

A New Black Box Factorization Algorithm - the Non-monic Case

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The sparse and black box representation of a polynomial

The **black box representation** of $f \in \mathbb{Z}[x_1, \dots, x_n]$ is a **program** that accepts a prime p and an evaluation point $\alpha \in \mathbb{Z}_p^n$ and outputs $f(\alpha) \bmod p$.

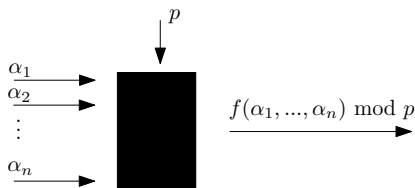


Figure: A modular black box for $f \in \mathbb{Z}[x_1, \dots, x_n]$.

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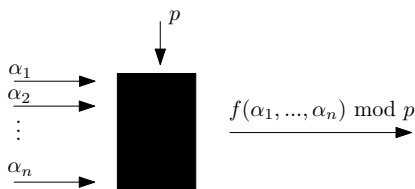


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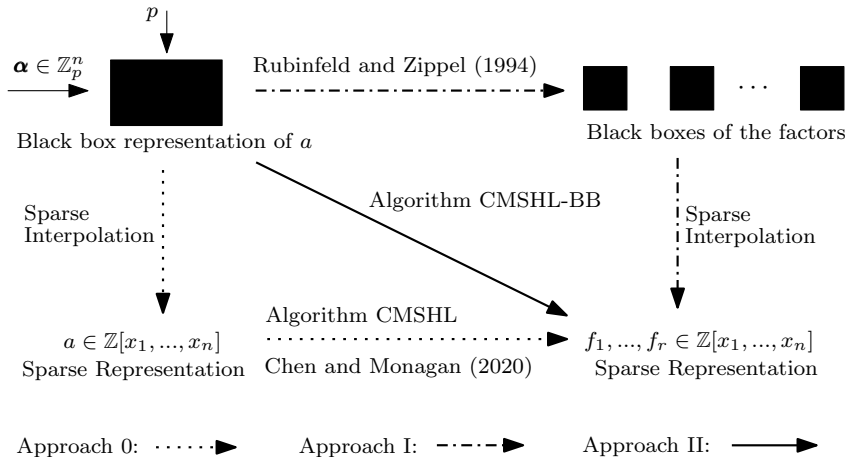
The **sparse representation** of $f \in \mathbb{Z}[x_1, \dots, x_n]$ consists of a list of coefficients $c_k \in \mathbb{Z}, c_k \neq 0$ and exponents $(e_{k_1}, \dots, e_{k_n}) \in \mathbb{N}^n$ such that

$$f = \sum_{k=1}^t c_k \cdot x_1^{e_{k_1}} \cdots x_n^{e_{k_n}},$$

where t is the number of non-zero terms of f .

Factoring $a \in \mathbb{Z}[x_1, \dots, x_n]$ represented by a black box

Given a polynomial $a \in \mathbb{Z}[x_1, \dots, x_n]$ represented by a black box, we aim to compute its factors in the sparse representation.



Previous work on multivariate polynomial factorization

Sparse Hensel lifting

- Yun (1974), Wang (1975), (1978): Multivariate Hensel lifting (MHL).
Recovers the factors one variable at a time. Solves MDP $\sigma_i g_{j-1} + \tau_i f_{j-1} = c_i$ for $\sigma_i, \tau_i \in \mathbb{Z}_p[x_1, \dots, x_{j-1}]$ one variable at a time (can be exponential in n).
- Zippel (1981), Kaltofen (1985): Sparse Hensel lifting (SHL).
- Monagan and Tuncer (2016): MTSHL. Solves MDP by sparse interpolation.
- Monagan and Tuncer (2018): Use bivariate Hensel lifts to compute σ_i, τ_i .
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Black box factorization

- Kaltofen and Trager (1990): Outputs black boxes of the factors.
- Rubinfeld and Zippel (1994): For factoring $a \in \mathbb{Z}[x_1, \dots, x_n]$.
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Other work

- Huang and Gao (2023): Non-sparse Hensel lifting
- Lecerf (2007): Dense multivariate polynomial factorization

Our Contributions

A new black box factorization algorithm CMSHL-BB:

- Accepts all cases of input polynomials, i.e. **non-monic**, **non-square-free** and **non-primitive** cases.
- A Maple + C hybrid implementation with timing benchmarks.
- A worst case complexity analysis with failure probabilities (Monte Carlo).

Example 1: Computing the determinant of a Toeplitz matrix

Let T_n be an $n \times n$ symmetric Toeplitz matrix

$$T_n = \begin{pmatrix} x_1 & x_2 & x_3 & \cdots & x_n \\ x_2 & x_1 & x_2 & & \\ x_3 & x_2 & x_1 & & \\ \vdots & & & \ddots & \vdots \\ x_n & & & \cdots & x_1 \end{pmatrix}.$$

For example,

$$\det(T_4) = (x_1^2 - x_1x_2 - x_1x_4 - x_2^2 + 2x_2x_3 + x_2x_4 - x_3^2)(x_1^2 + x_1x_2 + x_1x_4 - x_2^2 - 2x_2x_3 + x_2x_4 - x_3^2).$$

n	$\# \det(T_n)$	$\# f_i$	s
8	1628	167, 167	38
9	6090	294, 153	50
10	23797	931, 931	229
11	90296	1730, 849	337
12	350726	5579, 5579	1465
13	1338076	10611, 4983	2297
14	5165957	34937, 34937	9705
15	19732508	66684, 30458	34081
16	—	221854, 221854	127690

Table: Number of terms of $\det(T_n)$ and its factors. s is the maximum number of bivariate images [Chen and Monagan (2022)].

Algorithm CMSHL-BB (Approach II):

- Space efficient since $\#f_i \ll \# \det(T_n)$.
- Less probes to the black box than Rubinfeld and Zippel's algorithm since $s < \#f_{\max}$.

Example 2: Non-monic case

$$B = \begin{bmatrix} uvw & v & uvw + v + w & \dots & uvw + v \\ v & uvw & uvw + 2v & \dots & uvw + v \\ w & v & uvw + v + w & \dots & v + w \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ w & v & uvw + v + w & \dots & 2vwx + 2ux + 3v + 4w \end{bmatrix}.$$

$$\begin{aligned} a = \det(B) &= -(-v^2w^2x^2 + uvwx^2 + vw^2x - uwx + v^2 - 2vw + w^2) \\ &\quad (v^2w^2x^2 + uvwx^2 + vw^2x + uwx - v^2 - 2vw - w^2) \\ &\quad (u^2v^2w^2 + u^2vwx + uv^2w + uvx - v^2 - 2vw - w^2) \\ &\quad (u^2v^2w^2 - u^2vwx - uv^2w + uvx - v^2 + 2vw - w^2). \end{aligned}$$

$$\# \text{expand}(a) = 120.$$

$$\text{lcoeff}(a, u) = -v^6w^6x^4 + v^4w^4x^6 + v^4w^6x^2 - v^2w^4x^4,$$

$$\begin{aligned} \text{lcoeff}(a, v) &= u^4w^8x^4 - 2u^4w^6x^2 - 3u^2w^6x^4 + u^4w^4 + 6u^2w^4x^2 \\ &\quad + w^4x^4 - 3u^2w^2 - 2w^2x^2 + 1. \end{aligned}$$

Technicalities for our algorithm design

Algorithm CMSHL-BB:

- 1 Non-monic: Use **non-monic bivariate Hensel lifts (BHL)**, modified from Monagan and Paluck (2022). Cubic cost: $O(d_1^2 d_j + d_1 d_j^2)$.
- 2 Non-square-free: Compute the **square-free part** of the **bivariate images** of $a(x_1, \beta^k, x_j)$ with dense interpolation, gcd computation and division. Cost of gcd: $O(d_1^2 d_j + d_1 d_j^2)$ [Brown (1971)].
- 3 Non-primitive: Compute the content recursively after recovering the primitive factors.

How does our algorithm work?

Prior Hensel lifting steps:

- 1 Choose a large prime p , e.g. $p = 2^{62} - 57$ and a positive integer $\tilde{N} < p$.
- 2 Choose $\alpha = (\alpha_2, \dots, \alpha_n) \in \mathbb{Z}^{n-1}$ from $[1, \tilde{N} - 1]^{n-1}$ randomly s.t.
 - α is Hilbertian
 - α satisfies the weak SHL assumption (Lemma 2.2)
- 3 Compute $a(x_1, \alpha)$ from the black box **B** with Chinese remaindering.
- 4 Factor $a(x_1, \alpha)$ over \mathbb{Z} as follows.

Let the factorization of a over \mathbb{Z} be of the form

$$a = hf_1^{e_1} f_2^{e_2} \cdots f_r^{e_r} \in \mathbb{Z}[x_1, \dots, x_n]$$

where $\deg(f_\rho, x_1) > 0$ ($1 \leq \rho \leq r$), f_ρ is irreducible over \mathbb{Z} and h is the content of a in x_1 . Then, with high probability (w.h.p.),

$$a(x_1, \alpha) = \hat{h} \hat{f}_1^{e_1} \hat{f}_2^{e_2} \cdots \hat{f}_r^{e_r} \in \mathbb{Z}[x_1],$$

where $\hat{f}_\rho(x_1, \alpha) = (1/\lambda_\rho) f_\rho(x_1, \alpha)$ for some $\lambda_\rho \in \mathbb{Z}$ and \hat{f}_ρ is irreducible in $\mathbb{Z}[x_1]$ ($1 \leq \rho \leq r$).

How does our algorithm work?

Definition

The **square-free part** of a is defined as

$$\text{sqf}(a) := f_1 f_2 \cdots f_r = \frac{a}{\gcd(a, \partial a / \partial x_1)}.$$

Let $\hat{f}_{\rho,1} := \hat{f}_{\rho}(x_1, \alpha) \bmod p$ and \mathbf{B} be the black box representation of a .

Algorithm CMSHL-BB (non-monic and non-square-free):

- Input: A prime p , the black box \mathbf{B} , $\alpha \in \mathbb{Z}^{n-1}$, $\deg(a, x_j)$ ($1 \leq j \leq n$)

(pre-computed), $\hat{f}_{\rho,1} \in \mathbb{Z}_p[x_1]$ ($1 \leq \rho \leq r$) s.t.

(i) $\gcd(\hat{f}_{k,1}, \hat{f}_{l,1}) = 1$ for $k \neq l$ in $\mathbb{Z}_p[x_1]$,

(ii) $\text{sqf}(a(x_1, \alpha)) = \prod_{\rho=1}^r \lambda_{\rho} \prod_{\rho=1}^r \hat{f}_{\rho,1} \bmod p$.

- Output: $\hat{f}_{\rho,n} \in \mathbb{Z}_p[x_1, \dots, x_n]$ ($1 \leq \rho \leq r$) s.t.

$$\text{sqf}(a(x_1, \dots, x_n)) = \prod_{\rho=1}^r \lambda_{\rho} \prod_{\rho=1}^r \hat{f}_{\rho,n} \bmod p. \quad \text{Or FAIL.}$$

Finally, use **rational number reconstruction** to recover the integer coefficients to

get $f_{\rho} \in \mathbb{Z}[x_1, \dots, x_n]$ ($1 \leq \rho \leq r$).

Algorithm CMSHL-BB: the j^{th} Hensel lifting step

Define $\hat{f}_{\rho,j} := \hat{f}_{\rho}(x_1, \dots, x_j, \alpha_{j+1}, \dots, \alpha_n) \bmod p$ for $2 \leq j \leq n$ (to be computed).

- 1: Let $\hat{f}_{\rho,j-1} = \sum_{i=0}^{\text{df}_{\rho}} \sigma_{\rho,i}(x_2, \dots, x_{j-1})x_1^i$ (for $1 \leq \rho \leq r$)
where $\sigma_{\rho,i} = \sum_{k=1}^{s_{\rho,i}} c_{\rho,ik} M_{\rho,ik}$ with $M_{\rho,ik}$ the monomials in $\sigma_{\rho,i}$ and $\text{df}_{\rho} = \deg(\hat{f}_{\rho,j-1}, x_1)$.
- 2: Pick $\beta = (\beta_2, \dots, \beta_{j-1}) \in \mathbb{Z}_p^{j-2}$ at random.
- 3: Evaluate (for $1 \leq \rho \leq r$): $S_{\rho} = \{S_{\rho,i} = \{m_{\rho,ik} = M_{\rho,ik}(\beta), 1 \leq k \leq s_{\rho,i}\}, 0 \leq i \leq \text{df}_{\rho}\}$.
- 4: **if** any $|S_{\rho,i}| \neq s_{\rho,i}$ **then return FAIL end if**
- 5: Let s be the maximum of $s_{\rho,i}$.
- 6: **for** k from 1 to s **do**
- 7: Let $Y_k = (x_2 = \beta_2^k, \dots, x_{j-1} = \beta_{j-1}^k)$.
- 8: $A_k \leftarrow a_j(x_1, Y_k, x_j) \in \mathbb{Z}_p[x_1, x_j]$. $\dots \mathcal{O}(s(d_1^2 d_j + d_1 d_j^2 + d_1 d_j C(\text{probe B})))$
- 9: **if** $\deg(A_k, x_1) \neq d_1$ **or** $\deg(A_k, x_j) \neq d_j$ **then return FAIL end if**
- 10: $g_k \leftarrow \gcd(A_k, \frac{\partial A_k}{\partial x_1}) \bmod p$. $\dots \mathcal{O}(s(d_1^2 d_j + d_1 d_j^2))$
- 11: **if** $\deg(g_k, x_1) \neq d_1 - \sum_{\rho=1}^r \text{df}_{\rho}$ **then return FAIL end if**
- 12: $A_{\text{sf}} \leftarrow \text{quo}(A_k, g_k) \bmod p$.
- 13: $A_{\text{sfm}} \leftarrow A_{\text{sf}} / (\text{lc}(\text{lc}(A_{\text{sf}}, x_1), x_j)) \bmod p$.
- 14: $F_{\rho,k} \leftarrow \hat{f}_{\rho,j-1}(x_1, Y_k) \in \mathbb{Z}_p[x_1]$ for $1 \leq \rho \leq r$.
- 15: **if** any $\deg(F_{\rho,k}) < \text{df}_{\rho}$ (for $1 \leq \rho \leq r$) **then return FAIL end if**
- 16: **if** $\gcd(F_{\rho,k}, F_{\phi,k}) \neq 1$ for any $\rho \neq \phi$ ($1 \leq \rho, \phi \leq r$) **then return FAIL end if**
- 17: $\hat{f}_{\rho,k} \leftarrow \text{BivariateHenselLift}(A_{\text{sfm}}(x_1, x_j), F_{\rho,k}(x_1), \alpha_j, p)$.
 $\dots \mathcal{O}(s(\tilde{d}_1 \tilde{d}_j^2 + \tilde{d}_1^2 \tilde{d}_j)) \subseteq \mathcal{O}(s(d_1 d_j^2 + d_1^2 d_j))$
- 18: **end for**

Algorithm CMSHL: the j^{th} Hensel lifting step

- 19: Let $\hat{f}_{\rho,k} = \sum_{l=1}^{t_\rho} \alpha_{\rho,kl} \tilde{M}_{\rho,l}(x_1, x_j) \in \mathbb{Z}_p[x_1, x_j]$ for $1 \leq k \leq s$
where $t_\rho = \#\hat{f}_{\rho,k}$ (for $1 \leq \rho \leq r$).
- 20: **for** ρ from 1 to r **do**
- 21: **for** l from 1 to t_ρ **do**
- 22: $i \leftarrow \deg(\tilde{M}_{\rho,l}, x_1)$.
- 23: **Solve the linear system** $\left\{ \sum_{k=1}^{s_{\rho,i}} m_{\rho,ik}^t c_{\rho,lk} = \alpha_{\rho,tl} \text{ for } 1 \leq t \leq s_{\rho,i} \right\}$ for $c_{\rho,lk}$.
- 24: **end for** $\mathcal{O}(s \tilde{d}_j (\sum_{\rho=1}^r \#\hat{f}_{\rho,j-1}))$
- 25: $\hat{f}_{\rho,j} \leftarrow \sum_{l=1}^{t_\rho} \left(\sum_{k=1}^{s_{\rho,i}} c_{\rho,lk} M_{\rho,ik}(x_2, \dots, x_{j-1}) \right) \tilde{M}_{\rho,l}(x_1, x_j)$.
- 26: **end for**
- 27: Pick $\beta = (\beta_2, \dots, \beta_j) \in \mathbb{Z}_p^{j-1}$ at random.
- 28: $A_\beta \leftarrow \text{sqf}(a_j(x_1, \beta)) \bmod p$ // via probes to \mathbf{B} , interpolation, and sqrfree compt.
- 29: **if** $\hat{f}_{\rho,j}(x_1, \beta) \mid A_\beta$ **and** $\deg(\hat{f}_{\rho,j}(x_1, \beta)) = df_\rho$ (for $1 \leq \rho \leq r$) **then**
 return $\hat{f}_{\rho,j}$ (for $1 \leq \rho \leq r$)
else return FAIL
end if

Non-monic bivariate Hensel lift (BHL)

Input: prime p , $\alpha \in \mathbb{Z}_p$, $a \in \mathbb{Z}_p[x, y]$, $\hat{f}_{\rho,0} \in \mathbb{Z}_p[x]$ for $1 \leq \rho \leq r$ s.t.

- (i) a is primitive in x ,
- (ii) $a(y = \alpha) = \zeta \prod_{\rho=1}^r \hat{f}_{\rho,0}$, where $\zeta \in \mathbb{Z}_p$,
- (iii) $\gcd(\hat{f}_{k,0}, \hat{f}_{l,0}) = 1$ for $k \neq l$.

Output: $\hat{f}_{\rho} \in \mathbb{Z}_p[x, y]$ for $1 \leq \rho \leq r$ s.t.

- (i) $a = \zeta \prod_{\rho=1}^r \hat{f}_{\rho}$ and
- (ii) $\hat{f}_{\rho}(y = \alpha) = \hat{f}_{\rho,0}$.

Otherwise, **FAIL**.

BHL is called in Step 17 in CMSHL:

Input: p , α_j , $\text{sqf}(a_j(x_1, Y_k, x_j))$, $\hat{f}_{\rho,j-1}(x_1, Y_k)$ for $1 \leq \rho \leq r$ s.t.

$$\text{sqf}(a_j(x_1, x_j = \alpha_j)) = \left(\prod_{\rho=1}^r \lambda_{\rho}\right) \prod_{\rho=1}^r \hat{f}_{\rho,j-1}(x_1).$$

Output: $\hat{f}_{\rho,j}(x_1, x_j)$ for $1 \leq \rho \leq r$ s.t.

$$\text{sqf}(a_j(x_1, x_j)) = \left(\prod_{\rho=1}^r \lambda_{\rho}\right) \prod_{\rho=1}^r \hat{f}_{\rho,j}(x_1, x_j) \text{ and}$$

$$\hat{f}_{\rho,j}(x_1, x_j = \alpha_j) = \hat{f}_{\rho,j-1}(x_1).$$

Example

Example

Consider $a = f_1 f_2 \in \mathbb{Z}[x_1, \dots, x_4]$ where

$$f_1 = (2x_2^2 x_3^3 + 4)x_1^8 + (4x_2^2 x_3^3 + 22x_2^2 x_4^3 + 1452x_2^2 x_4)x_1 + x_2^2 x_3 x_4 - 4x_3,$$

$$f_2 = (3x_2 + 39x_4 + 3x_3)x_1^8 + (5x_2 x_3^2 x_4 + 33x_2 x_3 x_4^2)x_1^2 - 363x_4^2 + 44.$$

In this case, $h = 1$ (a has no content in x_1 , neither integer content) and $\text{sqf}(a) = a$. Let $\alpha = (2, 3, 9)$ and $p = 2^{31} - 1$,

$$\begin{aligned} a(x_1, \alpha) &= 80520x_1^{16} + 3706560x_1^{10} + \dots - 3430775304x_1 - 2818464 \\ &= \underbrace{4}_{\lambda_1} \underbrace{(55x_1^8 + 29214x_1 + 24)}_{\hat{f}_1} \underbrace{(366x_1^8 + 16848x_1^2 - 29359)}_{\hat{f}_2} \\ &= f_1(x_1, \alpha) f_2(x_1, \alpha). \end{aligned}$$

Example ctd..

After the 1st Hensel lifting step (a bivariate Hensel lift only),

$$\hat{f}_{1,2} = (1073741837x_2^2 + 1)x_1^8 + 1073749127x_2^2x_1 + 1610612742x_2^2 + 2147483644,$$

$$\hat{f}_{2,2} = (3x_2 + 360)x_1^8 + 8424x_2x_1^2 + 2147454288.$$

After the 2nd Hensel lifting step,

$$\begin{aligned}\hat{f}_{1,3} &= (1073741824x_2^2x_3^3 + 1)x_1^8 + (x_2^2x_3^3 + 1073749100x_2^2)x_1 \\ &\quad + 536870914x_2^2x_3 + 2147483646x_3,\end{aligned}$$

$$\hat{f}_{2,3} = (3x_2 + 3x_3 + 351)x_1^8 + (45x_2x_3^2 + 2673x_2x_3)x_1^2 + 2147454288.$$

The last Hensel lifting step outputs $\hat{f}_{\rho,4}$ ($\rho = 1, 2$) s.t.

$$a_4 = \text{sqf}(a_4) = (\lambda_1\lambda_2)\hat{f}_{1,4}\hat{f}_{2,4} \pmod p \quad \text{with}$$

$$\begin{aligned}\hat{f}_{1,4} &= (1073741824x_2^2x_3^3 + 1)x_1^8 + (x_2^2x_3^3 + 1073741829x_2^2x_4^3 \\ &\quad + 363x_2^2x_4)x_1 + 536870912x_2^2x_3x_4 + 2147483646x_3\end{aligned}$$

$$\begin{aligned}\hat{f}_{2,4} &= (3x_2 + 39x_4 + 3x_3)x_1^8 + (5x_2x_3^2x_4 + 33x_2x_3x_4^2)x_1^2 \\ &\quad + 2147483284x_4^2 + 44.\end{aligned}$$

Example ctd..

Rational number reconstruction in Maple:

```
> ff[1] := iratrecon(fhat[1,4], p);
```

$$\begin{aligned} ff_1 := & \frac{1}{2}x_2^2x_1^8x_3^3 + x_1^8 + x_1x_2^2x_3^3 + \frac{11}{2}x_2^2x_1x_4^3 + 363x_2^2x_1x_4 \\ & + \frac{1}{4}x_2^2x_3x_4 - x_3 \end{aligned}$$

λ_1 is the least common multiple of the denominators of coefficients of ff_1 .

Multiply ff_1 by $\lambda_1 = 4$, we get the true factor $f_1 \in \mathbb{Z}[x_1, \dots, x_n]$:

```
> f[1] := numer(ff[1]);
```

$$\begin{aligned} f_1 := & 2x_2^2x_1^8x_3^3 + 4x_1^8 + 4x_1x_2^2x_3^3 + 22x_2^2x_1x_4^3 + 1452x_2^2x_1x_4 \\ & + x_2^2x_3x_4 - 4x_3 \end{aligned}$$

A Hybrid Maple + C Implementation of Method II

CPU timings (in seconds) for our algorithm, compared with Maple and Magma's current best determinant and factorization algorithms.

n	4	5	6	7	8	9
$N = 2n$	8	10	12	14	16	18
$\#f_i$	7,7	12,7	32,32	56,30	167,167	153,294
	7,7	12,7	32,32	56,30	167,167	153,294
$\# \det(A)$	120	701	5162	79740	1716810	7490224
CMSHL total	0.092	0.257	0.972	3.618	19.677	40.219
total probes	721	2112	6453	19584	85189	145065
Maple det	0.057	0.455	7.880	382.80	> 64 gigs	-
Maple factor	0.140	0.109	0.326	1.270	-	-
Maple total	0.197	0.564	8.206	384.07	-	-
Magma det	0.140	1.680	6.290	594.60	> 3h	-
Magma factor	0.800	0.120	0.480	33.140	-	-
Magma total	0.940	1.800	6.770	627.74	-	-

More Timings for Large Matrices

Table: Timings (in seconds) for computing determinants of large matrices.

	heron3d	heron4d	robotarms	heron5d
n	7	11	8	16
$N \times N$	13×13	63×63	20×20	399×399
r	6	4	3	8
$\#f_i$	3,23,3,3,1,3	22,1,6,131	2124,4,7	823,130,22,3,3,3,3,1
e_i	1,2,1,1,7,1	2,37,7,4	1,4,4	8,8,20,46,46,46,1831
$\#\det(A)$	525	37666243	178053	-
$\max \lambda_\rho$	1	1	169	1
CMSHL tot	1.096	81.376	1083.335	155054.324
probes tot	8560	339840	540834	36008392
Maple det	0.614	O/M	N/A	N/A
Maple fac	0.084	O/M	N/A	N/A
Maple tot	0.698	-	-	-

$$\det(A_{heron3d}) = 64as^7(as - bs + cs)(as - bs - cs)(as + bs + cs)(as + bs - cs) \underbrace{(as^4es^2 + as^2bs^2cs^2 - \dots - cs^2es^2fs^2 + 144vo^2)^2}_{23 \text{ terms}}$$

More Timings for Large Matrices

Table: Breakdown of timings (in seconds) at H.L. x_n .

	heron3d	heron4d	robotarms	heron5d
n	7	11	8	16
$N \times N$	13×13	63×63	20×20	399×399
H.L. x_n tot	0.229	16.612	441.593	10361.995
s	13	85	806	571
BB tot	0.046	12.801	421.366	9940.302
BB eval	0.028	5.428	415.676	4809.717
BB det	0.011	6.507	7.193	5087.231
Eval $\hat{f}_{\rho,j-1}$	0.011	0.132	0.374	0.467
BHL	0.005	0.023	0.298	0.196
VSolve	0.003	0.001	0.333	0.021

Theorem

(Theorem 4.3) Let p be a large prime and $\tilde{N} < p$, $\tilde{N} \in \mathbb{Z}^+$. Let $a \in \mathbb{Z}[x_1, \dots, x_n]$ and $\alpha = (\alpha_2, \dots, \alpha_n) \in \mathbb{Z}_p^{n-1}$ be randomly chosen such that $0 < \alpha_i < \tilde{N}$. Suppose α is Hilbertian and condition (i) of the input of CMSHL is satisfied. Then, with a high probability of success, the total number of arithmetic operations in \mathbb{Z}_p in the worst case for lifting $\hat{f}_{\rho,1}$ to $\hat{f}_{\rho,n}$ using Algorithm CMSHL in $n-1$ steps is










$$O(nd_1 d_{\max} s_{\max} C(\text{probe } \mathbf{B})) + O\left((n-2)s_{\max} d_{\max} \left(\sum_{\rho=1}^r \#\hat{f}_{\rho,j-1} + d_1^2 + d_1 d_{\max}\right)\right) \quad (1)$$







where $d_{\max} = \max_{3 \leq i \leq n}(\deg(\text{sqf}(a), x_i))$, $s_{\max} = \max(s_j)$ where s_j is the number s defined at step 7 of Algorithm 1 and $C(\text{probe } \mathbf{B})$ is the cost of one probe to the black box \mathbf{B} . The total number of probes to the black box is $O(nd_1 d_{\max} s_{\max})$.

- Large integers
- Multi-point evaluations to speed up
- Parallelization

Thank you for attending!

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Computing the content

After recovering $f_1, \dots, f_r \in \mathbb{Z}[x_1, \dots, x_n]$, we create another black box \mathbf{C} for the content. Then the content is factored recursively.

- Let $F = f_1^{e_1} \dots f_r^{e_r} \in \mathbb{Z}[x_1, \dots, x_n]$.
- `MakeCont := proc(B, F, X, p)`
 `local alpha1 := rand(p)();`
 `proc(Y, alpha, p) nY := nops(Y);`
 `alphaF[1] := alpha1;`
 `for i to nY do alphaF[i+1] := alpha[i]; od;`
 `Feval := Eval(F, [X[1]=alpha1, seq(Y[i]=alpha[i], i=1..nY)])`
`mod p;`
 `if Feval = 0 then return FAIL; fi;`
 `c := B([X[1],op(Y)], alphaF, p) / Feval mod p;`
 `end;`
`end;`
- `C := MakeCont(B, F, X, p);`