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Numerical Analysis Hilary Term 2018  
Lecture 4: Gaussian Elimination

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**Setup:** given a square  $n$  by  $n$  matrix  $A$  and vector with  $n$  components  $b$ , find  $x$  such that

$$Ax = b.$$

Equivalently find  $x = (x_1, x_2, \dots, x_n)^T$  for which

$$\begin{aligned} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n &= b_1 \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n &= b_2 \\ &\vdots \\ a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nn}x_n &= b_n. \end{aligned} \tag{1}$$

**Lower-triangular matrices:** the matrix  $A$  is **lower triangular** if  $a_{ij} = 0$  for all  $1 \leq i < j \leq n$ . The system (1) is easy to solve if  $A$  is lower triangular.

$$\begin{array}{rcll} a_{11}x_1 & = b_1 & \implies & x_1 = \frac{b_1}{a_{11}} & \Downarrow \\ a_{21}x_1 + a_{22}x_2 & = b_2 & \implies & x_2 = \frac{b_2 - a_{21}x_1}{a_{22}} & \Downarrow \\ \vdots & & & & \Downarrow \\ & & & & b_i - \sum_{j=1}^{i-1} a_{ij}x_j \\ a_{i1}x_1 + a_{i2}x_2 + \dots + a_{ii}x_i & = b_i & \implies & x_i = \frac{b_i - \sum_{j=1}^{i-1} a_{ij}x_j}{a_{ii}} & \Downarrow \\ \vdots & & & & \Downarrow \end{array}$$

This works if, and only if,  $a_{ii} \neq 0$  for each  $i$ . The procedure is known as **forward substitution**.

**Computational work estimate:** one floating-point operation (flop) is one scalar multiply/division/addition/subtraction as in  $y = a * x$  where  $a$ ,  $x$  and  $y$  are computer representations of real scalars.<sup>1</sup>

Hence the work in forward substitution is 1 flop to compute  $x_1$  plus 3 flops to compute  $x_2$  plus ... plus  $2i - 1$  flops to compute  $x_i$  plus ... plus  $2n - 1$  flops to compute  $x_n$ , or in total

$$\sum_{i=1}^n (2i - 1) = 2 \left( \sum_{i=1}^n i \right) - n = 2 \left( \frac{1}{2}n(n + 1) \right) - n = n^2 + \text{lower order terms}$$

flops. We sometimes write this as  $n^2 + O(n)$  flops or more crudely  $O(n^2)$  flops.

**Upper-triangular matrices:** the matrix  $A$  is **upper triangular** if  $a_{ij} = 0$  for all  $1 \leq j < i \leq n$ . Once again, the system (1) is easy to solve if  $A$  is upper triangular.

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<sup>1</sup>This is an abstraction: e.g., some hardware can do  $y = a * x + b$  in one FMA flop ("Fused Multiply and Add") but then needs several FMA flops for a single division. For a trip down this sort of rabbit hole, look up the "Fast inverse square root" as used in the source code of the video game "Quake III Arena".

$$\begin{array}{rcccl}
& & \vdots & & \uparrow \\
a_{ii}x_i + \cdots + a_{in-1}x_{n-1} + a_{1n}x_n & = & b_i & \implies & x_i = \frac{b_i - \sum_{j=i+1}^n a_{ij}x_j}{a_{ii}} & \uparrow \\
& & \vdots & & \uparrow \\
a_{n-1n-1}x_{n-1} + a_{n-1n}x_n & = & b_{n-1} & \implies & x_{n-1} = \frac{b_{n-1} - a_{n-1n}x_n}{a_{n-1n-1}} & \uparrow \\
& & & & & \uparrow \\
a_{nn}x_n & = & b_n & \implies & x_n = \frac{b_n}{a_{nn}} & \uparrow
\end{array}$$

Again, this works if, and only if,  $a_{ii} \neq 0$  for each  $i$ . The procedure is known as **backward** or **back substitution**. This also takes approximately  $n^2$  flops.

For computation, we need a reliable, systematic technique for reducing  $Ax = b$  to  $Ux = c$  with the same solution  $x$  but with  $U$  (upper) triangular  $\implies$  Gauss elimination.

**Example**

$$\begin{bmatrix} 3 & -1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 12 \\ 11 \end{bmatrix}.$$

Multiply first equation by  $1/3$  and subtract from the second  $\implies$

$$\begin{bmatrix} 3 & -1 \\ 0 & \frac{7}{3} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 12 \\ 7 \end{bmatrix}.$$

**Gauss(ian) Elimination (GE):** this is most easily described in terms of overwriting the matrix  $A = \{a_{ij}\}$  and vector  $b$ . At each stage, it is a systematic way of introducing zeros into the lower triangular part of  $A$  by subtracting multiples of previous equations (i.e., rows); such (elementary row) operations do not change the solution.

for columns  $j = 1, 2, \dots, n - 1$   
 for rows  $i = j + 1, j + 2, \dots, n$

$$\begin{aligned} \text{row } i &\leftarrow \text{row } i - \frac{a_{ij}}{a_{jj}} * \text{row } j \\ b_i &\leftarrow b_i - \frac{a_{ij}}{a_{jj}} * b_j \end{aligned}$$

end  
 end

**Example.**

$$\begin{bmatrix} 3 & -1 & 2 \\ 1 & 2 & 3 \\ 2 & -2 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 12 \\ 11 \\ 2 \end{bmatrix} : \text{ represent as } \left[ \begin{array}{ccc|c} 3 & -1 & 2 & 12 \\ 1 & 2 & 3 & 11 \\ 2 & -2 & -1 & 2 \end{array} \right]$$

$$\Rightarrow \begin{aligned} \text{row } 2 &\leftarrow \text{row } 2 - \frac{1}{3}\text{row } 1 \\ \text{row } 3 &\leftarrow \text{row } 3 - \frac{2}{3}\text{row } 1 \end{aligned} \left[ \begin{array}{ccc|c} 3 & -1 & 2 & 12 \\ 0 & \frac{7}{3} & \frac{7}{3} & 7 \\ 0 & -\frac{4}{3} & -\frac{7}{3} & -6 \end{array} \right]$$

$$\Rightarrow \begin{aligned} \text{row } 3 &\leftarrow \text{row } 3 + \frac{4}{7}\text{row } 2 \end{aligned} \left[ \begin{array}{ccc|c} 3 & -1 & 2 & 12 \\ 0 & \frac{7}{3} & \frac{7}{3} & 7 \\ 0 & 0 & -1 & -2 \end{array} \right]$$

Back substitution:

$$\begin{aligned} x_3 &= 2 \\ x_2 &= \frac{7 - \frac{7}{3}(2)}{\frac{7}{3}} = 1 \\ x_1 &= \frac{12 - (-1)(1) - 2(2)}{3} = 3. \end{aligned}$$

**Cost of Gaussian Elimination:** note,  $\text{row } i \leftarrow \text{row } i - \frac{a_{ij}}{a_{jj}} * \text{row } j$  is  
 for columns  $k = j + 1, j + 2, \dots, n$

$$a_{ik} \leftarrow a_{ik} - \frac{a_{ij}}{a_{jj}} a_{jk}$$

end

This is approximately  $2(n - j)$  flops as the **multiplier**  $a_{ij}/a_{jj}$  is calculated with just one flop;  $a_{jj}$  is called the **pivot**. Overall therefore, the cost of GE is approximately

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

flops. The calculations involving  $b$  are

$$\sum_{j=1}^{n-1} 2(n - j) = 2 \sum_{l=1}^{n-1} l = 2 \frac{n(n - 1)}{2} = n^2 + O(n)$$

flops, just as for the triangular substitution.