0 Introduction

We have compiled some benchmarks that compare the relative performance of NTL (http://shoup.net/ntl/) and FLINT (http://www.flintlib.org/) on some fundamental benchmarks.

0.1 Methodology

All tests were carried out on a very lightly loaded machine with a 64-bit Intel “Haswell” CPU (Intel Xeon CPU E5-2698 v3 at 2.30GHz) with plenty of memory (over 250GB). The operating system was Cent OS. The compiler was gcc v4.8.5.

We compared NTL v10.3.0 with FLINT v2.5.2. These were both built using GMP v6.1.0. All packages were configured using their default configuration flags (except that NTL was configured with the \texttt{TUNE=x86} option). GMP’s configuration script correctly identified the machine as \texttt{haswell-pc-linux-gnu} and used assembly code and other other parameters tuned to the Haswell micro-architecture.

For each basic operation, the test program generated random inputs of a given size using NTL’s pseudo-random generator, and then converted these NTL objects to corresponding FLINT objects. So in all cases, both libraries were working on identical objects. Also, the test program iterated the basic operation sufficiently many time to ensure that at least 3 seconds passed (for the NTL execution), to ensure fairly accurate timing. Time itself was measured using \texttt{getrusage} (system plus user time).

Test programs may be downloaded here: http://shoup.net/ntl/benchtools.tar.

1 Multiplication in $\mathbb{Z}_p[X]$ 

Fig. 1 compares the relative speed of NTL’s \texttt{ZZ.pX mul} routine with FLINT’s \texttt{fmpz_mod_poly_mul} routine. The polynomials were generated at random to have degree less than $n$, and the modulus $p$ was chosen to be a random, odd $k$-bit number.\footnote{NTL’s behavior is somewhat sensitive to whether $p$ is even or odd, and since odd numbers correspond to the case where $p$ is prime, we stuck with those.} The unlabeled columns correspond to $n$-values half-way between the adjacent labeled columns. For example, just to be clear: the entry in the 3rd row and 7th column corresponds to $k = 1024$ and $n = 2048$; the entry in the 3rd row and 8th column corresponds to $k = 1024$ and $n = 2048 + 1024 = 3072$. 

\begin{table}[h]
\centering
\begin{tabular}{|c|c|c|c|c|c|c|c|c|}
\hline
$k$ & $n$ & NTL & FLINT & NTL & FLINT & NTL & FLINT & NTL & FLINT \\
\hline
1024 & 2048 & 1.00 & 1.00 & 1.00 & 1.00 & 1.00 & 1.00 & 1.00 & 1.00 \\
\hline
1024 & 3072 & 1.00 & 1.00 & 1.00 & 1.00 & 1.00 & 1.00 & 1.00 & 1.00 \\
\hline
\end{tabular}
\caption{Comparison of Multiplication Speed in $\mathbb{Z}_p[X]$}
\end{table}
The numbers in the table shown are ratios:

<table>
<thead>
<tr>
<th></th>
<th>FLINT time</th>
<th>NTL time</th>
</tr>
</thead>
</table>

So ratios greater than 1 mean NTL is faster, and ratios less than 1 mean FLINT is faster. The ratios are also color coded. Ratios between 1/2 and 2 are gray (essentially a tie), while ratios greater than 1.2 are green (NTL clearly wins) and those less than 1/2 are red (FLINT clearly wins). Emphasis is added to ratios that are greater than 2 (and 4), or less than 1/2 (and 1/4).

The ratios in the upper right-hand corner of the table essentially compare NTL’s multi-modular FFT algorithm with FLINT’s Kronecker-substitution algorithm. The ratios in the lower left-hand corner of the table essentially compare NTL’s Schönhage-Strassen algorithm with FLINT’s Schönhage-Strassen algorithm.

<table>
<thead>
<tr>
<th>k/1024</th>
<th>n/1024</th>
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<tbody>
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<tr>
<td>1/2</td>
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</tr>
<tr>
<td>8</td>
<td>0.89</td>
</tr>
<tr>
<td>16</td>
<td>0.86</td>
</tr>
</tbody>
</table>

Figure 1: Multiplication in \(\mathbb{Z}_p[X]\): \(n = \) degree bound, \(k = \) #bits in \(p\)

## 2 Multiplication in \(\mathbb{Z}_p[X]/(f)\)

Fig. 2 compares the relative performance of NTL’s \(\mathbb{Z}_p[X]\) \texttt{MulMod} routine with FLINT’s corresponding routine. The NTL routine takes as input precomputations based on \(f\), specifically, a \(\mathbb{Z}_p[X]\texttt{Modulus}\) object. The corresponding FLINT routine is \texttt{fmpz_mod_poly_mulmod_preinv}. The modulus \(p\) was chosen to be a random, odd \(k\)-bit number. The polynomial \(f\) was a random monic polynomial of degree \(n\), while the two multiplicands were random polynomials of degree less than \(n\).

NTL is using a multi-modular FFT strategy throughout, while FLINT is using Kronecker-substitution in the upper right region and Schönhage-Strassen in the lower left region.

The numbers in this table — and all the other tables in this report — have precisely the same meaning as in the table in Fig. 1.

<table>
<thead>
<tr>
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<th>n/1024</th>
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<td>1/2</td>
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<tr>
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<tr>
<td>4</td>
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<td>8</td>
<td>0.78</td>
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<tr>
<td>16</td>
<td>0.57</td>
</tr>
</tbody>
</table>

Figure 2: Multiplication in \(\mathbb{Z}_p[X]/(f)\): \(n = \) degree bound, \(k = \) #bits in \(p\)
3 Squaring in \( \mathbb{Z}_p[X]/(f) \)

Fig. 3 compares the relative performance of NTL’s \( \mathbb{Z}_p[X] \text{SqrMod} \) routine with FLINT’s corresponding routine. The NTL routine takes as input precomputations based on \( f \), specifically, a \( \mathbb{Z}_p[X]\text{Modulus} \) object. The corresponding FLINT routine is \( \text{fmpz\_mod\_poly\_mulmod\_preinv} \). This routine internally checks if the multiplicands point to the same object, and optimizes accordingly. The modulus \( p \) was chosen to be a random, odd \( k \)-bit number. The polynomial \( f \) was a random monic polynomial of degree \( n \), while the polynomial to be squared was a random polynomial of degree less than \( n \).

NTL is using a multi-modular FFT strategy throughout, while FLINT is using Kronecker-substitution in the upper right region and Schönhage-Strassen in the lower left region.

Squaring in \( \mathbb{Z}_p[X]/(f) \) is a critical operation that deserves special attention, as it is the bottleneck in many exponentiation algorithms in \( \mathbb{Z}_p[X]/(f) \).

\[
\begin{array}{c|ccccccccc}
 k/1024 & \frac{1}{4} & \frac{1}{2} & 1 & \frac{1}{2} & 1 & 2 & 4 & 8 & 16 \\
 \hline
 \frac{1}{4} & 2.82 & 2.50 & 3.41 & 2.88 & 3.74 & 2.92 & 3.75 & 2.97 & 3.90 & 3.11 & 4.15 & 3.19 & 4.37 \\
 \frac{1}{2} & 2.30 & 2.10 & 2.48 & 2.36 & 2.82 & 2.72 & 3.27 & 3.28 & 4.34 & 3.46 & 4.79 & 3.88 & 4.98 \\
 1 & 1.76 & 1.61 & 1.86 & 1.74 & 1.99 & 1.95 & 2.28 & 2.55 & 3.00 & 3.72 & 4.74 & 3.49 & 4.57 \\
 2 & 1.42 & 1.30 & 1.45 & 1.36 & 1.52 & 1.46 & 1.63 & 1.77 & 2.02 & 2.35 & 2.64 & 2.95 & 3.52 \\
 4 & 1.63 & 1.53 & 1.65 & 1.54 & 1.65 & 1.49 & 1.60 & 1.58 & 1.72 & 1.78 & 1.99 & 2.06 & 2.16 \\
 8 & 0.87 & 0.91 & 0.89 & 0.90 & 0.90 & 0.89 & 0.89 & 1.00 & 0.97 & 0.98 & 0.94 & 1.03 & 1.08 \\
 16 & 0.64 & 0.63 & 0.65 & 0.64 & 0.65 & 0.64 & 0.65 & 0.66 & 0.65 & 0.66 & 0.69 & 0.69 & 0.71 \\
\end{array}
\]

Figure 3: Squaring in \( \mathbb{Z}_p[X]/(f) \): \( n \) = degree bound, \( k \) = #bits in \( p \)

4 Pre-conditioned multiplication in \( \mathbb{Z}_p[X]/(f) \)

In some situations, one needs to compute \( ab \mod f \), where not only is \( f \) fixed for many operations, but so is \( b \). This arises, for example, in a repeated squaring exponentiation over \( \mathbb{Z}_p[X]/(f) \). As a second example, this arises in computing successive powers of a polynomial mod \( f \), which happens in building the matrix used in Berlekamp’s polynomial factoring algorithm, or in Brent and Kung’s modular composition algorithm. A third example would be scalar/vector products over \( \mathbb{Z}_p[X]/(f) \).

Fig. 4 compares the relative performance of NTL’s \( \mathbb{Z}_p[X] \text{pre-conditioned MulMod} \) routine with FLINT’s corresponding routine. The NTL routine takes as input precomputations based on \( f \) and \( b \), specifically, a \( \mathbb{Z}_p[X]\text{Modulus} \) object and a \( \mathbb{Z}_p[X]\text{Multiplier} \) object. There is no directly comparable FLINT routine — the best choice is the same routine we used above: \( \text{fmpz\_mod\_poly\_mulmod\_preinv} \). The modulus \( p \) was chosen to be a random, odd \( k \)-bit number. The polynomial \( f \) was a random monic polynomial of degree \( n \), while the two multiplicands were random polynomials of degree less than \( n \).

NTL is using a multi-modular FFT strategy throughout, while FLINT is using Kronecker-substitution in the upper right region and Schönhage-Strassen in the lower left region.

5 Factoring in \( \mathbb{Z}_p[X] \)

Fig. 5 compares the relative performance of NTL’s \( \mathbb{Z}_p[X] \text{CanZass} \) factoring routine and the corresponding FLINT routine \( \text{fmpz\_mod\_poly\_factor\_kaltofen\_shoup} \). Both routines implement the
same algorithm, and for the range of parameters that were benchmarked, both routines are the best each library has to offer.

The modulus $p$ was chosen to be a random $k$-bit prime. The polynomial to be factored was a random monic polynomial of degree $n$.

\begin{table}[h]
\centering
\begin{tabular}{|c|c|c|c|c|c|c|c|}
\hline
$k/1024$ & $1/4$ & $1/2$ & 1 & 2 & 4 & 8 & 16 \\
\hline
$1/2$ & 3.84 & 3.58 & 4.06 & 4.03 & 4.63 & 4.56 & 5.39 & 5.72 & 7.62 & 6.21 & 8.18 & 6.91 & 8.77 \\
2 & 2.27 & 2.08 & 2.30 & 2.21 & 2.46 & 2.39 & 2.65 & 2.89 & 3.25 & 3.92 & 4.31 & 5.09 & 5.82 \\
4 & 2.61 & 2.51 & 2.67 & 2.55 & 2.60 & 2.46 & 2.63 & 2.52 & 2.75 & 2.90 & 3.16 & 3.40 & 3.58 \\
8 & 1.40 & 1.38 & 1.43 & 1.37 & 1.43 & 1.41 & 1.47 & 1.50 & 1.60 & 1.58 & 1.54 & 1.67 & 1.75 \\
16 & 1.03 & 1.01 & 1.04 & 1.04 & 1.05 & 1.06 & 1.08 & 1.04 & 1.04 & 1.05 & 1.11 & 1.13 & \\
\hline
\end{tabular}
\caption{Pre-conditioned multiplication in $\mathbb{Z}_p[X]/(f)$: $n = \text{degree bound}$, $k = \#\text{bits in } p$}
\end{table}

\section{Single precision: Multiplication in $\mathbb{Z}_p[X]$}

Fig. 6 compares the relative speed of NTL’s $\texttt{zz.pX mul}$ routine with FLINT’s $\texttt{nmod.poly.mul}$ routine. The polynomials were generated at random to have degree less than $n$, and the modulus $p$ was chosen to be a random, odd $k$-bit number.

Note that in this in the following seven sections, we are working with “single precision” moduli $p$, i.e., moduli that fit into a single machine word. On 64-bit machines, NTL limits such moduli to 60 bits, while FLINT supports moduli up to the full 64 bits.

NTL is using a multi-modular FFT throughout, while FLINT is using Kronecker substitution throughout.

\section{Single precision: Multiplication in $\mathbb{Z}_p[X]/(f)$}

Fig. 7 compares the relative performance of NTL’s $\texttt{zz.pX MulMod}$ routine with FLINT’s corresponding routine. The NTL routine takes as input precomputations based on $f$, specifically, a $\texttt{zz.pXModulus}$ object. The corresponding FLINT routine is $\texttt{nmod_poly_mulmod_preinv}$. The modulus $p$ was chosen to be a random, odd $k$-bit number. The polynomial $f$ was a random monic polynomial of degree $n$, while the two multiplicands are random polynomial of degree less than $n$.

NTL is using a multi-modular FFT throughout, while FLINT is using Kronecker substitution throughout.
### 8 Single precision: Squaring in $\mathbb{Z}_p[X]/(f)$

Fig. 8 compares the relative performance of NTL’s `zz.pX SqrMod` routine with FLINT’s corresponding routine. The NTL routine takes as input precomputations based on $f$, specifically, a `zz.pXModulus` object. The corresponding FLINT routine is `nmod_poly_mulmod_preinv`. This routine internally checks if the multiplicands point to the same object, and optimizes accordingly. The modulus $p$ was chosen to be a random, odd $k$-bit number. The polynomial $f$ was a random monic polynomial of degree $n$, while the polynomial to be squared was a random polynomial of degree less than $n$.

NTL is using a multi-modular FFT throughout, while FLINT is using Kronecker substitution throughout.

### 9 Single precision: Pre-conditioned multiplication in $\mathbb{Z}_p[X]/(f)$

As in §4, we consider the computation of $ab \mod f$, where not only is $f$ fixed for many operations, but so is $b$.

Fig. 9 compares the relative performance of NTL’s `zz.pX pre-conditioned MulMod` routine with
Figure 8: Single precision: Squaring in $\mathbb{Z}_p[X]/(f)$: $n =$ degree bound, $k =$ #bits in $p$

FLINT’s corresponding routine. The NTL routine takes as input precomputations based on $f$ and $b$, specifically, a `zz_pXModulus` object and a `zz_pXMultiplier` object. There is no directly comparable FLINT routine — the best choice is the same routine we used above: `nmod_poly_mulmod_preinv`. The modulus $p$ was chosen to be a random, odd $k$-bit number. The polynomial $f$ was a random monic polynomial of degree $n$, while the two multiplicands are random polynomial of degree less than $n$.

NTL is using a multi-modular FFT throughout, while FLINT is using Kronecker substitution throughout.

Figure 9: Single precision: Pre-conditioned multiplication in $\mathbb{Z}_p[X]/(f)$: $n =$ degree bound, $k =$ #bits in $p$

10 Single precision: Factoring in $\mathbb{Z}_p[X]$

Fig. 10 compares the relative performance of NTL’s `ZZ_pX.CanZass` factoring routine and the corresponding FLINT routine `nmod_poly_factor_kaltofen_shoup`. Both routines implement the same algorithm, and for the range of parameters that were benchmarked, both routines are the best each library has to offer.
The modulus $p$ was chosen to be a random $k$-bit prime. The polynomial to be factored was a random monic polynomial of degree $n$.

<table>
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<tr>
<th>$k$</th>
<th>$n/1024$</th>
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<tbody>
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<td>15</td>
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<tr>
<td>30</td>
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</tbody>
</table>

Figure 10: Single precision: Factoring a degree $n$ polynomial in $\mathbb{Z}_p[X]$, $k = \#\text{bits in } p$

11 Single precision: Matrix multiplication over $\mathbb{Z}_p$

Recently, NTL has had its single-precision matrix arithmetic significantly upgraded to be more cache friendly and to take advantage of SIMD instructions on x86 machines. It has also been upgraded to exploit multi-core machines; however, all experiments reported here are single-core.

On modern Intel processors, NTL’s implementation works (very roughly) as follows. For $p$ up to 23-bits in length, floating point AVX instructions are used. For $p$ up to 31-bits in length, ordinary 64-bit integer multiplication is used. For larger $p$, 128-bit integer multiplication is used.

Fig. 11 compares the relative performance of NTL’s mat\_zz\_p\_mul routine and the corresponding FLINT routine nmod\_mat\_mul. Both routines use a subcubic Strassen recursion.

The modulus $p$ was chosen to be a random $k$-bit odd number. The matrices were random $n \times n$ matrices.

<table>
<thead>
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<th>$k$</th>
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</tr>
</tbody>
</table>

Figure 11: Single precision: Multiplication of $n \times n$ matrices over $\mathbb{Z}_p$, $k = \#\text{bits in } p$

12 Single precision: Matrix inversion over $\mathbb{Z}_p$

Fig. 12 compares the relative performance of NTL’s mat\_zz\_p\_inv routine and the corresponding FLINT routine nmod\_mat\_inv. FLINT reduces to (subcubic) matrix multiplication, while NTL uses a direct (cubic) implementation. The fact that FLINT’s algorithm is asymptotically faster is seen in the table: the ratios get smaller as $n$ increases.
The modulus $p$ was chosen to be a random $k$-bit prime. The matrices were random $n \times n$ invertible matrices.

\[
\begin{array}{|c|cccccc|}
\hline
k & \frac{1}{2} & 1 & \frac{n}{1024} & 2 & 4 & 8 \\
\hline
5 & 5.09 & 4.01 & 4.10 & 4.06 & 3.63 & 3.20 & 2.44 & 2.40 \\
10 & 6.26 & 6.02 & 5.71 & 5.44 & 5.23 & 4.22 & 4.08 & 3.82 & 3.60 \\
25 & 2.20 & 2.16 & 1.96 & 1.95 & 1.76 & 1.77 & 1.54 & 1.59 & 1.36 \\
30 & 1.78 & 1.75 & 1.68 & 1.65 & 1.50 & 1.49 & 1.35 & 1.36 & 1.19 \\
35 & 1.78 & 1.75 & 1.62 & 1.64 & 1.50 & 1.44 & 1.34 & 1.35 & 1.15 \\
40 & 1.78 & 1.75 & 1.69 & 1.64 & 1.52 & 1.43 & 1.33 & 1.35 & 1.15 \\
45 & 1.77 & 1.75 & 1.62 & 1.64 & 1.50 & 1.44 & 1.35 & 1.35 & 1.15 \\
50 & 1.80 & 1.76 & 1.62 & 1.62 & 1.51 & 1.45 & 1.37 & 1.32 & 1.19 \\
55 & 1.85 & 1.78 & 1.64 & 1.58 & 1.46 & 1.46 & 1.36 & 1.31 & 1.19 \\
60 & 1.81 & 1.74 & 1.60 & 1.60 & 1.49 & 1.43 & 1.35 & 1.30 & 1.17 \\
\hline
\end{array}
\]

Figure 12: Single precision: Inversion of $n \times n$ matrices over $\mathbb{Z}_p$, $k = \#\text{bits in } p$

13 Single precision: Nullspace computation over $\mathbb{Z}_p$

Fig. 13 compares the relative performance of NTL’s mat_zz_p kernel routine and the corresponding FLINT routine nmod_mat_nullspace. FLINT reduces to matrix multiplication, while NTL uses a direct implementation.

The modulus $p$ was chosen to be a random $k$-bit prime. The matrices were (roughly) random $n \times n$ matrices of rank $n/2$.

\[
\begin{array}{|c|cccccc|}
\hline
k & \frac{1}{2} & 1 & \frac{n}{1024} & 2 & 4 & 8 \\
\hline
5 & 3.70 & 3.04 & 3.00 & 3.04 & 2.77 & 2.55 & 2.97 & 2.09 & 2.10 \\
10 & 4.36 & 4.09 & 4.09 & 3.72 & 3.75 & 3.33 & 3.60 & 3.14 & 2.90 \\
15 & 5.85 & 6.46 & 5.89 & 5.15 & 6.84 & 5.79 & 5.33 & 5.35 & 4.69 \\
20 & 5.94 & 6.42 & 5.96 & 5.18 & 7.58 & 5.80 & 5.51 & 5.37 & 4.63 \\
25 & 1.87 & 1.75 & 1.55 & 1.53 & 1.35 & 1.37 & 1.30 & 1.26 & 1.06 \\
30 & 1.65 & 1.49 & 1.39 & 1.31 & 1.19 & 1.18 & 1.11 & 1.09 & 0.95 \\
35 & 1.60 & 1.50 & 1.39 & 1.29 & 1.22 & 1.15 & 1.10 & 1.03 & 0.92 \\
40 & 1.60 & 1.48 & 1.39 & 1.31 & 1.22 & 1.15 & 1.09 & 1.02 & 0.92 \\
45 & 1.61 & 1.50 & 1.39 & 1.29 & 1.22 & 1.15 & 1.10 & 1.03 & 0.92 \\
50 & 1.60 & 1.46 & 1.34 & 1.32 & 1.19 & 1.16 & 1.07 & 1.05 & 0.94 \\
55 & 1.62 & 1.47 & 1.35 & 1.32 & 1.18 & 1.16 & 1.08 & 1.04 & 0.91 \\
60 & 1.63 & 1.46 & 1.34 & 1.32 & 1.19 & 1.17 & 1.08 & 1.06 & 0.94 \\
\hline
\end{array}
\]

Figure 13: Single precision: Nullspace computation of $n \times n$ matrices over $\mathbb{Z}_p$, $k = \#\text{bits in } p$

14 Multiplication in $\mathbb{Z}[X]$

Fig. 14 compares the relative speed of NTL’s ZZX mul routine with FLINT’s fmpz_poly_mul routine. The polynomials were generated at random to have degree less than $n$, and coefficients in the range $0, \ldots, 2^k - 1$. 

8
In the upper right region, NTL is using a multi-modular FFT, while FLINT is using Kronecker substitution. In the lower left region, both are using Schönhage-Strassen.

<table>
<thead>
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Figure 14: Multiplication in $\mathbb{Z}[X]$: $n =$ degree bound, $k =$ #bits in each coefficient

15 Concluding remarks

We attempt to draw some conclusions from these benchmarks.

- NTL could perhaps be improved by improving its Schönhage-Strassen implementation, by using Kronecker substitution in place of multi-modular FFT in some parameter ranges, and by fine tuning some of its algorithm crossover points.

- FLINT could perhaps be improved by using a multi-modular FFT in place of Kronecker substitution in some parameter ranges.