

Improving the Goertzel–Blahut algorithm: An example

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In this preprint, we present an example illustrating the novel method for the discrete Fourier transform (DFT) computation based on the Goertzel–Blahut algorithm introduced in the paper “Improving the Goertzel–Blahut algorithm” (IEEE Signal Processing Letters, vol. 23, no. 6, pp. 824–827, 2016).

EXAMPLE

Consider the DFT of length $n = 15$ over the field $GF(2^4)$. The finite field $GF(2^4)$ is defined by an element α , which is a root of the primitive polynomial $x^4 + x + 1$. Let us take the primitive element α as a transform kernel. The binary conjugacy classes of $GF(2^m)$ are (α^0) , $(\alpha^1, \alpha^2, \alpha^4, \alpha^8)$, $(\alpha^3, \alpha^6, \alpha^{12}, \alpha^9)$, $(\alpha^7, \alpha^{14}, \alpha^{13}, \alpha^{11})$, (α^5, α^{10}) .

A. Goertzel–Blahut algorithm

The first step of the Goertzel–Blahut algorithm is

$$\begin{aligned}
 f(x) &= (x+1)q_0(x) + r_0(x), & r_0(x) &= r_{0,0} \\
 f(x) &= (x^4+x+1)q_1(x) + r_1(x), & r_1(x) &= r_{3,1}x^3 + r_{2,1}x^2 + r_{1,1}x + r_{0,1} \\
 f(x) &= (x^4+x^3+x^2+x+1)q_2(x) + r_2(x), & r_2(x) &= r_{3,2}x^3 + r_{2,2}x^2 + r_{1,2}x + r_{0,2} \\
 f(x) &= (x^4+x^3+1)q_3(x) + r_3(x), & r_3(x) &= r_{3,3}x^3 + r_{2,3}x^2 + r_{1,3}x + r_{0,3} \\
 f(x) &= (x^2+x+1)q_4(x) + r_4(x), & r_4(x) &= r_{1,4}x + r_{0,4}
 \end{aligned}$$

where

$$\begin{aligned}
 r_{0,0} &= \sum_{i=0}^{14} f_i \\
 r_{0,1} &= f_0 + f_4 + f_7 + f_8 + f_{10} + f_{12} + f_{13} + f_{14} \\
 r_{1,1} &= f_1 + f_4 + f_5 + f_7 + f_9 + f_{10} + f_{11} + f_{12} \\
 r_{2,1} &= f_2 + f_5 + f_6 + f_8 + f_{10} + f_{11} + f_{12} + f_{13} \\
 r_{3,1} &= f_3 + f_6 + f_7 + f_9 + f_{11} + f_{12} + f_{13} + f_{14} \\
 r_{0,2} &= f_0 + f_4 + f_5 + f_9 + f_{10} + f_{14} \\
 r_{1,2} &= f_1 + f_4 + f_6 + f_9 + f_{11} + f_{14} \\
 r_{2,2} &= f_2 + f_4 + f_7 + f_9 + f_{12} + f_{14} \\
 r_{3,2} &= f_3 + f_4 + f_8 + f_9 + f_{13} + f_{14} \\
 r_{0,3} &= f_0 + f_4 + f_5 + f_6 + f_7 + f_9 + f_{11} + f_{12} \\
 r_{1,3} &= f_1 + f_5 + f_6 + f_7 + f_8 + f_{10} + f_{12} + f_{13} \\
 r_{2,3} &= f_2 + f_6 + f_7 + f_8 + f_9 + f_{11} + f_{13} + f_{14} \\
 r_{3,3} &= f_3 + f_4 + f_5 + f_6 + f_8 + f_{10} + f_{11} + f_{14} \\
 r_{0,4} &= f_0 + f_2 + f_3 + f_5 + f_6 + f_8 + f_9 + f_{11} + f_{12} + f_{14} \\
 r_{1,4} &= f_1 + f_2 + f_4 + f_5 + f_7 + f_8 + f_{10} + f_{11} + f_{13} + f_{14}.
 \end{aligned}$$

The second step of the Goertzel–Blahut algorithm is

$$\begin{array}{rclcl}
F_0 & = & f(\alpha^0) & = & r_{0,0} \\
\hline
F_1 & = & f(\alpha^1) & = & r_1(\alpha^1) = r_{3,1}\alpha^3 + r_{2,1}\alpha^2 + r_{1,1}\alpha^1 + r_{0,1} \\
F_2 & = & f(\alpha^2) & = & r_1(\alpha^2) = r_{3,1}\alpha^6 + r_{2,1}\alpha^4 + r_{1,1}\alpha^2 + r_{0,1} \\
F_4 & = & f(\alpha^4) & = & r_1(\alpha^4) = r_{3,1}\alpha^{12} + r_{2,1}\alpha^8 + r_{1,1}\alpha^4 + r_{0,1} \\
F_8 & = & f(\alpha^8) & = & r_1(\alpha^8) = r_{3,1}\alpha^9 + r_{2,1}\alpha^1 + r_{1,1}\alpha^8 + r_{0,1} \\
\hline
F_3 & = & f(\alpha^3) & = & r_2(\alpha^3) = r_{3,2}\alpha^9 + r_{2,2}\alpha^6 + r_{1,2}\alpha^3 + r_{0,2} \\
F_6 & = & f(\alpha^6) & = & r_2(\alpha^6) = r_{3,2}\alpha^3 + r_{2,2}\alpha^{12} + r_{1,2}\alpha^6 + r_{0,2} \\
F_{12} & = & f(\alpha^{12}) & = & r_2(\alpha^{12}) = r_{3,2}\alpha^6 + r_{2,2}\alpha^9 + r_{1,2}\alpha^{12} + r_{0,2} \\
F_9 & = & f(\alpha^9) & = & r_2(\alpha^9) = r_{3,2}\alpha^{12} + r_{2,2}\alpha^3 + r_{1,2}\alpha^9 + r_{0,2} \\
\hline
F_7 & = & f(\alpha^7) & = & r_3(\alpha^7) = r_{3,3}\alpha^6 + r_{2,3}\alpha^{14} + r_{1,3}\alpha^7 + r_{0,3} \\
F_{14} & = & f(\alpha^{14}) & = & r_3(\alpha^{14}) = r_{3,3}\alpha^{12} + r_{2,3}\alpha^{13} + r_{1,3}\alpha^{14} + r_{0,3} \\
F_{13} & = & f(\alpha^{13}) & = & r_3(\alpha^{13}) = r_{3,3}\alpha^9 + r_{2,3}\alpha^{11} + r_{1,3}\alpha^{13} + r_{0,3} \\
F_{11} & = & f(\alpha^{11}) & = & r_3(\alpha^{11}) = r_{3,3}\alpha^3 + r_{2,3}\alpha^7 + r_{1,3}\alpha^{11} + r_{0,3} \\
\hline
F_5 & = & f(\alpha^5) & = & r_4(\alpha^5) = r_{1,4}\alpha^5 + r_{0,4} \\
F_{10} & = & f(\alpha^{10}) & = & r_4(\alpha^{10}) = r_{1,4}\alpha^{10} + r_{0,4}
\end{array}$$

or

$$F_0 = (1) (r_{0,0}) = V_0 (r_{0,0}),$$

$$\begin{pmatrix} F_1 \\ F_2 \\ F_4 \\ F_8 \end{pmatrix} = \begin{pmatrix} 1 & \alpha^1 & \alpha^2 & \alpha^3 \\ 1 & \alpha^2 & \alpha^4 & \alpha^6 \\ 1 & \alpha^4 & \alpha^8 & \alpha^{12} \\ 1 & \alpha^8 & \alpha^1 & \alpha^9 \end{pmatrix} \begin{pmatrix} r_{0,1} \\ r_{1,1} \\ r_{2,1} \\ r_{3,1} \end{pmatrix} = V_1 \begin{pmatrix} r_{0,1} \\ r_{1,1} \\ r_{2,1} \\ r_{3,1} \end{pmatrix},$$

$$\begin{pmatrix} F_3 \\ F_6 \\ F_{12} \\ F_9 \end{pmatrix} = \begin{pmatrix} 1 & \alpha^3 & \alpha^6 & \alpha^9 \\ 1 & \alpha^6 & \alpha^{12} & \alpha^3 \\ 1 & \alpha^{12} & \alpha^9 & \alpha^6 \\ 1 & \alpha^9 & \alpha^3 & \alpha^{12} \end{pmatrix} \begin{pmatrix} r_{0,2} \\ r_{1,2} \\ r_{2,2} \\ r_{3,2} \end{pmatrix} = V_2 \begin{pmatrix} r_{0,2} \\ r_{1,2} \\ r_{2,2} \\ r_{3,2} \end{pmatrix},$$

$$\begin{pmatrix} F_7 \\ F_{14} \\ F_{13} \\ F_{11} \end{pmatrix} = \begin{pmatrix} 1 & \alpha^7 & \alpha^{14} & \alpha^6 \\ 1 & \alpha^{14} & \alpha^{13} & \alpha^{12} \\ 1 & \alpha^{13} & \alpha^{11} & \alpha^9 \\ 1 & \alpha^{11} & \alpha^7 & \alpha^3 \end{pmatrix} \begin{pmatrix} r_{0,3} \\ r_{1,3} \\ r_{2,3} \\ r_{3,3} \end{pmatrix} = V_3 \begin{pmatrix} r_{0,3} \\ r_{1,3} \\ r_{2,3} \\ r_{3,3} \end{pmatrix},$$

$$\begin{pmatrix} F_5 \\ F_{10} \end{pmatrix} = \begin{pmatrix} 1 & \alpha^5 \\ 1 & \alpha^{10} \end{pmatrix} \begin{pmatrix} r_{0,4} \\ r_{1,4} \end{pmatrix} = V_4 \begin{pmatrix} r_{0,4} \\ r_{1,4} \end{pmatrix}.$$

B. Novel method based on the Goertzel–Blahut algorithm

The Moore–Vandermonde matrix V_1 factorization [13] is

$$\begin{aligned}
V_1 &= \begin{pmatrix} 1 & \alpha^1 & \alpha^2 & \alpha^3 \\ 1 & \alpha^2 & \alpha^4 & \alpha^6 \\ 1 & \alpha^4 & \alpha^8 & \alpha^{12} \\ 1 & \alpha^8 & \alpha^1 & \alpha^9 \end{pmatrix} = \left(\begin{array}{cc|cc} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ \hline 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \end{array} \right) \left(\begin{array}{cc|cc} 1 & 0 & \alpha & 0 \\ 0 & 1 & 0 & \alpha^2 \\ \hline 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{array} \right) \\
&\times \left(\begin{array}{cc|cc} 1 & \alpha^5 & 0 & 0 \\ 1 & \alpha^{10} & 0 & 0 \\ \hline 0 & 0 & 1 & \alpha^5 \\ 0 & 0 & 1 & \alpha^{10} \end{array} \right) \left(\begin{array}{cccc} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{array} \right) \left(\begin{array}{cccc} 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{array} \right) \\
&= \begin{pmatrix} \alpha^0 & \alpha^5 & \alpha^1 & \alpha^6 \\ \alpha^0 & \alpha^{10} & \alpha^2 & \alpha^{12} \\ \alpha^0 & \alpha^5 & \alpha^4 & \alpha^9 \\ \alpha^0 & \alpha^{10} & \alpha^8 & \alpha^3 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{pmatrix} = S_1 P_1,
\end{aligned}$$

where P_1 is the matrix of preadditions.

From Lemma 1 [SPL16] it follows that

$$\begin{aligned}
V_0 &= 1 \\
V_1 &= S_1 P_1 \\
V_2 &= V_1 N_{2,1} = S_1 P_1 N_{2,1} \\
V_3 &= V_1 N_{3,1} = S_1 P_1 N_{3,1}
\end{aligned}$$

where $N_{2,1}$ and $N_{3,1}$ are the basis transformation matrices:

$$N_{2,1} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 1 & 1 \end{pmatrix}, \quad N_{3,1} = \begin{pmatrix} 1 & 1 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 1 & 1 & 1 \end{pmatrix}.$$

The Moore–Vandermonde matrix V_4 factorization is very simple:

$$V_4 = \begin{pmatrix} 1 & \alpha^5 \\ 1 & \alpha^{10} \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & \alpha^5 \\ 0 & 1 \end{pmatrix}.$$

Further, we obtain

$$\begin{aligned}
V &= \begin{pmatrix} V_0 & & & & \\ & V_1 & & & \\ & & V_2 & & \\ & & & V_3 & \\ & & & & V_4 \end{pmatrix} \\
&= \begin{pmatrix} 1 & & & & \\ & S_1 & & & \\ & & S_1 & & \\ & & & S_1 & \\ & & & & V_4 \end{pmatrix} \begin{pmatrix} 1 & & & & \\ & P_1 & & & \\ & & P_1 & & \\ & & & P_1 & \\ & & & & I_2 \end{pmatrix} \begin{pmatrix} 1 & & & & \\ & I_4 & & & \\ & & N_{2,1} & & \\ & & & N_{3,1} & \\ & & & & I_2 \end{pmatrix} = SPN,
\end{aligned}$$

where I_2 is the 2×2 identity matrix and I_4 is the 4×4 identity matrix.

TABLE II
COMPLEXITY OF THE 15-POINT DFT COMPUTATION

method	multiplications	additions
Cyclotomic algorithm [18]	16	77
Cyclotomic algorithm with common subexpression elimination algorithm [7]	16	74
Novel method [SPL16]	13	70

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