

McBits:

fast constant-time

code-based cryptography

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with a solid track record.

... all of the above *at once*.

## Examples of the competition

Some cycle counts on `h9ivy`

(Intel Core i5-3210M, Ivy Bridge)

from [bench.cr.yp.to](http://bench.cr.yp.to):

`mceliece encrypt` 61440

(2008 Biswas–Sendrier,  $2^{80}$ )

`g1s254 DH` 77468

(binary elliptic curve; CHES 2013)

`kumfp127g DH` 116944

(hyperelliptic; Eurocrypt 2013)

`curve25519 DH` 182632

(conservative elliptic curve)

`mceliece decrypt` 1219344

`ronald1024 decrypt` 1340040



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All load/store addresses  
and all branch conditions  
are public. Eliminates  
cache-timing attacks etc.

Similar improvements for CFS.

## Constant-time fanaticism

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“How can this be competitive in speed? Are you really simulating field multiplication with hundreds of bit operations instead of simple log tables?”

Yes, we are.

Not as slow as it sounds!

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Low-end smartphone CPU:

128-bit XOR every cycle.

Ivy Bridge:

256-bit XOR every cycle,

or three 128-bit XORs.

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Typical decoding algorithms  
have add, mult roughly balanced.

Coming next: how to save  
many adds and *most* mults.  
Nice synergy with bitslicing.

## The additive FFT

Fix  $n = 4096 = 2^{12}$ ,  $t = 41$ .

Big final decoding step

is to find all roots in  $\mathbf{F}_{2^{12}}$

of  $f = c_{41}x^{41} + \cdots + c_0x^0$ .

For each  $\alpha \in \mathbf{F}_{2^{12}}$ ,

compute  $f(\alpha)$  by Horner's rule:

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Our cost: **6.01** adds, **2.09** mults.

Asymptotics:

normally  $t \in \Theta(n / \lg n)$ ,

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Wait a minute.

Didn't we learn in school

that FFT evaluates

an  $n$ -coeff polynomial

at  $n$  points

using  $n^{1+o(1)}$  operations?

Isn't this better than  $n^2 / \lg n$ ?

Standard radix-2 FFT:

Want to evaluate

$$f = c_0 + c_1x + \cdots + c_{n-1}x^{n-1}$$

at all the  $n$ th roots of 1.

Write  $f$  as  $f_0(x^2) + xf_1(x^2)$ .

Observe big overlap between

$$f(\alpha) = f_0(\alpha^2) + \alpha f_1(\alpha^2),$$

$$f(-\alpha) = f_0(\alpha^2) - \alpha f_1(\alpha^2).$$

$f_0$  has  $n/2$  coeffs;

evaluate at  $(n/2)$ nd roots of 1

by same idea recursively.

Similarly  $f_1$ .

Useless in char 2:  $\alpha = -\alpha$ .

Standard workarounds are painful.

FFT considered impractical.

1988 Wang–Zhu,

independently 1989 Cantor:

“additive FFT” in char 2.

Still quite expensive.

1996 von zur Gathen–Gerhard:

some improvements.

2010 Gao–Mateer:

much better additive FFT.

We use Gao–Mateer,

plus some new improvements.

Gao and Mateer evaluate

$$f = c_0 + c_1x + \cdots + c_{n-1}x^{n-1}$$

on a size- $n$   $\mathbf{F}_2$ -linear space.

Main idea: Write  $f$  as

$$f_0(x^2 + x) + xf_1(x^2 + x).$$

Big overlap between  $f(\alpha) =$

$$f_0(\alpha^2 + \alpha) + \alpha f_1(\alpha^2 + \alpha)$$

and  $f(\alpha + 1) =$

$$f_0(\alpha^2 + \alpha) + (\alpha + 1)f_1(\alpha^2 + \alpha).$$

“Twist” to ensure  $1 \in$  space.

Then  $\{\alpha^2 + \alpha\}$  is a

size- $(n/2)$   $\mathbf{F}_2$ -linear space.

Apply same idea recursively.

We generalize to

$$f = c_0 + c_1x + \cdots + c_t x^t$$

for any  $t < n$ .

⇒ several optimizations,  
not all of which are automated  
by simply tracking zeros.

For  $t = 0$ : copy  $c_0$ .

For  $t \in \{1, 2\}$ :

$f_1$  is a constant.

Instead of multiplying  
this constant by each  $\alpha$ ,  
multiply only by generators  
and compute subset sums.

# Syndrome computation

Initial decoding step: compute

$$s_0 = r_1 + r_2 + \cdots + r_n,$$

$$s_1 = r_1\alpha_1 + r_2\alpha_2 + \cdots + r_n\alpha_n,$$

$$s_2 = r_1\alpha_1^2 + r_2\alpha_2^2 + \cdots + r_n\alpha_n^2,$$

$\vdots$ ,

$$s_t = r_1\alpha_1^t + r_2\alpha_2^t + \cdots + r_n\alpha_n^t.$$

$r_1, r_2, \dots, r_n$  are received bits scaled by Goppa constants.

Typically precompute matrix mapping bits to syndrome.

Not as slow as Chien search but still  $n^{2+o(1)}$  and huge secret key.

Compare to multipoint evaluation:

$$f(\alpha_1) = c_0 + c_1\alpha_1 + \cdots + c_t\alpha_1^t,$$

$$f(\alpha_2) = c_0 + c_1\alpha_2 + \cdots + c_t\alpha_2^t,$$

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Amazing consequence:

syndrome computation is as few ops as multipoint evaluation.

Eliminate precomputed matrix.

Transposition principle:

If a linear algorithm

computes a matrix  $M$

then reversing edges and

exchanging inputs/outputs

computes the transpose of  $M$ .

1956 Bordewijk;

independently 1957 Lupanov

for Boolean matrices.

1973 Fiduccia analysis:

preserves number of mults;

preserves number of adds plus

number of nontrivial outputs.

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Built new interpreter, allowing some code compression.

Still big; still some overhead.

Better solution:  
started at additive FFT,  
wrote down transposition  
with same loops etc.

Small code, no overhead.

Speedups of additive FFT  
translate easily  
to transposed algorithm.

Further savings:  
merged first stage with  
scaling by Goppa constants.

## Secret permutation

Additive FFT  $\Rightarrow f$  values at field elements *in a standard order*.

This is not the order needed in code-based crypto!

Must apply a secret permutation, part of the secret key.

Same issue for syndrome.

Solution: Batchier sorting.

Almost done with faster solution: Beneš network.



## Results

60493 Ivy Bridge cycles:

8622 for permutation.

20846 for syndrome.

7714 for BM.

14794 for roots.

8520 for permutation.

Code will be public domain.

We're still speeding it up.

More information:

[cr.yp.to/papers.html#mcbits](http://cr.yp.to/papers.html#mcbits)