

# Sparse multivariate polynomial factorization: A high-performance design and implementation.

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This is joint work with Baris Tuncer.

# Factoring polynomials using Wang's Hensel lifting

```
> e1 := (a = f*g);
```

$$\begin{aligned}e1 &:= x^5 + 20x^2y^6z + 5x^4y^3z + 30xy^4z^3 + 12xy^4z^2 + 4x^3y^3 \\&\quad + 3x^3yz^2 + 18y^2z^4 + 6x^2yz^2 \\&= (x^2 + 5xy^3z + 3yz^2)(x^3 + 4xy^3 + 6yz^2)\end{aligned}$$

```
> e2 := eval(e1,z=3);
```

$$\begin{aligned}e2 &:= x^5 + 60x^2y^6 + 15x^4y^3 + 4x^3y^3 + 918xy^4 + 27x^3y + 54x^2y + 1458y^2 \\&= (x^2 + 15xy^3 + 27y)(x^3 + 4xy^3 + 54y)\end{aligned}$$

```
> e3 := eval(e2,y=-5);
```

$$\begin{aligned}e3 &:= x^5 - 1875x^4 - 635x^3 + 937230x^2 + 573750x + 36450 \\&= (x^2 - 1875x - 135)(x^3 - 500x - 270)\end{aligned}$$

Notes: Let  $h$  be any factor of  $a$  and let  $B > \max(||h||_\infty, ||a||_\infty)$ .  
Multivariate Hensel Lifting (MHL) is done modulo  $m = p^L > 2B$ .

# Wang's Multivariate Hensel Lifting (MHL)

Input  $a \in \mathbb{Z}_p[x_1, \dots, x_j]$ ,  $\alpha = (\alpha_2, \dots, \alpha_j)$ ,  $(f_0, g_0) \in \mathbb{Z}_p[x_1, \dots, x_{j-1}]$  s.t.

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

Idea:  $f = f_0 + f_1(x_j - \alpha_j) + f_2(x_j - \alpha_j)^2 + \dots + f_{df}(x_j - \alpha_j)^{df}$

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Initialize:  $(f, g) := (f_0, g_0)$  and  $\text{error} := a - fg$

For  $k = 1, 2, \dots$ , while  $\deg(f, x_j) + \deg(g, x_j) < \deg(a, x_j)$  do

$c_k := \text{coeff}(\text{error}, (x_j - \alpha_j)^k)$

If  $c_k \neq 0$  then

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

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

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

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

Implemented in Magma, Maple, Macsyma, Mathematica and Singular.

**Algorithms for Computer Algebra**, Geddes, Czapor, and Labahn, 1992.

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**Wang's algorithm is highly sequential.**  $O(d^n)$  if the  $\alpha_i$ 's are non-zero.

# The Taylor Coefficients : MT CASC 2016

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

The solutions  $f_k$  to the MDP are Taylor coefficients of  $f(z)$  about  $z = \alpha_3$ .

If  $\alpha_3 = 0$  then  $f(z) = \underbrace{(x^3 - 2)}_{f_0} + \underbrace{(y^3 - xy)}_{f_2} z^2 + \underbrace{1}_{f_4} z^4.$

If  $\alpha_3 = 2$  then

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

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**Lemma 1 [MT 2016]** If  $\alpha_j$  is chosen at random from a sufficiently large set then

$$\text{supp}(f_0) \supseteq \text{supp}(f_1) \supseteq \cdots \supseteq \text{supp}(f_k) \text{ w.h.p.}$$

# MTSHL : Monagan and Tuncer's Sparse Hensel Lifting

Solve  $f_k g_0 + g_k f_0 = c_k$  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**  $f_k = \sum_{i=0}^{df} \left( \sum_{j=1}^{r_i} \color{blue}{a_{ij}} M_{ij} \right) x_1^i$  and  $g_k = \sum_{i=0}^{dg} \left( \sum_{j=1}^{s_i} \color{blue}{b_{ij}} N_{ij} \right) x_1^i$ .

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

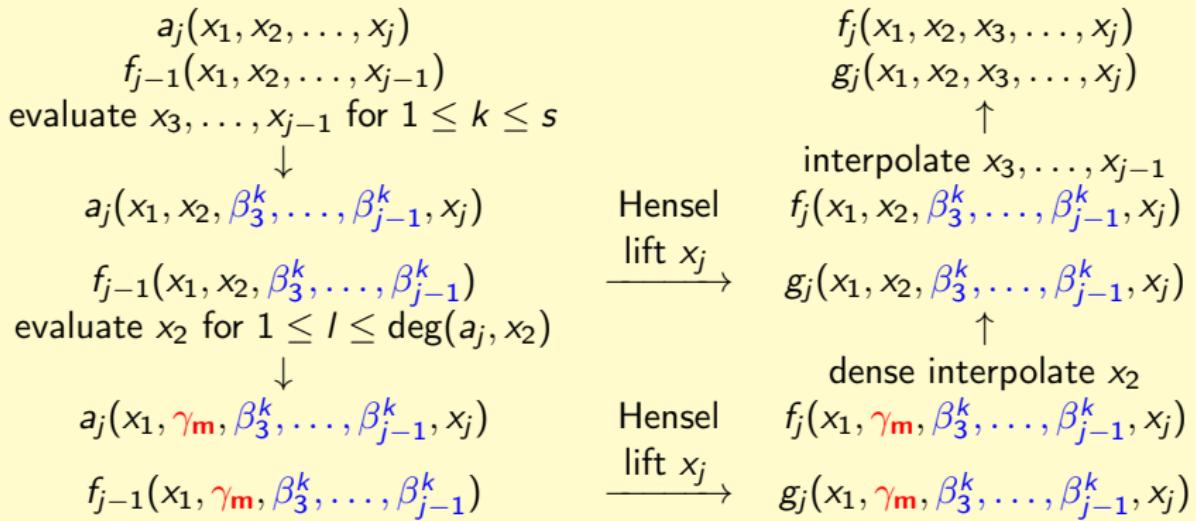
For  $1 \leq j \leq s$  solve  $\sigma_j g_0(\beta^j) + \tau_j f_0(\beta^j) = c_k(\beta^j)$  for  $\sigma_j, \tau_j \in \mathbb{Z}_p[x_1]$ .

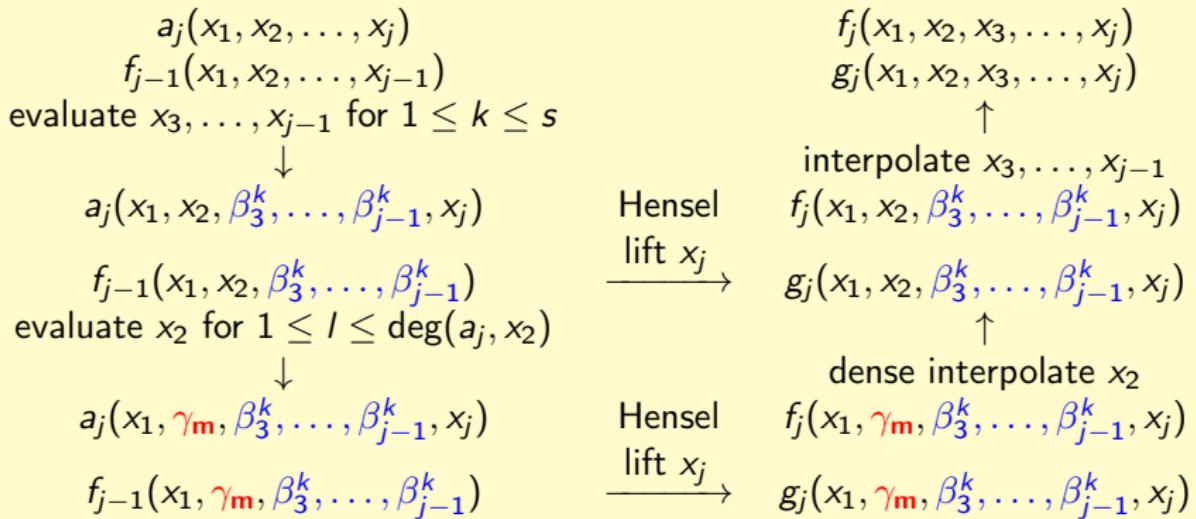
Needs  $\gcd(g_0(\beta^j), f_0(\beta^j)) = 1 \forall 1 \leq j \leq s$ .

Solve the shifted Vandermonde systems

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

Needs  $M_{ij}(\beta) \neq M_{ik}(\beta) \forall i, j \neq k$  and  $N_{ij}(\beta) \neq N_{ik}(\beta) \forall i, j \neq k$ .





Parallelized evaluations  $a(x_1, x_2, \beta_3^k, \dots, \beta_{j-1}^k, x_j)$  using Cilk C  
 Do evaluations  $x_2 = \gamma_m$  and **bivariate Hensel lifts** in parallel.  
 Optimize the bivariate Hensel lifts in  $\mathbb{Z}_p[x_1, x_j] \text{ mod } (x_j - \alpha_j)^k$ .

# Benchmark 1 – many variables, modest degree

On a server with 2 Intel Xeon E2660 8 core CPUs – 2.2 GHz(base) – 3.0 GHz(turbo)

| n | d | t    | Maple 2018 |         | New times (1 core) |          |         | New times (16 cores) |          |                 |
|---|---|------|------------|---------|--------------------|----------|---------|----------------------|----------|-----------------|
|   |   |      | best       | worst   | total              | (hensel) | (eval)  | total                | (hensel) | (eval)          |
| 6 | 7 | 500  | 0.411      | 28.84   | 0.098              | (0.015)  | (0.042) | 0.074                | (0.019)  | (0.008 – 5.2x)  |
| 6 | 7 | 1000 | 1.140      | 58.46   | 0.414              | (0.025)  | (0.247) | 0.180                | (0.027)  | (0.030 – 8.2x)  |
| 6 | 7 | 2000 | 3.066      | 99.88   | 1.593              | (0.041)  | (1.132) | 0.285                | (0.042)  | (0.121 – 9.4x)  |
| 6 | 7 | 4000 | 7.173      | 162.49  | 5.072              | (0.069)  | (4.070) | 0.814                | (0.074)  | (0.380 – 10.7x) |
| 6 | 7 | 8000 | 15.61      | NA      | 12.75              | (0.122)  | (10.95) | 1.896                | (0.130)  | (0.939 – 11.7x) |
| 9 | 7 | 500  | 1.171      | 7564.9  | 0.105              | (0.013)  | (0.040) | 0.101                | (0.024)  | (0.010 – 4.0x)  |
| 9 | 7 | 1000 | 3.704      | 10010.4 | 0.524              | (0.025)  | (0.297) | 0.233                | (0.026)  | (0.030 – 11.4x) |
| 9 | 7 | 2000 | 13.43      | NA      | 2.838              | (0.042)  | (1.973) | 0.483                | (0.045)  | (0.193 – 10.2x) |
| 9 | 7 | 4000 | 51.77      | NA      | 18.35              | (0.078)  | (14.84) | 2.325                | (0.083)  | (1.350 – 11.0x) |
| 9 | 7 | 8000 | NA         | NA      | 116.6              | (0.139)  | (102.5) | 11.50                | (0.145)  | (7.947 – 12.9x) |

Table 1: Timings (real time in seconds) for Hensel lift of  $x_n$ .

Legend:  $n = \#\text{variables}$ ,  $d = \deg(f, x_j) = \deg(g, x_j)$ ,  $t = \#f = \#g$ .

## Benchmark 2: higher degree

| n | d  | t    | Maple 2018 |         | New time (1 core) |          |         | New time (16 cores) |                |         |
|---|----|------|------------|---------|-------------------|----------|---------|---------------------|----------------|---------|
|   |    |      | best       | worst   | total             | (hensel) | (eval)  | total               | (hensel)       | (eval)  |
| 6 | 10 | 500  | 0.571      | 92.49   | 0.099             | (0.029)  | (0.025) | 0.081               | (0.024 – 1.2x) | (0.006) |
| 6 | 15 | 500  | 0.751      | 7956.5  | 0.134             | (0.070)  | (0.016) | 0.093               | (0.034 – 2.1x) | (0.006) |
| 6 | 20 | 500  | 0.919      | 48610.1 | 0.238             | (0.168)  | (0.017) | 0.130               | (0.065 – 2.6x) | (0.005) |
| 6 | 40 | 500  | 1.615      | NA      | 1.207             | (1.128)  | (0.015) | 0.282               | (0.203 – 5.6x) |         |
| 6 | 60 | 500  | 3.343      | NA      | 4.580             | (4.486)  | (0.015) | 0.732               | (0.631 – 7.1x) | (0.011) |
| 6 | 80 | 500  | 4.485      | NA      | 13.76             | (13.65)  | (0.016) | 1.674               | (1.554 – 8.8x) | (0.012) |
| 6 | 10 | 2000 | 5.237      | 976.94  | 1.775             | (0.089)  | (1.067) | 0.374               | (0.055 – 1.6x) | (0.121) |
| 6 | 15 | 2000 | 7.166      | 23128.5 | 1.616             | (0.221)  | (0.706) | 0.413               | (0.107 – 2.1x) | (0.061) |
| 6 | 20 | 2000 | 9.195      | NA      | 1.635             | (0.451)  | (0.480) | 0.431               | (0.150 – 3.0x) | (0.040) |
| 6 | 40 | 2000 | 15.98      | NA      | 4.008             | (2.993)  | (0.260) | 0.854               | (0.505 – 5.9x) | (0.038) |
| 6 | 60 | 2000 | 42.32      | NA      | 14.25             | (13.15)  | (0.292) | 1.926               | (1.500 – 8.8x) | (0.052) |
| 6 | 80 | 2000 | 57.33      | NA      | 26.34             | (25.25)  | (0.217) | 3.340               | (2.839 – 8.9x) | (0.050) |

Table 2: Timings (real time in seconds) for increasing degree  $d$ .

Legend:  $n = \#\text{variables}$ ,  $d = \deg(f, x_j) = \deg(g, x_j)$ ,  $t = \#f = \#g$ .

# Software Design Issues

For high performance we prefer arrays of machine integers

Data structure for  $f = 3x^4 + 5x^2y^3z^4 - 2yz^6 + 7$  in  $\mathbb{Z}_p[x, y, z]$ .

array of coefficients (64 bit integers): 

|   |   |    |   |
|---|---|----|---|
| 3 | 5 | -2 | 7 |
|---|---|----|---|

array of monomials (64 bit integers): 

|       |       |       |       |
|-------|-------|-------|-------|
| 4 0 0 | 2 3 4 | 0 1 6 | 0 0 9 |
|-------|-------|-------|-------|

For polynomials in more variables use `__int128_t` and `__uint128_t` in gcc.

# Software Design Issues

For multi-core performance using Cilk C we like “in-place” algorithms which reduce memory management.

Solve  $f_k g_0 + g_k f_0 = c_k$  given for  $f_k, g_k \in \mathbb{Z}_p[x_1]$  given  $sg_0 + tf_0 = 1$ .

```
for( k=1; k<da; k++ ) {  
    ...  
    polcopy64s( ck, dck, T );           // T = ck  
    dT = poldiv64s( T, f0, dt, df0, p ); // T = Tk mod f0  
    dfk = polmul64s( T, s, dT, ds, fk, p ); // fk = T * s  
    dfk = poldiv64s( fk, f0, dfk, df0, p ); // fk = fk mod f0  
    ...  
}
```

All routines for  $\mathbb{Z}_p[x_1]$  and  $\mathbb{Z}_p[x_1, \dots, x_n]$  do not allocate memory.  
The array T is reused in the next iteration.

Thank you for attending!