

Efficient implementation of polynomial arithmetic in a multiple-level programming environment

Xin Li & Marc Moreno Maza



24 July 2008

Aldor & AXIOM Workshop, RISC, Linz, Austria.

Motivations (I) Generic Code

Need for generic code of polynomial arithmetic.

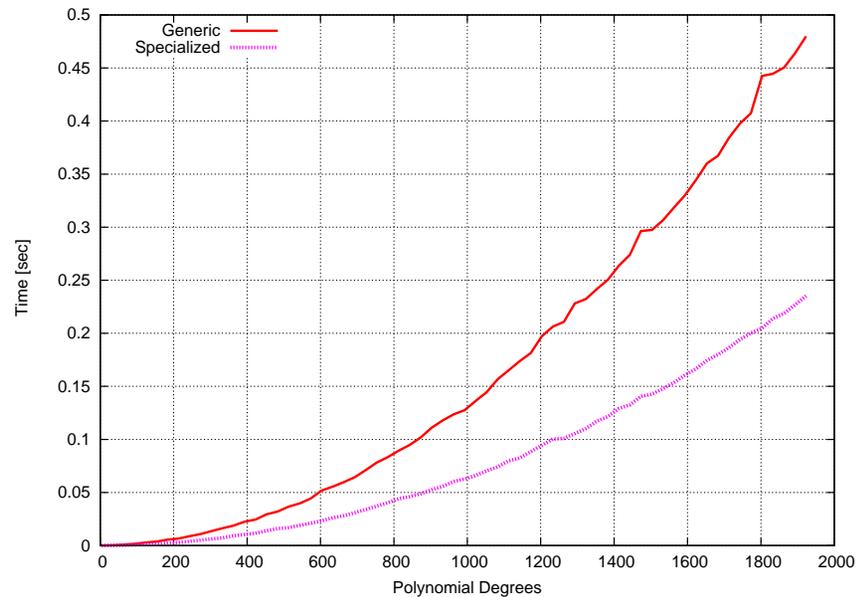
- **Suited to generic programming.**
- **Code reuse, maintainability.**
- **Reasonable speed after compiler optimization.**

We use *AXIOM* and ALDOR.

- **Parametric polymorphism.**
- **Dependent type.**

Motivations (II) Specialized Code

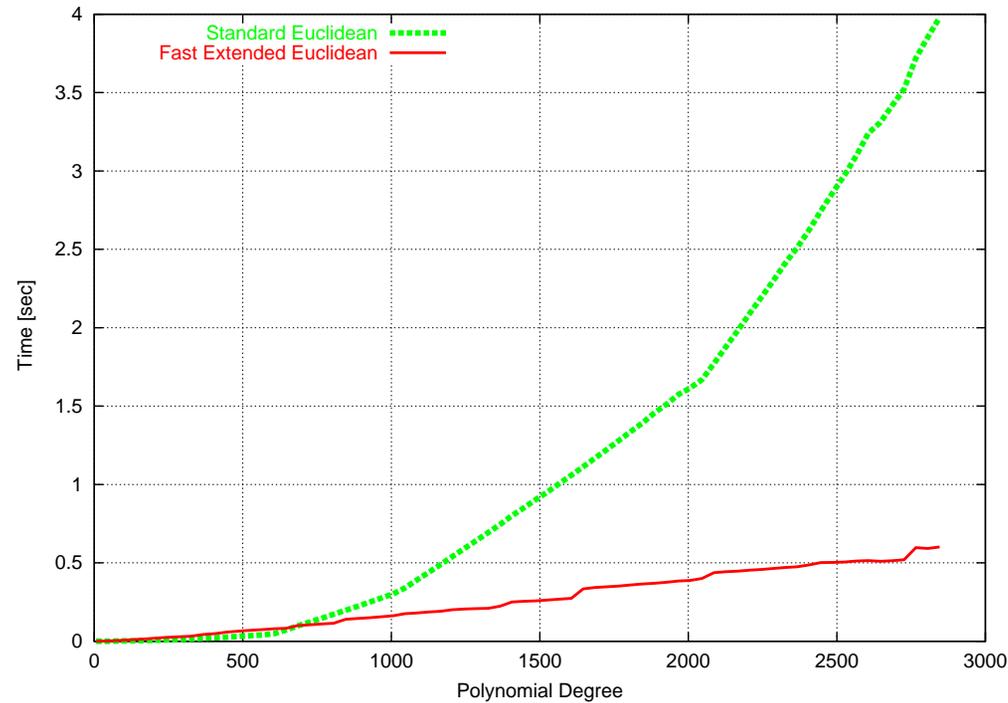
- “Particular” domains deserve their own specialized implementations.
- More efficient.



Univariate Generic F.E.E.A vs. Specialized F.E.E.A Modulo a 27-bit Prime.

Motivations (III) Fast Algorithms

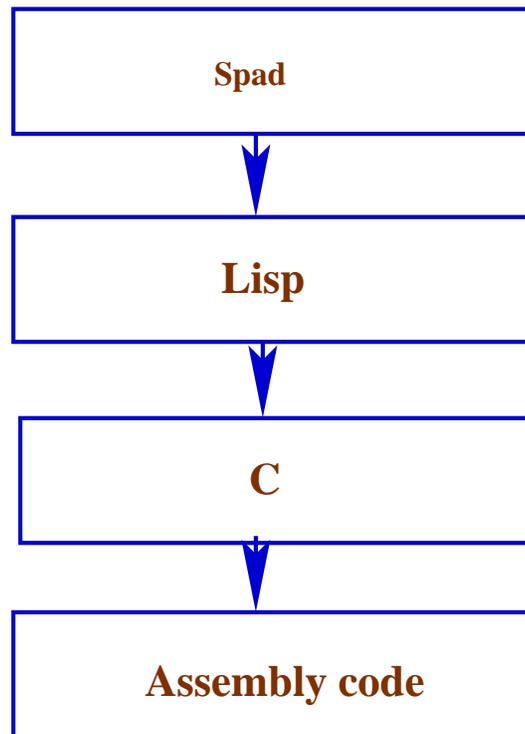
- We are interested in fast polynomial arithmetic.



Univariate Fast E.E.A. vs. Quadratic E.E.A. (both generic code.) Modulo a 27-bit Prime.

Motivations (IV) Implementation Issue

- Focus on implementation issues in *AXIOM*.
- Open *AXIOM* has a multiple-language level construction.



- Mix code at each level for high performance.

The SPAD level

In *AXIOM* at SPAD level:

- A two-level type system – *categories and domains*.

E.g. Ring is the AXIOM category.

- The domain $SUP(R)$, where R has Ring type, implements `UnivariatePolynomialCategory(R)` with sparse data representation.

- The domain $SMP(R, V)$, where R has Ring type, V has VariableSet type, implements `RecursivePolynomialCategory(R, V)` with recursive sparse data representation.

Our implementation for better support dense algorithms:

- The domain $DUP(R)$ implements the same category as $SUP(R)$ does. The data representation is dense.

- The domain $DRMP(R, V)$ implements the same category as $SMP(R, V)$ does. The data representation is recursive dense.

- The benefit of dense domains for dense algorithms in later slides.

The GNU Common Lisp (*GCL*) level

- *Lisp* is a symbolic-expression-oriented language.
- To speed up the performance of *GCL* code.
 - Using statically typed feature.
 - Suppressing run-time type checking.
 - Choosing adapted data structures.

Example-1: Array access in GCL

SPAD code: `array1.i := array2.i`

Compiled Lisp code: `(SETELT |array1| |i| (ELT |array2| |i|))`

Hand-written Lisp code: `(setf (aref array1 i) (aref array2 i))`

```
case t_vector:  
if (index >= seq->v.v_fillp) goto E;  
return(aref(seq, index));
```

Without array bound checking, “aref” is slightly faster.

Example-2: Vector-based recursive dense polynomials

- **Implement $Z/pZ[x_1, \dots, x_n]$ domain using the vector-based representation proposed by R. Fateman.**
- **Each polynomial encoded by a vector, each slot is pointer to another polynomial or a number.**
- **At *SPAD* level using Union type for this disjunction.**
- **At *Lisp* level using the property of dynamic typing.**

Spad Union type is a “cons”.

“cdr” keeps the type info. “car” keeps the value.

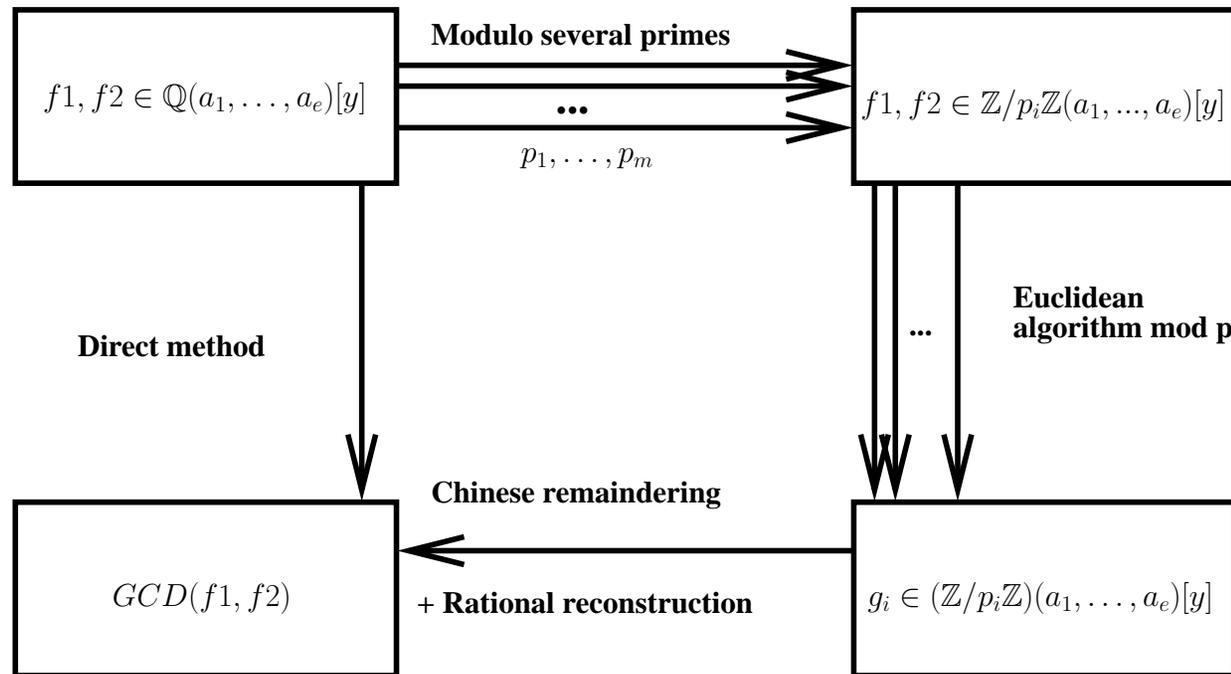
```
struct cons {  
FIRSTWORD;  
object c_cdr;  
object c_car; };
```

Benchmark (I)

- MMA – **Multivariate Modular Arithmetic domain.**
- $MMA(p, V)$ implements the same operations as $DRMP(PF(p), V)$.
- **Benchmark: van Hoeij and Monagan Modular GCD algorithm [1].**

Input: $f_1, f_2 \in \mathbb{Q}(a_1, \dots, a_e)[y]$

Output: $GCD(f_1, f_2)$



[1] A Modular GCD algorithm over Number Fields presented with Multiple Extensions. van Hoeij, Michael Monagan. ISSAC'02 proceedings, 109-116, (2002).

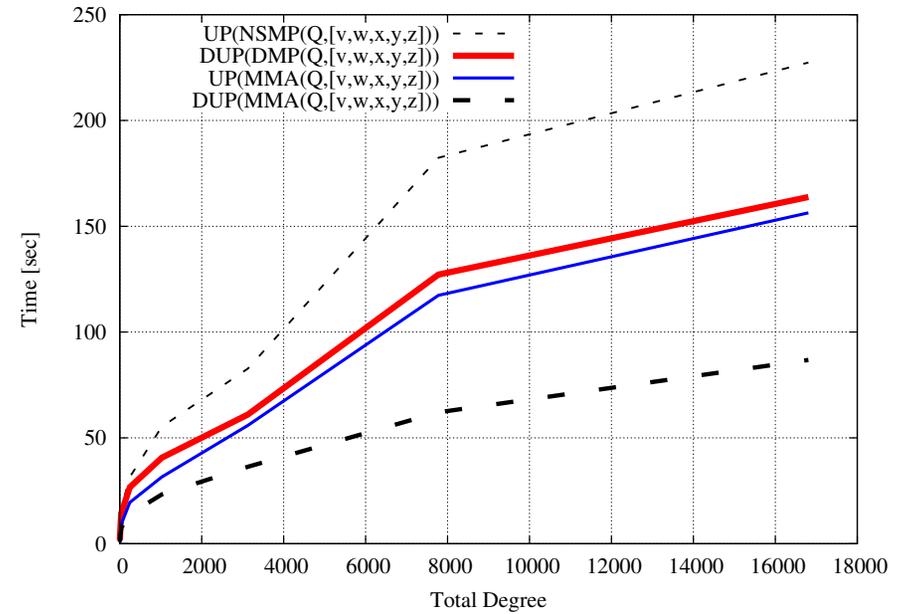
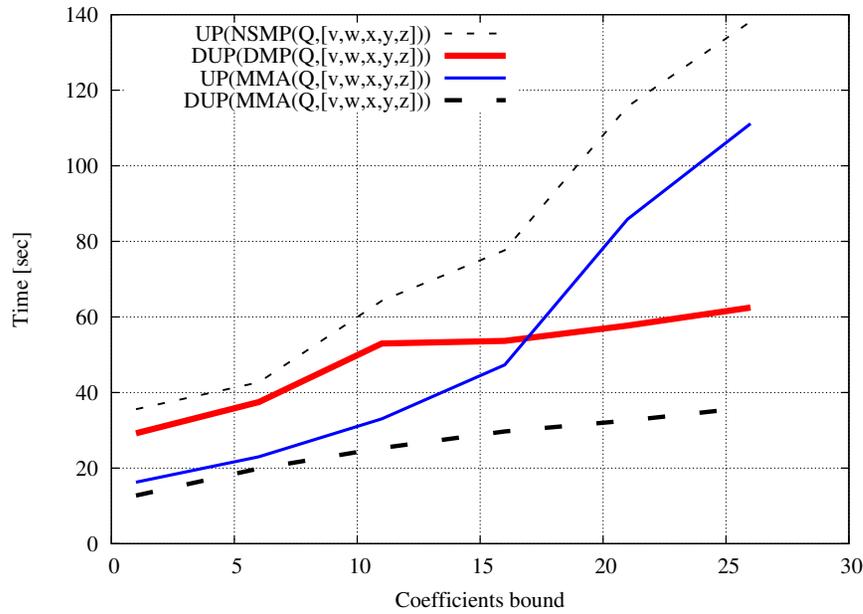
Benchmarks (II)

In *AXIOM* we implemented/used:

$\mathbb{Q}(a_1, a_2, \dots, a_e)$	$\mathbb{K}[y]$
NSMP in SPAD	SUP in SPAD
DMPR in SPAD	DUP in SPAD
MMA in LISP	SUP in SPAD
MMA in LISP	DUP in SPAD

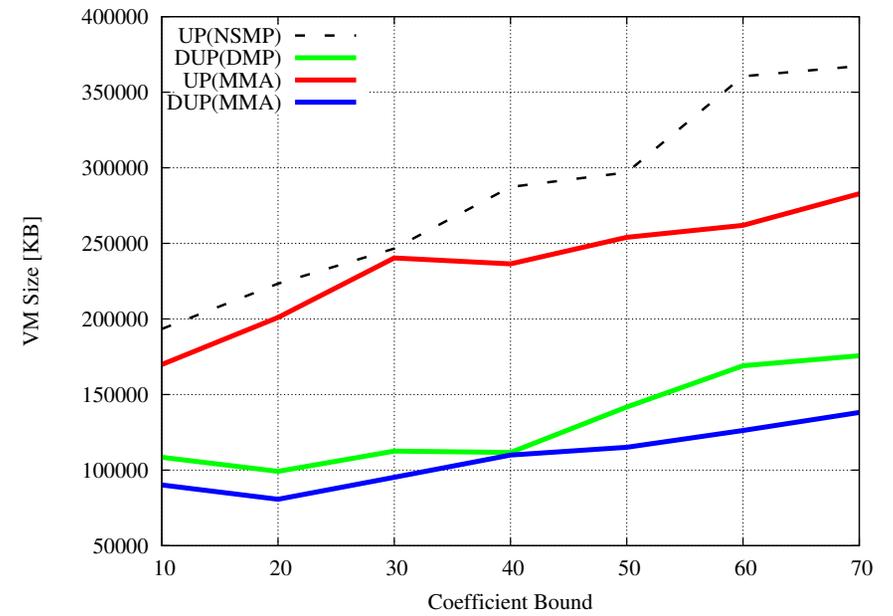
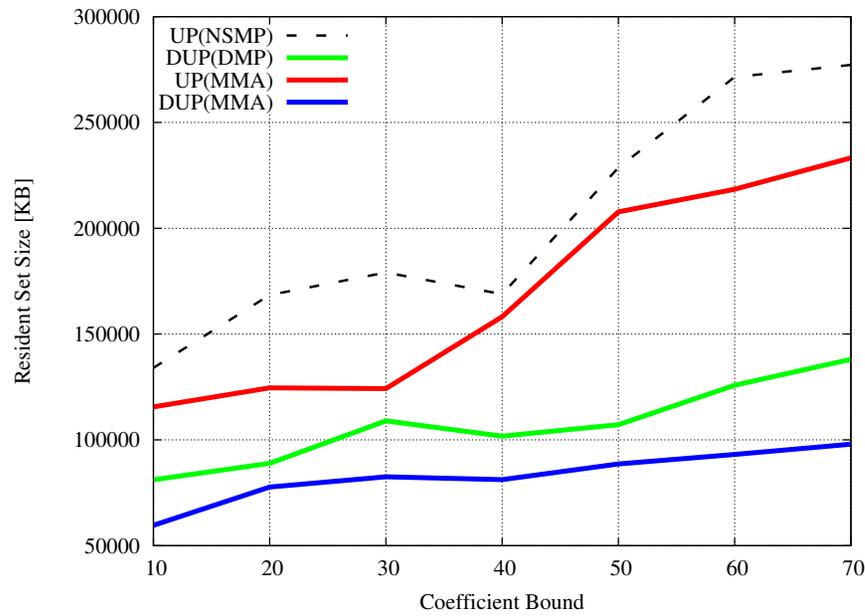
Benchmarks (III)

Timing:



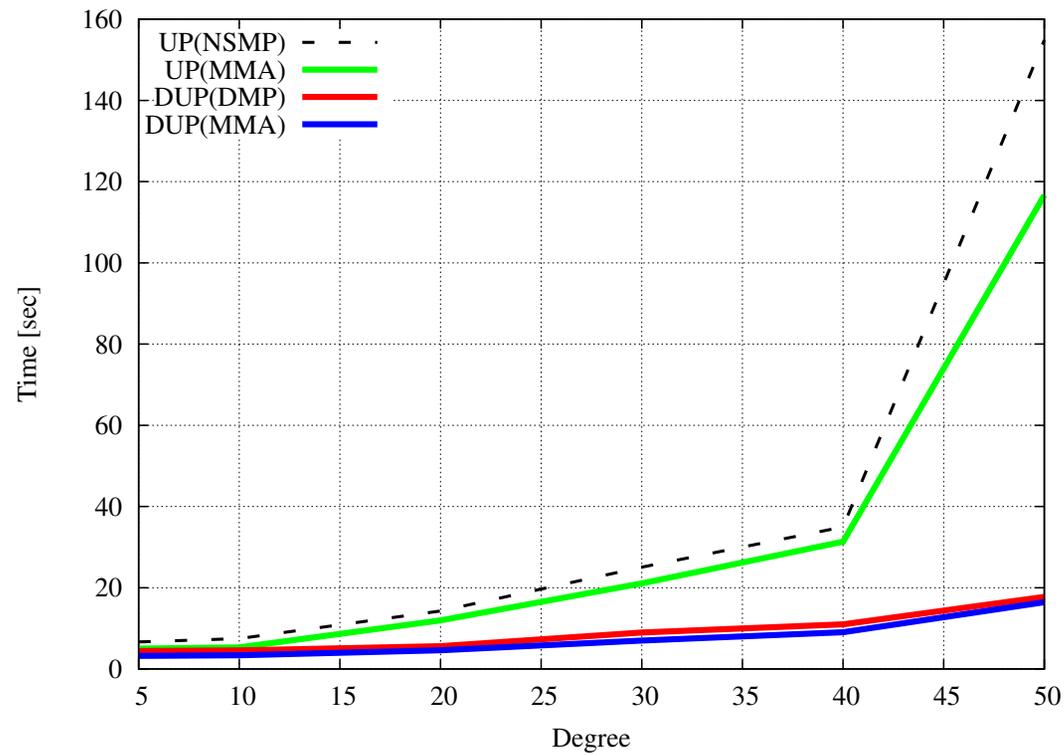
Benchmarks (IV)

Memory Consumption:



Benchmarks (V)

Garbage Collection Time:



The *C* level

We go to *C* level:

- **Implementing efficiency-critical operations requires direct access to machine arithmetic and no much symbolic expression manipulation.**
- ***GCL* is written in *C*, we can extend the *GCL* kernel at *C* level.**
- **Direct use of *C*\C++ or *Assembly* based libraries: NTL, GMP,**

Example-1: Modular integer reduction on special Fourier primes.

- $p = c \cdot 2^k + 1$, $c, k \in \mathbb{Z}^+$, $(c, 2) = 1$, p is a prime;
- **Input:** $Z, p \in \mathbb{Z}$, where $Z \leq (p - 1)^2$.
Output: $r' := (c \cdot Z \text{ rem } p) - \delta$, where $\delta < (c + 1) \cdot p$.
 - Consider $c \cdot Z = q \cdot p + r$.
 - Define $q' := \lfloor \frac{Z}{2^k} \rfloor$.
 - Define $r' := c \cdot (Z - q' \cdot 2^k) - q'$.
 - We have $r' = r - \delta$.
- δ is the difference between r and r' .
- This algorithm only needs “shifts” and “adds/subs”.

Example-2: *FFT* on special Fourier primes.

- **Special Fourier Prime modular reduction used in *FFT*.**
 - **Output: $r' = (c \cdot Z \text{ rem } p) - \delta$.**
 - **Cancel c by its inverse.**
 - **Control δ s in middle stages.**
 - **Remove δ s at the end.**
- **It's more efficient than Montgomery trick and floating point trick.**

Benchmark of FFT

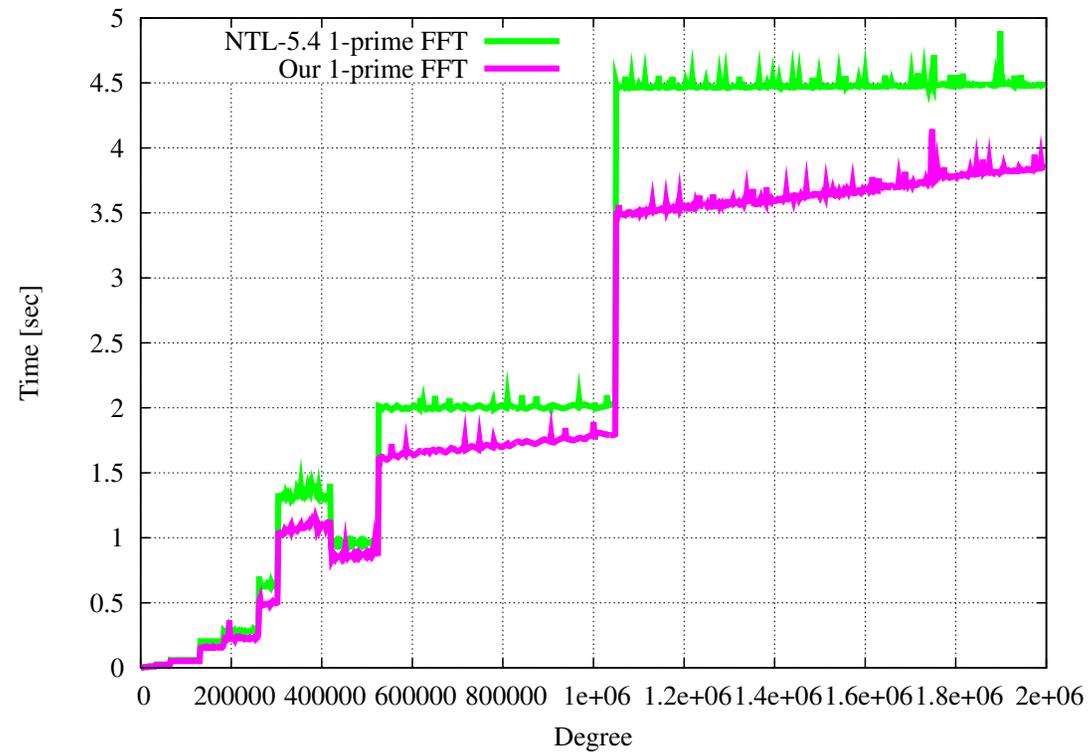


Figure 1: *FFT*-based univariate polynomial multiplication over $\mathbb{Z}/p\mathbb{Z}$,
 $p = 5 * 2^{25} + 1$.

Benchmark of FFT

Event Type	Incl.
Instruction Fetch	 166 165 831
Data Read Access	 82 331 696
Data Write Access	 28 625 645
L1 Instr. Fetch Miss	138
L1 Data Read Miss	 1 104 799
L1 Data Write Miss	 530 870
L2 Instr. Fetch Miss	77
L2 Data Read Miss	 64 220
L2 Data Write Miss	 62 029
L1 Miss Sum	 1 635 807
L2 Miss Sum	 126 326
Cycle Estimation	 195 156 501

NTL-FFT

Event Type	Incl.
Instruction Fetch	 110 363 043
Data Read Access	 33 976 545
Data Write Access	 14 838 926
L1 Instr. Fetch Miss	130
L1 Data Read Miss	 696 219
L1 Data Write Miss	 236 576
L2 Instr. Fetch Miss	71
L2 Data Read Miss	 154 837
L2 Data Write Miss	 47 911
L1 Miss Sum	 932 925
L2 Miss Sum	 202 819
Cycle Estimation	 139 974 193

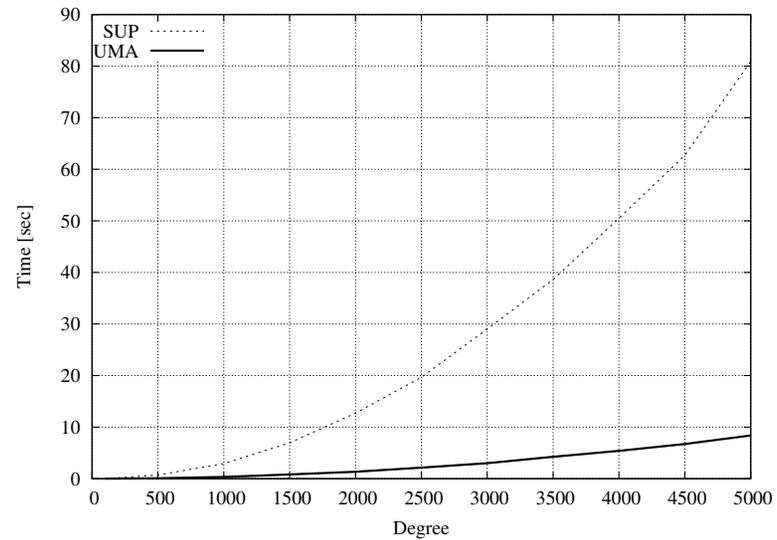
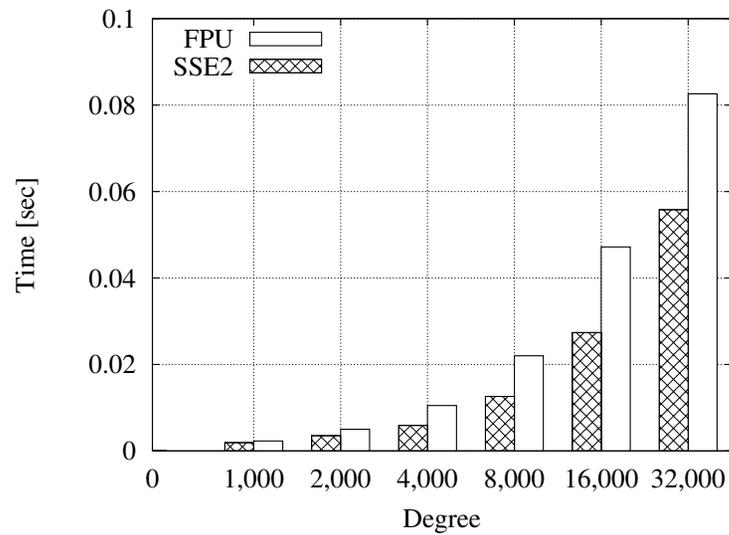
OUR-FFT

Cache setting:

- desc: I1 cache: 16384 B, 32 B, 8-way associative.
- desc: D1 cache: 65536 B, 64 B, 2-way associative.
- desc: L2 cache: 1048576 B, 64 B, 8-way associative.

The ASSEMBLY level

- For using specific hardware features or existing code.



Conclusion and future-work.

- By a few examples, we show that properly mixing code at each *AXIOM* language level may achieve better performance.
- Familiar with strength/weakness of language tools and their optimizer techniques are essential for high performance.
- We hope to construct automatic performance-tuning packages for fast polynomial arithmetic in near future.