

## Outline

### TSTE87 ASIC for DSP Lecture 11 2014

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Efficient polynomial multiplication

## Introduction

- ▶ In 1960 Anatolii Alexeevitch Karatsuba attended a lecture by Andrey Kolmogorov at Moscow State University, where Kolmogorov presented his conjecture that a multiplication of two  $n$ -digit numbers requires  $O(n^2)$  elementary operations
- ▶ The following week, Karatsuba presented an algorithm requiring  $O(n^{\log_2 3})$  elementary operations
- ▶ Kolmogorov then wrote and published a paper with Karatsuba as the author without telling him

## What was the idea?

- ▶ Consider a polynomial multiplication

$$(a_1x + a_0)(b_1x + b_0) = c_2x^2 + c_1x + c_0 \quad (1)$$

where we want to compute  $c_2$  to  $c_0$

- ▶ A straightforward realization require the computation of

$$a_1b_1, a_0b_1, a_1b_0, a_0b_0 \quad (2)$$

i.e., four multiplications

- ▶ Instead, by computing (e.g.)  $(a_1 + a_0)(b_1 + b_0)$ ,  $a_1b_1$ , and  $a_0b_0$  it is enough to perform three multiplications

$$c_2 = a_1b_1 \quad (3)$$

$$c_0 = a_0b_0 \quad (4)$$

$$c_1 = (a_1 + a_0)(b_1 + b_0) - c_0 - c_2 \quad (5)$$

## But two term polynomials are not so common!?

- ▶ Wrong!
- ▶ Fixed-point multiplication using  $B$ -bit numbers (use polynomial variable  $2^{B/2}$ )

$$(a_1 2^{B/2} + a_0)(b_1 2^{B/2} + b_0) = c_2 2^B + c_1 2^{B/2} + c_0 \quad (6)$$

$a_1, b_1$  are the most significant parts,  $a_0, b_0$  are the least significant parts

- ▶ Complex multiplication using polynomial variable  $j$

$$(a_1 j + a_0)(b_1 j + b_0) = c_2 j^2 + c_1 j + c_0 = c_0 - c_2 + j c_1 \quad (7)$$

- ▶ FIR filters using polynomial variable  $z$

$$(H_1(z^2)z + H_0(z^2))(X_1(z^2)z + X_0(z^2)) = c_2 z^2 + c_1 z + c_0 \quad (8)$$

where  $H_1, X_1$  are the odd indexed values of impulse response and input data, respectively, and  $H_0, X_0$  are the even indexed values

## Oh, I see...

- ▶ On matrix form this becomes

$$\begin{bmatrix} a_0 b_0 \\ (a_1 + a_0)(b_1 + b_0) \\ a_1 b_1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} c_0 \\ c_1 \\ c_2 \end{bmatrix} \quad (13)$$

- ▶ Now, to determine  $c_0$  to  $c_2$  invert the matrix as

$$\begin{bmatrix} c_0 \\ c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & -1 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} a_0 b_0 \\ (a_1 + a_0)(b_1 + b_0) \\ a_1 b_1 \end{bmatrix} \quad (14)$$

- ▶ This can be further rewritten as

$$\begin{bmatrix} c_0 \\ c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & -1 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} a_0 & 0 & 0 \\ 0 & a_1 + a_0 & 0 \\ 0 & 0 & a_1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} b_0 \\ b_1 \end{bmatrix} \quad (15)$$

## OK, but how can you determine these?

- ▶ Let us start with the property that an  $M$ -term polynomial is uniquely defined by the value of  $M$  points
- ▶ In addition, instead of using the polynomial itself one can use any derivative degree of it
- ▶ Let us consider the two-term polynomial multiplication

$$(a_1 x + a_0)(b_1 x + b_0) = c_2 x^2 + c_1 x + c_0 \quad (9)$$

- ▶ Now, evaluate the polynomial for the points  $x = 0, 1, \infty$

$$x = 0 \Rightarrow a_0 b_0 = c_0 \quad (10)$$

$$x = 1 \Rightarrow (a_1 + a_0)(b_1 + b_0) = c_2 + c_1 + c_0 \quad (11)$$

$$x = \infty \Rightarrow a_1 b_1 = c_2 \quad (12)$$

## Let us play around for a bit

- ▶ What other equations can one use?
- ▶ Evaluate for  $x = 2$  instead of  $x = 1$

$$(2a_1 + a_0)(2b_1 + b_0) = 4c_2 + 2c_1 + c_0 \quad (16)$$

$$\begin{bmatrix} a_0 b_0 \\ (2a_1 + a_0)(2b_1 + b_0) \\ a_1 b_1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 2 & 4 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} c_0 \\ c_1 \\ c_2 \end{bmatrix} \quad (17)$$

$$\begin{bmatrix} c_0 \\ c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ -\frac{1}{2} & \frac{1}{2} & -2 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} a_0 & 0 & 0 \\ 0 & 2a_1 + a_0 & 0 \\ 0 & 0 & a_1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 2 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} b_0 \\ b_1 \end{bmatrix} \quad (18)$$

- ▶ Quite OK assuming shifts are free, but not better in any way

## More alternatives

- ▶ Let us take a look at some alternatives

| Point            | Matrix row     | Multiplication(s)                 |
|------------------|----------------|-----------------------------------|
| $x = 0$          | $[1 \ 0 \ 0]$  | $a_0 b_0$                         |
| $x = 1$          | $[1 \ 1 \ 1]$  | $(a_1 + a_0)(b_1 + b_0)$          |
| $x = -1$         | $[1 \ -1 \ 1]$ | $(-a_1 + a_0)(-b_1 + b_0)$        |
| $x = \infty$     | $[0 \ 0 \ 1]$  | $a_1 b_1$                         |
| $x = 2$          | $[1 \ 2 \ 4]$  | $(2a_1 + a_0)(2b_1 + b_0)$        |
| $\partial x = 0$ | $[0 \ 1 \ 0]$  | $a_0 b_1 + a_1 b_0$               |
| $\partial x = 1$ | $[0 \ 1 \ 2]$  | $(a_1 + a_0)b_1 + a_1(b_0 + b_1)$ |

- ▶ Which ones should we choose?
  - ▶ Matrix should be invertible
  - ▶ Matrix inverse should include "nice" values
  - ▶ Preferably only one multiplication per row

## FIR filter aspects

- ▶ Consider an even  $N$ -order (odd-length) linear-phase FIR filter
- ▶ Direct realization require  $\frac{N}{2} + 1 \approx \frac{N}{2}$  multiplications per sample
- ▶ Polyphase components  $H_0$  and  $H_1$  are symmetric,  $H_0 + H_1$  is not
- ▶ Leading to  $\frac{N/4 + N/4 + N/2}{2} = \frac{N}{2}$  multiplications per sample
- ▶ For odd  $N$ ,  $H_0 + H_1$  is symmetric while  $H_0$  and  $H_1$  are not
- ▶ Leading to  $\frac{N/2 + N/2 + N/4}{2} = \frac{5N}{8}$  multiplications per sample
- ▶ Increase compared to direct realization

## FIR filter aspects

- ▶ For FIR filters, the polynomial weights are the polyphase components
- ▶ Hence, if a symmetric FIR filter is used, the resulting polyphase components and sums of polyphase components may also be symmetric (or not)
- ▶ For FIR filters it is therefore of interest to find structures using as many symmetric sub filters as possible

## FIR filter aspects

- ▶ Instead select to use  $x = -1, 0, 1$  leading to

$$\begin{bmatrix} a_0 b_0 \\ (a_1 + a_0)(b_1 + b_0) \\ (a_0 - a_1)(b_0 - b_1) \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 1 \\ 1 & -1 & 1 \end{bmatrix} \begin{bmatrix} c_0 \\ c_1 \\ c_2 \end{bmatrix} \quad (19)$$

- ▶ Now, to determine  $c_0$  to  $c_2$  invert the matrix as

$$\begin{bmatrix} c_0 \\ c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \frac{1}{2} & -\frac{1}{2} \\ -1 & \frac{1}{2} & \frac{1}{2} \end{bmatrix} \begin{bmatrix} a_0 b_0 \\ (a_1 + a_0)(b_1 + b_0) \\ (a_0 - a_1)(b_0 - b_1) \end{bmatrix} \quad (20)$$

- ▶ This can be further rewritten as

$$\begin{bmatrix} c_0 \\ c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & -1 \\ -1 & 1 & 1 \end{bmatrix} \begin{bmatrix} a_0 & 0 & 0 \\ 0 & \frac{a_1 + a_0}{2} & 0 \\ 0 & 0 & \frac{a_0 - a_1}{2} \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} b_0 \\ b_1 \end{bmatrix} \quad (21)$$

- ▶ Complexity now  $\frac{N/2 + N/4 + N/4}{2} = \frac{N}{2}$  multiplications per sample

## Higher order polynomials

- ▶ For two term polynomials there are a limited number of cases resulting in three multiplications and only ones in the inverse
- ▶ However, for higher order polynomials (more terms) it becomes more challenging/interesting
- ▶ First, note that two polynomials with a power of two terms, say  $2^N$  require  $2^{N \log_2 3} = 3^N$  multiplications by applying the scheme recursively instead of  $4^N$  multiplications
- ▶ Second, note that multiplying two polynomials with  $A$  and  $B$  terms respectively generates a polynomial with  $A + B - 1$  terms
- ▶ Hence,  $A + B - 1$  equations and therefore multiplications are required (but there may be non-trivial constant multiplications involved)

## Three term polynomials

- ▶ This matrix is non-trivial to implement since it includes fractions of three and six
- ▶ One simple way is to just multiply the matrix by six leading to exact results with a scaling error of six (three as one can shift)
- ▶ However, to avoid this let us consider the derivative of the polynomial multiplication

$$(a_2x^2 + a_1x + a_0)(b_2x^2 + b_1x + b_0) = c_4x^4 + c_3x^3 + c_2x^2 + c_1x + c_0 \quad (24)$$

- ▶ The derivative of this is

$$\begin{aligned} & (2a_2x + a_1)(b_2x^2 + b_1x + b_0) + \\ & (a_2x^2 + a_1x + a_0)(2b_2x + b_1) = \\ & 4c_4x^3 + 3c_3x^2 + 2c_2x + c_1 \end{aligned} \quad (25)$$

- ▶ Evaluate at  $x = 0$  leads

$$a_1b_0 + a_0b_1 = c_1 \quad (26)$$

## Three term polynomials

- ▶ For three terms let us evaluate the polynomials at  $x = 0, 1, -1, -2, \infty$  leading to

$$\begin{bmatrix} a_0b_0 \\ (a_2 + a_1 + a_0)(b_2 + b_1 + b_0) \\ (a_2 - a_1 + a_0)(b_2 - b_1 + b_0) \\ (4a_2 - 2a_1 + a_0)(4b_2 - 2b_1 + b_0) \\ a_2b_2 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 & 1 \\ 1 & -2 & 4 & -8 & 16 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} c_0 \\ c_1 \\ c_2 \\ c_3 \\ c_4 \end{bmatrix} \quad (22)$$

- ▶ The inverse of the matrix is now

$$\begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 1/2 & 1/3 & -1 & 1/6 & -2 \\ -1 & 1/2 & 1/2 & 0 & -1 \\ -1/2 & 1/6 & 1/2 & -1/6 & 2 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix} \quad (23)$$

- ▶ Leading to

$$\begin{bmatrix} a_0b_0 \\ (a_2 + a_1 + a_0)(b_2 + b_1 + b_0) \\ (a_2 - a_1 + a_0)(b_2 - b_1 + b_0) \\ a_1b_0 + a_0b_1 \\ a_2b_2 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 & 1 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} c_0 \\ c_1 \\ c_2 \\ c_3 \\ c_4 \end{bmatrix} \quad (27)$$

- ▶ With the inverse

$$\begin{bmatrix} c_0 \\ c_1 \\ c_2 \\ c_3 \\ c_4 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ -1 & 1/2 & 1/2 & 0 & -1 \\ 0 & 1/2 & -1/2 & -1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} a_0b_0 \\ (a_2 + a_1 + a_0)(b_2 + b_1 + b_0) \\ (a_2 - a_1 + a_0)(b_2 - b_1 + b_0) \\ a_1b_0 + a_0b_1 \\ a_2b_2 \end{bmatrix} \quad (28)$$

### Three term polynomials

- ▶ However, as  $a_1b_0 + a_0b_1$  can not be realized using a single filter one can write

$$\begin{bmatrix} c_0 \\ c_1 \\ c_2 \\ c_3 \\ c_4 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 0 \\ -1 & \frac{1}{2} & \frac{1}{2} & 0 & 0 & -1 \\ 0 & \frac{1}{2} & -\frac{1}{2} & -1 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} a_0b_0 \\ (a_2 + a_1 + a_0)(b_2 + b_1 + b_0) \\ (a_2 - a_1 + a_0)(b_2 - b_1 + b_0) \\ a_1b_0 \\ a_0b_1 \\ a_2b_2 \end{bmatrix} \quad (29)$$

- ▶ Instead one can use more multiplications and end up with matrices only containing powers of two as e.g.

$$\begin{bmatrix} c_0 \\ c_1 \\ c_2 \\ c_3 \\ c_4 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 & 0 \\ -1 & \frac{1}{2} & \frac{1}{2} & 0 & 0 & -1 \\ 0 & \frac{1}{2} & -\frac{1}{2} & -1 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} a_0 + a_1 + a_2 \\ a_0 - a_1 + a_2 \\ \text{diag} \\ a_1 \\ a_2 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 1 \\ 1 & -1 & 1 \\ 0 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} b_0 \\ b_1 \\ b_2 \end{bmatrix} \quad (30)$$

### Winograd-based FIR filters

- ▶ The  $c$  term is just a constant for a given FIR filter

$$c = \sum_{k=0}^{\frac{M+1}{2}-1} h_{2k} h_{2k+1} \quad (32)$$

- ▶ The  $d_n$  term can be computed recursively requiring only a single multiplication per iteration

$$d_n = \sum_{k=0}^{\frac{M+1}{2}-1} X_{n-2k} X_{n-(2k+1)} \quad (33)$$

Note that two different results for even and odd  $n$  must be stored, also for one multiplication per iteration the product must be stored

### Winograd's inner product algorithm

- ▶ For the previously discussed algorithms at least two samples are needed to iterate once
- ▶ In 1968 Winograd proposed the following way of computing an inner product

$$y_n = \underbrace{\sum_{k=0}^{\frac{M+1}{2}-1} (X_{n-(2k+1)} + h_{2k})(X_{n-2k} + h_{2k+1})}_{b_n} - \underbrace{\sum_{k=0}^{\frac{M+1}{2}-1} h_{2k} h_{2k+1}}_c - \underbrace{\sum_{k=0}^{\frac{M+1}{2}-1} X_{n-2k} X_{n-(2k+1)}}_{d_n} \cdot \quad (31)$$

- ▶ Let us consider the different terms from an FIR filter perspective

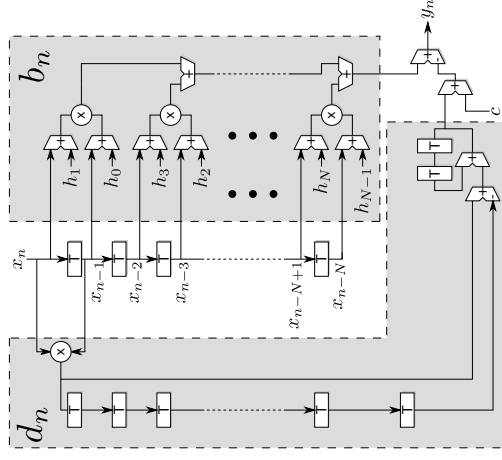
### Winograd-based FIR filters

- ▶ Finally, the  $b_n$  term require  $\frac{M+1}{2}$  multiplications per iteration

$$b_n = \sum_{k=0}^{\frac{M+1}{2}-1} (X_{n-(2k+1)} + h_{2k})(X_{n-2k} + h_{2k+1}) \quad (34)$$

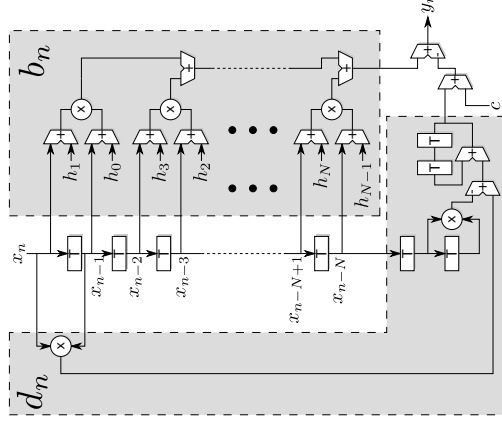
## Winograd-based FIR filters

- ▶ The resulting FIR filter using one multiplication per  $d_n$  iteration looks like



## Winograd-based FIR filters

- ▶ The resulting FIR filter using two multiplications per  $d_n$  iteration looks like



## Complexity per sample comparison

Odd-order FIR filters not considering possible symmetry

| Architecture                                | Mult                | Add/Sub            | Delays           |
|---|---------------------|--------------------|------------------|
| Direct form                                 | $N + 1$             | $N$                | $N$              |
| Poly-phase two-parallel                     | $N + 1$             | $N$                | $\frac{N}{2}$    |
| Polynomial two-parallel                     | $\frac{3(N+1)}{4}$  | $\frac{3N+5}{4}$   | $\frac{3N-1}{4}$ |
| Polynomial two-parallel (fewer reg.)        | $\frac{3(N+1)}{4}$  | $N + 1$            | $\frac{N}{2}$    |
| Winograd sequential (two mult per $d_n$ )   | $\frac{N+1}{2} + 2$ | $\frac{3(N+3)}{2}$ | $N + 4$          |
| Winograd sequential (one mult per $d_n$ )   | $\frac{N+1}{2} + 1$ | $\frac{3(N+3)}{2}$ | $2N + 2$         |
| Winograd two-parallel (two mult per $d_n$ ) | $\frac{N+1}{2} + 2$ | $\frac{3(N+3)}{2}$ | $\frac{N+7}{2}$  |
| Winograd two-parallel (one mult per $d_n$ ) | $\frac{N+1}{2} + 1$ | $\frac{3(N+3)}{2}$ | $N + 1$          |

## Squaring-based multiplication

$$(a + b)^2 = a^2 + b^2 + 2ab \Rightarrow ab = \frac{(a + b)^2 - a^2 - b^2}{2} \quad (35)$$

- ▶ Can be efficient when implementing multiplication using lookup tables
- ▶ Can also be used for FIR filters, complex multipliers, matrix multiplication etc