

# Computational Physics

## Lecture 3 – Matrix Methods

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Slides available from <https://bb.imperial.ac.uk/>

# Goals

By the end of today's lecture, you should be able to

- Solve matrix equations numerically
- Obtain the inverse of a matrix numerically
- Implement a simple elimination scheme to do the above
- Numerically decompose a matrix into a product of upper and lower triangular matrices

# Outline

- 1 The problem
- 2 Gauss-Jordan Elimination
- 3 LU Decomposition

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# Matrix equations

Want to solve a set of  $M$  linear equations in  $N$  unknowns  $x_{j=1..N}$ :

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1,N}x_N = b_1 \quad (1)$$

$$a_{21}x_1 + a_{22}x_2 + \dots + a_{2,N}x_N = b_2 \quad (2)$$

$$\dots \quad (3)$$

$$a_{M,1}x_1 + a_{M,2}x_2 + \dots + a_{M,N}x_N = b_M \quad (4)$$

These are most easily represented as matrices and vectors

$$\mathbf{Ax} = \mathbf{b} \quad (5)$$

$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & \dots \\ a_{21} & a_{22} & \dots \\ \dots & \dots & \end{bmatrix}, \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \dots \end{bmatrix}, \mathbf{b} = \begin{bmatrix} b_1 \\ b_2 \\ \dots \end{bmatrix}. \quad (6)$$

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

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in order to obtain different solutions for  $\mathbf{x}$ .

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in order to obtain different solutions for  $\mathbf{x}$ .
- 2 The inverse of  $A$ .  
Note that this is not the same as 1, and usually not the  
best way to achieve 1!

# Which method is best for my problem?

Many, many, many<sup>many</sup> canned matrix solution and inversion codes exist

- Highly optimised
- Well-know public free examples: LAPACK, BLAS
- Practically, you will usually want to use one of these

<http://www.netlib.org/lapack/>

<http://www.netlib.org/blas/>

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## 1 Gauss-Jordan elimination

- Computes set number of solutions and inverse in one hit
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- super-simple

# Which method is best for my problem?

However, a few simple methods are worth knowing about:

## 1 Gauss-Jordan elimination

- Computes set number of solutions and inverse in one hit
- All needs to be done over again for new **b** values
- super-simple

## 2 LU decomposition and back-substitution

- Faster when just solutions are needed
- Fast and accurate to apply to new **b** values
- Similar speed to Gauss-Jordan for inverses
- more complicated but almost always better choice

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# How you would proceed without a computer...

## Question:

What is the most naive way to solve  $A\mathbf{x} = \mathbf{b}$ ?

## Answer:

Replace rows (single equations) by linear combinations of each other until you've transformed  $A$  into the identity  $\mathbf{1}$

$\implies$  you will have transformed  $A\mathbf{x}$  into just  $\mathbf{x}$  and  $\mathbf{b}$  into  $\mathbf{b}A^{-1}$ ,  
i.e. you will have

$$\mathbf{x} = \mathbf{b}A^{-1}$$

and be able to just read the solution off the righthand side.

# Carrying out the transformation

Set the problem up as follows:

$$\begin{bmatrix} a_{11} & a_{12} & \dots \\ a_{21} & a_{22} & \dots \\ \dots & \dots & \dots \end{bmatrix} \begin{bmatrix} \begin{pmatrix} x_{11} \\ x_{21} \\ \dots \end{pmatrix} \begin{pmatrix} x_{12} \\ x_{22} \\ \dots \end{pmatrix} \dots \end{bmatrix} = \begin{bmatrix} \begin{pmatrix} b_{11} \\ b_{21} \\ \dots \end{pmatrix} \begin{pmatrix} b_{12} \\ b_{22} \\ \dots \end{pmatrix} \dots \end{bmatrix} \quad (7)$$

So, how do you achieve the transformation  $A \rightarrow \mathbf{1}$ ? You need to work one column at a time. For a given column  $j$ ,

- 1 Divide the  $j$ th row by  $a_{jj}$ , ensuring that  $a_{jj} \rightarrow 1$ . OK, so now you have the  $j$ th entry on the diagonal of the identity.
- 2 Subtract an appropriate multiple of the  $j$ th row from all the other entries in the column, so that all the other entries in column get set to zero. Now you have the non-diagonal elements for the  $j$ th column.



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# Carrying out the transformation

Set the problem up as follows:

$$\begin{bmatrix} 1 & a_{12}/a_{11} & \dots \\ a_{21} & a_{22} & \dots \\ \dots & \dots & \dots \end{bmatrix} \left[ \begin{pmatrix} x_{11} \\ x_{21} \\ \dots \end{pmatrix} \begin{pmatrix} x_{12} \\ x_{22} \\ \dots \end{pmatrix} \right] = \left[ \begin{pmatrix} b_{11}/a_{11} \\ b_{21} \\ \dots \end{pmatrix} \begin{pmatrix} b_{12}/a_{11} \\ b_{22} \\ \dots \end{pmatrix} \right]$$

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# Carrying out the transformation

Set the problem up as follows:

$$\begin{bmatrix} 1 & a_{12}/a_{11} & \dots \\ 0 & a_{22} - a_{21}a_{12}/a_{11} & \dots \\ \dots & \dots & \dots \end{bmatrix} \left[ \begin{pmatrix} x_{11} \\ x_{21} \\ \dots \end{pmatrix} \begin{pmatrix} x_{12} \\ x_{22} \\ \dots \end{pmatrix} \right]$$

$$= \left[ \begin{pmatrix} b_{11}/a_{11} & b_{12}/a_{11} \\ b_{21} - a_{21}b_{11}/a_{11} & b_{22} - a_{21}b_{12}/a_{11} \\ \dots & \dots \end{pmatrix} \right]$$

So, how do you achieve the transformation  $A \rightarrow \mathbf{1}$ ? You need to work one column at a time. For a given column  $j$ ,

- ➊ Divide the  $j$ th row by  $a_{jj}$ , ensuring that  $a_{jj} \rightarrow 1$ . OK, so now you have the  $j$ th entry on the diagonal of the identity.
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- Now you have the non-diagonal elements for the  $j$ th column.

## Carrying out the transformation

Perform this algorithm yourself by hand for a  $2 \times 2$  example (2–3 mins, feel free to discuss):

$$\begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 6 \\ 2 \end{bmatrix}$$
$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \text{??} \quad (8)$$

Work one column at a time. For a given column  $j$ ,

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$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} \frac{10}{3} \\ -\frac{2}{3} \end{bmatrix} \quad (8)$$

Work one column at a time. For a given column  $j$ ,

- 1 Divide the  $j$ th row by  $a_{jj}$ , ensuring that  $a_{jj} \rightarrow 1$ . OK, so now you have the  $j$ th entry on the diagonal of the identity.
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# Pivoting

Well, you're almost done.

- Notice that you're dividing by  $a_{jj}$  at each step.
- What if  $a_{jj} = 0$ ? Everything breaks.
- You need to **pivot**.
- $\rightarrow$  pre-shuffle the rows so that you never get a zero on the diagonal.
- Actually the most accuracy comes from shuffling so that the *biggest* value of each column ends up on the diagonal.
- Some codes renormalise each row to maximum value 1 to start with  $\rightarrow$  *implicit pivoting*

Pivoting is a bit fiddly and annoying, as you need to keep track of the row swaps – but it's essential for making a robust matrix solver.

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# Triangular Matrices

*Upper triangular* matrix  $U$  has diagonal and upper elements.

*Lower triangular* matrix  $L$  has diagonal and lower elements.

$$U = \begin{bmatrix} u_{11} & u_{12} & \dots \\ 0 & u_{22} & \dots \\ 0 & 0 & \dots \end{bmatrix}, L = \begin{bmatrix} \ell_{11} & 0 & 0 \\ \ell_{21} & \ell_{22} & 0 \\ \ell_{31} & \ell_{32} & \dots \end{bmatrix}. \quad (9)$$

Imagine if we could write  $A = LU$ . Then,

$$A\mathbf{x} = LU\mathbf{x} = L(U\mathbf{x}) = \mathbf{b}, \quad (10)$$

i.e. if we can solve

$$L\mathbf{y} = \mathbf{b} \quad (11)$$

$$U\mathbf{x} = \mathbf{y} \quad (12)$$

then we can solve  $A\mathbf{x} = \mathbf{b}$ .

# Forward and back-substitution

Actually,  $L\mathbf{y} = \mathbf{b}$  and  $U\mathbf{x} = \mathbf{y}$  are very easy to solve:

$$y_1 = \frac{b_1}{\ell_{11}} \quad (13)$$

$$y_i = \frac{1}{\ell_{ii}} \left( b_i - \sum_{j=1}^{i-1} \ell_{ij} y_j \right), i \neq 1 \quad (14)$$

$$x_N = \frac{y_N}{u_{N,N}} \quad (15)$$

$$x_i = \frac{1}{u_{ii}} \left( y_i - \sum_{j=i+1}^N u_{ij} x_j \right), i \neq N \quad (16)$$

So, all that remains is to work out how to get  $L$  and  $U$ ...

# The actual decomposition

If we actually write out  $A = LU$  we get:

$$i \leq j: \quad \ell_{i1}u_{1j} + \ell_{i2}u_{2j} + \dots + \ell_{ij}u_{jj} = a_{ij} \quad (17)$$

$$i \geq j: \quad \ell_{i1}u_{1j} + \ell_{i2}u_{2j} + \dots + \ell_{ij}u_{jj} = a_{ij} \quad (18)$$

This is  $N^2$  equations in  $N^2 + N$  unknowns  $\implies N$  free entries to choose at will.

So, just choose  $L$  to have entries of 1 on the diagonal, for simplicity.

The system of equations can then be easily solved with *Crout's Algorithm*.

# Crout's Algorithm

Essentially just reshuffles the rows in the equation  $A = LU$ :

- 1 Set all  $\ell_{jj} = 1$  (i.e. for  $1 \leq i \leq N$ ).
- 2 For each  $j = 1..N$ :
  - a) Use first equation (17) with  $\ell_{jj} = 1$  to get

$$u_{ij} = a_{ij} - \sum_{k=1}^{i-1} \ell_{ik} u_{kj} \quad (19)$$

- b) Use second equation (18) to get

$$\ell_{ij} = u_{jj}^{-1} \left( a_{ij} - \sum_{k=1}^{j-1} \ell_{ik} u_{kj} \right) \quad (20)$$

If working up in  $j$ , each  $u$  and  $\ell$  entry gets calculated before it is needed.

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Note that pivoting is still needed for robustness! (Luckily you get an example in your assignment that is OK without.)

## Putting it all together

Now you have  $L$  and  $U$ .

- $\implies$  can solve for  $\mathbf{y}$  using forward substitution.
- $\implies$  can solve for  $\mathbf{x}$  using back-substitution.
- $\rightarrow$  linear system is solved.

You can get  $\det(A)$  just by multiplying diagonals of  $U$

$$\det(A) = \prod_{i=1}^N u_{ii} \quad (21)$$

... and you can get the  $j$ th column of  $A^{-1}$  just by solving for  $\mathbf{x}$  with  $\mathbf{b}$  set to the unit vector  $\mathbf{b} = \mathbf{e}_j$ .

# Housekeeping

- Problem Sheet for Lecture 3 available now
  - Manipulate a few matrices
  - Implement Gauss-Jordan Elimination (+ its variant Gaussian Elimination)
- Later you will get the Assignment
  - Q2 covers LU decomposition

- Next lecture: Interpolation



## Bonus: Coding search tips

- Try googling errors directly, minus any project-specific output
- <https://stackoverflow.com/>
- Know whether you are using Python 2.7 or 3.x  
<https://docs.python.org/3/>  
<https://docs.python.org/2/>
- **Read the documentation of the command carefully**
- Read the documentation of anything you don't understand in the documentation of the command carefully
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- Read the... (just read a lot)
- Copy the documented example into your session. Run it. Keep changing it to be more like your code until it breaks. That's your error.