Numerical Methods. Subjects and goals:

Problems:

- Solving systems of linear equations
- Solving systems of non-linear equations
- Numerical integration and integral equations
- Numerical solution of ordinary differential equations
- Numerical solution of partial differential equations

For each problem, we will discuss

- examples of applications in which the problem occur
- outline the most important numerical solution methods including when to apply which method
- pitfalls in numerical solutions

System of linear equations:

$$a_{00}x_0 + a_{01}x_1 + a_{02}x_2 + \dots + a_{0,N-1}x_{N-1} = b_0$$

$$a_{10}x_0 + a_{11}x_1 + a_{12}x_2 + \dots + a_{1,N-1}x_{N-1} = b_1$$

$$a_{20}x_0 + a_{21}x_1 + a_{22}x_2 + \dots + a_{2,N-1}x_{N-1} = b_2$$

$$\dots$$

$$a_{M-1,0}x_0 + a_{M-1,1}x_1 + \dots + a_{M-1,N-1}x_{N-1} = b_{M-1}$$

Here the N unknowns x_j , $j=0,1,\ldots,N-1$ are related by M equations. The coefficients a_{ij} with $i=0,1,\ldots,M-1$ and $j=0,1,\ldots,N-1$ are known numbers, as are the *right-hand side* quantities b_i , $i=0,1,\ldots,M-1$.

Matrix-vector notation:

$$a_{00}x_0 + a_{01}x_1 + a_{02}x_2 + \dots + a_{0,N-1}x_{N-1} = b_0$$

$$a_{10}x_0 + a_{11}x_1 + a_{12}x_2 + \dots + a_{1,N-1}x_{N-1} = b_1$$

$$a_{20}x_0 + a_{21}x_1 + a_{22}x_2 + \dots + a_{2,N-1}x_{N-1} = b_2$$

$$\dots$$

$$a_{M-1,0}x_0 + a_{M-1,1}x_1 + \dots + a_{M-1,N-1}x_{N-1} = b_{M-1}$$

$$\mathbf{A} = \begin{bmatrix} a_{00} & a_{01} & \dots & a_{0,N-1} \\ a_{10} & a_{11} & \dots & a_{1,N-1} \\ & \dots & & & \\ a_{M-1,0} & a_{M-1,1} & \dots & a_{M-1,N-1} \end{bmatrix} \qquad \mathbf{b} = \begin{bmatrix} b_0 \\ b_1 \\ \dots \\ b_{M-1} \end{bmatrix}$$

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

We start with: N = M

General Numerical Recipes notation:

$$\mathbf{C} = \mathbf{A} \cdot \mathbf{B} \iff c_{ik} = \sum_{j} a_{ij} b_{jk}$$

$$\mathbf{b} = \mathbf{A} \cdot \mathbf{x} \iff b_i = \sum_{j} a_{ij} x_j$$

$$\mathbf{d} = \mathbf{x} \cdot \mathbf{A} \iff d_j = \sum_{i} x_i a_{ij}$$

$$q = \mathbf{x} \cdot \mathbf{y} \iff q = \sum_{i} x_i y_i$$

Boldface capital letters are matrices, boldface small letters are vectors, non-boldface small letters are scalars

LU decomposition:

Suppose we are able to write the matrix **A** as a product of two matrices,

$$\mathbf{L} \cdot \mathbf{U} = \mathbf{A} \tag{2.3.1}$$

where L is *lower triangular* (has elements only on the diagonal and below) and U is *upper triangular* (has elements only on the diagonal and above). For the case of a 4×4 matrix A, for example, equation (2.3.1) would look like this:

$$\begin{bmatrix} \alpha_{00} & 0 & 0 & 0 \\ \alpha_{10} & \alpha_{11} & 0 & 0 \\ \alpha_{20} & \alpha_{21} & \alpha_{22} & 0 \\ \alpha_{30} & \alpha_{31} & \alpha_{32} & \alpha_{33} \end{bmatrix} \cdot \begin{bmatrix} \beta_{00} & \beta_{01} & \beta_{02} & \beta_{03} \\ 0 & \beta_{11} & \beta_{12} & \beta_{13} \\ 0 & 0 & \beta_{22} & \beta_{23} \\ 0 & 0 & 0 & \beta_{33} \end{bmatrix} = \begin{bmatrix} a_{00} & a_{01} & a_{02} & a_{03} \\ a_{10} & a_{11} & a_{12} & a_{13} \\ a_{20} & a_{21} & a_{22} & a_{23} \\ a_{30} & a_{31} & a_{32} & a_{33} \end{bmatrix}$$

$$\mathbf{A} \cdot \mathbf{x} = (\mathbf{L} \cdot \mathbf{U}) \cdot \mathbf{x} = \mathbf{L} \cdot (\mathbf{U} \cdot \mathbf{x}) = \mathbf{b}$$

$$\begin{bmatrix} \alpha_{00} & 0 & 0 & 0 \\ \alpha_{10} & \alpha_{11} & 0 & 0 \\ \alpha_{20} & \alpha_{21} & \alpha_{22} & 0 \\ \alpha_{30} & \alpha_{31} & \alpha_{32} & \alpha_{33} \end{bmatrix} \cdot \begin{bmatrix} \beta_{00} & \beta_{01} & \beta_{02} & \beta_{03} \\ 0 & \beta_{11} & \beta_{12} & \beta_{13} \\ 0 & 0 & \beta_{22} & \beta_{23} \\ 0 & 0 & 0 & \beta_{33} \end{bmatrix} = \begin{bmatrix} a_{00} & a_{01} & a_{02} & a_{03} \\ a_{10} & a_{11} & a_{12} & a_{13} \\ a_{20} & a_{21} & a_{22} & a_{23} \\ a_{30} & a_{31} & a_{32} & a_{33} \end{bmatrix}$$
(N=4)

Solve $L \cdot y = b$ (forward substitution):

$$y_0 = \frac{b_0}{\alpha_{00}}$$

$$y_i = \frac{1}{\alpha_{ii}} \left[b_i - \sum_{j=0}^{i-1} \alpha_{ij} y_j \right] \qquad i = 1, 2, \dots, N-1$$

$$\begin{bmatrix} \alpha_{00} & 0 & 0 & 0 \\ \alpha_{10} & \alpha_{11} & 0 & 0 \\ \alpha_{20} & \alpha_{21} & \alpha_{22} & 0 \\ \alpha_{30} & \alpha_{31} & \alpha_{32} & \alpha_{33} \end{bmatrix} \cdot \begin{bmatrix} \beta_{00} & \beta_{01} & \beta_{02} & \beta_{03} \\ 0 & \beta_{11} & \beta_{12} & \beta_{13} \\ 0 & 0 & \beta_{22} & \beta_{23} \\ 0 & 0 & 0 & \beta_{33} \end{bmatrix} = \begin{bmatrix} a_{00} & a_{01} & a_{02} & a_{03} \\ a_{10} & a_{11} & a_{12} & a_{13} \\ a_{20} & a_{21} & a_{22} & a_{23} \\ a_{30} & a_{31} & a_{32} & a_{33} \end{bmatrix}$$
(N=4)

Solve $\mathbf{L} \cdot \mathbf{y} = \mathbf{b}$ (forward substitution):

$$y_{0} = \frac{b_{0}}{a_{i}}$$

$$y_{i} = \sum_{\alpha_{i}}^{1} \left[b_{i} - \sum_{j=0}^{i-1} \alpha_{ij} y_{j} \right] \qquad i = 1, 2, \dots, N-1$$

Solve $U \cdot x = y$ (back substitution):

$$x_{N-1} = \frac{y_{N-1}}{\beta_{N-1,N-1}}$$

$$x_i = \frac{1}{\beta_{ii}} \left[y_i - \sum_{j=i+1}^{N-1} \beta_{ij} x_j \right] \qquad i = N-2, N-3, \dots, 0$$

$$\alpha_{ii} \equiv 1$$
 $i = 0, \dots, N-1$

Solve for L and U:

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ \alpha_{10} & 1 & 0 & 0 \\ \alpha_{20} & \alpha_{21} & 1 & 0 \\ \alpha_{30} & \alpha_{31} & \alpha_{32} & 1 \end{bmatrix} \cdot \begin{bmatrix} \beta_{00} & \beta_{01} & \beta_{02} & \beta_{03} \\ 0 & \beta_{11} & \beta_{12} & \beta_{13} \\ 0 & 0 & \beta_{22} & \beta_{23} \\ 0 & 0 & 0 & \beta_{33} \end{bmatrix} = \begin{bmatrix} a_{00} & a_{01} & a_{02} & a_{03} \\ a_{10} & a_{11} & a_{12} & a_{13} \\ a_{20} & a_{21} & a_{22} & a_{23} \\ a_{30} & a_{31} & a_{32} & a_{33} \end{bmatrix}$$

$$\beta_{00} = a_{00}$$

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ \alpha_{10} & 1 & 0 & 0 \\ \alpha_{20} & \alpha_{21} & 1 & 0 \\ \alpha_{30} & \alpha_{31} & \alpha_{32} & 1 \end{bmatrix} \cdot \begin{bmatrix} \beta_{00} & \beta_{01} & \beta_{02} & \beta_{03} \\ 0 & \beta_{11} & \beta_{12} & \beta_{13} \\ 0 & 0 & \beta_{22} & \beta_{23} \\ 0 & 0 & 0 & \beta_{33} \end{bmatrix} = \begin{bmatrix} a_{00} & a_{01} & a_{02} & a_{03} \\ a_{10} & a_{11} & a_{12} & a_{13} \\ a_{20} & a_{21} & a_{22} & a_{23} \\ a_{30} & a_{31} & a_{32} & a_{33} \end{bmatrix}$$

$$\alpha_{10} = \frac{a_{10}}{\beta_{00}}$$

$$\begin{bmatrix} \alpha_{00} & 0 & 0 & 0 \\ \alpha_{10} & \alpha_{11} & 0 & 0 \\ \alpha_{20} & \alpha_{21} & \alpha_{22} & 0 \\ \alpha_{30} & \alpha_{31} & \alpha_{32} & \alpha_{33} \end{bmatrix} \cdot \begin{bmatrix} \beta_{00} & \beta_{01} & \beta_{02} & \beta_{03} \\ 0 & \beta_{11} & \beta_{12} & \beta_{13} \\ 0 & 0 & \beta_{22} & \beta_{23} \\ 0 & 0 & 0 & \beta_{33} \end{bmatrix} = \begin{bmatrix} a_{00} & a_{01} & a_{02} & a_{03} \\ a_{10} & a_{11} & a_{12} & a_{13} \\ a_{20} & a_{21} & a_{22} & a_{23} \\ a_{30} & a_{31} & a_{32} & a_{33} \end{bmatrix}$$

$$\alpha_{20} = \frac{a_{20}}{\beta_{00}}$$

$$\begin{bmatrix} \alpha_{00} & 0 & 0 & 0 \\ \alpha_{10} & \alpha_{11} & 0 & 0 \\ \alpha_{20} & \alpha_{21} & \alpha_{22} & 0 \\ \alpha_{30} & \alpha_{31} & \alpha_{32} & \alpha_{33} \end{bmatrix} \cdot \begin{bmatrix} \beta_{00} & \beta_{01} & \beta_{02} & \beta_{03} \\ 0 & 0 & \beta_{11} & \beta_{12} & \beta_{13} \\ 0 & 0 & \beta_{22} & \beta_{23} \\ 0 & 0 & 0 & \beta_{33} \end{bmatrix} = \begin{bmatrix} a_{00} & a_{01} & a_{02} & a_{03} \\ a_{10} & a_{11} & a_{12} & a_{13} \\ a_{20} & a_{21} & a_{22} & a_{23} \\ a_{30} & a_{31} & a_{32} & a_{33} \end{bmatrix}$$

$$\alpha_{30} = \frac{a_{30}}{\beta_{00}}$$

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ \alpha_{10} & 1 & 0 & 0 \\ \alpha_{20} & \alpha_{21} & 1 & 0 \\ \alpha_{30} & \alpha_{31} & \alpha_{32} & 1 \end{bmatrix} \cdot \begin{bmatrix} \beta_{00} & \beta_{01} & \beta_{02} & \beta_{03} \\ 0 & \beta_{11} & \beta_{12} & \beta_{13} \\ 0 & 0 & \beta_{22} & \beta_{23} \\ 0 & 0 & 0 & \beta_{33} \end{bmatrix} = \begin{bmatrix} a_{00} & a_{01} & a_{02} & a_{03} \\ a_{10} & a_{11} & a_{12} & a_{13} \\ a_{20} & a_{21} & a_{22} & a_{23} \\ a_{30} & a_{31} & a_{32} & a_{33} \end{bmatrix}$$

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ \alpha_{10} & 1 & 0 & 0 \\ \alpha_{20} & \alpha_{21} & 1 & 0 \\ \alpha_{30} & \alpha_{31} & \alpha_{32} & 1 \end{bmatrix} \cdot \begin{bmatrix} \beta_{00} & \beta_{01} & \beta_{02} & \beta_{03} \\ 0 & \beta_{11} & \beta_{12} & \beta_{13} \\ 0 & 0 & \beta_{22} & \beta_{23} \\ 0 & 0 & \beta_{33} \end{bmatrix} = \begin{bmatrix} a_{00} & a_{01} & a_{02} & a_{03} \\ a_{10} & a_{11} & a_{12} & a_{13} \\ a_{20} & a_{21} & a_{22} & a_{23} \\ a_{30} & a_{31} & a_{32} & a_{33} \end{bmatrix}$$

$$\beta_{11} = a_{11} - \alpha_{10}\beta_{01}$$

$$\beta_{11} = a_{11} - \alpha_{10}\beta_{01}$$

 $\beta_{01} = a_{01}$

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ \alpha_{10} & 1 & 0 & 0 \\ \alpha_{20} & \alpha_{21} & 1 & 0 \\ \alpha_{30} & \alpha_{31} & \alpha_{32} & 1 \end{bmatrix} \cdot \begin{bmatrix} \beta_{00} & \beta_{01} & \beta_{02} & \beta_{03} \\ 0 & \beta_{11} & \beta_{12} & \beta_{13} \\ 0 & 0 & \beta_{22} & \beta_{23} \\ 0 & 0 & 0 & \beta_{33} \end{bmatrix} = \begin{bmatrix} a_{00} & a_{01} & a_{02} & a_{03} \\ a_{10} & a_{11} & a_{12} & a_{13} \\ a_{20} & a_{21} & a_{22} & a_{23} \\ a_{30} & a_{31} & a_{32} & a_{33} \end{bmatrix}$$

$$\alpha_{21} = \frac{a_{21} - \alpha_{20}\beta_{01}}{\beta_{11}}$$

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ \alpha_{10} & 1 & 0 & 0 \\ \alpha_{20} & \alpha_{21} & 1 & 0 \\ \alpha_{30} & \alpha_{31} & \alpha_{32} & 1 \end{bmatrix} \cdot \begin{bmatrix} \beta_{00} & \beta_{01} & \beta_{02} & \beta_{03} \\ 0 & \beta_{11} & \beta_{12} & \beta_{13} \\ 0 & 0 & \beta_{22} & \beta_{23} \\ 0 & 0 & \beta_{33} \end{bmatrix} = \begin{bmatrix} a_{00} & a_{01} & a_{02} & a_{03} \\ a_{10} & a_{11} & a_{12} & a_{13} \\ a_{20} & a_{21} & a_{22} & a_{23} \\ a_{30} & a_{31} & a_{32} & a_{33} \end{bmatrix}$$

$$\alpha_{31} = \frac{a_{31} - \alpha_{30} \beta_{01}}{\beta_{11}}$$

For each j = 0, 1, 2, ..., N - 1 do these two procedures:

for
$$i = 0, 1, ..., j$$

$$\beta_{ij} = a_{ij} - \sum_{k=0}^{l-1} \alpha_{ik} \beta_{kj}$$

for
$$i = j + 1, j + 2, ..., N - 1$$

$$\alpha_{ij} = \frac{1}{\beta_{jj}} \left(a_{ij} - \sum_{k=0}^{j-1} \alpha_{ik} \beta_{kj} \right)$$

What if $\beta_{jj} = 0$? (also a problem for a "small" β_{jj}

Pivoting (row swaps) like in Gaussian elimination. Terrible to bookkeep, but implemented in the Numerical Recipes routines Ludcmp.

Computational complexity (big O() notation)

LU decomposition:

For each j = 0, 1, 2, ..., N - 1 do these two procedures:

for
$$i = 0, 1, ..., j$$

$$\beta_{ij} = a_{ij} - \sum_{k=0}^{i-1} \alpha_{ik} \beta_{kj}$$
 for $i = j + 1, j + 2, ..., N - 1$
$$\alpha_{ij} = \frac{1}{\beta_{jj}} \left(a_{ij} - \sum_{k=0}^{j-1} \alpha_{ik} \beta_{kj} \right)$$

 $O(N_3)$

Forward substitution:

$$y_{0} = \frac{b_{0}}{\omega_{0}}$$

$$y_{i} = \sum_{j=0}^{i-1} a_{ij} y_{j}$$

$$i = 1, 2, ..., N-1$$

$$i = 1, 2, ..., N-1$$

Back substitution:

$$x_{N-1} = \frac{y_{N-1}}{\beta_{N-1,N-1}}$$

$$x_i = \frac{1}{\beta_{ii}} \left[y_i - \sum_{j=i+1}^{N-1} \beta_{ij} x_j \right] \qquad i = N-2, N-3, \dots, 0$$
O(N²)

Advantage of LU decomposition when comparing to Gaussian elimination?

Assume we have to solve a system of linear equations with the **same matrix** and a **new right hand side.**

Then we only need to do forward and back substitutions (assuming we stored L and U). Hence $O(N^2)$ rather than $O(N^3)$.

Exercises

- 1) Consider an NxN matrix A where we have computed L and U. How can we then easily compute the determinant of A?
- 2) Assume that we have 2 NxN matrices A and B, and we have LU-decompositions A=LU and B=L'U'. Let now C=AB. How can we solve Cx=b and what is the resulting computational complexity?
- 3) Compute by hand the LU decomposition of

$$A = \left(\begin{array}{c} 1 & 3 & 5 \\ -2 & 0 & -1 \\ 2 & 3 & 1 \end{array}\right)$$