# Cyclotomic Basis for Computing the Discrete Fourier Transform 

G. Jerônimo da Silva Jr. and R. M. Campello de Souza<br>Department of Electronics and Systems<br>UFPE, CP7800, 50711-970<br>Recife PE, Brazil<br>e-mails: gilsonjr@gmail.com, ricardo@ufpe.br


#### Abstract

This paper presents a new fast algorithm for computing an $N$-point discrete Fourier transform. The algorithm meets the Heideman multiplicative complexity lower bound for $N=\{3,4,6,8,12\}$ and is based upon the decomposition of the elements of the transform matrix into a cyclotomic basis.


Keywords- FFT; DFT; multiplicative complexity; cyclotomic basis.

## I. Introduction

Transforms are mathematical tools used in many applications of Engineering. A particularly important example is the continuous Fourier transform and its discrete version in the time and frequency domains, the discrete Fourier transform (DFT) [1], [2].

In practical application scenarios one is always looking for efficient ways, in terms of arithmetic complexity, to compute a DFT. This is a problem that fascinated many engineers and mathematicians for centuries [3] and its study has led to the development of well known fast Fourier transform (FFT) algorithms, such as the algorithms of Coley-Tukey, Good-Thomas and Winograd [4]-[6]. In [7] Heideman presented a minimum multiplicative complexity for computing an N -point DFT. Very few algorithms meet this minimum for some blocklengths and there is not a systematic way to derive them. This is the main motivation for the work reported in this paper, namely, to reach the minimum multiplicative complexity with a systematic and general algorithm.

The discrete Fourier transform of the sequence $v=\left(v_{n}\right)$, $n=0, \ldots, N-1$, is the sequence $V=\left(V_{k}\right), k=0, \ldots, N-1$, defined by

$$
\begin{equation*}
V_{k} \Delta \sum_{n=0}^{N-1} v_{n} W_{N}^{k n} \tag{1}
\end{equation*}
$$

where $W_{N} \triangleq e^{-j \frac{2 \pi}{N}}$ is an element of order $N$ in $\mathbb{C}$ and $j \triangleq$ $\sqrt{-1}$. Expression (1) may be written in matrix form as

$$
\begin{equation*}
\mathbf{V}=\mathbf{W} \mathbf{v} \tag{2}
\end{equation*}
$$

by defining

$$
\mathbf{V} \triangleq\left[\begin{array}{c}
V_{0}  \tag{3}\\
V_{1} \\
\vdots \\
V_{N-1}
\end{array}\right]
$$

$$
\mathbf{v} \xlongequal{\Delta}\left[\begin{array}{c}
v_{0}  \tag{4}\\
v_{1} \\
\vdots \\
v_{N-1}
\end{array}\right]
$$

and

$$
\mathbf{W}=\left[\begin{array}{cccc}
1 & 1 & \cdots & 1  \tag{5}\\
1 & W_{N} & \cdots & W_{N}^{N-1} \\
\vdots & \vdots & \ddots & \vdots \\
1 & W_{N}^{N-1} & \cdots & W_{N}^{(N-1)(N-1)}
\end{array}\right]
$$

Throughout the paper it is assumed that $\mathbf{v}$ is a real-valued sequence.

## II. Cyclotomic Basis

The cyclotomic polynomial of order $N$ over $\mathbb{C}$, denoted by $\Phi_{N}(x)$, is defined as the monic polynomial the roots of which are all elements of order $N$ in the complex field. It can be written as

$$
\begin{equation*}
\Phi_{N}(x)=\prod_{\operatorname{ord}(\theta)=N}(x-\theta) . \tag{6}
\end{equation*}
$$

It can be shown, using the Möbius inversion formula [8], that

$$
\begin{equation*}
\Phi_{N}(x)=\prod_{d \mid N}\left(x^{d}-1\right)^{\mu(N / d)} \tag{7}
\end{equation*}
$$

where $\mu(n)$ is the Möbius function [9]. The degree of $\Phi_{N}(x)$ is given by Euler's totient function $\phi(N)$, defined as the number of positive integers less than $N$ and relatively prime to $N$. Cyclotomic polynomials have integer coefficients which, for $N$ smaller than 105 , are equal to 0,1 or -1 [4].

Definition 1: A cyclotomic basis $\left(C B_{N}\right)$ is a set in the complex field such that any element of a cyclic multiplicative group of order $N\left(C M G_{N}\right)$ can be represented by a linear combination, with rational coefficients, of the elements of $C B_{N}$.

We can see that $\alpha=W_{N}$ is a root of $\Phi_{N}(x)$. Therefore, from equation $\Phi(\alpha)=0$, we can write all elements of $C M G_{N}=\left\{\alpha^{0}, \ldots, \alpha^{N-1}\right\}$ as linear combination, with integer coefficients, of the elements $\left\{1, \alpha, \alpha^{2}, \ldots, \alpha^{\phi(N)-1}\right\}$. This motivates the following definition.

Definition 2: A canonical cyclotomic basis (CCB) is the cyclotomic basis formed by $C C B_{N}=\left\{1, \alpha, \alpha^{2}, \ldots, \alpha^{\phi(N)-1}\right\}$ relative to $C M G_{N}$, where $\alpha$ is a generator of $C M G_{N}$.

Example 1: For $N=6, \alpha=W_{6}$ is a root of $\Phi_{6}(x)=$ $x^{2}-x+1$, or $\Phi(\alpha)=\alpha^{2}-\alpha+1=0$. The set $C C B_{6}=\{1, \alpha\}$ is a canonical cyclotomic basis of the multiplicative group $C M G_{6}=\left\{1, \alpha, \alpha^{2}, \alpha^{3}, \alpha^{4}, \alpha^{5}\right\}$. The representation of the elements of $C M G_{6}$ as a linear combination of $\{1, \alpha\}$ is

$$
\begin{array}{cc}
\alpha^{0}=1 & \alpha^{3}=-1 \\
\alpha^{1}=\alpha & \alpha^{4}=-\alpha \\
\alpha^{2}=-1+\alpha & \alpha^{5}=1-\alpha
\end{array}
$$

or simply

$$
\left[\begin{array}{l}
\alpha^{0} \\
\alpha^{1} \\
\alpha^{2} \\
\alpha^{3} \\
\alpha^{4} \\
\alpha^{5}
\end{array}\right]=\left[\begin{array}{cc}
1 & 0 \\
0 & 1 \\
-1 & 1 \\
-1 & 0 \\
0 & -1 \\
1 & -1
\end{array}\right]\left[\begin{array}{l}
\alpha^{0} \\
\alpha^{1}
\end{array}\right]
$$

$C B_{6}=\left\{1, \alpha^{5}\right\}$ is another canonical basis for $C M G_{6}$.
In general we can write, for all $C M G_{N}$,

$$
\left[\begin{array}{c}
\alpha^{0}  \tag{8}\\
\vdots \\
\alpha^{N-1}
\end{array}\right]=R_{c}\left[\begin{array}{c}
\alpha^{0} \\
\vdots \\
\alpha^{\phi(N)-1}
\end{array}\right]
$$

where $\alpha=W_{N}$ and $R_{c}(N \times \Phi(N))$ is said to be the rectangular canonical matrix, the elements of which are integers. This is always possible if we write $\alpha^{n}=\alpha^{n}\left(\bmod \Phi_{N}(\alpha)\right)$, for all $n=0,1, \ldots, N-1$.

We can make a basis change to represent a $C M G_{N}$ under another $C B_{N}$. To do this, simply express the elements of the new basis as

$$
\left[\begin{array}{c}
\alpha^{i_{1}}  \tag{9}\\
\vdots \\
\alpha^{i_{\phi(N)}}
\end{array}\right]=Q\left[\begin{array}{c}
\alpha^{0} \\
\vdots \\
\alpha^{\phi(N)-1}
\end{array}\right]
$$

where the $n$th row of $Q$ is the $i_{n}$ th row of $R_{c}$ for all $n=$ $1, \ldots, \phi(N) . Q$ is a square matrix and if it is nonsingular, then

$$
\left[\begin{array}{c}
\alpha^{0}  \tag{10}\\
\vdots \\
\alpha^{\phi(N)-1}
\end{array}\right]=Q^{-1}\left[\begin{array}{c}
\alpha^{i_{1}} \\
\vdots \\
\alpha^{i_{\phi(N)}}
\end{array}\right]
$$

and a $C M G_{N}$ can be expressed by the new basis

$$
\left[\begin{array}{c}
\alpha^{0}  \tag{11}\\
\vdots \\
\alpha^{N-1}
\end{array}\right]=R_{c} Q^{-1}\left[\begin{array}{c}
\alpha^{i_{1}} \\
\vdots \\
\alpha^{i_{\phi(N)}}
\end{array}\right]
$$

where $R=R_{c} Q^{-1}$ is a general rectangular decomposition matrix. Therefore, we can decompose a $C M G_{N}$ into any cyclotomic basis, $C B_{N}=\left\{\alpha^{i_{1}}, \ldots, \alpha^{i_{\phi(N)}}\right\}$, provided $Q$ has an inverse. The following example illustrates this point.

Example 2: For $N=8$ we have, from (7), $\Phi_{8}(x)=1+$ $x^{4}$. Computing the rectangular canonical matrix $R_{c}(8 \mathrm{x} 4)$ we obtain the following decomposition using $C C B_{8}=$ $\left\{1, \alpha, \alpha^{2}, \alpha^{3}\right\}$,

$$
\left[\begin{array}{l}
\alpha^{0} \\
\alpha^{1} \\
\alpha^{2} \\
\alpha^{3} \\
\alpha^{4} \\
\alpha^{5} \\
\alpha^{6} \\
\alpha^{7}
\end{array}\right]=\left[\begin{array}{cccc}
1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 1 & 0 \\
0 & 0 & 0 & 1 \\
-1 & 0 & 0 & 0 \\
0 & -1 & 0 & 0 \\
0 & 0 & -1 & 0 \\
0 & 0 & 0 & -1
\end{array}\right]\left[\begin{array}{c}
1 \\
\alpha \\
\alpha^{2} \\
\alpha^{3}
\end{array}\right]
$$

We can make a basis change to represent $C M G_{8}$ upon a new cyclotomic basis $C B_{8}=\left\{1, \alpha^{6}, \alpha, \alpha^{7}\right\}$. Using the $R_{c}$ matrix, we can write

$$
\left[\begin{array}{c}
1 \\
j \\
\alpha \\
\alpha^{*}
\end{array}\right]=\left[\begin{array}{cccc}
1 & 0 & 0 & 0 \\
0 & 0 & -1 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 0 & -1
\end{array}\right]\left[\begin{array}{c}
1 \\
\alpha \\
\alpha^{2} \\
\alpha^{3}
\end{array}\right],
$$

where $\alpha^{*}$ denotes the complex conjugate of $\alpha$. Since the matrix $Q$ is invertible, the decomposition matrix is $R=R_{c} Q^{-1}$, leading to

$$
\left[\begin{array}{c}
\alpha^{0} \\
\alpha^{1} \\
\alpha^{2} \\
\alpha^{3} \\
\alpha^{4} \\
\alpha^{5} \\
\alpha^{6} \\
\alpha^{7}
\end{array}\right]=\left[\begin{array}{cccc}
1 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 \\
0 & -1 & 0 & 0 \\
0 & 0 & 0 & -1 \\
-1 & 0 & 0 & 0 \\
0 & 0 & -1 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 0 & 1
\end{array}\right]\left[\begin{array}{c}
1 \\
j \\
\alpha \\
\alpha^{*}
\end{array}\right]
$$

It is also possible to choose an appropriate $C B_{N}$ so as to separate the real and imaginary parts, as shown in the next example.

Example 3: For $N=6$, we can take as a cyclotomic basis the set $\left\{1 / 2,\left(\alpha-\alpha^{5}\right) / 2\right\}$ and, using the matrix $R_{c}$ from Example 1, it is possible to write

$$
\left[\begin{array}{c}
1 / 2 \\
\left(\alpha^{1}-\alpha^{5}\right) / 2
\end{array}\right]=\frac{1}{2}\left[\begin{array}{cc}
1 & 0 \\
-1 & 2
\end{array}\right]\left[\begin{array}{l}
\alpha^{0} \\
\alpha^{1}
\end{array}\right]
$$

Computing $R_{c} Q^{-1}$ yields

$$
\left[\begin{array}{c}
1 \\
\alpha^{1} \\
\alpha^{2} \\
\alpha^{3} \\
\alpha^{4} \\
\alpha^{5}
\end{array}\right]=\left[\begin{array}{cc}
2 & 0 \\
1 & 1 \\
-1 & 1 \\
-2 & 0 \\
-1 & -1 \\
1 & -1
\end{array}\right]\left[\begin{array}{c}
1 / 2 \\
\left(\alpha-\alpha^{5}\right) / 2
\end{array}\right]
$$

and it is clear that $\left(\alpha-\alpha^{5}\right) / 2=j \operatorname{Im}(\alpha)$, where $\operatorname{Im}($. denotes the imaginary part of the argument.

This example inspires two new definitions.
Definition 3: A sine and cosine cyclotomic basis ( $\sin / \mathrm{cos}-$ $\mathrm{CB})$ is the basis, with $\phi(N)$ elements, $C B_{N}=\{1,(\alpha-$ $\left.\left.\alpha^{N-1}\right) / 2,\left(\alpha+\alpha^{N-1}\right) / 2,\left(\alpha^{2}-\alpha^{N-2}\right) / 2, \ldots\right\}$, or simply $C B_{N}=\left\{1, j \operatorname{Im}(\alpha), \operatorname{Re}(\alpha), j \operatorname{Im}\left(\alpha^{2}\right), \ldots\right\}$.

Definition 4: A cosine cyclotomic basis (cos-CB) is the basis, with $\phi(N)$ elements, $N \equiv 0(\bmod 4), C B_{N}=$ $\left\{1, \alpha^{N / 4},\left(\alpha+\alpha^{N-1}\right) / 2,\left(\alpha^{1+N / 4}+\alpha^{N / 4-1}\right) / 2, \ldots\right\}$, or simply $C B_{N}=\left\{1,-j, \operatorname{Re}(\alpha), j \operatorname{Re}(\alpha), \operatorname{Re}\left(\alpha^{2}\right), j \operatorname{Re}\left(\alpha^{2}\right), \ldots\right\}$.

The advantage of choosing a $\sin / \cos -\mathrm{CB}$ or a cos-CB is that the basis are purely real or imaginary. In this case we have

$$
\left[\begin{array}{c}
\alpha^{0}  \tag{12}\\
\vdots \\
\alpha^{N-1}
\end{array}\right]=R\left[\begin{array}{c}
\gamma_{1} \\
\vdots \\
\gamma_{\phi(N)}
\end{array}\right]
$$

where $R$ is a general rectangular decomposition matrix with integer elements (depending on the basis choice), and $\gamma_{1}, \gamma_{2}, \ldots, \gamma_{\phi(N)}$ is the cyclotomic basis formed from purely real or imaginary constants. If $r_{i j}, i=0,1, \ldots, N-1$, $j=1, \ldots, \phi(N)$, denotes the elements of $R$, then

$$
\begin{equation*}
\alpha^{i}=\sum_{j=1}^{\phi(N)} r_{i j} \gamma_{j} \tag{13}
\end{equation*}
$$

Equation (13) characterizes decompositions over cyclotomic basis. If $N \equiv 0(\bmod 4)$ a cos-CB can be used and $\{1,-j\} \in$ $C B_{N}$. Otherwise, a sine/cos-CB can be used and only $\{1\} \in$ $C B_{N}$.

## III. A New Fast Fourier Transform Algorithm

The algorithm is constructed by decomposing the DFT matrix $W$ over a cyclotomic basis. Equation (2) can be rewritten as

$$
\mathbf{V}=\underbrace{\left[\begin{array}{cccc}
1 & 1 & \cdots & 1  \tag{14}\\
1 & \alpha & \cdots & \alpha^{N-1} \\
\vdots & \vdots & \ddots & \vdots \\
1 & \alpha^{N-1} & \cdots & \alpha^{(N-1)(N-1)}
\end{array}\right]}_{\text {decompose in } C B_{N}} \mathbf{v}
$$

where $\alpha=W_{N}$. Therefore, it becomes

$$
\begin{equation*}
\mathbf{V}=\left(\gamma_{1} A_{1}+\gamma_{2} A_{2}+\ldots+\gamma_{\phi(N)} A_{\phi(N)}\right) \mathbf{v} \tag{15}
\end{equation*}
$$

or

$$
\begin{equation*}
\mathbf{V}=\sum_{i=1}^{\phi(N)} \gamma_{i} A_{i} \mathbf{v} \tag{16}
\end{equation*}
$$

where the elements of the matrices $A_{i}$ are rational numbers. In most cases, these are small integers.

Apparently, there is not a significant improvement in the multiplicative complexity for computing $V$, since it is related to the products $\gamma_{i} A_{i} \mathbf{v}$. However, these products can be made in a very effective way.

Theorem 1: Let $\gamma$ be a purely real or imaginary constant, $A$ an integer matrix and $\mathbf{v}$ a vector of variables. The computation of $\mathbf{s}=\gamma A \mathbf{v}$ requires $\operatorname{rank}(A)$ real multiplications.

Proof: With $l=\operatorname{rank}(A)$, there are $l$ linearly independent rows of A , namely $r_{1}, r_{2}, \ldots, r_{l}$. There are $l$ real multiplications

$$
\begin{equation*}
s_{j_{i}}=\gamma\left(r_{i} \mathbf{v}\right), \tag{17}
\end{equation*}
$$

for each $i=1,2, \ldots, l$, and all other rows can be expressed by

$$
\begin{equation*}
r_{m}=\sum_{i=1}^{l} b_{m i} r_{i} \tag{18}
\end{equation*}
$$

where $b_{m i} \in \mathbb{Q}$. Therefore, any component $s_{m}$ can be computed by

$$
\begin{gather*}
s_{m}=\gamma\left(r_{m} \mathbf{v}\right)  \tag{19}\\
s_{m}=\gamma \sum_{i=1}^{l} b_{m i} r_{i} \mathbf{v}  \tag{20}\\
s_{m}=\sum_{i=1}^{l} b_{m i}\left(\gamma r_{i} \mathbf{v}\right),  \tag{21}\\
s_{m}=\sum_{i=1}^{l} b_{m i} s_{j_{i}} \tag{22}
\end{gather*}
$$

where multiplications by $b_{m i}$ are considered trivial, and all $s_{m}$ can be computed by trivial combinations of $l$ multiplications.

Using Theorem 1 and (16), the multiplicative complexity, $M_{r}$, is given by

$$
\begin{equation*}
M_{r}=\sum_{i=2}^{\phi(N)} \operatorname{rank}\left(A_{i}\right) \tag{23}
\end{equation*}
$$

since that $\gamma_{1}=1$ (trivial multiplication). If cos-CB is used, $\gamma_{2}=j$, and

$$
M_{r}=\sum_{i=1}^{\phi(N) / 2-1} \operatorname{rank}\left(\left[\begin{array}{c}
A_{2 i+1}  \tag{24}\\
A_{2(i+1)}
\end{array}\right]\right),
$$

since that $\gamma_{2(i+1)}=-j \gamma_{2 i+1}$ in this basis.
Example 4: We start with the simplest nontrivial example, the 3-point DFT. To find the decomposition based on
the $\sin / \cos -\mathrm{CB}$, we use the procedure describe in section II, which results in the expression

$$
\left[\begin{array}{l}
\alpha^{0} \\
\alpha^{1} \\
\alpha^{2}
\end{array}\right]=\left[\begin{array}{cc}
1 & 0 \\
-0.5 & 1 \\
-0.5 & -1
\end{array}\right]\left[\begin{array}{c}
1 \\
-j \sin (2 \pi / 3)
\end{array}\right]
$$

Applying it to the $3 \times 3$ DFT matrix yields

$$
\begin{gathered}
\mathbf{W}=\left[\begin{array}{ccc}
1 & 1 & 1 \\
1 & -0.5 & -0.5 \\
1 & -0.5 & -0.5
\end{array}\right] \\
-j \sin (2 \pi / 3)\left[\begin{array}{ccc}
0 & 0 & 0 \\
0 & 1 & -1 \\
0 & -1 & 1
\end{array}\right] .
\end{gathered}
$$

Considering the array $\mathbf{L}$ of all linearly independent rows of $A_{1}$ and $A_{2}$, we have

$$
\mathbf{S}=\mathbf{L} \mathbf{v}=\left[\begin{array}{ccc}
1 & 1 & 1 \\
1 & -0.5 & -0.5 \\
0 & 1 & -1
\end{array}\right]\left[\begin{array}{l}
v_{0} \\
v_{1} \\
v_{2}
\end{array}\right]=\left[\begin{array}{c}
S_{0} \\
S_{1} \\
S_{2}
\end{array}\right]
$$

Since that $\mathbf{V}=\mathbf{W v}$, defining

$$
m_{1}=-j \sin (2 \pi / 3) S_{2}
$$

we may write

$$
\mathbf{V}=\left[\begin{array}{c}
S_{0} \\
S_{1}+m_{1} \\
S_{1}-m_{1}
\end{array}\right]
$$

Therefore, the matrix $\mathbf{W}$ can be expressed as
$\mathbf{W}=\left[\begin{array}{ccc}1 & 0 & 0 \\ 0 & 1 & 1 \\ 0 & 1 & -1\end{array}\right]\left[\begin{array}{ccc}1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & \gamma_{2}\end{array}\right]\left[\begin{array}{ccc}1 & 1 & 1 \\ 1 & -0.5 & -0.5 \\ 0 & 1 & -1\end{array}\right]$.
This algorithm implements the 3-point DFT with the minimum number of multiplications, a better performance than the Winograd algorithm, which implements it with two multiplications [6].

Example 5: We derive the 8-point FFT algorithm using the same procedure as in the previous example. In this case the cos-CB is used,

$$
\left[\begin{array}{l}
\alpha^{0} \\
\alpha^{1} \\
\alpha^{2} \\
\alpha^{3} \\
\alpha^{4} \\
\alpha^{5} \\
\alpha^{6} \\
\alpha^{7}
\end{array}\right]=\left[\begin{array}{cccc}
1 & 0 & 0 & 0 \\
0 & 0 & 1 & 1 \\
0 & 1 & 0 & 0 \\
0 & 0 & -1 & 1 \\
-1 & 0 & 0 & 0 \\
0 & 0 & -1 & -1 \\
0 & -1 & 0 & 0 \\
0 & 0 & 1 & -1
\end{array}\right]\left[\begin{array}{c}
1 \\
-j \\
\frac{\sqrt{2}}{2} \\
-j \frac{\sqrt{2}}{2}
\end{array}\right]
$$

Using it to decompose the 8 -point DFT matrix, leads to

$$
\begin{aligned}
& \mathbf{W}=\left[\begin{array}{cccccccc}
1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\
1 & 0 & 0 & 0 & -1 & 0 & 0 & 0 \\
1 & 0 & -1 & 0 & 1 & 0 & -1 & 0 \\
1 & 0 & 0 & 0 & -1 & 0 & 0 & 0 \\
1 & -1 & 1 & -1 & 1 & -1 & 1 & -1 \\
1 & 0 & 0 & 0 & -1 & 0 & 0 & 0 \\
1 & 0 & -1 & 0 & 1 & 0 & -1 & 0 \\
1 & 0 & 0 & 0 & -1 & 0 & 0 & 0
\end{array}\right] \\
& -j\left[\begin{array}{cccccccc}
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 & 0 & 0 & -1 & 0 \\
0 & 1 & 0 & -1 & 0 & 1 & 0 & -1 \\
0 & 0 & -1 & 0 & 0 & 0 & 1 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 & 0 & 0 & -1 & 0 \\
0 & -1 & 0 & 1 & 0 & -1 & 0 & 1 \\
0 & 0 & -1 & 0 & 0 & 0 & 1 & 0
\end{array}\right] \\
& +\frac{\sqrt{2}}{2}\left[\begin{array}{cccccccc}
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 1 & 0 & -1 & 0 & -1 & 0 & 1 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & -1 & 0 & 1 & 0 & 1 & 0 & -1 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & -1 & 0 & 1 & 0 & 1 & 0 & -1 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 1 & 0 & -1 & 0 & -1 & 0 & 1
\end{array}\right] \\
& -j \frac{\sqrt{2}}{2}\left[\begin{array}{cccccccc}
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 1 & 0 & 1 & 0 & -1 & 0 & -1 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 1 & 0 & 1 & 0 & -1 & 0 & -1 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & -1 & 0 & -1 & 0 & 1 & 0 & 1 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & -1 & 0 & -1 & 0 & 1 & 0 & 1
\end{array}\right] .
\end{aligned}
$$

The algorithm is obtained from (16). As in example 4, we use the array of all linearly independent rows of all matrices $A_{i}$ to write

$$
\mathbf{S}=\left[\begin{array}{cccccccc}
1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\
1 & 0 & 0 & 0 & -1 & 0 & 0 & 0 \\
1 & 0 & -1 & 0 & 1 & 0 & -1 & 0 \\
1 & -1 & 1 & -1 & 1 & -1 & 1 & -1 \\
0 & 0 & 1 & 0 & 0 & 0 & -1 & 0 \\
0 & 1 & 0 & -1 & 0 & 1 & 0 & -1 \\
0 & 1 & 0 & -1 & 0 & -1 & 0 & 1 \\
0 & 1 & 0 & 1 & 0 & -1 & 0 & -1
\end{array}\right] \mathbf{v}
$$

$$
\mathbf{S}=\left[\begin{array}{c}
S_{0} \\
S_{1} \\
S_{2} \\
S_{3} \\
S_{4} \\
S_{5} \\
S_{6} \\
S_{7}
\end{array}\right]
$$

Let

$$
m_{1}=\frac{\sqrt{2}}{2} S_{4}
$$

and

$$
m_{2}=-j \frac{\sqrt{2}}{2} S_{5}
$$

The DFT is computed by

$$
\mathbf{V}=\left[\begin{array}{c}
S_{0} \\
S_{1}-j S_{4}+m_{1}+m_{2} \\
S_{2}-j S_{5} \\
S_{1}+j S_{4}-m_{1}+m_{2} \\
S_{3} \\
S_{1}-j S_{4}+m_{1}-m_{2} \\
S_{2}+j S_{5} \\
S_{1}+j S_{4}-m_{1}-m_{2}
\end{array}\right],
$$

or

$$
\mathbf{V}=\left[\begin{array}{cccccccc}
1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 & -j & 0 & 1 & 1 \\
0 & 0 & 1 & 0 & 0 & -j & 0 & 0 \\
0 & 1 & 0 & 0 & j & 0 & -1 & 1 \\
0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 & -j & 0 & 1 & -1 \\
0 & 0 & 1 & 0 & 0 & j & 0 & 0 \\
0 & 1 & 0 & 0 & j & 0 & -1 & -1
\end{array}\right]\left[\begin{array}{c}
S_{0} \\
S_{1} \\
S_{2} \\
S_{3} \\
S_{4} \\
S_{5} \\
m_{1} \\
m_{2}
\end{array}\right]
$$

Table I shows the multiplicative complexity, for several values of $N$, of the FFT proposed in this paper and that of some well known algorithms. The Heideman lower bound is also indicated on the table.

It can be observed in Table I, that the CBD-FFT has the minimum number of multiplications for $N=3,4,6,8,12$. The algorithm also presents a better performance than the classical Coley-Tukey and Good-Thomas algorithms for most small values of $N$ up to 96 .

## IV. Conclusions

A new algebraic approach, called cyclotomic basis decomposition, was introduced and used as a tool to provide good algorithms, in terms of multiplicative complexity, for computing the discrete Fourier transform. The FFT based on this new approach presents a better performance than

TABLE I
REAL MULTIPLICATIVE COMPLEXITY OF: CBD - CYCLOTOMIC BASIS decomposition FFT; HLB - Heideman lower bound; CT/GT Combination of the two most popular FFT algorithms, fully optimized, the Cooley-Tukey and Good-Thomas algorithms; SW-FFT - Small Winograd FFT.

| N | (CBD-FFT) | (HLB) | (CT/GT) | (SW-FFT) |
| :---: | :---: | :---: | :---: | :---: |
| 3 | 1 | 1 | 8 | 2 |
| 4 | 0 | 0 | 0 | 0 |
| 5 | 5 | 4 | 32 | 5 |
| 6 | 2 | 2 | 16 | - |
| 7 | 13 | 7 | 72 | 8 |
| 8 | 2 | 2 | 4 | 2 |
| 9 | 10 | 8 | 60 | 10 |
| 10 | 10 | 8 | 64 | - |
| 12 | 4 | 4 | 32 | - |
| 16 | 12 | 10 | 20 | 10 |
| 20 | 20 | 16 | 128 | - |
| 24 | 14 | 12 | 108 | - |
| 32 | 54 | 32 | 88 | - |
| 48 | 64 | 38 | 252 | - |
| 64 | 224 | 84 | 208 | - |
| 96 | 258 | 105 | 648 | - |

standard FFT algorithms known in the literature, for various blocklengths.

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