

# $p$ -adic precision and Gröbner bases

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## 1 Row-echelon form and $p$ -adic precision

- Step-by-step analysis
- Loss in precision in the row-echelon form computation

## 2 Matrix-F5 algorithm and $p$ -adic computations

- On Matrix-F5 algorithm
- Issues with finite precision
- Which GB can be computed ?
- Continuity and optimality

# Definition of the precision

## Finite-precision $p$ -adics

Elements of  $\mathbb{Q}_p$  can be written  $\sum_{i=-l}^{+\infty} a_i p^i$ , with  $a_i \in \llbracket 0, p-1 \rrbracket$ ,  $l \in \mathbb{Z}$  and  $p$  a prime number.

While working with a computer, we usually only can consider the beginning of this power serie expansion: we only consider elements of the following form  $\sum_{i=l}^{d-1} a_i p^i + O(p^d)$ , with  $l \in \mathbb{Z}$ .

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## Definition

The **order**, or the **absolute precision** of  $\sum_{i=k}^{d-1} a_i p^i + O(p^d)$  is  $d$ . Its **relative precision** corresponds to the number of its significant figures, and thus, is given by  $d - \min \{i \in \mathbb{Z}, a_i \neq 0\}$ .

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## Example

The order of  $3 * 7^{-1} + 4 * 7^0 + 5 * 7^1 + 6 * 7^2 + O(7^3)$  is 3, and its relative precision is  $4 = 3 - (-1)$ .

# Motivation for $p$ -adic algorithm

Why should one work with  $p$ -adic numbers ?

- Going from  $\mathbb{F}_p$  to  $\mathbb{Z}_p$  and then back to  $\mathbb{F}_p$  enables more computation ;

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- Polynomial factorization with Hensel lemma ;
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## My personal (long-term) motivation

Computing moduli spaces of  $p$ -adic Galois representations.

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## $p$ -adic precision vs real precision

The quintessential idea of the step-by-step analysis is the following :

Proposition ( $p$ -adic errors don't add)

*Indeed,*

$$(a + O(p^k)) + (b + O(p^k)) = a + b + O(p^k).$$

*That is to say, if  $a$  and  $b$  are known up to precision  $O(p^k)$ , then so is  $a + b$ .*

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**Remark**

It is quite the opposite to when dealing with real numbers, because of **Round-off error** :

$$(1 + 5 * 10^{-2}) + (2 + 6 * 10^{-2}) = 3 + 1 * 10^{-1} + 1 * 10^{-2}.$$

That is to say, if  $a$  and  $b$  are known up to precision  $10^{-n}$ , then  $a + b$  is known up to  $10^{(-n + 1)}$ .



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# Precision formulae

## Proposition (addition)

$$(x_0 + O(p^{k_0})) + (x_1 + O(p^{k_1})) = x_0 + x_1 + O(p^{\min(k_0, k_1)})$$

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$$(x_0 + O(p^{k_0})) * (x_1 + O(p^{k_1})) = x_0 * x_1 + O(p^{\min(k_0 + v_p(x_1), k_1 + v_p(x_0))})$$

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## Proposition (division)

$$\frac{xp^a + O(p^b)}{yp^c + O(p^d)} = x * y^{-1} p^{a-c} + O(p^{\min(d+a-2c, b-c)})$$

In particular,

$$\frac{1}{p^c y + O(p^d)} = y^{-1} p^{-c} + O(p^{d-2c})$$

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# The result for the Gauss method

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- its coefficients are known up to  $O(p^k)$ .
- $\text{val}(\Delta) < k$ , with  $\Delta = \det((M_{i,j})_{1 \leq i \leq n, 1 \leq j \leq n})$ .



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## Theorem

Let  $M \in M_{n,m}(\mathbb{Z}_p)$  such that :

- its coefficients are known up to  $O(p^k)$ .
- $\text{val}(\Delta) < k$ , with  $\Delta = \det((M_{i,j})_{1 \leq i \leq n, 1 \leq j \leq n})$ .

Then the **loss of precision** to compute a row-echelon form of  $M$  is  $\leq \text{val}(\Delta)$ .

# Proof of the theorem

## Gauss' method

$$M = \begin{bmatrix} m_{1,1} + O(p^k) & m_{1,2} + O(p^k) & \cdots & m_{1,m} + O(p^k) \\ m_{2,1} + O(p^k) & m_{2,2} + O(p^k) & \cdots & m_{2,m} + O(p^k) \end{bmatrix}$$

We assume that,

$$\det \left( \begin{bmatrix} m_{1,1} + O(p^k) & m_{1,2} + O(p^k) \\ m_{2,1} + O(p^k) & m_{2,2} + O(p^k) \end{bmatrix} \right) \neq O(p^k).$$



# Proof of the theorem

## Gauss' method

$$M \simeq \begin{bmatrix} p^{a_1} + O(p^k) & m_{1,2} + O(p^k) & \cdots & m_{1,m} + O(p^k) \\ p^{a_2} + O(p^k) & m_{2,2} + O(p^k) & \cdots & m_{2,m} + O(p^k) \end{bmatrix} \quad L_2 \leftarrow L_2 - \frac{M_{2,1}}{M_{1,1}} L_1$$



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Indeed,  $M_{2,1} - \frac{M_{2,1}}{M_{1,1}} * M_{1,1} = 0$  (formally).

In addition,  $\frac{M_{2,1}}{M_{1,1}} = \frac{p^{a_2} + O(p^k)}{p^{a_1} + O(p^k)} = p^{a_2 - a_1} + O(p^{k-a_1})$ , therefore

$$L_2 - \frac{M_{2,1}}{M_{1,1}} L_1 = L_2 + (p^{a_2 - a_1} + O(p^{k-a_1})) L_1 .$$

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In the end, we get :

$$M = \begin{bmatrix} p^{a_1} + O(p^k) & m_{1,2} + O(p^k) & \dots & m_{1,m} + O(p^k) \\ \boxed{0} & p^{a_2} + O(p^{k-a_1}) & \dots & \boxed{m_{2,m} + O(p^{k-a_1})} \end{bmatrix}$$

The loss of precision on the second row is  $a_1$  .



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$$\text{val}(\det \left( \begin{bmatrix} m_{1,1} + O(p^k) & m_{1,2} + O(p^k) \\ m_{2,1} + O(p^k) & m_{2,2} + O(p^k) \end{bmatrix} \right)) = a_1 + a_2, \text{ with } a_i > 0.$$

The loss in precision is upper-bounded by

$$\text{val}(\det((M_{i,j})_{1 \leq i \leq 2, 1 \leq j \leq 2}))$$





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# The Macaulay matrix

## Notations

For  $k$  a field,  $n, s \in \mathbb{N}$ , and  $R = k[X_1, \dots, X_n]$ , we denote by  $R_d$  the homogeneous polynomials of degree  $d$  of  $R$ .

Let  $\omega$  be a monomial order on  $R$ .

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Let  $\omega$  be a monomial order on  $R$ .

## Proposition (D. Lazard 83)

For an homogeneous ideal  $I = (f_1, \dots, f_s) \subset R$  ( $f_1, \dots, f_s$  being homogeneous),  $d \in \mathbb{N}$ ,

$$I \cap R_d = \langle x^\alpha f_i, |\alpha| + \deg(f_i) = d \rangle,$$

as  $k$ -vector spaces .

# The Macaulay matrix

## Definition (Macaulay's matrix)

We denote by  $Mac_d(f_1, \dots, f_s)$  the matrix :

$$\begin{array}{c}
 x^{\alpha_{1,1}} f_1 \\
 \vdots \\
 x^{\alpha_{1, \binom{n+d-d_1-1}{n-1}}} f_1 \\
 x^{\alpha_{2,1}} f_2 \\
 \vdots \\
 x^{\alpha_{s, \binom{n+d-d_s-1}{n-1}}} f_s
 \end{array}
 \left[ \begin{array}{c}
 x^{\alpha} f_i \text{ written in the basis of the } x^{d_i} \\
 \vdots \\
 \vdots
 \end{array} \right]$$

Its rows  $x^{\alpha} f_i$  are written in the basis  $x^{d_1}, \dots, x^{\binom{n+d-1}{n-1}}$ , with  $|\alpha| + \deg(f_i) = d$ . Also,  $x^{\alpha_{i,j}} < x^{\alpha_{i,j+1}}$ .



# An algorithm

## The idea of the Matrix-F5 algorithm

The idea is to successively row-echelon the matrices  $Mac_d(f_1, \dots, f_i)$ , **iteratively** with  $d$  and  $i$ .

If you know the **profile** of  $Mac_d(f_1, \dots, f_i)$ , then you know what are the leading terms in  $LT((f_1, \dots, f_i)_d)$  and so, you can remove useless rows in  $Mac_{d'}(f_1, \dots, f_{i'})$  with  $d' > d$  and  $i' > i$ .

# An algorithm

## A Matrix-F5 algorithm

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### Algorithm 1 Matrix-F5 algorithm

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Let  $F = (f_1, \dots, f_s) \in R^s$ , of degree  $d_1, \dots, d_s$ , and  $D \in \mathbb{N}$ .

$G \leftarrow F$

**for**  $d \in \llbracket 0, D \rrbracket$  **do**

**for**  $i \in \llbracket 1, s \rrbracket$  **do**

    Build  $\widetilde{Mac}_d f_1, \dots, f_i$  ;

    Remove the rows  $x^\alpha f_i$  such that  $x^\alpha$  is the leading term of a row of  $\widetilde{Mac}_{d-d_i, i-1}$ ;

    Compute the row-echelon form  $\widetilde{Mac}_{d, i}$ ;

    Add to  $G$  the rows with a new leading monomial.

**end for**

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# The position of the leading terms ideals

## Problem with testing nullity

A major issue can happen when dealing with finite-precision numbers : not being able **to decide whether there is no non-zero pivot** on a column or whether the precision is not enough.

## Being able to compute the leading terms ideals

$$\begin{bmatrix} 1 + O(p^k) & 1 + O(p^k) & 1 + O(p^k) & 0 \\ 1 + O(p^k) & 1 + O(p^k) & 0 & 1 + O(p^k) \end{bmatrix} \quad L_2 \leftarrow L_2 - \frac{M_{2,1}}{M_{1,1}} L_1$$



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What is the leading term for the second row ?

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# Moreno-Socias conjecture

## Definition (weakly- $\omega$ ideal)

$I$  is said to be a weakly- $\omega$  ideal if :

- for all  $x^\alpha$  a leading monomial according to  $\omega$  of the reduced Gröbner basis of  $I$ ,
- for all  $x^\beta$  such that  $|\alpha| = |\beta|$  and  $x^\beta > x^\alpha$ ,

we have  $x^\beta \in LM(I)$ .

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## Conjecture (Moreno-Socias)

*If  $k$  is an infinite field,  $s \in \mathbb{N}$ ,  $d_1, \dots, d_s \in \mathbb{N}$ , then there is a non-empty Zariski-open subset  $U$  in  $R_{d_1} \times \dots \times R_{d_s}$  such that for all  $(f_1, \dots, f_s) \in U$ ,  $I = (f_1, \dots, f_s)$  is a weakly-grevlex ideal.*

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## Remark

If the conjecture holds, then regular sequences generating a weakly-grevlex ideal are generic.



# An algorithm suited for weakly- $\omega$ ideal

## Proposition ("weak" Matrix-F5 algorithm)

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Then, we can proceed :

# An algorithm suited for weakly- $\omega$ ideal

## Proposition ("weak" Matrix-F5 algorithm)

We assume :

- $(f_1, \dots, f_s)$  is a regular, **(H1)**
- the  $\langle f_1, \dots, f_l \rangle$  are weakly- $\omega$  ideals, **(H2)**
- precision on the  $f_i$ 's is enough.

Then, we can proceed :

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Then, we can proceed :

- At first, we proceed like in the normal Matrix-F5 algorithm ;
- But, as soon as a column with no non-zero pivot is encountered, we **halt** the row-echelon computation. Instead, we **can replace** the non-reduced rows by (already reduced) multiples of the rows of  $\text{Mac}_{d-1,i}$ , so as to get a matrix under row-echelon form.

## 3 quadrics in 6 variables

### An example

With 3 generic quadrics in 6 variables, what we get after reducing the Macaulay matrix in degree 3 is the following :

$$\left[ \begin{array}{c|c} \text{a } 9 \times 9 \text{ invertible block} & \text{(loss in precision : determinant of the } 9 \times 9 \text{ matrix)} \\ \hline 0 & \text{9 rows, } \in \langle x_k L, L \text{ row of of the matrix with } d = 2 \rangle \end{array} \right].$$

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- Issues with finite precision
- Which GB can be computed ?
- Continuity and optimality

# On regularity

## Proposition

Let  $F = (f_1, \dots, f_s)$  satisfying **H1** and **H2**. Then:

- $(h_1, \dots, h_s) \mapsto LM(\langle f_1, \dots, f_s \rangle)$  is **constant** around  $F$ .

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- We can give **explicit neighbourhood** of  $F$  with the Macaulay matrices.

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- $(h_1, \dots, h_s) \mapsto GBR(\langle f_1, \dots, f_s \rangle)$  may **not be stable** at  $F$ .

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## Remark

Moreno-Socias conjecture implies that **Structure** is generic for grevlex.

# Further concerns

## Implementation

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- F5 algorithm for tropical Gröbner bases;

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- F5 algorithm for tropical Gröbner bases;
- Border bases.

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