

# **Speeding-up Newton iteration using variants of polynomial multiplication**

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## Newton iteration

- computing symbolic solutions
- to polynomial / differential equations
- at high precision

**Example** Consider the equation, with coefficients in  $\mathbb{Z}/101\mathbb{Z}$   
[Bostan, Morain, Salvy, Schost, 08]

$$(x^6 + x^4 + 1)f'(x)^2 = 1 + 75f(x)^4 + 16f(x)^6, \quad f(0) = 0, \quad f'(0) = 1.$$

We want to find the first few terms of the power series solution

$$f = x + 68x^5 + 66x^7 + 60x^9 + 84x^{11} + \dots$$

## Newton iteration is fast

- $M(n)$  denotes the cost of polynomial multiplication in degree  $n$
- then, for most problems,  $O(M(n))$  to get  $n$  terms
- compared to (usually)  $O(n^2)$

## Objective: make it faster

- reducing the constant in the big-Oh
- using tricks such as short product (Mulders) or middle product (Hanrot, Quercia, Zimmermann)
- for moderate degrees

## This talk

- first order differential equations

# Related work

## Newton for ODE's

- [Brent, Kung, 78]  
Focused on first-order equations.
- [Watt, 88]  
Recast differential equations as fixed point problems.
- [Hoeven, 02]  
Used a similar idea + fast “relaxed multiplication”.
- [Bostan, Chyzak, Ollivier, Salvy, Schost, 07]  
Focused in particular on higher order equations.

## Other contexts

- [Hanrot *et al.*, 04]  
middle product for inverse, square-root
- [Hanrot-Zimmermann, 04], [Bernstein, 04], [Bostan-Schost, 08]  
tricks for the FFT model

# Motivation

Previous example from a **point-counting** algorithm in elliptic cryptography: computing a degree  $n$  morphism

$$\Phi : E \rightarrow E', \quad (x, y) \mapsto (\varphi(x), y\varphi'(x)).$$

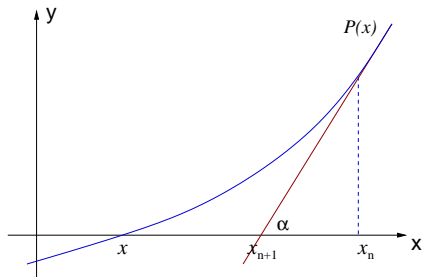
- not-so-naive algorithm  $O(n^2)$
- Newton  $O(M(n))$

Newton wins for record-size computations (degree  $> 1000$ ).

However, *“if we want the cryptologists to buy our stuff, we’d better be competitive in crypto size”*:

- small degree (about **300**).

# Newton iteration for numerical root-finding



$$x_{n+1} = x_n - \frac{P(x_n)}{P'(x_n)}.$$

The number of correct digits approximately *doubles* at each iteration.

# Newton iteration for ODE's

Given the equation  $G(x, f, f') = 0$  and  $f \bmod x^n$ , we want  $f \bmod x^{2n}$ .

- evaluate

$$a = \frac{\partial G}{\partial u}(x, f, f'), \quad b = \frac{\partial G}{\partial t}(x, f, f'), \quad c = -G(x, f, f') \quad \bmod x^{2n}$$

- use *inverse* and *exponential* to compute

$$d = \frac{b}{a}, \quad e = \frac{c}{a}, \quad j = \exp(\int d) \quad \bmod x^{2n},$$

we obtain

$$f = f + \frac{\int e j}{j} \quad \bmod x^{2n}.$$

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# Power series inverse

Consider power series

$$f = \sum_{i \geq 0} f_i x^i \quad \text{and} \quad \tilde{g} = \sum_{i \geq 0} g_i x^i$$










such that  $f_0 = 1, \tilde{g} = \frac{1}{f}$ .

**Newton iteration:**

- suppose that we know  $g = \tilde{g} \bmod x^n$
- then we get  $G = \tilde{g} \bmod x^{2n}$  as

$$g(2 - fg) \bmod x^{2n}$$

# Various multiplications

Type	Lengths & Graph rep.
plain product $M(n)$	A: $(0,n)$  B: $(0,n)$  C: $(0,2n)$ 
middle product $M(n) + O(n)$	A: $(0,2n)$  B: $(0,n)$  C: $(n,2n)$ 
short product $m(n)$	A: $(0,n)$  B: $(0,n)$  C: $(0,n)$ 
⋮	⋮

# Updating inverses

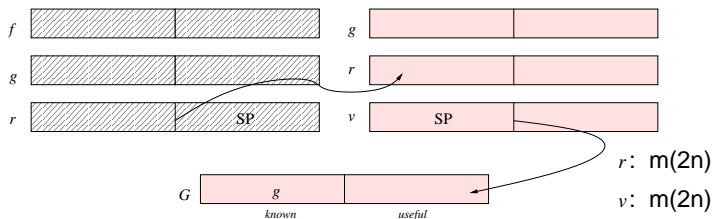


Figure: Naive inverse

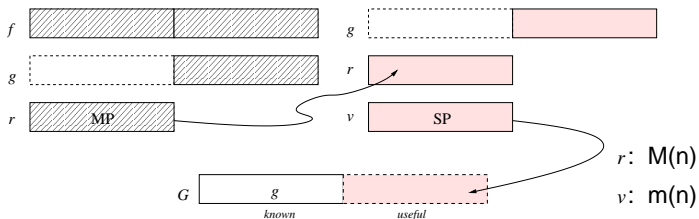
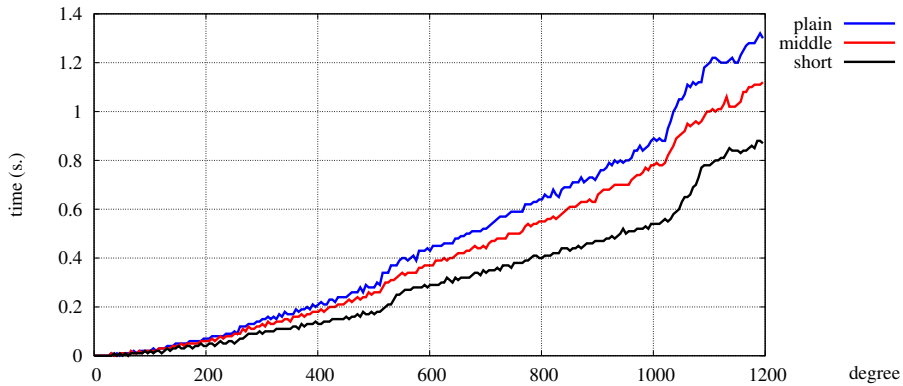
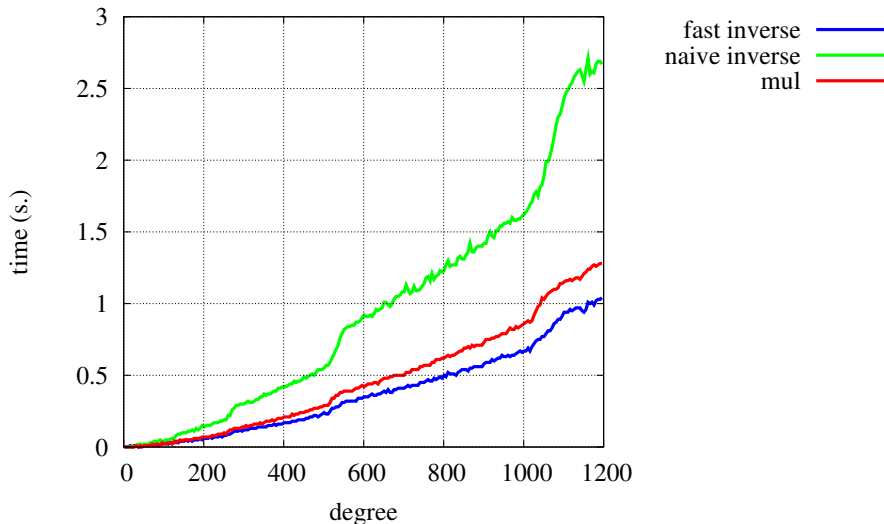


Figure: Updating inverse

# Fast multiplication



# Comparison between naive and fast inverse



# Precision issues for evaluation

Recall: we need

$$a = \frac{\partial G}{\partial u}(x, f, f'), \quad b = \frac{\partial G}{\partial t}(x, f, f'), \quad c = -G(x, f, f') \pmod{x^{2n}}$$

**Objective:** avoid computing useless quantities, as for the inverse

## Starting points

- $c$  starts with  $n$  zeros
- $a$  and  $b$  needed only modulo  $x^n$

## Propagation

- length analysis: **high-deg**, **low-deg**.

$$A = a_0 + a_1x + \cdots + a_nx^n + \cdots + a_{2n}x^{2n} + \cdots + a_ix^i$$

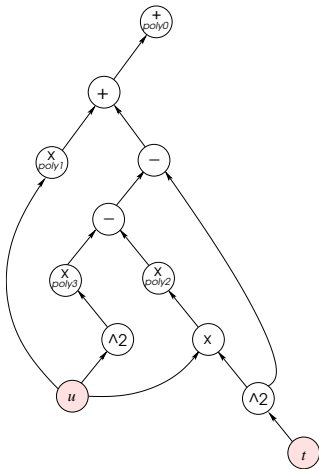
$$(low-deg, high-deg) = (n, 2n)$$

- apply variants of multiplications (middle, short product, ...)

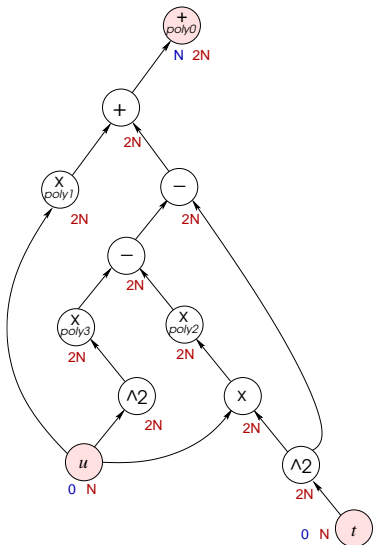
# Example

We consider  $G(x, f, f') = 0$ , with

$$G(x, t, u) = (1 + x + x^2)u^2 - (2 + x)ut^2 - t^2 + 5u + 3.$$



# Assigning high-degrees



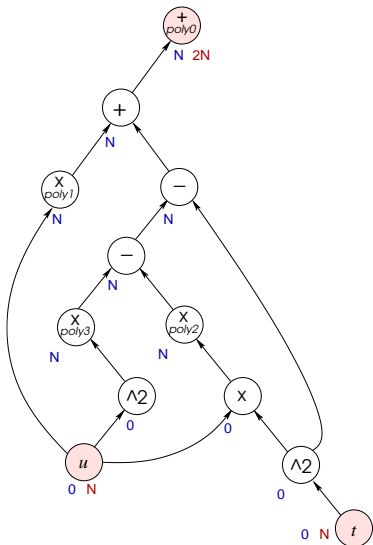
↑ Assign "high-deg"

For "+", "-",  
"high-deg" is max of the arguments.

For "x", " $\wedge$ ",  
"high-deg" is the same as the output.



# Assigning low-degrees

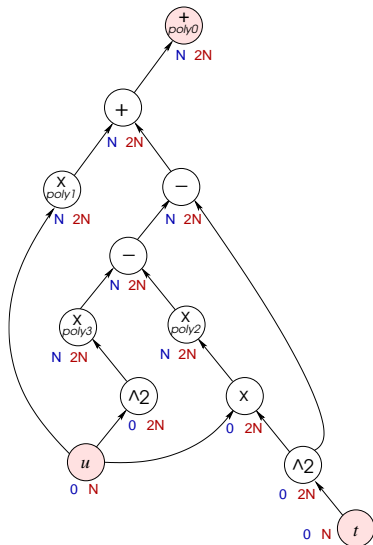


↓ Assign "low-deg"-

For "+", "-",  
"low-deg"s are the same as the result.

For "x", " $\wedge$ ",  
"low-deg"s are set to "0"s.

# Choosing which multiplication



↑ Assign "high-deg"

For "+", "-",  
"high-deg" is max of the arguments.

For "x", "^",  
"high-deg" is the same as the output.

↓ Assign "low-deg"

For "+", "-",  
"low-deg"s are the same as the result.

For "x", "^",  
"low-deg"s are set to "0"s.

# Turning graph to code

## Java code generator

- input: a DAG for  $G$
- outputs C code

## Main steps

- **Workspace allocation.**  
allocate the memory for temporary results.
- **Initialization.**  
initialize constants and polynomials given in the graph  $\mathcal{G}$
- **Evaluation.**  
evaluate  $c$  by following the graph  $\mathcal{G}$   
evaluate  $a, b$  by following the graph  $\mathcal{G}'$

# Output code overview

```
void G_unsigned_long(unsigned long * __restrict__ C,
                     unsigned long * __restrict__ A, unsigned long * __restrict__ B,
                     const unsigned long * __restrict__ t, const unsigned long * __restrict__ u,
                     const unsigned long p, const unsigned long ip, const unsigned long jp, int N){
/*----- Workspace allocation -----*/
unsigned long *wk=(unsigned long *) malloc(30*N*sizeof(unsigned long));

/*----- Initialization -----*/
unsigned long *poly0=(unsigned long *)malloc(1*sizeof(unsigned long));
unsigned long *poly0_pre=(unsigned long *)malloc(1*sizeof(unsigned long));
poly0[0]=3;
poly0_pre[0]=mulredcred(p, ip, jp, 3);
...
/*----- Evaluation C -----*/
mul_plain_unsigned_long(wk+N*0, u, u, p, ip, N);
constant_mul_unsigned_long(wk+N*2, wk+N*0, poly3_pre, 3, 1*N, 2*N, 0*N, 2*N, p, ip);
...
constant_add_unsigned_long(C, wk+N*16, poly0, 1, 1*N, 2, 1*N, 2*N, p, ip);

/*----- Evaluation A -----*/
zero_unsigned_long(wk+N*18, p, ip, N);
...
sub_unsigned_long(A, wk+N*22, wk+N*23, p, ip, 0*N, 1*N);

/*----- Evaluation B -----*/
add_unsigned_long(wk+N*24, u, u, p, ip, 0*N, 1*N);
...
/*----- Workspace, polys free -----*/
free(wk);
free(poly0);
free(poly0_pre);...}
```

# Timings

$$G(x, t, u) = (1 + x + x^2)u^2 - (2 + x)ut^2 - t^2 + 5u + 3.$$

