

Lecture 18

$Bx = b$, B is non-singular

$$Bx = L \overset{y}{\underbrace{Ux}} = b$$

upper triangular matrix
lower triangular matrix

- 1. solve $Ly = b$
2. After getting y ,
 $Ux = y$

Back to the example! (lect. 17)

2. Solve $Ux = y$

example from
the book!

$$\begin{pmatrix} 2 & 0 & 4 & 0 & -2 \\ 0 & 1 & -6 & 1 & 3 \\ 0 & 0 & 1 & 0 & -3 \\ 0 & 0 & 0 & 1 & -21 \\ 0 & 0 & 0 & 0 & 7 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix} = \begin{pmatrix} 7 \\ -25/2 \\ 7/2 \\ 23/2 \\ -7/2 \end{pmatrix}$$

Now start with the last line:

$$7x_5 = -\frac{7}{2} \Rightarrow x_5 = -\frac{1}{2}$$

$$x_4 - 21\left(-\frac{1}{2}\right) = \frac{23}{2} \Rightarrow x_4 = 1$$

\vdots

Solution $x^* = (-1 \ 0 \ 2 \ 1 \ -\frac{1}{2})$

- Using LU factorisation of B implies the solution of two "easier" systems $Ly = b$, $Ux = y$ instead of solving $Bx = b$ (since L and U are triangular).
- However the difficulty, efficiency of both approaches is similar
- We need perturbations of rows and/or columns in order to obtain non-zero diagonal elements. Since B is non-singular, each row/column contains at least one non-zero element.

Obviously, the more zero-elements matrix B has, the faster the solution is obtained.

Def. A matrix that contains zero-elements is called a **sparse** matrix.

Trick! Apply further row/column permutations aiming at keeping L and U as sparse as possible.

Problem! How to find the best such permutation is a harder problem than solving the original LP (makes no sense to find it deterministically).

Use a heuristic, that is, a result/converg. cannot be proved, but works well in practice.

We will look at the "minimum degree ordering heuristic".

The effect of making L and U more sparse (L and U contains much more zero-elements).

This can have a dramatic effect for high dimensions.

Approximately: the cost of solving our original problem depends cubically on the dimension of B for Gaussian elimination or LU -factorization.

◦ But only linear for sufficiently sparse LU -matrices.

Idea for minimum degree
ordering heuristic (MDOH).

- o Choose in each step of Gaussian elimination/LU-factorisation within the uneliminated part of B the sparsest row (with the smallest number of non-zero elements) and
- o From the non-zero elements of this row, the one whose column is the sparsest one.
- o Then apply a permutation such that the chosen row/column define a diagonal element.

Example

Apply MDOH heuristic to B.

$$B = \begin{array}{ccccc|c} & 1 & 2 & 3 & 4 & 5 & \\ \begin{array}{c} 1 \\ 2 \\ 3 \\ 4 \\ 5 \end{array} & \begin{pmatrix} 2 & 0 & 4 & 0 & -2 \\ 3 & 1 & 0 & 1 & 0 \\ -1 & 0 & -1 & 0 & -2 \\ 0 & -1 & 0 & 0 & -6 \\ 0 & 0 & 1 & 0 & 4 \end{pmatrix} \end{array}$$

- o Row 4 is the sparsest row
- o Check columns 2 and 5 (non-zero elements). Sparsest column is 2.
- o Permutate rows 4 and 1 and columns 2 and 1 s.t. -1 becomes

the $(1,1)$ -diagonal element

$$\begin{array}{c} 4 \\ 2 \\ 3 \\ 1 \\ 5 \end{array} \begin{array}{c} 2 \\ 1 \\ 3 \\ 4 \\ 5 \end{array} \left(\begin{array}{ccccc|c} -1 & 0 & 0 & 0 & 0 & -6 \\ \textcircled{1} & 3 & 0 & 1 & 0 & 0 \\ 0 & -1 & -1 & 0 & 0 & -2 \\ 0 & 2 & 4 & 0 & 0 & -2 \\ 0 & 0 & 1 & 0 & 0 & 4 \end{array} \right)$$

Eliminate $\textcircled{1}$ by adding $2 + 4$
and leave the $\textcircled{1}$ in its place.

Gauss elimination leads to :

$$\begin{array}{c}
 4 \\
 2 \\
 3 \\
 1 \\
 5
 \end{array}
 \begin{pmatrix}
 & 2 & 1 & 3 & 4 & 5 \\
 -1 & 0 & 0 & 0 & -6 \\
 1 & 3 & 0 & 1 & -6 \\
 0 & -1 & -1 & 0 & -2 \\
 0 & 2 & 4 & 0 & -2 \\
 0 & 0 & 1 & 0 & 4
 \end{pmatrix}$$

Uneliminated
part

Notation: the # of non-zero elem. of a row or column in the uneliminated part is called the **degree** of a row/column.

o Row 5 has degree 2 (min.) in the uneliminated part (red square).

• We swap 5 and 2. Check columns 3 and 5. Column 3 has min. degree.

$$\begin{array}{c}
 4 \\
 5 \\
 3 \\
 1 \\
 2
 \end{array}
 \begin{array}{c}
 2 \\
 3 \\
 1 \\
 4 \\
 5
 \end{array}
 \left(\begin{array}{ccccc}
 -1 & 0 & 0 & 0 & -6 \\
 0 & 1 & 0 & 0 & 4 \\
 0 & -1 & -1 & 0 & -2 \\
 0 & 4 & 2 & 0 & -2 \\
 1 & 0 & 3 & 1 & -6
 \end{array} \right)$$

Gauss \Rightarrow

$$\begin{array}{c}
 4 \\
 5 \\
 3 \\
 1 \\
 2
 \end{array}
 \begin{array}{c}
 2 \\
 3 \\
 1 \\
 4 \\
 5
 \end{array}
 \left(\begin{array}{ccccc}
 -1 & 0 & 0 & 0 & -6 \\
 0 & 1 & 0 & 0 & 4 \\
 0 & -1 & -1 & 0 & 2 \\
 0 & 4 & 2 & 0 & -18 \\
 1 & 0 & 3 & 1 & -6
 \end{array} \right)$$

uneliminated part.

Row 3 has min. degree. Column 1 has min. degree. Since there are no zeros there are no necessary permutations.

Gauss \Rightarrow

$$\begin{array}{c}
 4 \\
 5 \\
 3 \\
 1 \\
 2
 \end{array}
 \begin{array}{c}
 2 \\
 3 \\
 1 \\
 4 \\
 5
 \end{array}
 \begin{pmatrix}
 -1 & 0 & 0 & 0 & -6 \\
 0 & 1 & 0 & 0 & 4 \\
 0 & -1 & -1 & 0 & 2 \\
 0 & 4 & 2 & 0 & -14 \\
 1 & 0 & 3 & 1 & 0
 \end{pmatrix}$$

Uneliminated part.

Row 1 has min. degree. Swap 4 and 5.

$$\Rightarrow \begin{matrix} & \overset{2}{} & \overset{3}{} & \overset{1}{} & \overset{5}{} & \overset{4}{} \\ \overset{4}{} & -1 & 0 & 0 & 0 & -6 \\ \overset{5}{} & 0 & 1 & 0 & 0 & 4 \\ \overset{3}{} & 0 & -1 & -1 & 0 & 2 \\ \overset{1}{} & 0 & 4 & 2 & -14 & 0 \\ \overset{2}{} & 1 & 0 & 3 & 0 & 1 \end{matrix}$$

Elimination is now complete since we have now non-zero elements in the diagonal.

For L and U we get the following:

$$L = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & \underline{-1} & 1 & 0 & 0 \\ 0 & \underline{4} & \underline{-2} & 1 & 0 \\ \underline{-1} & 0 & \underline{-3} & 0 & 1 \end{pmatrix}$$

$$U = \begin{pmatrix} -1 & 0 & 0 & \underline{-6} & 0 \\ 0 & 1 & 0 & \underline{4} & 0 \\ 0 & 0 & -1 & \underline{2} & 0 \\ 0 & 0 & 0 & -14 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

The number of non-zero non-diagonal elements in L and U :

$$5 + 3 = \underline{8}$$

We had 12 non-zero non-diagonal elements in L and U before our heuristic method so this is an improvement.

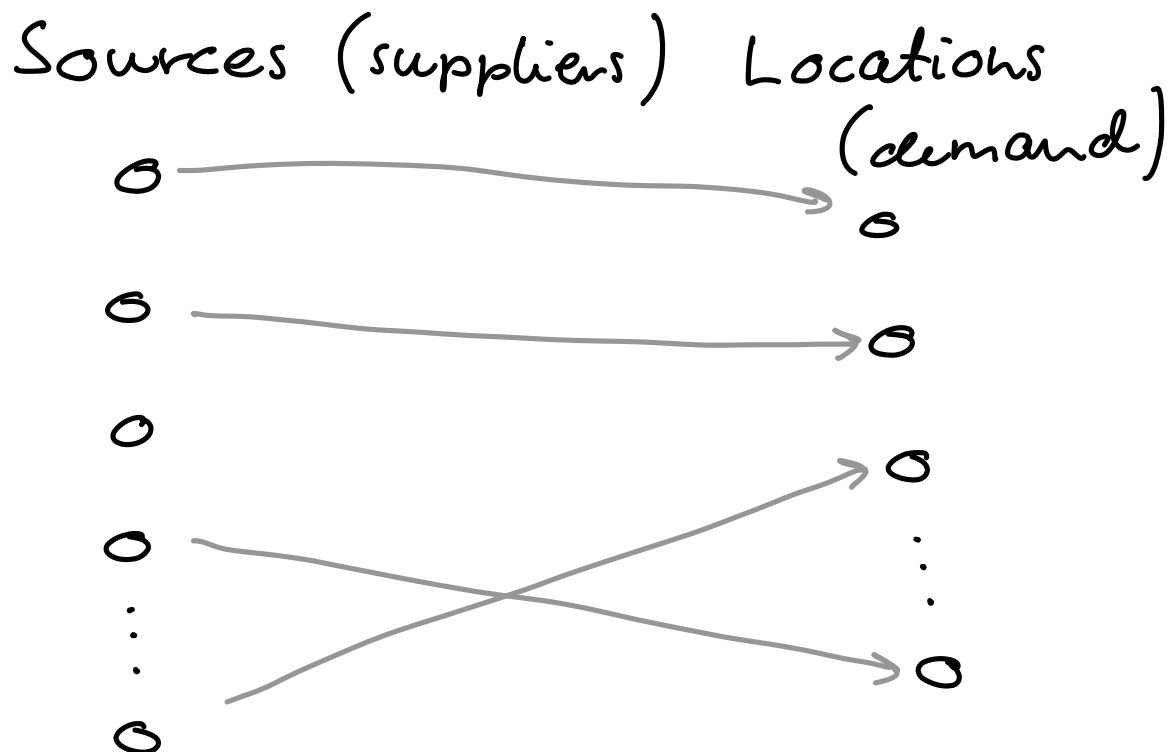
↑ This can have a dramatic effect on large problems in terms of efficiency!!

Note that a heuristic will not always provide a better result. It might sometimes even be worse.

Chapter 14 Network flow problems

We consider an LP problem with a particular structure.

General situation of a network:



E.g. transportation, electricity,
communication network etc.

Possible objectives: min. of costs,
time etc.

Here: min. cost network flow problem.