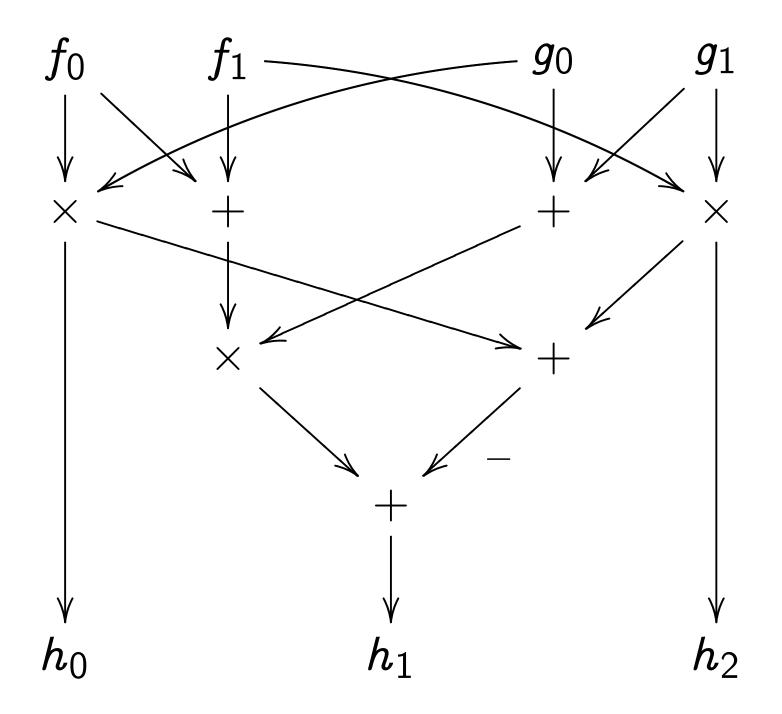
#### The tangent FFT

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See online version of paper, particularly for bibliography: http://cr.yp.to/papers.html#tangentfft

#### Algebraic algorithms



- × multiplies its two inputs.
- + adds its two inputs.
- + subtracts its two inputs.

This "**R**-algebraic algorithm" computes product  $h_0+h_1x+h_2x^2$  of  $f_0+f_1x$ ,  $g_0+g_1x\in \mathbf{R}[x]$ .

More precisely: It computes the coeffs of the product (on standard basis  $1, x, x^2$ ) given the coeffs of the factors (on standard bases 1, x and 1, x).

3 mults, 4 adds.

Compare to obvious algorithm:

4 mults, 1 add.

(1963 Karatsuba)

#### Algebraic complexity

Are 3 mults, 4 adds better than 4 mults, 1 add?

In this talk: No!

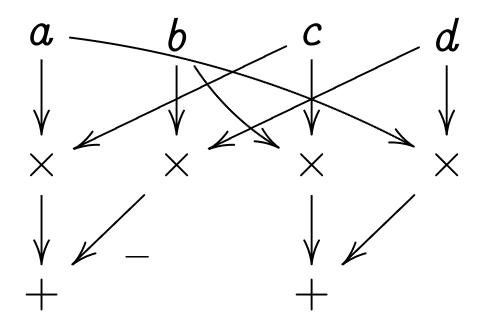
Cost measure for this talk: "total **R**-algebraic complexity."

- + ("add"): cost 1.
- $+^{-}$  (also "add"): cost 1.
- $\times$  ("mult"): cost 1.

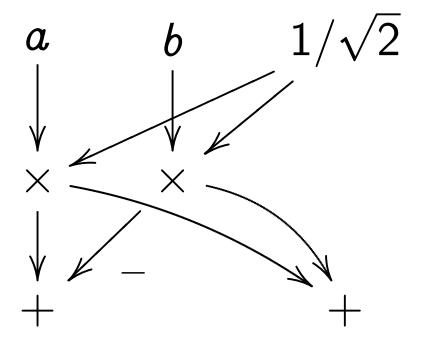
Constant in R: cost 0.

- 3 mults, 4 adds: cost 7.
- 4 mults, 1 add: cost 5.

Cost 6 to multiply in  $\mathbf{C}$  (on standard basis 1, i):



Cost 4 to multiply by  $\sqrt{i}$ :



Can use (e.g.) Pentium M's 80-bit floating-point instructions to approximate operations in **R**.

Each cycle, Pentium M follows
≤ 1 floating-point instruction.
So #Pentium M cycles
> total **R**-algebraic complexity.

Usually can achieve #cycles  $\approx$  total **R**-algebraic complexity. Analysis of "usually" and " $\approx$ " is beyond this talk.

Many other cost measures.

Some measures emphasize adds. e.g. 64-bit fp on one core of Core 2 Duo: #cycles  $\approx \max\{\#\mathbf{R}\text{-adds}, \#\mathbf{R}\text{-mults}\}/2$ . Typically more adds than mults.

Some measures emphasize mults.
e.g. Dedicated hardware
for floating-point arithmetic:
mults more expensive than adds.

But "cost" in this talk means  $\#\mathbf{R}$ -adds  $+ \#\mathbf{R}$ -mults.

#### Fast Fourier transforms

Define  $\zeta_n \in \mathbf{C}$  as  $\exp(2\pi i/n)$ . Define  $T_n : \mathbf{C}[x]/(x^n-1) \hookrightarrow \mathbf{C}^n$  as  $f \mapsto f(1), f(\zeta_n), \ldots, f(\zeta_n^{n-1})$ .

Can very quickly compute  $T_n$ .

First publication of fast algorithm: 1866 Gauss.

Easy to see that Gauss's FFT uses  $O(n \lg n)$  arithmetic operations if  $n \in \{1, 2, 4, 8, \ldots\}$ .

Several subsequent reinventions, ending with 1965 Cooley/Tukey.

Inverse map is also very fast.

Multiplication in  $\mathbb{C}^n$  is very fast.

1966 Sande, 1966 Stockham: Can very quickly multiply in  $\mathbf{C}[x]/(x^n-1)$  or  $\mathbf{C}[x]$  or  $\mathbf{R}[x]$  by mapping  $\mathbf{C}[x]/(x^n-1)$  to  $\mathbf{C}^n$ . "Fast convolution."

Given  $f,g\in {f C}[x]/(x^n-1)$ : compute fg as  $T_n^{-1}(T_n(f)T_n(g))$ .

Given  $f,g \in \mathbf{C}[x]$ ,  $\deg fg < n$ : compute fg from its image in  $\mathbf{C}[x]/(x^n-1)$ .

Cost  $O(n \lg n)$ .

#### A closer look at costs

More precise analysis of Gauss FFT (and Cooley-Tukey FFT):

 $\mathbf{C}[x]/(x^n-1) \hookrightarrow \mathbf{C}^n$  using  $n \mid \mathbf{C}$  adds (costing 2 each),  $(n \mid \mathbf{g} \mid n)/2 \mid \mathbf{C}$ -mults (6 each), if  $n \in \{1, 2, 4, 8, \ldots\}$ .

Total cost  $5n \lg n$ .

After peephole optimizations:

cost  $5n \lg n - 10n + 16$  if  $n \in \{4, 8, 16, 32, \ldots\}$ .

Either way,  $5n \lg n + O(n)$ . This talk focuses on the 5. What about cost of convolution?

 $5n \lg n + O(n)$  to compute  $T_n(f)$ ,  $5n \lg n + O(n)$  to compute  $T_n(g)$ , O(n) to multiply in  ${\bf C}^n$ , similar  $5n \lg n + O(n)$  for  $T_n^{-1}$ .

Total cost  $15n\lg n + O(n)$  to compute  $fg \in \mathbf{C}[x]/(x^n-1)$  given  $f,g \in \mathbf{C}[x]/(x^n-1)$ .

Total cost  $(15/2)n\lg n + O(n)$  to compute  $fg \in \mathbf{R}[x]/(x^n-1)$  given  $f,g \in \mathbf{R}[x]/(x^n-1)$ : map  $\mathbf{R}[x]/(x^n-1) \hookrightarrow \mathbf{R}^2 \oplus \mathbf{C}^{n/2-1}$  (Gauss) to save half the time.

1968 R. Yavne: Can do better! Cost  $4n \lg n + O(n)$  to map  $\mathbf{C}[x]/(x^n-1) \hookrightarrow \mathbf{C}^n$ , if  $n \in \{1, 2, 4, 8, 16, \ldots\}$ .

1968 R. Yavne: Can do better! Cost  $4n \lg n + O(n)$  to map  $\mathbf{C}[x]/(x^n-1) \hookrightarrow \mathbf{C}^n$ , if  $n \in \{1, 2, 4, 8, 16, \ldots\}$ .

2004 James Van Buskirk:

Can do better!

Cost  $(34/9)n \lg n + O(n)$ .

Expositions of the new algorithm:

Frigo, Johnson,

in IEEE Trans. Signal Processing;

Lundy, Van Buskirk,

in Computing;

Bernstein, this AAECC paper.

### Understanding the FFT

If  $f \in \mathbf{C}[x]$  and  $f \mod x^4 - 1 = f_0 + f_1 x + f_2 x^2 + f_3 x^3$  then  $f \mod x^2 - 1 = (f_0 + f_2) + (f_1 + f_3)x$ ,  $f \mod x^2 + 1 = (f_0 - f_2) + (f_1 - f_3)x$ .

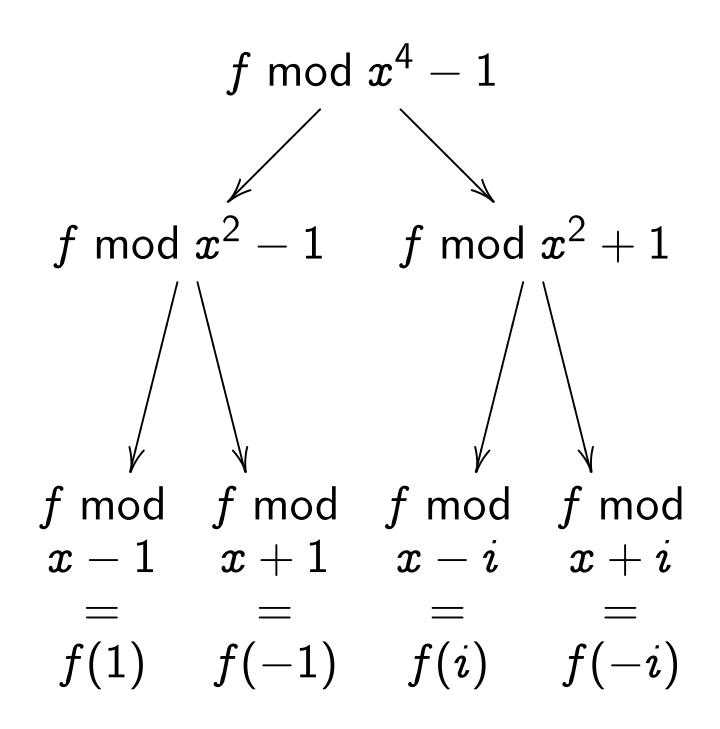
Given  $f \mod x^4 - 1$ , cost 8 to compute  $f \mod x^2 - 1$ ,  $f \mod x^2 + 1$ .

" $\mathbf{C}[x]$ -morphism  $\mathbf{C}[x]/(x^4-1) \hookrightarrow$   $\mathbf{C}[x]/(x^2-1) \oplus \mathbf{C}[x]/(x^2+1)$ ."

If  $f \in \mathbf{C}[x]$  and  $f \mod x^{2n} - r^2 =$  $f_0 + f_1 x + \cdots + f_{2n-1} x^{2n-1}$  then  $f \mod x^n - r =$  $(f_0 + rf_n) + (f_1 + rf_{n+1})x$  $+(f_2+rf_{n+2})x^2+\cdots$  $f \mod x^n + r =$  $(f_0 - rf_n) + (f_1 - rf_{n+1})x$  $+(f_2-rf_{n+2})x^2+\cdots$ 

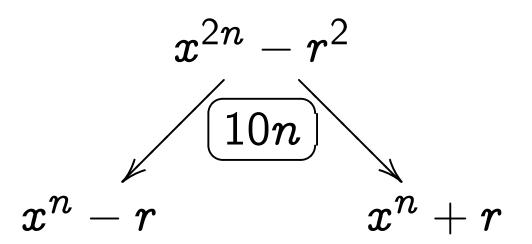
Given  $f_0, f_1, \ldots, f_{2n-1} \in \mathbf{C}$ ,  $\cos t \leq 10n$  to compute  $f_0 + rf_n, f_1 + rf_{n+1}, \ldots, f_0 - rf_n, f_1 - rf_{n+1}, \ldots$  Note: can compute in place.

### The FFT: Do this recursively!

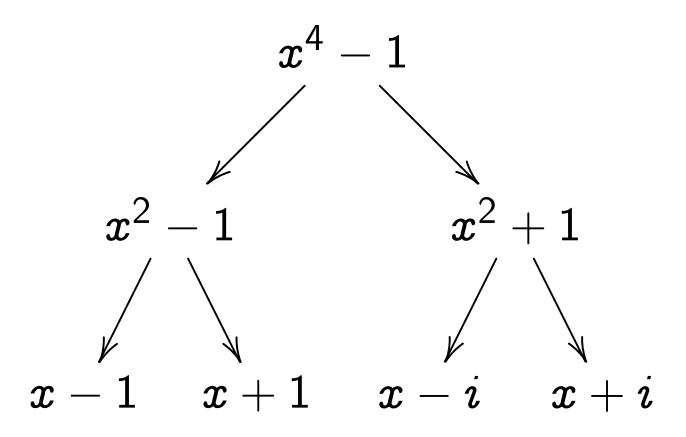


(expository idea: 1972 Fiduccia)

Modulus tree for one step:



Modulus tree for full size-4 FFT:



#### Alternative: the twisted FFT

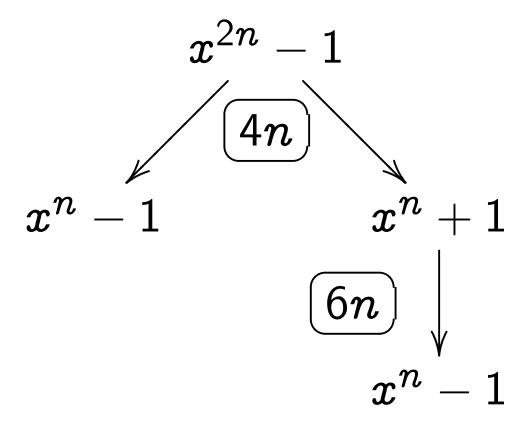
If 
$$f \in \mathbf{C}[x]$$
 and  $f \mod x^n + 1 =$   $g_0 + g_1x + g_2x^2 + \cdots$  then  $f(\zeta_{2n}x) \mod x^n - 1 =$   $g_0 + \zeta_{2n}g_1x + \zeta_{2n}^2g_2x^2 + \cdots$ 

"C-morphism 
$${f C}[x]/(x^n+1) \hookrightarrow {f C}[x]/(x^n-1)$$
 by  $x\mapsto \zeta_{2n}x$ ."

Modulus tree:

$$egin{array}{c} x^n+1 \ \hline 6n \ x^n-1 \end{array}$$

Merge with the original FFT trick:



"Twisted FFT" applies this modulus tree recursively.

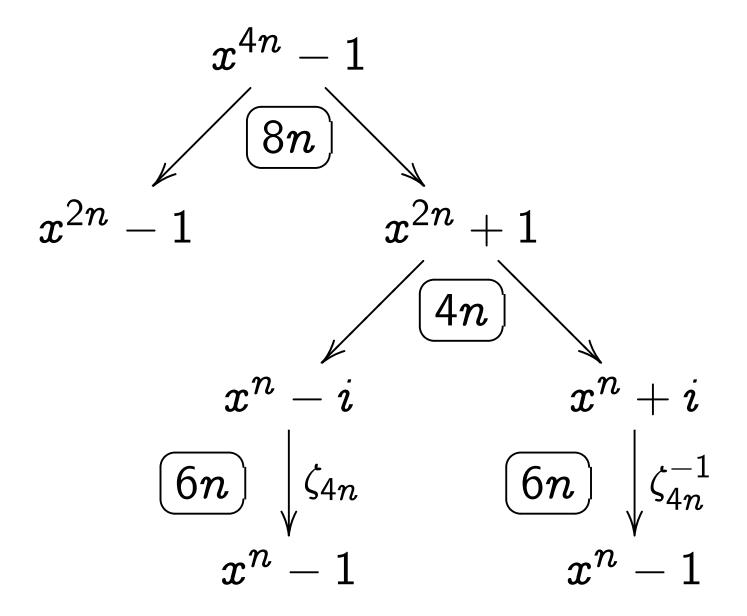
Cost  $5n \lg n + O(n)$ , just like the original FFT.

## The split-radix FFT

FFT and twisted FFT end up with same number of mults by  $\zeta_n$ , same number of mults by  $\zeta_{n/2}$ , same number of mults by  $\zeta_{n/4}$ , etc.

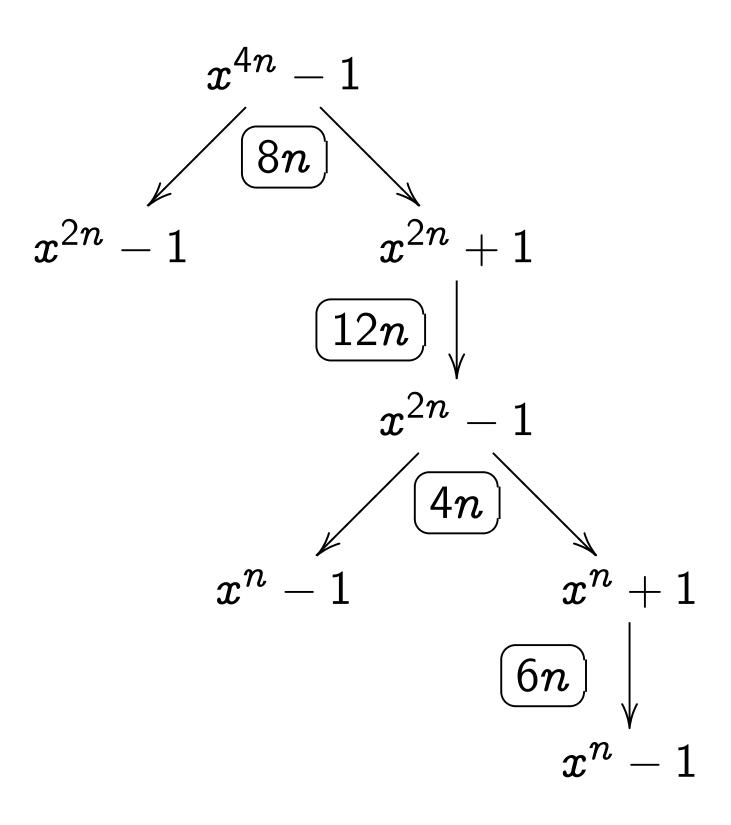
Is this necessary? No! Split-radix FFT: more easy mults. "Don't twist until you see the whites of their i's."

(Can use same idea to speed up Schönhage-Strassen algorithm for integer multiplication.)



Split-radix FFT applies this modulus tree recursively. Cost  $4n \lg n + O(n)$ .

Compare to how twisted FFT splits 4n into 2n, n, n:



## The tangent FFT

Several ways to achieve cost 6 for mult by  $e^{i\theta}$ .

One approach: Factor  $e^{i\theta}$  as  $(1+i\tan\theta)\cos\theta$ . Cost 2 for mult by  $\cos\theta$ . Cost 4 for mult by  $1+i\tan\theta$ .

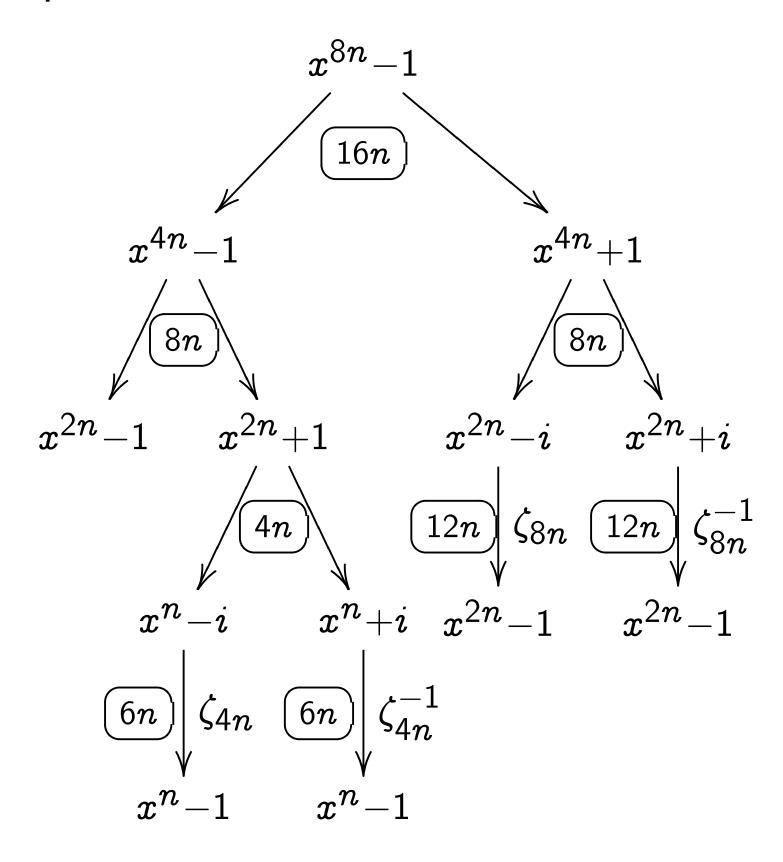
For stability and symmetry, use  $\max\{|\cos\theta|, |\sin\theta|\}$  instead of  $\cos\theta$ .

Surprise (Van Buskirk): Can merge some cost-2 mults! Rethink basis of  $\mathbf{C}[x]/(x^n-1)$ . Instead of  $1,x,\ldots,x^{n-1}$  use  $1/s_{n,0},x/s_{n,1},\ldots,x^{n-1}/s_{n,n-1}$  where  $s_{n,k}=\max\{\left|\cos\frac{2\pi k}{n}\right|,\left|\sin\frac{2\pi k}{n}\right|\}$   $\max\{\left|\cos\frac{2\pi k}{n/4}\right|,\left|\sin\frac{2\pi k}{n/4}\right|\}$   $\max\{\left|\cos\frac{2\pi k}{n/4}\right|,\left|\sin\frac{2\pi k}{n/4}\right|\}$   $\max\{\left|\cos\frac{2\pi k}{n/16}\right|,\left|\sin\frac{2\pi k}{n/16}\right|\}$ 

Now  $(g_0,g_1,\ldots,g_{n-1})$  represents  $g_0/s_{n,0}+\cdots+g_{n-1}x^{n-1}/s_{n,n-1}.$ 

Note that  $s_{n,k}=s_{n,k+n/4}$ . Note that  $\zeta_n^k(s_{n/4,k}/s_{n,k})$  is  $\pm (1+i\tan\cdots)$  or  $\pm (\cot\cdots+i)$ .

# Look at how split-radix splits 8n into 2n, 2n, 2n, n, n:



New basis saves 12n:

4n in  $\zeta_{8n}$  twist, 4n in  $\zeta_{8n}^{-1}$  twist, 2n in  $\zeta_{4n}$  twist, 2n in  $\zeta_{4n}^{-1}$  twist.

New basis costs 8n:

4n to change basis of  $x^{2n}+1$ , 4n to change basis of top-left  $x^{2n}-1$ .

Overall 68n instead of 72n.

Recurse:  $(34/9)n \lg n + O(n)$ , as in 2004 Van Buskirk.

Open: Can 34/9 be improved?