

# Online order basis algorithm and its application to block Wiedemann algorithm\*

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\*. This document has been written using the GNU T<sub>E</sub>X<sub>MACS</sub> text editor (see [www.texmacs.org](http://www.texmacs.org)).

## 1. Order basis

- a. Definition
- b. Algorithms

## 2. Application to block Wiedemann algorithm

## 3. Contributions :

- a. Fast iterative order basis
- b. Fast online order basis
- c. Timings

Let  $\mathbb{K}$  be a field and  $F \in \mathbb{K}[[x]]^{m \times n}$ .

Let  $(F, \sigma)$  be the  $\mathbb{K}[x]$ -module  $\{v \in \mathbb{K}[[x]]^{1 \times m} \text{ such that } vF = 0 \text{ mod } x^\sigma\}$ .

**Definition.** An  $(F, \sigma)$  *order basis*  $P$  is a basis of  $(F, \sigma)$  of minimal degree.

## Minimal degree ?

1. Row degree :

$$\text{rdeg}(P_1, \dots, P_n) = \max(\deg P_i), \quad \text{rdeg}\left(\begin{pmatrix} \text{row 1} \\ \vdots \\ \text{row } n \end{pmatrix}\right) = (\text{rdeg}(\text{row } i))_{i=1 \dots n}$$

Partial order  $(v_1, \dots, v_m) \leq (w_1, \dots, w_m) \Leftrightarrow \forall i \quad v_i \leq w_i$  on the sorted vector

2. Shifted row degree :  $\text{rdeg}_{\vec{s}}(P_1, \dots, P_n) = \max(\deg P_i + s_i)$  where  $\vec{s} = (s_1, \dots, s_n)$

$\rightsquigarrow (F, \sigma, \vec{s})$  order basis : minimal for the  $\vec{s}$ -row degree

Let  $\mathbb{K}$  be a field and  $F \in \mathbb{K}[[x]]^{m \times n}$ .

Let  $(F, \sigma)$  be the  $\mathbb{K}[x]$ -module  $\{v \in \mathbb{K}[[x]]^{1 \times m} \text{ such that } vF = 0 \text{ mod } x^\sigma\}$ .

**Definition.** An  $(F, \sigma)$  *order basis*  $P$  is a basis of  $(F, \sigma)$  of minimal degree.

**Lemma.** There exists a basis  $P$  of minimal degree.

**Remark.** Existence but no unicity ( $\rightsquigarrow$  Popov form).

**Example.** (taken from Zhou thesis)

$$\underbrace{\begin{pmatrix} 1 & 0 & 1 & 1 \\ x & 1 & 1+x & 0 \\ 1 & x^2+x^3 & x & 0 \\ x^2 & 0 & x^3+x^4 & 0 \end{pmatrix}}_{(F, 8, \vec{0})\text{-order basis over } \mathbb{F}_2} \underbrace{\begin{pmatrix} x+x^2+x^3+x^4+x^5+x^6 \\ 1+x+x^5+x^6+x^7 \\ 1+x^2+x^4+x^5+x^6+x^7 \\ 1+x+x^3+x^7 \end{pmatrix}}_{F \text{ in } \mathbb{F}_2[[x]]^{4 \times 1}} = 0^{4 \times 1} \text{ mod } x^8$$

## 1. Base case $\sigma = 1$

Algorithm Basis :  $F \bmod x \longrightarrow$  a  $(F, 1)$  order basis

**Basic idea :**

If  $\begin{pmatrix} K \\ S \end{pmatrix} (F \bmod x) = \begin{pmatrix} 0 \\ R \end{pmatrix}$  with  $R$  full rank

then  $\begin{pmatrix} K \\ xS \end{pmatrix} (F \bmod x) = \begin{pmatrix} 0 \\ xR \end{pmatrix} = (0) \bmod x$

$\rightarrow \begin{pmatrix} K \\ xS \end{pmatrix}$  is a basis of the module  $(F, 1)$ .

## 2. Splitting the order basis problem

### Theorem.

Let  $P_1$  be a  $(F, \sigma_1, \vec{s})$  order basis of  $\vec{s}$ -degree  $\vec{u}$ .

Let  $P_2$  be a  $((P_1 F \operatorname{div} x^{\sigma_1}), \sigma_2, \vec{u})$  order basis of  $\vec{u}$ -degree  $\vec{v}$ .

Then  $P_2 P_1$  is a  $(F, \sigma_1 + \sigma_2, \vec{s})$  order basis of  $\vec{s}$ -degree  $\vec{v}$ .

### Remarks.

- $P_1 F = x^{\sigma_1} M$  where  $M = (P_1 F \operatorname{div} x^{\sigma_1}) \in \mathbb{K}[[x]]^{m \times n}$

$$P_2 P_1 F = x^{\sigma_1} P_2 M = x^{\sigma_1} (x^{\sigma_2} M') = (0) \bmod x^{\sigma_1 + \sigma_2}$$

- The module  $(F, \sigma_1 + \sigma_2, \vec{s})$  is a subset of  $(F, \sigma_1, \vec{s})$  of basis  $P_1$

$\rightsquigarrow$  Express the module  $(F, \sigma_1 + \sigma_2, \vec{s})$  on the basis  $P_1 \rightarrow$  reduce the problem

- Need of  $\vec{s}$ -row degree:

The row total degree of  $vP$  is the exactly the  $\vec{s}$ -row degree of  $v$  where  $\vec{s}$  is the row degree of  $P$  □

**Input** :  $F \in \mathbb{K}[x]^{m \times n}, \sigma \in \mathbb{N}$

**Output** : A  $(F, \sigma)$  order basis  $P$  of  $F$

## 1. Quadratic algorithm M-Basis

Iterative :  $(F, 1) \rightarrow (F, 2) \rightarrow (F, 3) \rightarrow \dots \rightarrow (F, \sigma)$

### Algorithm M-Basis

1.  $P_0 := \text{Basis}(F \bmod x)$
2. **for**  $k = 1, \dots, \sigma - 1$  **do**
3.      $F' := x^{-k} P_{k-1} F$
4.      $M_k := \text{Basis}(F' \bmod x)$
5.      $P_k := M_k P_{k-1}$
6. **return**  $P_{\sigma-1}$

In terms of polynomial multiplication, naive multiplication  $P_{\sigma-1} = M_{\sigma-1} (\dots M_3 (M_2 M_1) )$  where each  $M_i$  is of degree one.

**Input** :  $F \in \mathbb{K}[x]^{m \times n}, \sigma \in \mathbb{N}$

**Output** : A  $(F, \sigma)$  order basis  $P$  of  $F$

## 2. Quasi-linear algorithm PM-Basis

Divide-and-conquer :  $(F, 1) \rightarrow (F, 2) \rightarrow (F, 4) \rightarrow \dots \rightarrow (F, \sigma/2) \rightarrow (F, \sigma)$

### Algorithm PM-Basis

- |   |                        |
|---|------------------------|
| 1. <b>if</b> $\sigma = 1$ <b>then</b>   |                        |
| 2. <b>return</b> Basis( $F \bmod x$ )   |                        |
| 3. <b>else</b>  |                        |
| 4. $P_{\text{low}} := \text{PM-Basis}(F, \lfloor \sigma/2 \rfloor)$                           | First subproblem       |
| 5. $F' := \text{MiddleProduct}(P_{\text{low}}, F, \lfloor \sigma/2 \rfloor \dots \sigma - 1)$ | Update problem         |
| 6. $P_{\text{high}} := \text{PM-Basis}(F', \lceil \sigma/2 \rceil)$                           | Second subproblem      |
| 7. <b>return</b> $P_{\text{high}} \cdot P_{\text{low}}$                                       | Solve original problem |

In terms of polynomial multiplication, binary multiplication tree.



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### Algorithm - Wiedemann '86

**Input:** A sparse matrix  $A \in \mathcal{M}_N(\mathbb{K})$

**Output:** Its minimal polynomial

1. Choose random  $u, v \in \mathbb{K}^{N \times 1}$
2. Compute the sequence  $S_i = {}^t u A^i v \in \mathbb{K}$  for  $i = 0 \dots 2N$
3. Return the minimal generating polynomial of the linear recursive sequence  $S$   
(using Berlekamp-Massey / Padé approximants)

### Applications :

- sparse linear system solving
- rank, determinant computation of sparse matrices

**Algorithm - Block Wiedemann [Coppersmith '94]****Input:** A sparse matrix  $A \in \mathcal{M}_N(\mathbb{K})$ **Output:** Its minimal polynomial

1. Choose random  $U, V \in \mathbb{K}^{N \times m}$
2. Compute the sequence  $S_i = U^t A^i V \in \mathbb{K}^{m \times m}$  for  $i = 0 \dots 2N/m^{**} + O(1)?^{**}$
3. Compute the *left matrix generating polynomial*  $\Pi$  of the recursive sequence  $S$

$$\forall j, \quad \sum_{i=0}^d \Pi_i S_{i+j} = 0^{m \times m} \quad \text{with } \Pi = \sum_{i=0}^d \Pi_i x^i$$

(using **Order basis** / matrix-type Padé approximants)

4. Return the minimal polynomial of  $A$  (using the matrix generating polynomial)

**Advantages of block Wiedemann algorithm :**

- Better probability of success if  $\mathbb{K}$  is a small field
- Enable parallelization of the algorithm (step 2)

## Algorithm - Block Wiedemann [Coppersmith '94]

1. Choose random  $U, V \in \mathbb{K}^{N \times m}$
2. Compute the sequence  $S_i = {}^t U A^i V \in \mathbb{K}^{m \times m}$  for  $i = 0 \dots 2N/m$
3. Compute the *left matrix generating polynomial* of  $S$  using order basis
4. return the minimal polynomial of  $A$

### Problem :

The bound  $2N/m$  is general but loose + Step 2 has **dominant cost**

↪ Early termination strategy

### Pros and cons of order basis algorithms ?

- for M-Basis
- for PM-Basis

**Algorithm - Block Wiedemann using M-Basis with early termination**

1. Choose random  $U, V \in \mathbb{K}^{N \times m}$
2. **for**  $i = 0 \dots 2N/m$ 
  - a. Update  $S$  from  $[S_0, \dots, S_{i-1}]$  to  $[S_0, \dots, S_i]$
  - b. Update order basis from order  $i-1$  to order  $i$  using M-Basis
  - c. **if** StopCriteria( $S$ , order basis) **then** break
3. return the minimal polynomial of  $A$

\*\*Will stop at  $i = \sigma_{KV} := \lceil \mu/m \rceil + \lceil \mu/n \rceil + O(1)$  - careful with formula : what is  $\mu$ ?\*\*

**Pros and cons of block Wiedemann using M-Basis:**

M-Basis

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Early termination possible  
Minimal knowledge required on  $S$

Slow Step 2.b (non negligible)

**Algorithm - Block Wiedemann using PM-Basis with early termination**

1. Choose random  $U, V \in \mathbb{K}^{N \times m}$
2. **for**  $\ell = 0 \dots \lceil \log_2(2N/m) \rceil$  \\ Assume  $2N/m = 2^k$ 
  - a. Update  $S$  from  $[S_0, \dots, S_{2^{\ell-1}-1}]$  to  $[S_0, \dots, S_{2^\ell-1}]$
  - b. Update order basis from order  $2^{\ell-1}$  to order  $2^\ell$  using PM-Basis
  - c. **if** StopCriteria( $S$ , order basis) **then** break
3. return the minimal polynomial of  $A$

**Pros and cons of block Wiedemann using PM-Basis:**

PM-Basis

---

Restrictive early termination (recursive algo)  
 May require more knowledge on  $S$

$\rightsquigarrow$  Let's start by an iterative PM-Basis  
 for better early termination

Fast Step 2.b (negligible)

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**First objective :**

Transform recursive PM-Basis into iterative algorithm

**Why ?** Enable early termination

**Simpler subproblem :****Algorithm RecursiveAlgo**

1. **if**  $\sigma = 1$  **then** Base Case
2. **else**
3.   Recursive Call 1
4.   Update input
5.   Recursive Call 2
6. **return** Multiply answers

→

**Algorithm BinaryMultiplicationTree**

1. **if**  $\sigma = 1$  **then** Base Case
2. **else**
3.   Recursive Call 1
4.   Recursive Call 2
5. **return** Multiply answers

**On the blackboard :**

Binary multiplication trees and their iterative version ( $\sigma = 2^k$ )



### Algorithm BMT

**Input:**  $M = [M_0, \dots, M_{2^k-1}]$

**Output:**  $M_{2^k-1} \cdots M_0$

1. **if**  $\#M = 1$  **then** return  $M[0]$
2. **else**
3.  $P_l = \text{BMT}([M_0, \dots, M_{2^{k-1}-1}])$
4.  $P_h = \text{BMT}([M_{2^{k-1}}, \dots, M_{2^k-1}])$
5. **return**  $P_h \cdot P_l$

→

### Algorithm iBMT

**Input:**  $M = [M_0, \dots, M_{2^k-1}]$

**Output:**  $M_{2^k-1} \cdots M_0$

1.  $P = []$
2. **for**  $i = 0 \dots 2^k - 1$  **do**
3. Add  $M_k$  at the beginning of  $P$
4. **for**  $i = 1 \dots \nu_2(k)$
5. Merge  $P[0], P[1]$  by multiplication
6. **return**  $P[0]$

Remarks : At step  $2^k$ , we have the product. Otherwise, only chunks of the product. These chunks are enough for what we need.

Derecursion of PM-Basis when  $\sigma = 2^r$  :**Algorithm PM-Basis****Input** :  $F \in \mathbb{K}[x]^{m \times n}, \sigma \in \mathbb{N}$ **Output** : A  $(F, \sigma)$  order basis  $P$  of  $F$ 1. **if**  $\sigma = 1$  **then**2.   **return** Basis( $F \bmod x$ )3. **else**4.    $P_{\text{low}} := \text{PM-Basis}(F, \lfloor \sigma/2 \rfloor)$ 

First subproblem

5.    $F' := \text{MiddleProduct}(P_{\text{low}}, F, \lfloor \sigma/2 \rfloor \dots \sigma - 1)$ 

Update problem

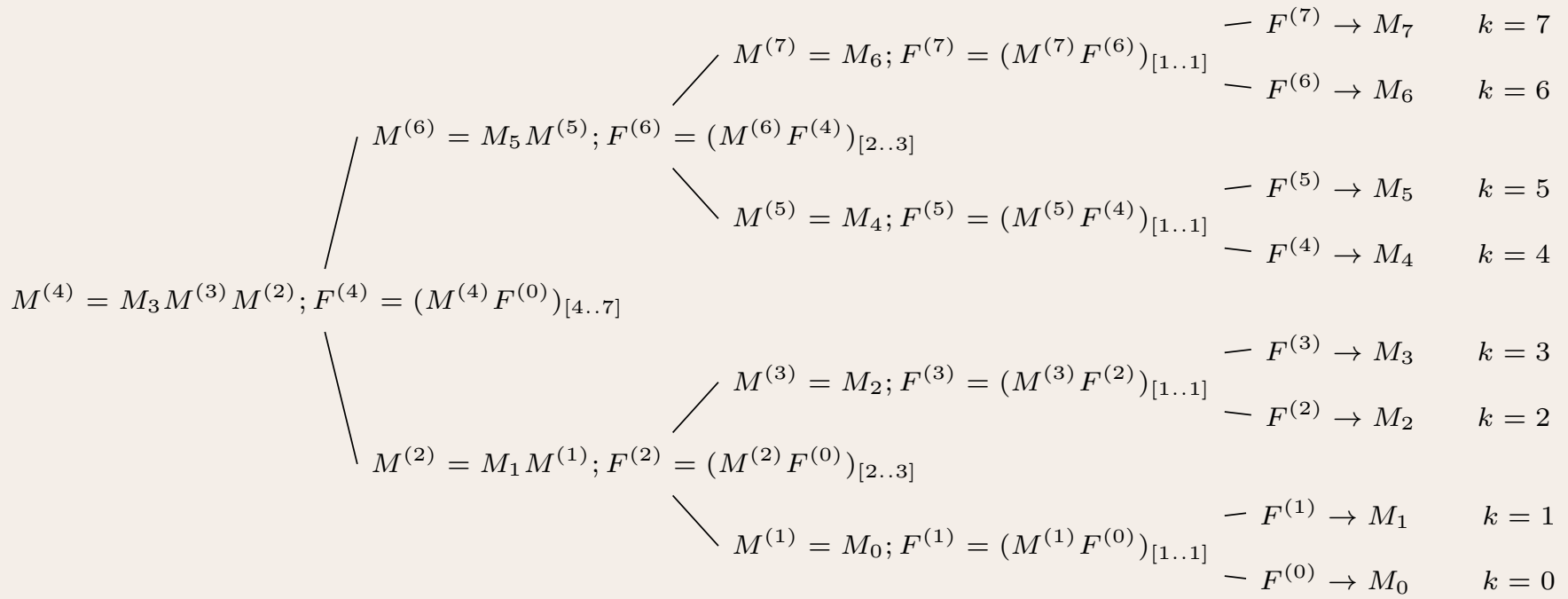
6.    $P_{\text{high}} := \text{PM-Basis}(F', \lceil \sigma/2 \rceil)$ 

Second subproblem

7. **return**  $P_{\text{high}} \cdot P_{\text{low}}$ 

Solve original problem

**On the blackboard** : Computation tree of PM-Basis for  $\sigma = 4$ Step  $i$  : Computations until  $M_i$  and after Step  $i - 1$



**Figure.** Computation tree of PM-Basis

**Definition** : If  $\deg(A) \leq d$ , then  $\text{MP}(A, B, d \dots h) := (A B)_{d \dots h}$ .

**Computations of PM-Basis( $F, 2^k$ ) :**

1.  $M_0 = \text{Basis}(F_0)$  Step 0

---

2.  $F^{(1)} = \text{MP}(M_0, F, 1\dots 1)$  Step 1

---

3.  $M_1 = \text{Basis}((F^{(1)})_0)$

---

4.  $M^{(2)} = M_1 \cdot M_0$
5.  $F^{(2)} = \text{MP}(M^{(2)}, F, 2\dots 3)$  Step 2

---

6.  $M_2 = \text{Basis}((F^{(2)})_0)$

---

7.  $F^{(3)} = \text{MP}(M_2, F^{(1)}, 1\dots 1)$  Step 3

---

8.  $M_3 = \text{Basis}((F^{(3)})_0)$

---

9.  $M^{(4)} = M_3 \cdot M_2$
10.  $M^{(4)} = M^{(4)} \cdot M^{(2)}$  Step 4
11.  $F^{(3)} = \text{MP}(M^{(4)}, F, 4\dots 7)$
12.  $M_4 = \text{Basis}((F^{(4)})_0)$

### Algorithm iPM-Basis - Step $k$

1.  $v = \nu_2(k)$
2. One step of iterative multiplication tree on  $M_i$
3.  $F^{(k)} = \text{MP}(M^{(k)}, F^{(k-2^v)}, 2^v \dots 2^{v+1} - 1)$
4.  $M_k = \text{Basis}(F^{(k)} \bmod x)$



**Algorithm - Block Wiedemann using iPM-Basis with early termination**

1. Choose random  $U, V \in \mathbb{K}^{N \times m}$
2. **for**  $\ell = 0 \dots \lceil \log_2(2N/m) \rceil$  \\ Assume  $2N/m = 2^k$ 
  - a. Update  $S$  from  $[S_0, \dots, S_{2^{\ell-1}-1}]$  to  $[S_0, \dots, S_{2^\ell-1}]$
  - b. **for**  $i = 2^{\ell-1} \dots 2^\ell - 1$ 
    - i. Update order basis from order  $i - 1$  to order  $i$  using iPM-Basis
    - ii. **if** StopCriteria( $S$ , order basis) **then** break
3. return the minimal polynomial of  $A$

**Problem :**

The bound  $2N/m$  is general but loose + Step 2 has **dominant cost**

**Pros and cons of order basis algorithms**

M-Basis	PM-Basis	iPM-Basis
Early termination Minimal knowledge on $S$	No early termination More knowledge on $S$	Early termination More knowledge on $S$
Slow Step 3	Fast Step 3	Fast Step 3

**At stake :** We could gain a constant factor up to 2.

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**Definition. (Online algorithm  $\sim$  minimal knowledge on the input)**

Let  $F = \sum_{i \in \mathbb{N}} F_i x^i \in \mathbb{K}[[x]]^{m \times n}$ .

An order basis algorithm is **online** if it reads at most  $F_0, \dots, F_k$  when computing  $M_k$ .

**Example :**

- M-Basis is online:

Computation of  $M_k$  requires  $(x^{-k} P_{k-1} F) \bmod x$ , which involve only  $F_0, \dots, F_k$ .

- iPM-Basis (and PM-Basis) are off-line

Step  $2^k$ :  $F^{(2^k)} = \text{MP}(M^{(2^k)}, F, 2^k \dots 2^{k+1} - 1)$  requires  $F_0, \dots, F_{2^{k+1}-1}$  !

**Definition.** (Online algorithm  $\sim$  minimal knowledge on the input)

Let  $F = \sum_{i \in \mathbb{N}} F_i x^i \in \mathbb{K}[[x]]^{m \times n}$ .

An order basis algorithm is **online** if it reads at most  $F_0, \dots, F_k$  when computing  $M_k$ .

**Goal :** Turn iPM-Basis into an online algorithm

#### Algorithm iPM-Basis - step $k$

1.  $v = \nu_2(k)$
2. One step of iterative multiplication tree on  $M_i$
3. Update the problem  $F^{(k)} = \text{MP}(M^{(k)}, F^{(k-2^v)}, 2^v \dots 2^{v+1} - 1)$
4.  $M_k = \text{Basis}(F^{(k)} \bmod x)$

**On the blackboard :**

- iPM-Basis middle products and how to make them online
- Necessity of an **online** middle product

**Definition :** If  $\deg(A) \leq d$ , then  $MP(A, B, d\dots h) := (A B)_{d\dots h}$ .

**Definition.** *An middle product algorithm is shifted online if at each step it requires minimal knowledge on  $A$  and  $B$ .*

*In practice, the  $i$ th coefficient of  $MP(A, B, d\dots h)$  must use at most  $A_0, \dots, A_{d+i}$  and  $B_0, \dots, B_{d+i}$ .*

**Example:**  $MP(A, B, 3\dots 6) = MP(A_{0\dots 3}, B_{0\dots 6}, 3\dots 6)$  on the blackboard

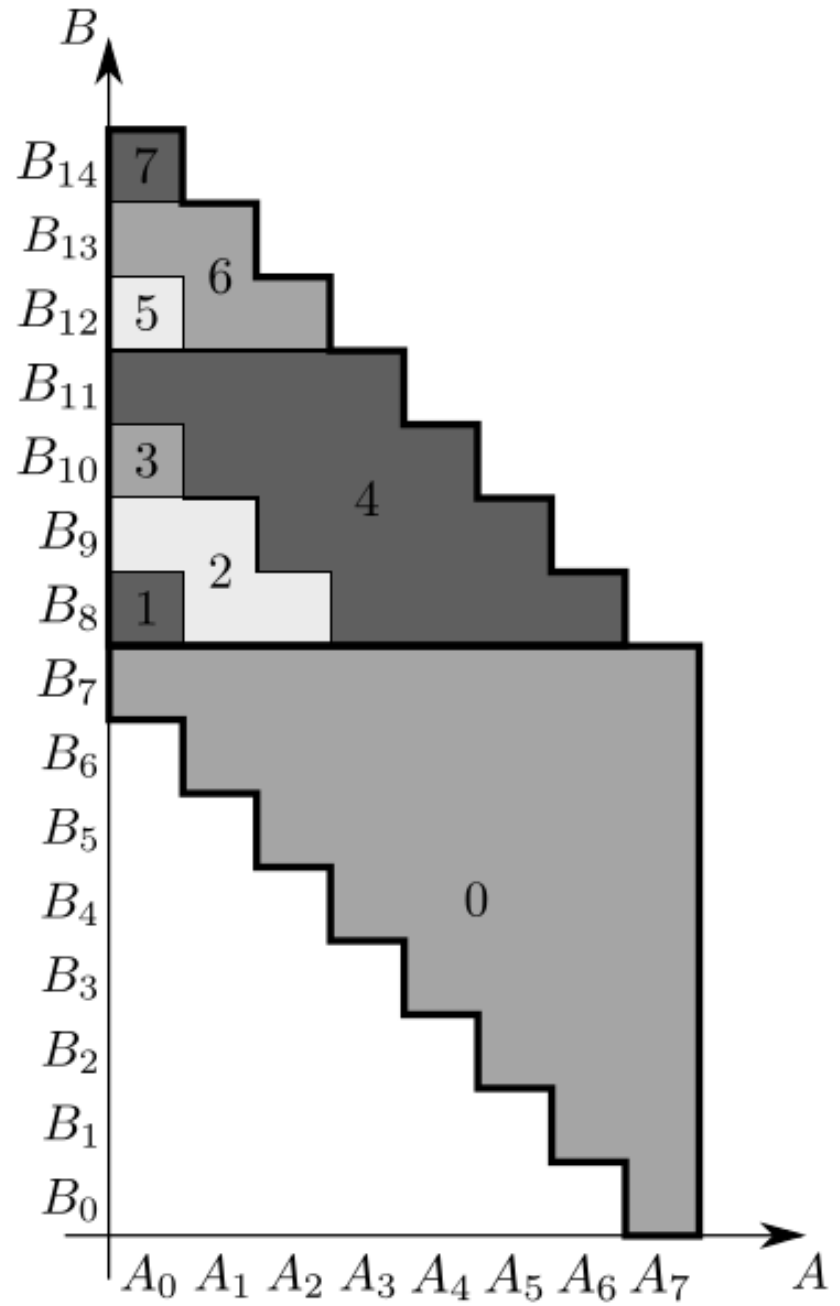
# Shifted online middle product

**Another example:**  $MP(A_{0\dots7}, B_{0\dots14}, 7\dots14)$

Shifted online algorithm :

Step  $i$  can read only  $B_0, \dots, B_{i+7}$

Step	Computation
0	$C = (A B_{0\dots7})_{7\dots14}$
1	$C += x \text{ MP}(A_0, B_8, 0)$
2	$C += x^2 \text{ MP}(A_{0\dots2}, B_{8\dots9}, 1\dots2)$
3	$C += x^3 \text{ MP}(A_0, B_{10}, 0)$
4	$C += x^4 \text{ MP}(A_{0\dots6}, B_{8\dots11}, 3\dots6)$
5	$C += x^5 \text{ MP}(A_0, B_{12}, 0)$
6	$C += x^6 \text{ MP}(A_{0\dots2}, B_{12\dots13}, 1\dots2)$
7	$C += x^7 \text{ MP}(A_0, B_{14}, 0)$



**Goal :** Turn iPM-Basis into an online algorithm

### Algorithm oPM-Basis - step $k$

1.  $v = \nu_2(k)$
2. One step of iterative multiplication tree on  $M_i$
3. Compute one more term of previous online middle products
4. Compute first term of  $F^{(k)} = \text{onlineMP}(M^{(k)}, F^{(k-2^v)}, 2^v \dots 2^{v+1} - 1)$
5.  $M_k = \text{Basis}(F^{(k)} \bmod x)$

**Theorem.** oPM-Basis is an online order basis algorithm which is quasi-linear in the order  $\sigma$ .

## Algorithm - Block Wiedemann using oPM-Basis with early termination

1. Choose random  $U, V \in \mathbb{K}^{N \times m}$
2. **for**  $i = 0 \dots 2N/m$ 
  - a. Update  $S$  from  $[S_0, \dots, S_{i-1}]$  to  $[S_0, \dots, S_i]$
  - b. Update order basis from order  $i-1$  to order  $i$  using oPM-Basis
  - c. **if** StopCriteria( $S$ , order basis) **then** break
3. return the minimal polynomial of  $A$

### Pros and cons of order basis algorithms in block Wiedemann :

M-Basis	PM-Basis	oPM-Basis
Early termination Minimal knowledge on $S$	No early termination More knowledge on $S$	Early termination Minimal knowledge on $S$
Slow Step 3	Fast Step 3	Fast Step 3

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**Related work : Polynomial matrix multiplication**

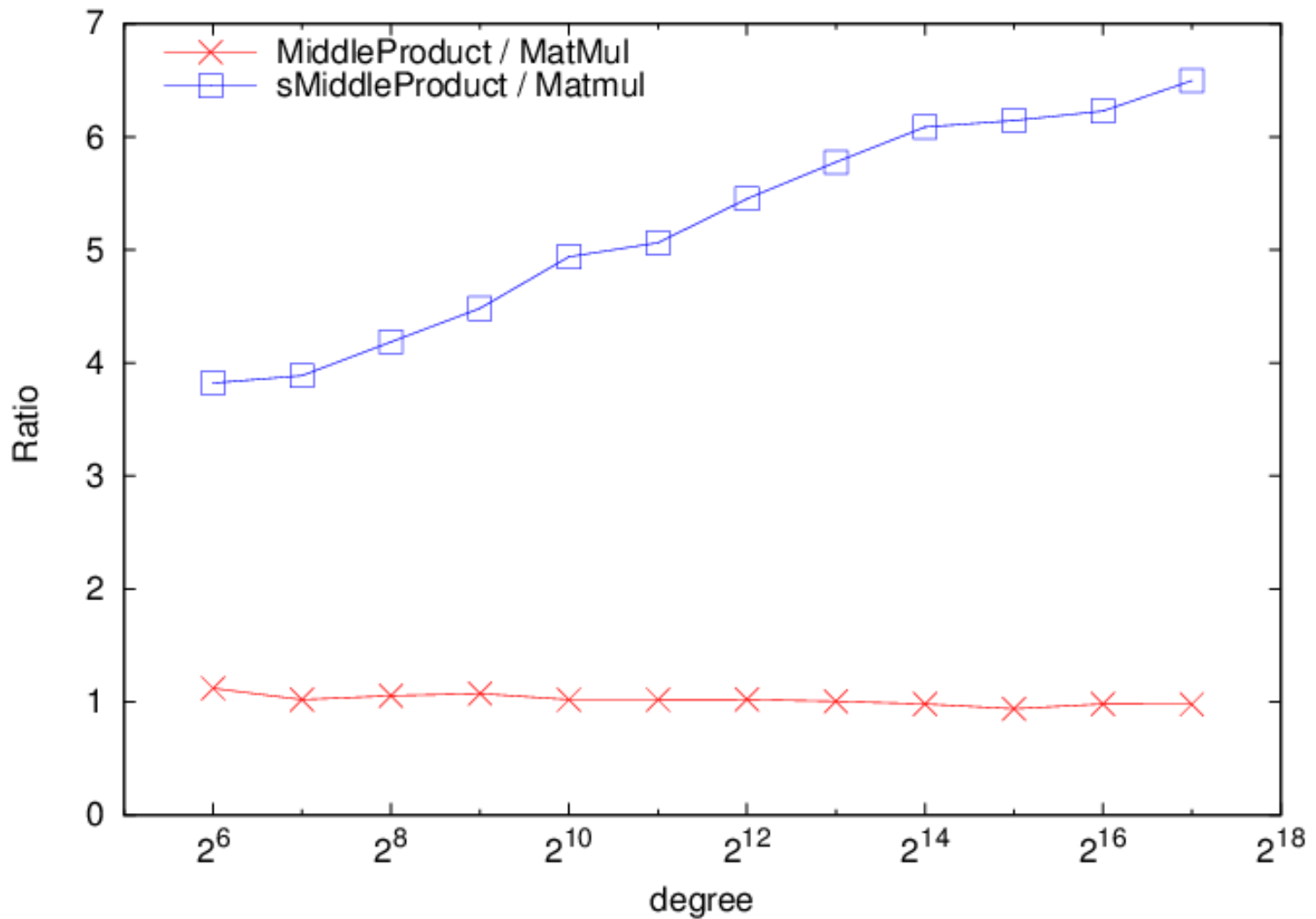
Tailored implementation for multi-core architectures

- Polynomial aspects : DFT cache friendly, SSE4 instructions
- Matrix aspects : use of BLAS, two data structures

$m, d$	MMX	FLINT 2.4	our code
16, 1024	<b>0.02 s</b>	0.26 s	0.03 s
16, 2048	<b>0.04 s</b>	0.70 s	0.06 s
16, 4096	<b>0.09 s</b>	1.68 s	0.13 s
16, 8192	<b>0.20 s</b>	4.52 s	0.28 s
128, 512	1.00 s	26.21 s	<b>0.82 s</b>
256, 256	4.00 s	36.71 s	<b>1.75 s</b>
512, 512	69.19 s	465.66 s	<b>19.64 s</b>
1024, 64	71.36 s	115.52 s	<b>13.95 s</b>
2048, 32	298.27 s	263.88 s	<b>48.90 s</b>

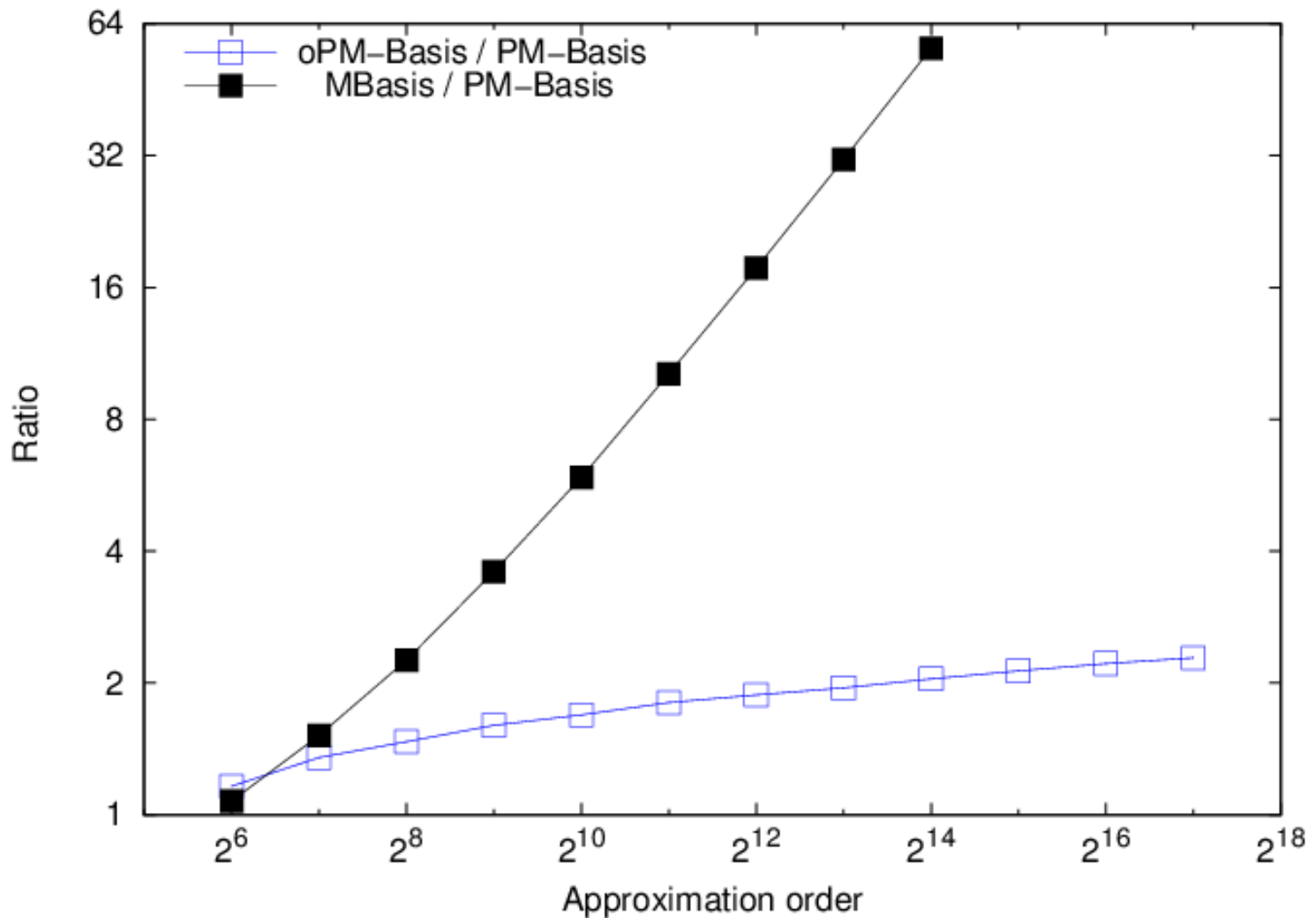
**Table.** Computation times of polynomial matrix multiplication in  $\mathbb{F}_p[x]^{m \times m}$  with degree  $d$  and  $p$  a 23-bit FFT prime compared to Mathemagix (MMX) and FLINT.





**Figure.** Relative performance of different middle products against polynomial matrix multiplication over  $\mathbb{F}_p[x]^{16 \times 16}$  with  $p$  a 23-bit FFT prime

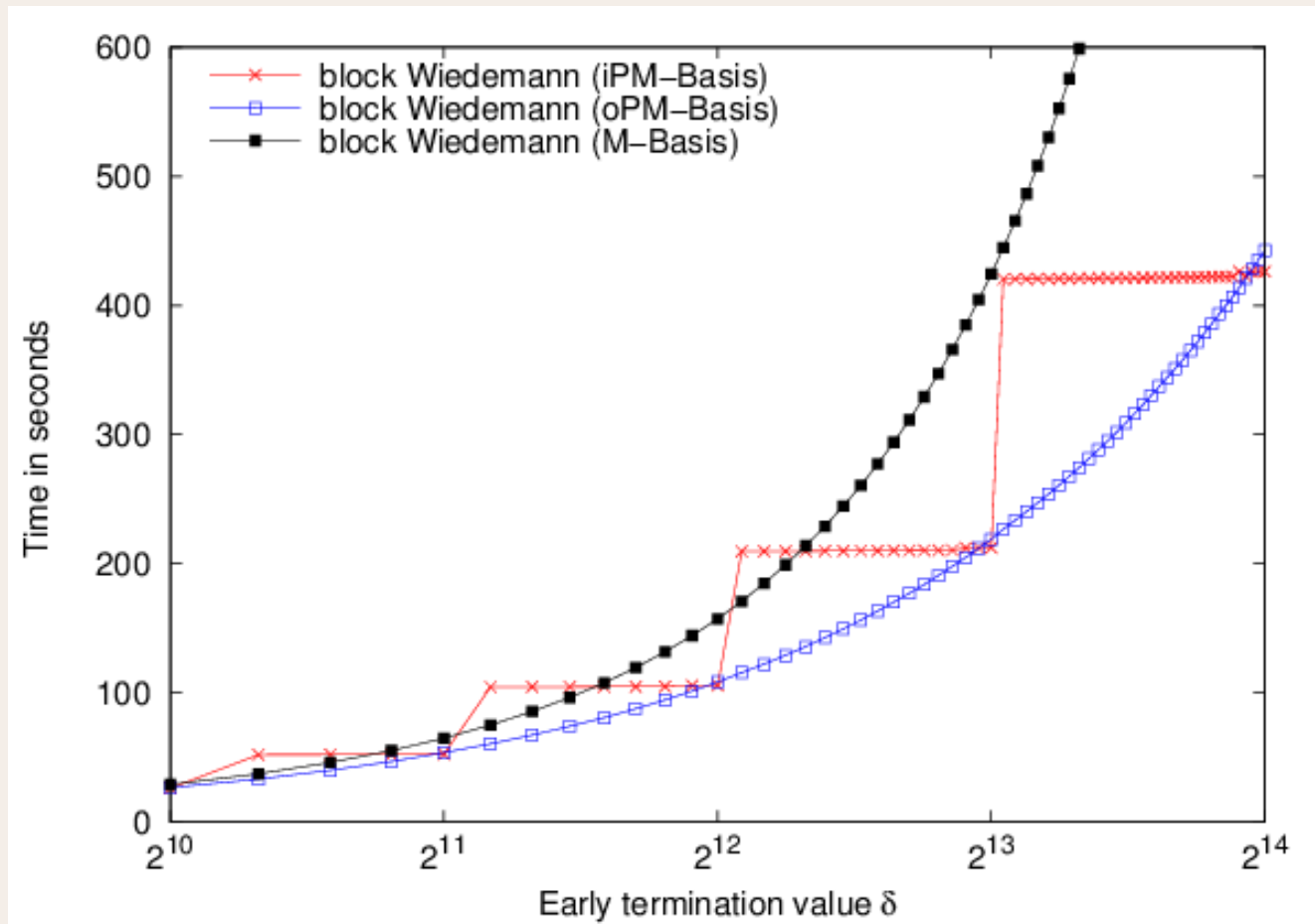




**Figure.** Relative performance of different middle products against polynomial matrix multiplication over  $\mathbb{F}_p[x]^{16 \times 16}$  with  $p$  a 23-bit FFT prime

# Timings

**Setting** :  $A \in GL_{2^{17}}(\mathbb{F}_p)$  with 20 elements / row,  $m = 16$ , loose bound  $\sigma = 2^{14}$



**Figure.** Computation times of early termination in block Wiedemann using order basis algorithm.