

Modélisation et résolutions numérique et symbolique

via les logiciels MAPLE et MATLAB (MODEL)

Jérémy Berthomieu Mohab Safey El Din Stef Graillat

`jeremy.berthomieu@lip6.fr`



Roadmap

- ① Chinese Remainder Theorem in \mathbb{Z} .
- ② Lagrange Interpolation in $\mathbb{K}[X]$.
- ③ Comparison between CRT and Lagrange Interpolation.
- ④ Eigenvalues problem.

Chinese Remainder Theorem (I)

Theorem 1

If n_0, \dots, n_d are **pairwise coprime** integers and a_0, \dots, a_d are integers modulo n_0, \dots, n_d , then system $(S) = \begin{cases} x = a_0 \pmod{n_0} \\ \vdots \\ x = a_d \pmod{n_d} \end{cases}$ has solutions in \mathbb{Z} .

Chinese Remainder Theorem (I)

Theorem 1

If n_0, \dots, n_d are **pairwise coprime** integers and a_0, \dots, a_d are integers modulo n_0, \dots, n_d , then system $(S) = \begin{cases} x = a_0 \pmod{n_0} \\ \vdots \\ x = a_d \pmod{n_d} \end{cases}$ has solutions in \mathbb{Z} .

If x_0 is a solution, then the other ones are $x = x_0 + kn_0 \cdots n_d$ for any $k \in \mathbb{Z}$.

Chinese Remainder Theorem (II)

Solution

Let v_0, \dots, v_d be such that $v_i n_0 \cdots n_{i-1} n_{i+1} \cdots n_d = 1 \pmod{n_i}$, i.e.

$$v_i \prod_{\substack{0 \leq j \leq d \\ j \neq i}} n_j = 1 \pmod{n_i},$$

then

$$x_0 = \sum_{0 \leq i \leq d} a_i v_i \prod_{\substack{0 \leq j \leq d \\ j \neq i}} n_j.$$

Chinese Remainder Theorem (III)

Examples

1

$$\begin{cases} x = 1 \pmod{2} \\ x = -1 \pmod{7}, \end{cases} \quad \begin{cases} 7v_1 = 1 \pmod{2} \\ 2v_2 = 1 \pmod{7} \end{cases}$$

Chinese Remainder Theorem (III)

Examples

1

$$\begin{cases} x = 1 \pmod{2} \\ x = -1 \pmod{7}, \end{cases} \quad \begin{cases} 7v_1 = 1 \pmod{2} \\ 2v_2 = 1 \pmod{7} \end{cases} \Rightarrow \begin{cases} v_1 = 1, \\ v_2 = 4. \end{cases}$$

Chinese Remainder Theorem (III)

Examples

1

$$\begin{cases} x = 1 \pmod{2} & 7v_1 = 1 \pmod{2} \Rightarrow v_1 = 1, \\ x = -1 \pmod{7}, & 2v_2 = 1 \pmod{7} \Rightarrow v_2 = 4. \end{cases}$$

$$x_0 = 1 \times 1 \times 7 - 1 \times 4 \times 2.$$

Chinese Remainder Theorem (III)

Examples

1

$$\begin{cases} x = 1 \pmod{2} \\ x = -1 \pmod{7}, \end{cases} \quad \begin{array}{l} 7v_1 = 1 \pmod{2} \Rightarrow v_1 = 1, \\ 2v_2 = 1 \pmod{7} \Rightarrow v_2 = 4. \end{array}$$

$$x_0 = -1.$$

Chinese Remainder Theorem (III)

Examples

1

$$\begin{cases} x = 1 \pmod{2} & 7v_1 = 1 \pmod{2} \Rightarrow v_1 = 1, \\ x = -1 \pmod{7}, & 2v_2 = 1 \pmod{7} \Rightarrow v_2 = 4. \end{cases}$$

$$x = -1 + 14k, k \in \mathbb{Z}.$$

Chinese Remainder Theorem (III)

Examples

1

$$\begin{cases} x = 1 \pmod{2} & 7v_1 = 1 \pmod{2} \Rightarrow v_1 = 1, \\ x = -1 \pmod{7}, & 2v_2 = 1 \pmod{7} \Rightarrow v_2 = 4. \end{cases}$$

$$x = -1 + 14k, k \in \mathbb{Z}.$$

2

$$\begin{cases} x = 1 \pmod{2} & 21v_1 = 1 \pmod{2} \\ x = 1 \pmod{3} & 14v_2 = 1 \pmod{3} \\ x = 5 \pmod{7}, & 6v_3 = 1 \pmod{7} \end{cases}$$

Chinese Remainder Theorem (III)

Examples

1

$$\begin{cases} x = 1 \pmod{2} & 7v_1 = 1 \pmod{2} \Rightarrow v_1 = 1, \\ x = -1 \pmod{7}, & 2v_2 = 1 \pmod{7} \Rightarrow v_2 = 4. \end{cases}$$

$$x = -1 + 14k, k \in \mathbb{Z}.$$

2

$$\begin{cases} x = 1 \pmod{2} & 21v_1 = 1 \pmod{2} \Rightarrow v_1 = 1, \\ x = 1 \pmod{3} & 14v_2 = 1 \pmod{3} \Rightarrow v_2 = 2, \\ x = 5 \pmod{7}, & 6v_3 = 1 \pmod{7} \Rightarrow v_3 = 6. \end{cases}$$

Chinese Remainder Theorem (III)

Examples

1

$$\begin{cases} x = 1 \pmod{2} & 7v_1 = 1 \pmod{2} \Rightarrow v_1 = 1, \\ x = -1 \pmod{7}, & 2v_2 = 1 \pmod{7} \Rightarrow v_2 = 4. \end{cases}$$

$$x = -1 + 14k, k \in \mathbb{Z}.$$

2

$$\begin{cases} x = 1 \pmod{2} & 21v_1 = 1 \pmod{2} \Rightarrow v_1 = 1, \\ x = 1 \pmod{3} & 14v_2 = 1 \pmod{3} \Rightarrow v_2 = 2, \\ x = 5 \pmod{7}, & 6v_3 = 1 \pmod{7} \Rightarrow v_3 = 6. \end{cases}$$

$$x_0 = 1 \times 1 \times 21 + 1 \times 2 \times 14 + 5 \times 6 \times 6.$$

Chinese Remainder Theorem (III)

Examples

1

$$\begin{cases} x = 1 \pmod{2} & 7v_1 = 1 \pmod{2} \Rightarrow v_1 = 1, \\ x = -1 \pmod{7}, & 2v_2 = 1 \pmod{7} \Rightarrow v_2 = 4. \end{cases}$$

$$x = -1 + 14k, k \in \mathbb{Z}.$$

2

$$\begin{cases} x = 1 \pmod{2} & 21v_1 = 1 \pmod{2} \Rightarrow v_1 = 1, \\ x = 1 \pmod{3} & 14v_2 = 1 \pmod{3} \Rightarrow v_2 = 2, \\ x = 5 \pmod{7}, & 6v_3 = 1 \pmod{7} \Rightarrow v_3 = 6. \end{cases}$$

$$x_0 = 229.$$

Chinese Remainder Theorem (III)

Examples

1

$$\begin{cases} x = 1 \pmod{2} & 7v_1 = 1 \pmod{2} \Rightarrow v_1 = 1, \\ x = -1 \pmod{7}, & 2v_2 = 1 \pmod{7} \Rightarrow v_2 = 4. \end{cases}$$

$$x = -1 + 14k, k \in \mathbb{Z}.$$

2

$$\begin{cases} x = 1 \pmod{2} & 21v_1 = 1 \pmod{2} \Rightarrow v_1 = 1, \\ x = 1 \pmod{3} & 14v_2 = 1 \pmod{3} \Rightarrow v_2 = 2, \\ x = 5 \pmod{7}, & 6v_3 = 1 \pmod{7} \Rightarrow v_3 = 6. \end{cases}$$

$$x = 19 + 42k, k \in \mathbb{Z}.$$

Chinese Remainder Theorem (IV)

Applications

- 1 RSA : $\mathbb{Z}/pq\mathbb{Z} \simeq \mathbb{Z}/p\mathbb{Z} \times \mathbb{Z}/q\mathbb{Z}$.
- 2 Multimodular computation, parallelism.
- 3 Fast Fourier Transform.
- 4 Generalization to any ring, e.g. Lagrange interpolation for polynomial rings $\mathbb{K}[X]$.

Lagrange Interpolation (I)

Theorem 2

If (x_0, \dots, x_d) are **different elements** of \mathbb{K} and a_0, \dots, a_d are in \mathbb{K} ,
then system $(S) = \begin{cases} P(x_0) = a_0 \\ \vdots \\ P(x_d) = a_d. \end{cases}$ has solutions in $\mathbb{K}[X]$.

Lagrange Interpolation (I)

Theorem 2

If (x_0, \dots, x_d) are **different elements** of \mathbb{K} and a_0, \dots, a_d are in \mathbb{K} ,

then system $(S) = \begin{cases} P(x_0) = a_0 \\ \vdots \\ P(x_d) = a_d. \end{cases}$ has solutions in $\mathbb{K}[X]$.

If P_0 is a solution, then the other ones are

$P = P_0 + Q \cdot (X - x_0) \cdots (X - x_d)$ for any $Q \in \mathbb{K}[X]$.

Lagrange Interpolation (II)

Solution

Let L_0, \dots, L_d be such that $L_i = \prod_{\substack{0 \leq j \leq d \\ j \neq i}} \frac{X - x_j}{x_i - x_j}$, i.e.

$$\prod_{\substack{0 \leq j \leq d \\ j \neq i}} \frac{X - x_j}{x_i - x_j} = 1 \pmod{(X - x_i)},$$

then

$$P_0 = \sum_{0 \leq i \leq d} a_i L_i = \sum_{0 \leq i \leq d} a_i \prod_{\substack{0 \leq j \leq d \\ j \neq i}} \frac{X - x_j}{x_i - x_j}.$$

→ P_0 is the **only** polynomial of **degree at most d** , solution of (S).

Lagrange Interpolation (III)

Algorithm

$$① \quad T \leftarrow \prod_{0 \leq i \leq d} X - x_i,$$

$$② \quad U \leftarrow \prod_{\substack{0 \leq j \leq d \\ j \neq i}} x_i - x_j,$$

$$③ \quad P \leftarrow \sum_{0 \leq i \leq d} \frac{T}{X - x_i} \frac{1}{U} y_i.$$

→ Complexity is in $O(d^2)$.

Lagrange Interpolation (IV)

Examples

1

$$\left\{ \begin{array}{l} P(-1) = 3 \\ P(1) = 1 \\ P(2) = 3, \end{array} \right. \quad \begin{array}{l} L_1 = \frac{X-1}{-1-1} \cdot \frac{X-2}{-1-2} \\ L_2 = \frac{X+1}{1+1} \cdot \frac{X-2}{1-2} \\ L_3 = \frac{X+1}{2+1} \cdot \frac{X-1}{2-1} \end{array}$$

Lagrange Interpolation (IV)

Examples

1

$$\begin{cases} P(-1) = 3 \\ P(1) = 1 \\ P(2) = 3, \end{cases} \quad \begin{aligned} L_1 &= \frac{X-1}{-1-1} \cdot \frac{X-2}{-1-2} \Rightarrow L_1 = \frac{X^2-3X+2}{6}, \\ L_2 &= \frac{X+1}{1+1} \cdot \frac{X-2}{1-2} \Rightarrow L_2 = \frac{X^2-X-2}{-2}, \\ L_3 &= \frac{X+1}{2+1} \cdot \frac{X-1}{2-1} \Rightarrow L_3 = \frac{X^2-1}{3}. \end{aligned}$$

Lagrange Interpolation (IV)

Examples

1

$$\begin{cases} P(-1) = 3 \\ P(1) = 1 \\ P(2) = 3, \end{cases} \quad \begin{aligned} L_1 &= \frac{X-1}{-1-1} \cdot \frac{X-2}{-1-2} \Rightarrow L_1 = \frac{X^2-3X+2}{6}, \\ L_2 &= \frac{X+1}{1+1} \cdot \frac{X-2}{1-2} \Rightarrow L_2 = \frac{X^2-X-2}{-2}, \\ L_3 &= \frac{X+1}{2+1} \cdot \frac{X-1}{2-1} \Rightarrow L_3 = \frac{X^2-1}{3}. \end{aligned}$$

$$P_0 = 3 \frac{X^2 - 3X + 2}{6} + \frac{X^2 - X - 2}{-2} + 3 \frac{X^2 - 1}{3}.$$

Lagrange Interpolation (IV)

Examples

1

$$\begin{cases} P(-1) = 3 \\ P(1) = 1 \\ P(2) = 3, \end{cases} \quad \begin{aligned} L_1 &= \frac{X-1}{-1-1} \cdot \frac{X-2}{-1-2} \Rightarrow L_1 = \frac{X^2-3X+2}{6}, \\ L_2 &= \frac{X+1}{1+1} \cdot \frac{X-2}{1-2} \Rightarrow L_2 = \frac{X^2-X-2}{-2}, \\ L_3 &= \frac{X+1}{2+1} \cdot \frac{X-1}{2-1} \Rightarrow L_3 = \frac{X^2-1}{3}. \end{aligned}$$

$$P_0 = X^2 - X + 1.$$

Lagrange Interpolation (IV)

Examples

1

$$\begin{cases} P(-1) = 3 \\ P(1) = 1 \\ P(2) = 3, \end{cases} \quad \begin{aligned} L_1 &= \frac{X-1}{-1-1} \cdot \frac{X-2}{-1-2} \Rightarrow L_1 = \frac{X^2-3X+2}{6}, \\ L_2 &= \frac{X+1}{1+1} \cdot \frac{X-2}{1-2} \Rightarrow L_2 = \frac{X^2-X-2}{-2}, \\ L_3 &= \frac{X+1}{2+1} \cdot \frac{X-1}{2-1} \Rightarrow L_3 = \frac{X^2-1}{3}. \end{aligned}$$

$$P = X^2 - X + 1 + Q \cdot (X + 1)(X - 1)(X - 2), \quad Q \in \mathbb{K}[X].$$

Lagrange Interpolation (IV)

Examples

1

$$\begin{cases} P(-1) = 3 \\ P(1) = 1 \\ P(2) = 3, \end{cases} \quad \begin{aligned} L_1 &= \frac{X-1}{-1-1} \cdot \frac{X-2}{-1-2} \Rightarrow L_1 = \frac{X^2-3X+2}{6}, \\ L_2 &= \frac{X+1}{1+1} \cdot \frac{X-2}{1-2} \Rightarrow L_2 = \frac{X^2-X-2}{-2}, \\ L_3 &= \frac{X+1}{2+1} \cdot \frac{X-1}{2-1} \Rightarrow L_3 = \frac{X^2-1}{3}. \end{aligned}$$

$$P = X^2 - X + 1 + Q \cdot (X + 1)(X - 1)(X - 2), \quad Q \in \mathbb{K}[X].$$

2

$$\begin{cases} P(-1) = 3 \\ P(1) = 3 \\ P(2) = 3, \end{cases}$$

Lagrange Interpolation (IV)

Examples

1

$$\begin{cases} P(-1) = 3 \\ P(1) = 1 \\ P(2) = 3, \end{cases} \quad \begin{aligned} L_1 &= \frac{X-1}{-1-1} \cdot \frac{X-2}{-1-2} \Rightarrow L_1 = \frac{X^2-3X+2}{6}, \\ L_2 &= \frac{X+1}{1+1} \cdot \frac{X-2}{1-2} \Rightarrow L_2 = \frac{X^2-X-2}{-2}, \\ L_3 &= \frac{X+1}{2+1} \cdot \frac{X-1}{2-1} \Rightarrow L_3 = \frac{X^2-1}{3}. \end{aligned}$$

$$P = X^2 - X + 1 + Q \cdot (X + 1)(X - 1)(X - 2), \quad Q \in \mathbb{K}[X].$$

2

$$\begin{cases} P(-1) = 3 \\ P(1) = 3 \\ P(2) = 3, \end{cases} \quad \begin{aligned} \Rightarrow L_1 &= \frac{X^2-3X+2}{6}, \\ \Rightarrow L_2 &= \frac{X^2-X-2}{-2}, \\ \Rightarrow L_3 &= \frac{X^2-1}{3}. \end{aligned}$$

Lagrange Interpolation (IV)

Examples

1

$$\begin{cases} P(-1) = 3 \\ P(1) = 1 \\ P(2) = 3, \end{cases} \quad \begin{aligned} L_1 &= \frac{X-1}{-1-1} \cdot \frac{X-2}{-1-2} \Rightarrow L_1 = \frac{X^2-3X+2}{6}, \\ L_2 &= \frac{X+1}{1+1} \cdot \frac{X-2}{1-2} \Rightarrow L_2 = \frac{X^2-X-2}{-2}, \\ L_3 &= \frac{X+1}{2+1} \cdot \frac{X-1}{2-1} \Rightarrow L_3 = \frac{X^2-1}{3}. \end{aligned}$$

$$P = X^2 - X + 1 + Q \cdot (X + 1)(X - 1)(X - 2), \quad Q \in \mathbb{K}[X].$$

2

$$\begin{cases} P(-1) = 3 \\ P(1) = 3 \\ P(2) = 3, \end{cases} \Rightarrow \begin{aligned} L_1 &= \frac{X^2-3X+2}{6}, \\ L_2 &= \frac{X^2-X-2}{-2}, \\ L_3 &= \frac{X^2-1}{3}. \end{aligned}$$

$$P_0 = 3 \frac{X^2 - 3X + 2}{6} + 3 \frac{X^2 - X - 2}{-2} + 3 \frac{X^2 - 1}{3}.$$

Lagrange Interpolation (IV)

Examples

1

$$\begin{cases} P(-1) = 3 \\ P(1) = 1 \\ P(2) = 3, \end{cases} \quad \begin{aligned} L_1 &= \frac{X-1}{-1-1} \cdot \frac{X-2}{-1-2} \Rightarrow L_1 = \frac{X^2-3X+2}{6}, \\ L_2 &= \frac{X+1}{1+1} \cdot \frac{X-2}{1-2} \Rightarrow L_2 = \frac{X^2-X-2}{-2}, \\ L_3 &= \frac{X+1}{2+1} \cdot \frac{X-1}{2-1} \Rightarrow L_3 = \frac{X^2-1}{3}. \end{aligned}$$

$$P = X^2 - X + 1 + Q \cdot (X + 1)(X - 1)(X - 2), \quad Q \in \mathbb{K}[X].$$

2

$$\begin{cases} P(-1) = 3 \Rightarrow L_1 = \frac{X^2-3X+2}{6}, \\ P(1) = 3 \Rightarrow L_2 = \frac{X^2-X-2}{-2}, \\ P(2) = 3, \Rightarrow L_3 = \frac{X^2-1}{3}. \end{cases}$$

$$P_0 = 3.$$

Lagrange Interpolation (IV)

Examples

1

$$\begin{cases} P(-1) = 3 \\ P(1) = 1 \\ P(2) = 3, \end{cases} \quad \begin{aligned} L_1 &= \frac{X-1}{-1-1} \cdot \frac{X-2}{-1-2} \Rightarrow L_1 = \frac{X^2-3X+2}{6}, \\ L_2 &= \frac{X+1}{1+1} \cdot \frac{X-2}{1-2} \Rightarrow L_2 = \frac{X^2-X-2}{-2}, \\ L_3 &= \frac{X+1}{2+1} \cdot \frac{X-1}{2-1} \Rightarrow L_3 = \frac{X^2-1}{3}. \end{aligned}$$

$$P = X^2 - X + 1 + Q \cdot (X + 1)(X - 1)(X - 2), \quad Q \in \mathbb{K}[X].$$

2

$$\begin{cases} P(-1) = 3 \\ P(1) = 3 \\ P(2) = 3, \end{cases} \quad \begin{aligned} \Rightarrow L_1 &= \frac{X^2-3X+2}{6}, \\ \Rightarrow L_2 &= \frac{X^2-X-2}{-2}, \\ \Rightarrow L_3 &= \frac{X^2-1}{3}. \end{aligned}$$

$$P = 3 + Q \cdot (X + 1)(X - 1)(X - 2), \quad Q \in \mathbb{K}[X].$$

Lagrange Interpolation (V)

Applications

- 1 Computation of characteristic polynomials.
- 2 Shamir's Secret Sharing.
- 3 Computation of the inverse of a Vandermonde matrix.
- 4 Evaluation-interpolation algorithm.

Comparison between CRT and Lagrange interpolation (I)

n_0, \dots, n_d pairwise coprime.

$$(S) = \begin{cases} x = a_0 \pmod{n_0} \\ \vdots \\ x = a_d \pmod{n_d}. \end{cases}$$

x_0, \dots, x_d pairwise distinct.

$$(S) = \begin{cases} P(x_0) = a_0 \\ \vdots \\ P(x_d) = a_d. \end{cases}$$

Comparison between CRT and Lagrange interpolation (I)

n_0, \dots, n_d pairwise coprime.

$$(S) = \begin{cases} x = a_0 \pmod{n_0} \\ \vdots \\ x = a_d \pmod{n_d}. \end{cases}$$

x_0, \dots, x_d pairwise distinct.

$$(S) = \begin{cases} P(x_0) = a_0 \\ \vdots \\ P(x_d) = a_d, \end{cases}$$
$$= \begin{cases} P = a_0 \pmod{(X - x_0)} \\ \vdots \\ P = a_d \pmod{(X - x_d)}. \end{cases}$$

Comparison between CRT and Lagrange interpolation (II)

$$v_i \prod_{\substack{0 \leq j \leq d \\ j \neq i}} n_j = 1 \bmod n_i.$$

$$L_i = \prod_{\substack{0 \leq j \leq d \\ j \neq i}} \frac{X - x_j}{x_i - x_j}.$$

Comparison between CRT and Lagrange interpolation (II)

$$v_i \prod_{\substack{0 \leq j \leq d \\ j \neq i}} n_j = 1 \pmod{n_i}.$$

$$\begin{aligned} L_i &= \prod_{\substack{0 \leq j \leq d \\ j \neq i}} \frac{X - x_j}{x_i - x_j} \\ &= 1 \pmod{X - x_i} \\ &= 0 \pmod{X - x_j}, \quad j \neq i. \end{aligned}$$

Comparison between CRT and Lagrange interpolation (II)

$$v_i \prod_{\substack{0 \leq j \leq d \\ j \neq i}} n_j = 1 \pmod{n_i} \\ = 0 \pmod{n_j}, j \neq i.$$

$$L_i = \prod_{\substack{0 \leq j \leq d \\ j \neq i}} \frac{X - x_j}{x_i - x_j} \\ = 1 \pmod{(X - x_i)} \\ = 0 \pmod{(X - x_j)}, j \neq i.$$

Comparison between CRT and Lagrange Interpolation (III)

Reminder

There is an **Euclidean division** in $\mathbb{K}[X]$:

$$\forall A, B \in \mathbb{K}[X], \exists Q, R \in \mathbb{K}[X], \quad A = BQ + R, \quad \deg R < \deg B.$$

Comparison between CRT and Lagrange Interpolation (III)

Reminder

There is an **Euclidean division** in $\mathbb{K}[X]$:

$$\forall A, B \in \mathbb{K}[X], \exists Q, R \in \mathbb{K}[X], \quad A = BQ + R, \quad \deg R < \deg B.$$

Example

If $A = X^3 + X + 1$ and $B = X^2 - 1$, then $Q = X$, $R = 2X + 1$.

$$X^3 + X + 1 = X(X^2 - 1) + (2X + 1).$$

Comparison between CRT and Lagrange Interpolation (IV)

Consequence

- 1 One can perform the **Euclidean algorithm** to compute the **GCD** of two polynomials of $\mathbb{K}[X]$.
- 2 Chinese Remainder Theorem (**CRT**) is valid in $\mathbb{K}[X]$!

Chinese Remainder Theorem in $\mathbb{K}[X]$

Rewritten Theorem

If R_1, \dots, R_r are **pairwise coprime** polynomials and A_1, \dots, A_r are polynomials modulo R_1, \dots, R_r , then system $(S) = \begin{cases} P = A_1 \pmod{R_1} \\ \vdots \\ P = A_r \pmod{R_r} \end{cases}$ has solutions in $\mathbb{K}[X]$.

Chinese Remainder Theorem in $\mathbb{K}[X]$

Rewritten Theorem

If R_1, \dots, R_r are **pairwise coprime** polynomials and A_1, \dots, A_r are polynomials modulo R_1, \dots, R_r , then system $(S) = \begin{cases} P = A_1 \pmod{R_1} \\ \vdots \\ P = A_r \pmod{R_r} \end{cases}$

has solutions in $\mathbb{K}[X]$.

If P_0 is a solution, then the other ones are $P = P_0 + QR_1 \cdots R_r$ for any $Q \in \mathbb{K}[X]$.

Special case

If each R_i has degree 1, then $R_i = X - x_i$ and $A_i = a_i \in \mathbb{K}$.

Furthermore, R_i, R_j are **coprime** if, and only if, $x_i \neq x_j$.

→ This is **Lagrange interpolation** problem!

Eigenvalues problem

Definition

Let A be a square matrix of size d with coefficients in \mathbb{K} . We say that λ is an **eigenvalue** of A if there is a vector $v \neq 0$ such that $Av = \lambda v$.

Problem

How can one determine the eigenvalues of A ?

How long does it take?

Characteristic polynomial (I)

Definition

Let A be a square matrix of size d with coefficients in \mathbb{K} . We say that λ is an **eigenvalue** of A if there is a vector $v \neq 0$ such that $Av = \lambda v$.

Proposition

The eigenvalues are the roots of a specific polynomial called the **characteristic polynomial of A** defined by

$$P_A(\lambda) = \det(A - \lambda \text{Id}) = \begin{vmatrix} a_{11} - \lambda & a_{12} & \cdots & a_{1d} \\ a_{21} & a_{22} - \lambda & \ddots & \vdots \\ \vdots & \ddots & \ddots & a_{d-1,d} \\ a_{d1} & \cdots & a_{d,d-1} & a_{dd} - \lambda \end{vmatrix}.$$

Characteristic polynomial always has degree d .
Eigenvalues are **not always** in \mathbb{K} !

Characteristic polynomial (II)

Proposition

$P_A(\lambda) = \det(A - \lambda \text{Id})$ and P_A has degree d .

Proof

Let λ be an eigenvalue of A and $v \neq 0$ be an **eigenvector** of A for λ , *i.e.* $Av = \lambda v$.

- One has $(A - \lambda \text{Id})v = 0$ and $v \neq 0$ **if, and only if**, $\det(A - \lambda \text{Id}) = 0$ **if, and only if**, λ is a root of P_A .
- When expanding the determinant, only one term has degree d , it is $(a_{11} - \lambda) \cdots (a_{dd} - \lambda)$.

Characteristic polynomial (III)

Examples

1

$$A = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix},$$

$$P_A(\lambda) = \begin{vmatrix} 2 - \lambda & 1 \\ 1 & 2 - \lambda \end{vmatrix} = (2 - \lambda)^2 - 1 = (\lambda - 1)(\lambda - 3),$$

2

$$B = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix},$$

$$P_B(\lambda) = \begin{vmatrix} 1 - \lambda & 1 \\ 0 & 1 - \lambda \end{vmatrix} = (1 - \lambda)^2 = (\lambda - 1)^2.$$

Characteristic polynomial: complexity (I)

Naive method

→ Gaussian elimination.

Characteristic polynomial: complexity (I)

Naive method

- Gaussian elimination.
 - Matrix $A - \lambda Id$ has coefficients in $\mathbb{K}[\lambda]$.
- If one **inverts** the pivot, one introduces rational fractions. They are very **costly**!
- Otherwise, one must take care that at each step, the computed determinant is a multiple of the target determinant.

Characteristic polynomial: complexity (II)

Example: Gaussian elimination, 1st attempt

$$A = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 2 & 2 \\ 1 & 2 & 3 \end{pmatrix}$$

$$\begin{aligned} P_A(\lambda) &= \begin{vmatrix} 1-\lambda & 1 & 1 \\ 1 & 2-\lambda & 2 \\ 1 & 2 & 3-\lambda \end{vmatrix} = \begin{vmatrix} 1-\lambda & 1 & 1 \\ 0 & \frac{\lambda^2-3\lambda+1}{1-\lambda} & \frac{-2\lambda+1}{1-\lambda} \\ 0 & \frac{-2\lambda+1}{1-\lambda} & \frac{\lambda^2-4\lambda+1}{1-\lambda} \end{vmatrix} \\ &= \begin{vmatrix} 1-\lambda & 1 & 1 \\ 0 & \frac{\lambda^2-3\lambda+1}{1-\lambda} & \frac{-2\lambda+1}{1-\lambda} \\ 0 & 0 & \frac{\lambda^4-7\lambda^3+12\lambda^2-4\lambda}{(1-\lambda)(\lambda^2-3\lambda+1)} \end{vmatrix}. \end{aligned}$$

Denominators introduce a memory overhead.

Characteristic polynomial: complexity (III)

Example: Gaussian elimination, 2nd attempt

$$\begin{aligned} P_A(\lambda) &= \begin{vmatrix} 1-\lambda & 1 & 1 \\ 1 & 2-\lambda & 2 \\ 1 & 2 & 3-\lambda \end{vmatrix} \\ &= \frac{1}{(1-\lambda)^2} \begin{vmatrix} 1-\lambda & 1 & 1 \\ 0 & \lambda^2-3\lambda+1 & -2\lambda+1 \\ 0 & -2\lambda+1 & \lambda^2-4\lambda+1 \end{vmatrix} \\ &= \frac{1}{(1-\lambda)^2(\lambda^2-3\lambda+1)} \begin{vmatrix} 1-\lambda & 1 & 1 \\ 0 & \lambda^2-3\lambda+1 & -2\lambda+1 \\ 0 & 0 & \lambda^4-7\lambda^3+12\lambda^2-4\lambda \end{vmatrix}. \end{aligned}$$

- Divisions of polynomials are still needed.
- The degrees of the polynomials are multiplied by two at each step of the Gaussian elimination.

Characteristic polynomial: complexity (III)

Example: Gaussian elimination, 2nd attempt

$$\begin{aligned} P_A(\lambda) &= \begin{vmatrix} 1-\lambda & 1 & 1 \\ 1 & 2-\lambda & 2 \\ 1 & 2 & 3-\lambda \end{vmatrix} \\ &= \frac{1}{(1-\lambda)^2} \begin{vmatrix} 1-\lambda & 1 & 1 \\ 0 & \lambda^2-3\lambda+1 & -2\lambda+1 \\ 0 & -2\lambda+1 & \lambda^2-4\lambda+1 \end{vmatrix} \\ &= \frac{1}{(1-\lambda)^2(\lambda^2-3\lambda+1)} \begin{vmatrix} 1-\lambda & 1 & 1 \\ 0 & \lambda^2-3\lambda+1 & -2\lambda+1 \\ 0 & 0 & \lambda^4-7\lambda^3+12\lambda^2-4\lambda \end{vmatrix}. \end{aligned}$$

- Divisions of polynomials are still needed.
 - The degrees of the polynomials are multiplied by two at each step of the Gaussian elimination.
- We deal with polynomials of degree 2^{d-1} at the end!

Characteristic polynomial: complexity (III)

Example: Gaussian elimination, 2nd attempt

$$\begin{aligned} P_A(\lambda) &= \begin{vmatrix} 1-\lambda & 1 & 1 \\ 1 & 2-\lambda & 2 \\ 1 & 2 & 3-\lambda \end{vmatrix} \\ &= \frac{1}{(1-\lambda)^2} \begin{vmatrix} 1-\lambda & 1 & 1 \\ 0 & \lambda^2-3\lambda+1 & -2\lambda+1 \\ 0 & -2\lambda+1 & \lambda^2-4\lambda+1 \end{vmatrix} \\ &= \frac{1}{(1-\lambda)^2(\lambda^2-3\lambda+1)} \begin{vmatrix} 1-\lambda & 1 & 1 \\ 0 & \lambda^2-3\lambda+1 & -2\lambda+1 \\ 0 & 0 & \lambda^4-7\lambda^3+12\lambda^2-4\lambda \end{vmatrix}. \end{aligned}$$

- Divisions of polynomials are still needed.
- The degrees of the polynomials are multiplied by two at each step of the Gaussian elimination.
- We deal with polynomials of degree 2^{d-1} at the end!
- The complexity would be exponential in d . This cannot be right.

Characteristic polynomial: algorithm (I)

Interpolation method

- Characteristic polynomial has degree d .

Characteristic polynomial: algorithm (I)

Interpolation method

- Characteristic polynomial has degree d .
- The **complexity** should be **polynomial** in d .

Characteristic polynomial: algorithm (I)

Interpolation method

- Characteristic polynomial has degree d .
- The **complexity** should be **polynomial** in d .
- It **suffices** to know $d + 1$ of its value to compute it.
- One computes $d + 1$ determinants of $A - \lambda_0 \text{Id}, \dots, A - \lambda_d \text{Id}$ for different $\lambda_0, \dots, \lambda_d$.

Characteristic polynomial: algorithm (I)

Interpolation method

- Characteristic polynomial has degree d .
- The **complexity** should be **polynomial** in d .
- It **suffices** to know $d + 1$ of its value to compute it.
- One computes $d + 1$ determinants of $A - \lambda_0 \text{Id}, \dots, A - \lambda_d \text{Id}$ for different $\lambda_0, \dots, \lambda_d$.
- By Lagrange interpolation, one can recover the characteristic polynomial.

Characteristic polynomial: algorithm (II)

Algorithm

- 1 Pick $\lambda_0, \dots, \lambda_d \in \mathbb{K}$ pairwise distinct.
- 2 **for** i **from** 0 **to** d
 $a_i \leftarrow \det(A - \lambda_i \text{Id})$.
- 3 Call LagrangeInterpol with $\lambda_0, \dots, \lambda_d, a_0, \dots, a_d$.

Characteristic polynomial: algorithm (III)

Complexity

- 1 Computation of $d + 1$ determinants of size d : $O(d^4)$.
 - 2 Interpolation of a polynomial of degree at most d : $O(d^2)$.
- Complexity is in $O(d^4)$.

Characteristic polynomial: algorithm (IV)

Warning

- If there are **less** than $d + 1$ elements in \mathbb{K} , this cannot be used.
- This is not an issue in $\mathbb{Q}, \mathbb{R}, \mathbb{C}$.
- This is an issue in $\mathbb{F}_p, \mathbb{F}_q$: the finite fields.
- In that case, one needs to pick up $\lambda_0, \dots, \lambda_d$ in a bigger field \mathbb{L} containing \mathbb{K} .

MAPLE

- ① Use of `LINALG` package, not `LINEARALGEBRA` (too young in MAPLE 10).
- ② Use `collect` and `simplify` when dealing with rational fractions and polynomials.
 - MAPLE will sort the terms by degree,
 - and will write the expression with only one denominator.
- ③ Use `?package` and `?function` to see the help page of the chosen package or function.

Sylvester Matrix

Definition

Let P and Q be two polynomials of $\mathbb{K}[X]$ of degree m and n .

$$P = p_m X^m + \cdots + p_0$$

$$Q = q_n X^n + \cdots + q_0.$$

The **Sylvester matrix** associated to P and Q is the $(m+n) \times (m+n)$ matrix

$$\begin{pmatrix} p_m & 0 & \cdots & 0 & q_n & 0 & \cdots & 0 \\ p_{m-1} & p_m & & \vdots & q_{n-1} & q_n & & \vdots \\ \vdots & p_{m-1} & \ddots & \vdots & \vdots & q_{n-1} & \ddots & \vdots \\ p_1 & \vdots & & 0 & q_1 & \vdots & & 0 \\ p_0 & p_1 & & p_m & q_0 & q_1 & & q_m \\ 0 & p_0 & & p_{m-1} & 0 & q_0 & & q_{m-1} \\ \vdots & 0 & & \vdots & \vdots & 0 & & \vdots \\ \vdots & \vdots & \ddots & p_1 & \vdots & \vdots & \ddots & q_1 \\ 0 & 0 & \cdots & p_0 & 0 & 0 & \cdots & q_0 \end{pmatrix}$$

Sylvester Matrix

Example

If $P = X^4 + 2X^3 + 3X^2 + 4X + 5$ and $Q = 9X^3 + 8X^2 + 7X + 6$, then their Sylvester Matrix is

$$\begin{pmatrix} 1 & 0 & 0 & 9 & 0 & 0 & 0 \\ 2 & 1 & 0 & 8 & 9 & 0 & 0 \\ 3 & 2 & 1 & 7 & 8 & 9 & 0 \\ 4 & 3 & 2 & 6 & 7 & 8 & 9 \\ 5 & 4 & 3 & 0 & 6 & 7 & 8 \\ 0 & 5 & 4 & 0 & 0 & 6 & 7 \\ 0 & 0 & 5 & 0 & 0 & 0 & 6 \end{pmatrix}$$