

# Intro to Analysis of Algorithms

## Linear Algebra / Parallel

### Chapter 7

Michael Soltys

CSU Channel Islands

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# Row-echelon form

$$\begin{bmatrix} 1 & * \dots * & * & * \dots * & * & * \dots * & * \\ & & 1 & * \dots * & * & * \dots * & * \\ & & & & & 1 & * \dots * & * \\ & & & 0 & & & & 1 & \dots \\ & & & & & & & & \vdots & \ddots \\ & & & & & & & & & \ddots \end{bmatrix}$$

# Elementary matrices

one of the following three forms:

$$I + aT_{ij} \quad i \neq j \quad \text{(elementary of type 1)}$$

$$I + T_{ij} + T_{ji} - T_{ii} - T_{jj} \quad \text{(elementary of type 2)}$$

$$I + (c - 1)T_{ii} \quad c \neq 0 \quad \text{(elementary of type 3)}$$

Gaussian Elimination is a divide and conquer algorithm, with a recursive call to smaller matrices.

If  $A$  is a  $1 \times m$  matrix,  $A = [a_{11} a_{12} \dots a_{1m}]$ , then:

$$GE(A) = \begin{cases} [1/a_{1i}] & \text{where } i = \min\{1, 2, \dots, m\} \text{ such that } a_{i1} \neq 0 \\ [1] & \text{if } a_{11} = a_{12} = \dots = a_{1m} = 0 \end{cases}$$

Suppose now that  $n > 1$ . If  $A = 0$ , let  $GE(A) = I$ . Otherwise, let:

$$GE(A) = \begin{bmatrix} 1 & 0 \\ 0 & GE((EA)[1|1]) \end{bmatrix} E$$

where  $E$  is a product of at most  $n + 1$  elementary matrices. Note that  $C[i|j]$  denotes the matrix  $C$  with row  $i$  and  $j$ .

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1: if  $n = 1$  then
2:     if  $a_{11} = a_{12} = \cdots = a_{1m} = 0$  then
3:         return  $[1]$ 
4:     else
5:         return  $[1/a_{1\ell}]$  where  $\ell = \min_{i \in [n]} \{a_{1i} \neq 0\}$ 
6:     end if
7: else
8:     if  $A = 0$  then
9:         return  $/$ 
10:    else
11:        if first column of  $A$  is zero then
12:            Compute  $E$  as in Case 1.
13:        else
14:            Compute  $E$  as in Case 2.
15:        end if
16:        return  $\begin{bmatrix} 1 & 0 \\ 0 & GE((EA)[1|1]) \end{bmatrix} E$ 
17:    end if
18: end if

```

# Gram-Schmidt

**Pre-condition:**  $\{v_1, \dots, v_n\}$  a basis for  $\mathbb{R}^n$

```
1:  $v_1^* \leftarrow v_1$ 
2: for  $i = 2, 3, \dots, n$  do
3:     for  $j = 1, 2, \dots, (i - 1)$  do
4:          $\mu_{ij} \leftarrow (v_i \cdot v_j^*) / \|v_j^*\|^2$ 
5:     end for
6:      $v_i^* \leftarrow v_i - \sum_{j=1}^{i-1} \mu_{ij} v_j^*$ 
7: end for
```

**Post-condition:**  $\{v_1^*, \dots, v_n^*\}$  an orthogonal basis for  $\mathbb{R}^n$

# Gauss lattice reduction

**Pre-condition:**  $\{v_1, v_2\}$  are linearly independent in  $\mathbb{R}^2$

```
1: loop
2:     if  $\|v_2\| < \|v_1\|$  then
3:         swap  $v_1$  and  $v_2$ 
4:     end if
5:      $m \leftarrow \lfloor v_1 \cdot v_2 / \|v_1\|^2 \rfloor$  (note that  $\lfloor x \rfloor = \lfloor x + 1/2 \rfloor$ )
6:     if  $m \neq 0$  then
7:         return  $v_1, v_2$ 
8:     else
9:          $v_2 \leftarrow v_2 - mv_1$ 
10:    end if
11: end loop
```

Given a matrix  $A$ , its *trace* is defined as the sum of the diagonal entries, i.e.,  $\text{tr}(A) = \sum_i a_{ii}$ . Using traces we can compute the *Newton's symmetric polynomials* which are defined as follows:  $s_0 = 1$ , and for  $1 \leq k \leq n$ , by:

$$s_k = \frac{1}{k} \sum_{i=1}^k (-1)^{i-1} s_{k-i} \text{tr}(A^i).$$

Then, it turns out that

$p_A(x) = s_0 x^n - s_1 x^{n-1} + s_2 x^{n-2} - \dots \pm s_n x^0$ , that is, Newton's symmetric polynomials compute the coefficients of the characteristic polynomial,  $p_A(x) = \det(xI - A)$ .



$$\begin{pmatrix} s_1 \\ s_2 \\ \vdots \\ s_n \end{pmatrix}, \quad \begin{pmatrix} 0 & 0 & 0 & \dots \\ \frac{1}{2}\text{tr}(A) & 0 & 0 & \dots \\ \frac{1}{3}\text{tr}(A^2) & \frac{1}{3}\text{tr}(A) & 0 & \dots \\ \frac{1}{4}\text{tr}(A^3) & \frac{1}{4}\text{tr}(A^2) & \frac{1}{4}\text{tr}(A) & \dots \\ \vdots & \vdots & \vdots & \ddots \end{pmatrix}, \quad \begin{pmatrix} \text{tr}(A) \\ \frac{1}{2}\text{tr}(A^2) \\ \vdots \\ \frac{1}{n}\text{tr}(A^n) \end{pmatrix}$$

Berkowitz's algorithm is also Divide and Conquer, and it computes the characteristic polynomial of  $A$  from the characteristic polynomial of its *principal minor*, i.e., the matrix  $M$  obtained from deleting the first row and column of  $A$ :

$$A = \begin{pmatrix} a_{11} & R \\ S & M \end{pmatrix},$$

$R$  is an  $1 \times (n - 1)$  row matrix and  $S$  is a  $(n - 1) \times 1$  column matrix and  $M$  is  $(n - 1) \times (n - 1)$ . Let  $p(x)$  and  $q(x)$  be the characteristic polynomials of  $A$  and  $M$  respectively. Suppose that the coefficients of  $p$  form the column vector:

$$p = \begin{pmatrix} p_n & p_{n-1} & \cdots & p_0 \end{pmatrix}^t,$$

where  $p_i$  is the coefficient of  $x^i$  in  $\det(xI - A)$ , and similarly for  $q$ . Then:

$$p = C_1 q,$$

where  $C_1$  is an  $(n + 1) \times n$  *Toeplitz* lower triangular matrix (Toeplitz means that the values on each diagonal are constant)

where the entries in the first column are defined as follows:  $c_{i1} = 1$  if  $i = 1$ ,  $c_{i1} = -a_{11}$  if  $i = 2$ , and  $c_{i1} = -(RM^{i-3}S)$  if  $i \geq 3$ . Berkowitz's algorithm consists in repeating this for  $q$ , and continuing so that  $p$  is expressed as a product of matrices. Thus:

$$p_A^{\text{BERK}} = C_1 C_2 \cdots C_n,$$

where  $C_i$  is an  $(n + 2 - i) \times (n + 1 - i)$  Toeplitz matrix defined as above except  $A$  is replaced by its  $i$ -th principal sub-matrix. Note that  $C_n = (1 \quad -a_{nn})^t$ .