

Towards High Performance Multivariate Factorization

Michael Monagan

Center for Experimental and Constructive Mathematics
Simon Fraser University
British Columbia

This is joint work with Baris Tuncer.

To factor a multivariate polynomial in $\mathbb{Z}[x_1, \dots, x_n]$ Hensel lifting must solve many polynomial diophantine equations in $\mathbb{Z}_p[x_2, \dots, x_j][x_1]$ for $2 \leq j \leq n$. These are equations of the form

$$\sigma A + \tau B = C$$

where A, B, C are given, $\gcd(A, B) = 1$, and we must solve for $\sigma, \tau \in \mathbb{Z}_p[x_2, \dots, x_j][x_1]$ s.t. $\deg_{x_1} \sigma < \deg_{x_1} B$. Usually this is the most costly part and Wang's algorithm can be exponential in n .

Outline

- Some factorization timing data
- Factoring $a \in \mathbb{Z}[x_1, \dots, x_n]$ using Hensel lifting
- Wang's solution
- Kaltofen's sparse solution v. our sparse solution
- Eliminating multi-precision arithmetic
- Eliminating the error computation

Factoring the determinants of Cyclic and Toeplitz matrices

Let C_n denote the $n \times n$ cyclic matrix and let T_n denote the $n \times n$ symmetric Toeplitz matrix below.

$$C_n = \begin{pmatrix} x_1 & x_2 & \dots & x_{n-1} & x_n \\ x_n & x_1 & \dots & x_{n-2} & x_{n-1} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ x_3 & x_4 & \dots & x_1 & x_2 \\ x_2 & x_3 & \dots & x_n & x_1 \end{pmatrix} \quad T_n = \begin{pmatrix} x_1 & x_2 & \dots & x_{n-1} & x_n \\ x_2 & x_1 & \dots & x_{n-2} & x_{n-1} \\ \ddots & \ddots & \ddots & \ddots & \ddots \\ x_{n-1} & x_{n-2} & \dots & x_1 & x_2 \\ x_n & x_{n-1} & \dots & x_2 & x_1 \end{pmatrix}$$

The determinants of C_n and T_n are homogeneous polynomials in n variables.

Example

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

$$\det C_4 = (x_4 + x_3 + x_1 + x_2)(x_4 - x_3 - x_1 + x_2) \\ (x_1^2 - 2x_1x_3 + x_2^2 - 2x_2x_4 + x_3^2 + x_4^2)$$

n	# d_n	#factors	Maple	MDP	MTSHL	Magma	Singular
7	427	30,56	0.035	30%	0.046	0.01	0.02
8	1628	167,167	0.065	43%	0.073	0.04	0.05
9	6090	153,294	0.166	73%	0.122	0.10	0.28
10	23797	931,931	0.610	76%	0.418	0.89	1.77
11	90296	849,1730	2.570	74%	1.138	1.96	8.01
12	350726	5579,5579	19.45	80%	13.16	72.17	84.04
13	1338076	4983,10611	84.08	84%	21.77	181.0	607.99
14	5165957	34937,34937	637.8	77%	249.9	6039.0	20333.45
15	19732508	30458,66684	4153.2	84%	1651.7	12899.2	—

Table: Factorization timings (seconds) for $\det T_n$ evaluated at $x_n = 1$

Notes: Intel Core i5-4670 CPU @ 3.40GHz, 16 gigs RAM.

Maple 2017: kernelopts(numcpus=1), Magma 2.22–5, Singular 3–1–6,

n	# d_n	#fmax	Maple	MDP	MTSHL	Magma	Singular
7	246	924	0.045	90%	0.026	0.01	0.02
8	810	86	0.059	46%	0.063	0.07	0.06
9	2704	1005	0.225	74%	0.120	0.74	0.24
10	7492	715	0.853	62%	0.500	8.44	2.02
11	32066	184756	7.160	91%	0.945	104.3	11.39
12	86500	621	19.76	76%	5.121	7575.1	30.27
13	400024	2704156	263.4	92%	27.69	30871.9	??
14	1366500	27132	1664.4	77%	523.07	$> 10^6$	288463.2
15	4614524	303645	18432.	82%	7496.9	—	—

Table: Factorization timings (seconds) for $\det C_n$ evaluated at $x_n = 1$

Notes: ?? = I cannot compute $\det(C_n)$ nor read in $\det(C_n)$ nor it's factors.

Multivariate polynomial factorization

Input $a = fg \in \mathbb{Z}[x_1, \dots, x_n]$ primitive and square-free.

Output f, g .

- ① Pick x_1 and factor $LC(a) = h_1^{e_1} h_2^{e_2} \dots$ in $\mathbb{Z}[x_2, \dots, x_n]$.
- ② Pick $\alpha = \alpha_2, \dots, \alpha_n \in \mathbb{Z}$ such that $LC(a)(\alpha) \neq 0$ and $\gcd(a(\alpha), \frac{\partial a}{\partial x_1}(\alpha)) = 1$ and $h_i(\alpha) \in \mathbb{Z}$ have distinct prime divisors.
- ③ Factor $a(x_1, \alpha) = f(x_1, \alpha) g(x_1, \alpha)$ over \mathbb{Z} .
- ④ Distribute the factors of $LC(a)$ on $f(x_1, \alpha)$ and $g(x_1, \alpha)$.
- ⑤ Pick a prime p and $L \in \mathbb{N}$ s.t. $p^L > 2\|f\|$ where $f|a$ and $\gcd(f(x_1, \alpha), g(x_1, \alpha)) = 1$ in $\mathbb{Z}_p[x_1]$.
- ⑥ Hensel lift x_2 then x_3 then ... then x_n doing arithmetic mod p^L .

P. Wang. An improved Multivariate Polynomial Factoring Algorithm.
Mathematics of Computation, 32:1215–1231. AMS (1978)

Algorithms for Computer Algebra, K.O. Geddes, S.R. Czapor, G. Labahn. Kluwer, 1992.

Wang's Multivariate Hensel Lifting (MHL)

Input $a \in \mathbb{Z}[x_1, \dots, x_n]$, $\alpha = (\alpha_2, \dots, \alpha_n)$, $(f_0, g_0) \in \mathbb{Z}[x_1]$ s.t.

(i) $a(x_1, \alpha) - f_0 g_0 = 0$ and (ii) $\gcd(f_0, g_0) = 1$ in $\mathbb{Z}_p[x_1]$.

Output f, g satifying $a = fg$ or FAIL.

1 If $n = 1$ output (f_0, g_0) .

2 Lift x_2, \dots, x_{n-1} : $(f_0, g_0) := \text{MHL}(a(x_n = \alpha_n), \alpha, (f_0, g_0))$

Idea: $f = \sum_{k=0}^{df} f_k (x_n - \alpha_n)^k$ and $g = \sum_{k=0}^{dg} g_k (x_n - \alpha_n)^k$.

Wang's Multivariate Hensel Lifting (MHL)

Input $a \in \mathbb{Z}[x_1, \dots, x_n]$, $\alpha = (\alpha_2, \dots, \alpha_n)$, $(f_0, g_0) \in \mathbb{Z}[x_1]$ s.t.

(i) $a(x_1, \alpha) - f_0 g_0 = 0$ and (ii) $\gcd(f_0, g_0) = 1$ in $\mathbb{Z}_p[x_1]$.

Output f, g satisfying $a = fg$ or FAIL.

1 If $n = 1$ output (f_0, g_0) .

2 Lift x_2, \dots, x_{n-1} : $(f_0, g_0) := \text{MHL}(a(x_n = \alpha_n), \alpha, (f_0, g_0))$

Idea: $f = \sum_{k=0}^{df} f_k (x_n - \alpha_n)^k$ and $g = \sum_{k=0}^{dg} g_k (x_n - \alpha_n)^k$.

3 Initialize: $(f, g) := (f_0, g_0)$; $\text{error} := a - f_0 g_0$.

4 For $k = 1, \dots, \deg_{x_1} a$ while $\text{error} \neq 0$ do

$T_k := \text{Taylor coeff}(\text{error}, (x_n - \alpha_n)^k)$

If $T_k \neq 0$ then

Solve $f_k g_0 + g_k f_0 = T_k$ for $f_k, g_k \in \mathbb{Z}_p[x_1, \dots, x_{n-1}]$.

Set $f := f + f_k (x_n - \alpha_n)^k$ and $g := g + g_k (x_n - \alpha_n)^k$

Set $\text{error} := a - fg$

5 If $\text{error} = 0$ output (f, g) else output FAIL.

$$f = x^3 - xyz^2 + y^3z^2 + z^4 - 2.$$

Taylor series for $f(z)$ about $z = \alpha_3 = 0$:

$$f(z) = \underbrace{(x^3 - 2)}_{f_0} + \underbrace{(y^3 - yx)}_{f_2} z^2 + \underbrace{1}_{f_4} z^4.$$

$$f = x^3 - xyz^2 + y^3z^2 + z^4 - 2.$$

Taylor series for $f(z)$ about $z = \alpha_3 = 0$:

$$f(z) = \underbrace{(x^3 - 2)}_{f_0} + \underbrace{(y^3 - yx)}_{f_2} z^2 + \underbrace{\frac{1}{4}}_{f_4} z^4.$$

Taylor series for $f(z)$ about $z = \alpha_3 = 2$:

$$\begin{aligned} f(z) &= \underbrace{(x^3 + 4y^3 - 4yx + 14)}_{f_0} + \underbrace{(4y^3 - 4xy + 32)}_{f_1}(z-2) + \\ &\quad \underbrace{(y^3 - xy + 24)}_{f_2}(z-2)^2 + \underbrace{\frac{8}{3}}_{f_3}(z-2)^3 + \underbrace{\frac{1}{4}}_{f_4}(z-2)^4 \end{aligned}$$

Note $f_k = f_z^{(k)}(z=2)/k!$.

Wang's Multivariate Diophantine Lifting (MDP)

Input $A, B, C \in \mathbb{Z}_p[x_1, \dots, x_n]$, $\alpha = \alpha_2, \dots, \alpha_n$)

Output $\sigma, \tau \in \mathbb{Z}_p[x_1, \dots, x_n]$ satisfying $\sigma A + \tau B = C$

- 1 If $n = 1$ solve $\sigma A + \tau B = C$ using the Euclidean algorithm in $\mathbb{Z}_p[x_1]$.

Let $\sigma = \sum_{k=0}^{df} \sigma_k (x_n - \alpha_n)^k$ and $\tau = \sum_{k=0}^{dg} \tau_k (x_n - \alpha_n)^k$.

- 2 $(\sigma_0, \tau_0) := \text{Multi Dio Lift}(A(x_n = \alpha_n), B(x_n = \alpha_n), C(x_n = \alpha_n), \alpha)$

Wang's Multivariate Diophantine Lifting (MDP)

Input $A, B, C \in \mathbb{Z}_p[x_1, \dots, x_n]$, $\alpha = \alpha_2, \dots, \alpha_n$)

Output $\sigma, \tau \in \mathbb{Z}_p[x_1, \dots, x_n]$ satisfying $\sigma A + \tau B = C$

- 1 If $n = 1$ solve $\sigma A + \tau B = C$ using the Euclidean algorithm in $\mathbb{Z}_p[x_1]$.

Let $\sigma = \sum_{k=0}^{df} \sigma_k (x_n - \alpha_n)^k$ and $\tau = \sum_{k=0}^{dg} \tau_k (x_n - \alpha_n)^k$.

- 2 $(\sigma_0, \tau_0) := \text{MultiDioLift}(A(x_n = \alpha_n), B(x_n = \alpha_n), C(x_n = \alpha_n), \alpha)$

- 3 Initialize: $(\sigma, \tau) := (\sigma_0, \tau_0)$ $\text{error} := C - \sigma A - \tau B$

- 4 For $k = 1, 2, \dots$ while $\text{error} \neq 0$ do

$T_k := \text{Taylor coeff}(\text{error}, (x_n - \alpha_n)^k)$

If $T_k \neq 0$ then

$\sigma_k, \tau_k := \text{MultiDioLift}(\sigma_0, \tau_0, T_k, \alpha)$

$\sigma, \tau := \sigma + \sigma_k (x_n - \alpha_n)^k, \tau + \tau_k (x_n - \alpha_n)^k$

$\text{error} := \text{error} - \sigma_k (x_n - \alpha_n)^k A - \tau_k (x_n - \alpha_n)^k B$

- 5 output (σ, τ) .

Wang's Multivariate Diophantine Lifting (MDP)

Input $A, B, C \in \mathbb{Z}_p[x_1, \dots, x_n]$, $\alpha = \alpha_2, \dots, \alpha_n$)

Output $\sigma, \tau \in \mathbb{Z}_p[x_1, \dots, x_n]$ satisfying $\sigma A + \tau B = C$

- 1 If $n = 1$ solve $\sigma A + \tau B = C$ using the Euclidean algorithm in $\mathbb{Z}_p[x_1]$.

Let $\sigma = \sum_{k=0}^{df} \sigma_k (x_n - \alpha_n)^k$ and $\tau = \sum_{k=0}^{dg} \tau_k (x_n - \alpha_n)^k$.

- 2 $(\sigma_0, \tau_0) := \text{MultiDioLift}(A(x_n = \alpha_n), B(x_n = \alpha_n), C(x_n = \alpha_n), \alpha)$

- 3 Initialize: $(\sigma, \tau) := (\sigma_0, \tau_0)$ $\text{error} := C - \sigma A - \tau B$

- 4 For $k = 1, 2, \dots$ while $\text{error} \neq 0$ do

$T_k := \text{Taylor coeff}(\text{error}, (x_n - \alpha_n)^k)$

If $T_k \neq 0$ then

$\sigma_k, \tau_k := \text{MultiDioLift}(\sigma_0, \tau_0, T_k, \alpha)$

$\sigma, \tau := \sigma + \sigma_k (x_n - \alpha_n)^k, \tau + \tau_k (x_n - \alpha_n)^k$

$\text{error} := \text{error} - \sigma_k (x_n - \alpha_n)^k A - \tau_k (x_n - \alpha_n)^k B$

- 5 output (σ, τ) .

Let $M(n)$ count calls to the Euc. alg. and $d \geq \max(\deg_{x_i} f, \deg_{x_i} g)$.

Then $M(1) = 1, M(n) \leq dM(n-1) \implies M(n) \leq d^{n-1}$.

Solve the MDP $f_k g_0 + g_k f_0 = T_k$ for $f_k, g_k \in \mathbb{Z}_p[x_1, \dots, x_{n-1}]$.

Interpolate x_2, \dots, x_{n-1} from images in $\mathbb{Z}_p[x_1]$ using sparse interpolation?

Solve the MDP $f_k \ g_0 + g_k \ f_0 = T_k$ for $f_k, g_k \in \mathbb{Z}_p[x_1, \dots, x_{n-1}]$.

Interpolate x_2, \dots, x_{n-1} from images in $\mathbb{Z}_p[x_1]$ using sparse interpolation?

Theorem [MT] The Strong Sparse Hensel Assumption

If α_n is random then, with high probability

$$\begin{aligned} \text{supp}(f_0) &\supseteq \text{supp}(f_1) \supseteq \cdots \supseteq \text{supp}(f_{df}) \text{ and} \\ \text{supp}(g_0) &\supseteq \text{supp}(g_1) \supseteq \cdots \supseteq \text{supp}(g_{dg}). \end{aligned}$$

Example ($n = 8$, $\#f = 10,000$, $df = 16$, $\alpha = 3$)

$|f_k| = 9877, 7043, 4932, 3374, 2310, 1545, 1001, 654, 418, 245, 141, 81, 34, 13, 5, 2, 1$.

Solve the MDP $f_k \ g_0 + g_k \ f_0 = T_k$ for $f_k, g_k \in \mathbb{Z}_p[x_1, \dots, x_{n-1}]$.

Interpolate x_2, \dots, x_{n-1} from images in $\mathbb{Z}_p[x_1]$ using sparse interpolation?

Theorem [MT] The Strong Sparse Hensel Assumption

If α_n is random then, with high probability

$$\begin{aligned} \text{supp}(f_0) &\supseteq \text{supp}(f_1) \supseteq \cdots \supseteq \text{supp}(f_{df}) \text{ and} \\ \text{supp}(g_0) &\supseteq \text{supp}(g_1) \supseteq \cdots \supseteq \text{supp}(g_{dg}). \end{aligned}$$

Example ($n = 8$, $\#f = 10,000$, $df = 16$, $\alpha = 3$)

$|f_k| = 9877, 7043, 4932, 3374, 2310, 1545, 1001, 654, 418, 245, 141, 81, 34, 13, 5, 2, 1$.

Kaltofen's Sparse Hensel Lifting.

Let $\text{supp}(f_0) = \{M_1, \dots, M_r\}$ and $\text{supp}(g_0) = \{N_1, \dots, N_s\}$

Equate the coefficients c_i of

$$\left(\sum_{i=1}^r c_i M_i \right) g_0 + \left(\sum_{i=1}^s c_{r+i} N_i \right) f_0 = T_k \quad \text{in } \mathbb{Z}_p[x_1, \dots, x_{n-1}]$$

in monomials in (x_1, \dots, x_{n-1}) and solve the $(r+s) \times (r+s)$ linear system.

MTSHL : Monagan and Tuncer's Sparse Hensel Lifting

Solve $f_k g_0 + g_k f_0 = T_k$ for $f_k, g_k \in \mathbb{Z}_p[x_1, \dots, x_{n-1}]$.

Let $f_{k-1} = \sum_{i=1}^{df} a_i(x_2, \dots, x_{n-1})x_1^i$ and $\text{supp}(a_i) = \{M_{i1}, \dots, M_{ir_i}\}$,

and $g_{k-1} = \sum_{i=1}^{dg} b_i(x_2, \dots, x_{n-1})x_1^i$ and $\text{supp}(b_i) = \{N_{i1}, \dots, N_{is_i}\}$.

Assume $f_k = \sum_{i=0}^{df} \left(\sum_{j=1}^{r_i} a_{ij} M_{ij} \right) x_1^i$ and $g_k = \sum_{i=0}^{dg} \left(\sum_{j=1}^{s_i} b_{ij} N_{ij} \right) x_1^i$.

MTSHL : Monagan and Tuncer's Sparse Hensel Lifting

Solve $f_k g_0 + g_k f_0 = T_k$ for $f_k, g_k \in \mathbb{Z}_p[x_1, \dots, x_{n-1}]$.

Let $f_{k-1} = \sum_{i=1}^{df} a_i(x_2, \dots, x_{n-1})x_1^i$ and $\text{supp}(a_i) = \{M_{i1}, \dots, M_{ir_i}\}$,

and $g_{k-1} = \sum_{i=1}^{dg} b_i(x_2, \dots, x_{n-1})x_1^i$ and $\text{supp}(b_i) = \{N_{i1}, \dots, N_{is_i}\}$.

Assume $f_k = \sum_{i=0}^{df} \left(\sum_{j=1}^{r_i} a_{ij} M_{ij} \right) x_1^i$ and $g_k = \sum_{i=0}^{dg} \left(\sum_{j=1}^{s_i} b_{ij} N_{ij} \right) x_1^i$.

Let $t = \max(\#a_i, \#b_i) \ll \#a + \#b$. Let $\beta = (\beta_2, \dots, \beta_{n-1}) \in \mathbb{Z}_p^{n-2}$.

Solve $\sigma_j g_0(\beta^j) + \tau_j f_0(\beta^j) = T_k(\beta^j)$ for $\sigma_j, \tau_j \in \mathbb{Z}_p[x_1]$. **for** $1 \leq j \leq t$.

Solve the Vandermonde systems

$$\begin{cases} \text{coeff}(f_k(\beta^j), x_n^i) = \text{coeff}(\sigma_j, x_n^i) \text{ for } 1 \leq j \leq \#a_i \\ \text{coeff}(g_k(\beta^j), x_n^i) = \text{coeff}(\tau_j, x_n^i) \text{ for } 1 \leq j \leq \#b_i \end{cases} \quad \begin{array}{l} \text{for } 0 \leq i < df, \\ \text{for } 0 \leq i < dg. \end{array}$$

Cost of MTSHL: number of arith. ops in \mathbb{Z}_p where $d = \max \deg_{x_i} a$

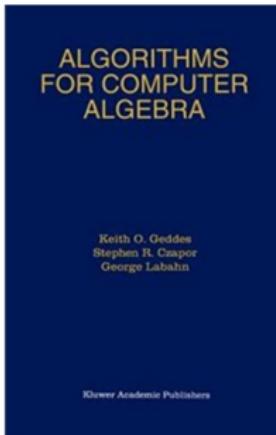
$$(n-1)d O\left(\underbrace{\#f \#g}_{\text{error}} + \underbrace{\#\alpha}_{\text{Taylor}} + \underbrace{(\#\alpha + \#f + \#g)}_{\text{eval}} t + \underbrace{d^2 t}_{\text{UniDio}} + \underbrace{dt^2}_{\text{Vandermonde}} \right).$$

$n/d/T$	Wang (MDP)	Kaltofen (MDP)	MTSHL (MDP)
4/35/100	13.07 (11.95)	1.75 (1.18)	1.51 (0.24)
5/35/100	88.10 (86.28)	3.75 (2.57)	1.16 (0.36)
7/35/100	800.0 (797.0)	5.04 (4.08)	1.58 (0.59)
9/35/100	4451.6 (4449.4)	8.13 (6.22)	2.94 (0.56)
4/35/500	33.96 (26.48)	642.2 (635.1)	11.29 (0.82)
5/35/500	472.1 (402.5)	1916.2 (1899.6)	26.0 (4.86)
7/35/500	3870.5 (3842.2)	2329.4 (2305.5)	43.1 (6.84)
9/35/500	> 60000	3866.3 (3805.9)	79.6 (9.71)

```

> M := product( xi, i = 1..n );
> f := x1d + M randpoly( [x1, ..., xn], terms = T, degree = d );

```



Chapter 6

Newton's Iteration and the Hensel Construction

Run the entire Hensel lifting mod p^L

where $p^L > 2 \max(\|a\|, \|f\|, \|g\|)$.

Why?

To avoid the expression swell when solving
 $\sigma g_0(\alpha) + \tau f_0(\alpha) = T_k(\alpha)$ in $\mathbb{Z}[x_1]$ in MDP.

Lemma [Gelfond] Let $a \in \mathbb{Z}[x_1, \dots, x_n]$ with $d_i = \deg_{x_i} a$.

If $f \mid a$ then

$$\|f\| \leq e^{d_1 + d_2 + \dots + d_n} \|a\| \text{ where } e = 2.718281828$$

Let p be a prime and let $f = \sum_{k=0}^{hf} f_k p^k$ and $g = \sum_{k=0}^{hg} g_k p^k$.

Example

$$\begin{aligned}f &= 2x_1 + (5 + 0 \cdot p + 2p^2)x_2 + (7 + 3p)x_3 \\&= \underbrace{(2x_1 + 5x_2 + 7x_3)}_{f_0} + \underbrace{3x_3}_{f_1} p + \underbrace{2x_2}_{f_2} p^2.\end{aligned}$$

$$\text{supp}(f_0) = \{x_1, x_2, x_3\} \supseteq \text{supp}(f_1) = \{x_3\} \not\supseteq \text{supp}(f_2) = \{x_2\}.$$

Let p be a prime and let $f = \sum_{k=0}^{hf} f_k p^k$ and $g = \sum_{k=0}^{hg} g_k p^k$.

Example

$$\begin{aligned}f &= 2x_1 + (5 + 0 \cdot p + 2p^2)x_2 + (7 + 3p)x_3 \\&= \underbrace{(2x_1 + 5x_2 + 7x_3)}_{f_0} + \underbrace{3x_3}_{f_1} p + \underbrace{2x_2}_{f_2} p^2.\end{aligned}$$

$$\text{supp}(f_0) = \{x_1, x_2, x_3\} \supseteq \text{supp}(f_1) = \{x_3\} \not\supseteq \text{supp}(f_2) = \{x_2\}.$$

Theorem If p is chosen at random from $[2^{b-1}, 2^b]$ for b sufficiently large then

$$\begin{aligned}\text{supp}(f_0) &\supseteq \text{supp}(f_1) \supseteq \text{supp}(f_2) \supseteq \dots \text{supp}(f_{hf}) \text{ and} \\ \text{supp}(g_0) &\supseteq \text{supp}(g_1) \supseteq \text{supp}(g_2) \supseteq \dots \text{supp}(g_{hg}) \text{ whp.}\end{aligned}$$

Idea: Run the entire Hensel lifting modulo a **machine** prime p to obtain $f_0, g_0 \in \mathbb{Z}_p[x_1, \dots, x_n]$ satisfying $a - f_0 g_0 \pmod{p} = 0$. Now do a p -adic lift on f_0, g_0 to get f, g .

Multivariate p -adic Lifting

Input $a \in \mathbb{Z}[x_1, \dots, x_n]$, $(f_0, g_0) \in \mathbb{Z}_p[x_1, \dots, x_n]$ satisfying $a - f_0g_0 \pmod{p} = 0$

prime p and a lifting bound L satisfying $p^L > \|f\|$ where $f|a$.

Output $f, g \in \mathbb{Z}[x_1, \dots, x_n]$ satisfying $a = fg$ or FAIL.

1 Initialize: $(f, g) := (f_0, g_0)$; $\text{error} := a - f_0g_0$;

2 For $k = 1, \dots, L$ while $\text{error} \neq 0$ do

$$c_k := \left(\frac{\text{error}}{p^k}\right) \pmod{p}$$

If $c_k \neq 0$ then

Solve the MDP $f_k g_0 + g_k f_0 = c_k$ for $f_k, g_k \in \mathbb{Z}_p[x_1, \dots, x_n]$.

Set $f_k := \text{mods}(\sigma, p)$ and $g_k := \text{mods}(\tau, p)$.

Set $f := f + f_k p^k$; $g := g + g_k p^k$; $\text{error} := a - fg$.

3 If $\text{error} = 0$ output (f, g) else output FAIL.

The MDP is solved in $\mathbb{Z}_p[x_1, \dots, x_n]$ where p is a machine prime!

$n/d/T$	m	L	MTSHL mod p^L	MTSHL mod $p + \text{lift}$
5/10/300	2	5	5.86 (5.10)	$0.197 + 0.241 = 0.438$
5/10/300	4	9	6.92 (6.10)	$0.514 + 0.553 = 1.067$
5/10/300	8	17	8.63 (7.596)	$0.477 + 2.076 = 2.553$
5/10/1000	2	5	14.44 (12.82)	$0.870 + 1.332 = 2.202$
5/10/1000	4	9	16.94 (15.18)	$0.921 + 2.632 = 3.553$
5/10/1000	8	17	19.03 (17.22)	$0.873 + 4.032 = 4.905$

Table: CPU timings (seconds) for p -adic lifting for $p = 2^{31} - 1$.

```
> f := x1d + randpoly( [x1, ..., xn], terms = T, degree = d , coeffs = rand(pm));
```

Eliminating the error multiplication $a - fg$

Let $f^{(k)} = \sum_{i=0}^{k-1} f_i(x_n - \alpha_n)^i$ and $g^{(k)} = \sum_{i=0}^{k-1} g_i(x_n - \alpha_n)^i$.

Main Idea: $\text{error}(\beta^j) = a(\beta^j) - f^{(k)}(\beta^j) g^{(k)}(\beta^j)$ in $\mathbb{Z}_p[x_1, x_n]$.

for $j = 1..t$ do evaluate $a(\beta^j), f_0(\beta^j), g_0(\beta^j)$.

for $j = 1..t$ **in parallel** do

Hensel lift $f_0(\beta^j)$ to $f(\beta^j) = \sum_{k=0}^d f_k(\beta^j)(x_n - \alpha_n)^k$.

And also $g_0(\beta^j)$ to $g(\beta^j)$.

for $k = 1..d$ in do solve for coefficients of f_k, g_k .

Set $f = \sum_{k=0}^d f_k(x_n - \alpha_n)^k$ and $g = \sum_{k=0}^d g_k(x_n - \alpha_n)^k$.

$$(n-1)d O\left(\underbrace{\#f \#g}_{\text{error}} + \underbrace{\#a}_{\text{Taylor}} + \underbrace{d^2 t}_{\text{Taylor}} + \underbrace{(\#a + \#f + \#g)t/d}_{\text{eval}} + \underbrace{d^2 t}_{\text{UniDio}} + \underbrace{dt^2}_{\text{Vandermonde}}\right).$$

We eliminated the the error computation in $\mathbb{Z}_p[x_1, \dots, x_{n-1}]$!