n$$

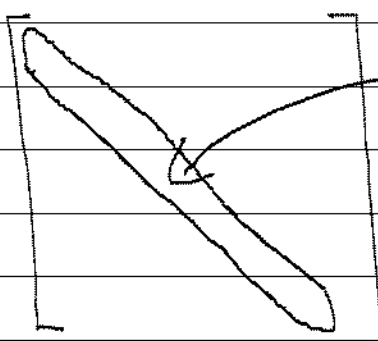
Def:  $A \in \mathbb{R}^{n \times n}$  is upper triangular if  $A$  is of the form

$$A = \begin{bmatrix} x & - & - & - & x \\ & \ddots & & & \vdots \\ 0 & & & & \end{bmatrix}, \text{ ie } a_{ij} = 0 \text{ if } i > j, \text{ all the entries}$$

$$\begin{bmatrix} \vdots & & & & \\ 0 & \dots & 0 & & \\ & & & \ddots & \\ & & & & x \end{bmatrix}$$

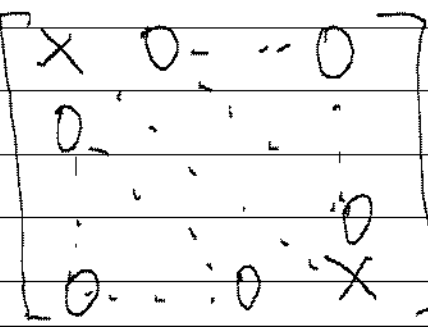
to zero

of  $A$  below the diagonal are equal

$A =$   diagonal of  $A$  (entries  $a_{ii}$   $1 \leq i \leq n$ )

$A$  is lower triangular if

$$A = \begin{bmatrix} x & 0 & \dots & 0 \\ \vdots & \ddots & \ddots & \vdots \\ \vdots & & \ddots & 0 \\ x & - & - & x \end{bmatrix}$$

$A$  is diagonal if  $A =$  

Solving  $Ax=b$  by back substitution if  $A$  is upper triangular

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1$$

$$a_{22}x_2 + \dots + a_{2n}x_n = b_2$$

"

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$$a_{n-1,n-1}x_{n-1} + a_{n-1,n}x_n = b_{n-1}$$

$$a_{nn}x_n = b_n$$

$$x_n = b_n / a_{nn}$$

$$x_k = \left( b_k - \sum_{j=k+1}^n a_{kj}x_j \right) / a_{kk}$$

start with  $k=n-1$  and

go all the way down to  $k=1$

This works when  $a_{kk} \neq 0$  for all  $1 \leq k \leq n$ , which is the case if and only if  $A$  is invertible

Algorithm (back substitution)

for  $k=n, 1$

$$x_k = b_k$$

for  $j=k+1, n$

$$x_k = x_k - a_{kj} x_j$$

end

Complexity

$$\boxed{\frac{n(n+1)}{2}}$$

$\left. \begin{matrix} \leftarrow 1 \\ \end{matrix} \right\}$

$n-k$

$$\sum_{k=1}^n n-k+1 =$$

$$= n(n+1)/2$$

$$x_k = x_k / a_{kk}$$

end

} 1 )

### Elementary operations

- 1) Multiply an equation by a non-zero number
- 2) Replace equation  $k$  by equation  $k + \alpha$  equation  $j$ , where  $\alpha$  is any number
- 3) Exchange two equations

Obs. The solutions of  $Ax=b$  does not change if we perform elementary operations

Gaussian Elimination: Perform elementary operations on  $Ax=b$  to be left with an upper triangular system

$$Ax=b \quad A^{(1)} = A \quad b^{(1)} = b$$

$$A^{(k)} = \begin{bmatrix} a_{11}^{(k)} & - & - & - & a_{1n}^{(k)} \\ 0 & & & & \vdots \\ & 0 & a_{kk}^{(k)} & - & a_{kn}^{(k)} \\ & & \vdots & & \vdots \\ 0 & \dots & 0 & a_{nk}^{(k)} & a_{nn}^{(k)} \end{bmatrix}$$

1) Find  $j$  such that  $k \leq j \leq n$  and  $|a_{jk}^{(k)}| \geq |a_{lk}^{(k)}|$  for all  $k \leq l \leq n$

2) Swap equations  $k$  and  $j$ . This is called pivoting and keeps the algorithm stable

3) Replace equation  $l$  by equation  $l - \frac{a_{lk}^{(k)}}{a_{kk}^{(k)}} \text{ equation } k$  for  $k+1 \leq l \leq n$

Example: 1) 
$$\begin{bmatrix} 1 & 2 & 1 \\ 2 & 2 & 3 \\ -1 & -3 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 3 \\ 2 \end{bmatrix}$$

Eq 2 
$$\left[ \begin{array}{ccc|c} 2 & 2 & 3 & 3 \\ 1 & 2 & 1 & 0 \\ -1 & -3 & 0 & 2 \end{array} \right]$$

Eq 2  $-\frac{1}{2}$  Eq 1

Eq 3  $+\frac{1}{2}$  Eq 1

$$\left[ \begin{array}{ccc|c} 2 & 2 & 3 & 3 \\ 0 & 1 & -1/2 & -3/2 \\ 0 & -2 & 3/2 & 7/2 \end{array} \right]$$

Eq 3  
Eq 2

$$\begin{bmatrix} 2 & 2 & 3 & : & 3 \\ 0 & -2 & 3/2 & : & 7/2 \\ 0 & 1 & -1/2 & : & -3/2 \end{bmatrix}$$

$$\begin{bmatrix} 2 & 2 & 3 & : & 3 \\ 0 & -2 & 3/2 & : & 7/2 \\ 0 & 0 & 1/4 & : & 1/4 \end{bmatrix}$$

$$Eq_3 + \frac{1}{2} Eq_2$$

Algorithm (Gaussian Elimination)

Complexity

for  $i = 1 : n-1$

$k = i$

for  $l = i+1 : n$

if  $|a_{li}| > |a_{ki}|$  then

$k = l$

endif

end

for  $j=i:n$

$x = a_{ij}$

$a_{ij} = a_{kj}$

$a_{kj} = x$

end

$x = b_i$

$b_i = b_k$

$b_k = x$

for  $l = i+1:n$

$$x = a_{li}/a_{ii}$$

$\{ 1$

$$a_{li} = 0$$

for  $j = i+1:n$

$$a_{lj} = a_{lj} - x a_{ij}$$

$\{ 1$

end

$$b_l = b_l - x b_i$$

$\{ 1$

end

end

Complexity

$\sum_{i=1}^{n-1}$

$$(n-i+2)(n-i) \approx \frac{n^3}{3} = O(n^3)$$

$(n-i+2)(n-i)$

$n-i+2$

$n-i$

Use algorithm to solve  $Ax=b$  when  $A$  is upper triangular

LU factorization A little extra work lead to finding a permutation  $P$  and lower and upper triangular matrices  $L$  and  $U$  such that  $PA=LU$

Cholesky factorization

Def:  $A^T$  is the transpose of  $A$

Def:  $A$  is symmetric  $A=A^T$

Def:  $A \in \mathbb{R}^{n \times n}$   $x \in \mathbb{R}^n$   $x^T A x \in \mathbb{R}$

$A$  is positive definite if  $x^T A x \geq 0$  for all  $x \in \mathbb{R}^n$   
and  $x^T A x = 0$  only if  $x = 0 = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$

Fact:  $A$  is symmetric positive definite. then there exists  
 $U$  upper triangular such that  $A = U^T U$

This is called the Cholesky factorization

Obs:  $A = U^T U$  with  $U$  upper triangular. then

$$a_{ij} = \sum_{k=1}^n [U^T]_{ik} [U]_{kj} = \sum_{k=1}^n u_{ki} u_{kj}$$

Assume

$$j \geq i$$

$$a_{ij} = \sum_{k=1}^i u_{ki} u_{kj}$$

If  $i < j$  then

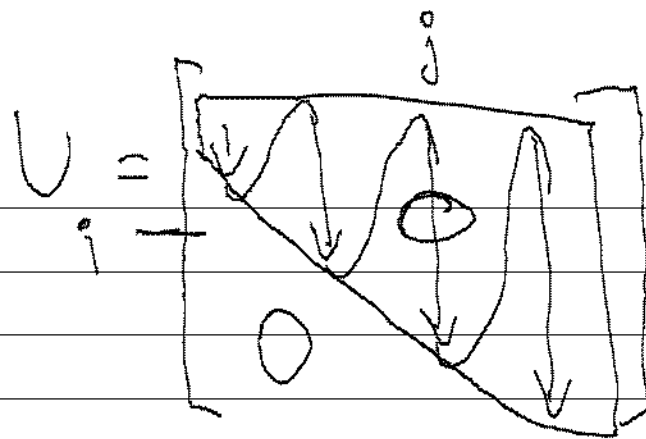
$$a_{ij} = u_{1i} u_{1j} + u_{2i} u_{2j} + \dots + u_{i-1,i} u_{i-1,j}$$

$$u_{ij} = (a_{ij} - \sum_{k=1}^{i-1} u_{ki} u_{kj}) / u_{ii}$$

If  $i = j$

$$a_{ij} = a_{ii} = u_{1i}^2 + u_{2i}^2 + \dots + u_{i-1,i}^2$$

$$u_{ii} = \sqrt{a_{ii} - \sum_{k=1}^{i-1} u_{ki}^2}$$



# Algorithm Cholesky

Complexity

for  $j = 1:n$

for  $i = 1:j-1$

$$u_{ij} = a_{ij}$$

for  $k = 1:i-1$

$$u_{ij} = u_{ij} - u_{ki} u_{kj}$$

end

$$u_{ij} = u_{ij} / u_{ii}$$

end

$$u_{jj} = a_{jj}$$

$$\left. \left. \left. \begin{matrix} j-1 \\ 1 \end{matrix} \right\} \right\} i-1 \right\} \frac{j(j-1)}{2}$$

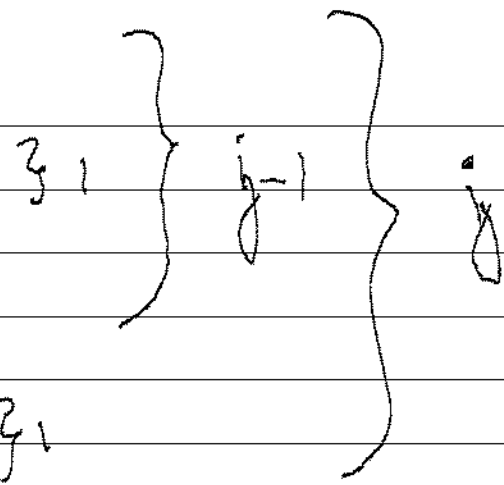
for  $k=1:j-1$

$$u_{jj} = u_{jj} - u_{kj}^2$$

end

$$u_{jj} = \sqrt{u_{jj}}$$

end



Complexity  $\sum_{j=1}^n \frac{j(j+1)}{2} \approx \frac{n^3}{6} = O(n^3)$