

## Key points in chapter 1

Suppose  $A$  is  $n \times n$  and suppose  $A\underline{v}_i = \lambda_i \underline{v}_i$ ,  $\underline{v}_i \neq \underline{0}$ ,  $i = 1, \dots, n$ .  
 $\underline{v}_i$  is an **eigenvector**,  $\lambda_i$  is the **eigenvalue**.

$$\rho(A) = \max\{|\lambda_1|, \dots, |\lambda_n|\} = \text{spectral radius of } A.$$

We have **vector norms** and we use the notation  $\|\underline{x}\|$ .

The **matrix norm** induced by a vector norm is

$$\|A\| = \max\{\|A\underline{x}\| : \|\underline{x}\| = 1\}.$$

For all such matrix norms  $\rho(A) \leq \|A\|$ .

The **matrix condition number** is defined by

$$\kappa(A) = \|A\| \|A^{-1}\|, \quad 1 \leq \kappa(A) \leq \infty.$$

We say  $\kappa(A) = \infty$  when  $A$  does not have an inverse.

$\kappa(A)$  is large when  $A$  is near to a matrix which has no inverse.

When this is the case the solution  $\underline{x}$  to  $A\underline{x} = \underline{b}$  is very sensitive to changes in  $A$  and/or  $\underline{b}$ .

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## The common vector norms

$$\|\underline{x}\|_2 = (x_1^2 + x_2^2 + \dots + x_n^2)^{1/2} = (\underline{x}^T \underline{x})^{1/2},$$

$$\|\underline{x}\|_\infty = \max\{|x_1|, \dots, |x_n|\},$$

$$\|\underline{x}\|_1 = |x_1| + |x_2| + \dots + |x_n|.$$

## Expressions for the common vector norms

$$\|A\|_\infty = \max_{1 \leq i \leq n} \sum_{j=1}^n |a_{ij}|,$$

$$\|A\|_1 = \max_{1 \leq j \leq n} \sum_{i=1}^n |a_{ij}|,$$

$$\|A\|_2 = (\rho(A^T A))^{1/2}.$$

$\|A\|_\infty$  and  $\|A\|_1$  are "easy to calculate".

$\|A\|_2$  is "expensive to calculate".

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## Chap 2: Direct methods for solving $A\underline{x} = \underline{b}$

### Upper triangular systems

$$\begin{array}{rcccccc} u_{11}x_1 & + & u_{12}x_2 & + & \dots & + & u_{1n}x_n & = & b_1 \\ & & + & u_{22}x_2 & + & \dots & + & u_{2n}x_n & = & b_2 \\ & & & & \ddots & & & & & \vdots \\ & & & & & & u_{n-1,n-1}x_{n-1} & + & u_{n-1,n}x_n & = & b_{n-1} \\ & & & & & & & & u_{nn}x_n & = & b_n \end{array}$$

Backward substitution:

$$\begin{aligned} x_n &= b_n / u_{nn} \\ x_i &= \left( b_i - \sum_{k=i+1}^n u_{ik} x_k \right) / u_{ii}, \quad i = n-1, \dots, 1. \end{aligned}$$

$\mathcal{O}(n^2/2)$  entries in  $U$  and about  $n^2$  operations to get  $\underline{x}$ .

With lower triangular systems we use forward substitution which has the same number of operations.

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## Solving $A\underline{x} = \underline{b}$ when $A = LU$

Here  $L$  is lower triangular and  $U$  is upper triangular.

$$A\underline{x} = LU\underline{x} = \underline{b}.$$

Algorithm:

Solve  $L\underline{y} = \underline{b}$  by forward substitution.

Solve  $U\underline{x} = \underline{y}$  by backward substitution.

The number of operations is about the same as computing

$$A^{-1}\underline{b}$$

if the inverse matrix  $A^{-1}$  is available.

It is rare to need to have  $A^{-1}$ . When we write  $\underline{x} = A^{-1}\underline{b}$  it should be considered as a way of describing the solution and not a preferred method to compute the solution.

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## Reduction to triangular form

At each stage in the basic reduction process we create zeros below the diagonal in a column and we have a vector of the **multipliers**.

$$\begin{aligned}
 A = \begin{pmatrix} x & x & x & x \\ x & x & x & x \\ x & x & x & x \\ x & x & x & x \end{pmatrix} &\rightarrow \begin{pmatrix} x & x & x & x \\ 0 & X & X & X \\ 0 & X & X & X \\ 0 & X & X & X \end{pmatrix} = A^{(1)}, \quad \underline{m}_1 = \begin{pmatrix} 0 \\ m_{21} \\ m_{31} \\ m_{41} \end{pmatrix}, \\
 &\rightarrow \begin{pmatrix} x & x & x & x \\ 0 & x & x & x \\ 0 & 0 & X & X \\ 0 & 0 & X & X \end{pmatrix} = A^{(2)}, \quad \underline{m}_2 = \begin{pmatrix} 0 \\ 0 \\ m_{32} \\ m_{42} \end{pmatrix}, \\
 &\rightarrow \begin{pmatrix} x & x & x & x \\ 0 & x & x & x \\ 0 & 0 & x & x \\ 0 & 0 & 0 & X \end{pmatrix} = U, \quad \underline{m}_3 = \begin{pmatrix} 0 \\ 0 \\ 0 \\ m_{43} \end{pmatrix}.
 \end{aligned}$$

x and X are potentially non-zero.

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## 3 × 3 example

Solve

$$\begin{pmatrix} 1 & 1 & 1 \\ 2 & 1 & -1 \\ 3 & 2 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 6 \\ 5 \\ 13 \end{pmatrix}.$$

Basic Gauss elimination involves the following sequence.

$$\begin{pmatrix} 1 & 1 & 1 & 6 \\ 2 & 1 & -1 & 5 \\ 3 & 2 & 1 & 13 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 1 & 1 & 6 \\ 0 & -1 & -3 & -7 \\ 0 & -1 & -2 & -5 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 1 & 1 & 6 \\ 0 & -1 & -3 & -7 \\ 0 & 0 & 1 & 2 \end{pmatrix}.$$

At the 1st stage we subtract multiples of row 1 from rows 2 and 3.

At the 2nd stage we subtract multiples of row 2 from row3.

Back substitution then gives  $x_3 = 2$ ,  $x_2 = 1$  and  $x_1 = 3$ . Here collecting the multipliers together gives  $A = LU$  with

$$L = \begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 3 & 1 & 1 \end{pmatrix} \quad \text{and} \quad U = \begin{pmatrix} 1 & 1 & 1 \\ 0 & -1 & -3 \\ 0 & 0 & 1 \end{pmatrix}.$$

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## The $A = LU$ factorization

When the basic reduction is possible we have the factorization

$$\begin{pmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ m_{21} & 1 & 0 & 0 \\ m_{31} & m_{32} & 1 & 0 \\ m_{41} & m_{42} & m_{43} & 1 \end{pmatrix} \begin{pmatrix} u_{11} & u_{12} & u_{13} & u_{14} \\ 0 & u_{22} & u_{23} & u_{24} \\ 0 & 0 & u_{33} & u_{34} \\ 0 & 0 & 0 & u_{44} \end{pmatrix}.$$

$$L = M_1^{-1} M_2^{-1} M_3^{-1} = I + \underline{m}_1 \underline{e}_1^T + \underline{m}_2 \underline{e}_2^T + \underline{m}_3 \underline{e}_3^T,$$

where each  $M_k = I - \underline{m}_k \underline{e}_k^T$  is a **Gauss transformation matrix** with the inverse being  $M_k^{-1} = I + \underline{m}_k \underline{e}_k^T$ .

Later we sometimes write  $l_{ij} = m_{ij}$  for the entries of the lower triangular matrix  $L$ . As the diagonal entries of  $L$  are all equal to 1 the matrix is said to be **unit lower triangular**.

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## Factorization of the principal sub-matrices

$$\begin{aligned}
 a_{11} &= u_{11}. \\
 \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} &= \begin{pmatrix} 1 & 0 \\ l_{21} & 1 \end{pmatrix} \begin{pmatrix} u_{11} & u_{12} \\ 0 & u_{22} \end{pmatrix} \\
 \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix} &= \begin{pmatrix} 1 & 0 & 0 \\ l_{21} & 1 & 0 \\ l_{31} & l_{32} & 1 \end{pmatrix} \begin{pmatrix} u_{11} & u_{12} & u_{13} \\ 0 & u_{22} & u_{23} \\ 0 & 0 & u_{33} \end{pmatrix}. \\
 \begin{pmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{pmatrix} &= \begin{pmatrix} 1 & 0 & 0 & 0 \\ l_{21} & 1 & 0 & 0 \\ l_{31} & l_{32} & 1 & 0 \\ l_{41} & l_{42} & l_{43} & 1 \end{pmatrix} \begin{pmatrix} u_{11} & u_{12} & u_{13} & u_{14} \\ 0 & u_{22} & u_{23} & u_{24} \\ 0 & 0 & u_{33} & u_{34} \\ 0 & 0 & 0 & u_{44} \end{pmatrix}.
 \end{aligned}$$

In general

$$\det(A_k) = u_{11} \cdots u_{kk}.$$

This factorization is possible if all the principle sub-matrices are invertible, i.e. the entries  $u_{kk} \neq 0$ ,  $k = 1, \dots, n - 1$ . We also need  $u_{nn} \neq 0$  to solve a system.

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