Historical Notes on the Fast Fourier Transform

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Abstract—The fast Fourier transform algorithm has a long and interesting history that has only recently been appreciated. In this paper, the contributions of many investigators are described and placed in historical perspective.

HISTORICAL REMARKS

THE fast Fourier transform (FFT) algorithm is a method for computing the finite Fourier transform of a series of N (complex) data points in approximately $N \log_2 N$ operations. The algorithm has a fascinating history. When it was described by Cooley and Tukey^[1] in 1965 it was regarded as new by many knowledgeable people who believed Fourier analysis to be a process requiring something proportional to N^2 operations with a proportionality factor which could be reduced by using the symmetries of the trigonometric functions. Computer programs using the N^2 -operation methods were, in fact, using up hundreds of hours of machine time. However, in response to the Cooley-Tukey paper, Rudnick^[5], of Scripps Institution of Oceanography, La Jolla, Calif., described his computer program which also takes a number of operations proportional to $N \log_2 N$ and is based on a method published by Danielson and Lanczos^[21]. It is interesting that the Danielson-Lanczos paper described the use of the method in X-ray scattering problems, an area where, for many years after 1942, the calculations of Fourier transforms presented a formidable bottleneck to researchers who were unaware of this efficient method. Danielson and Lanczos refer to Runge^{[6],[7]} for the source of their method. These papers and the lecture notes of Runge and König[8] describe the procedure in terms of sine-cosine series. The greatest emphasis, however, was on the computational economy that could be derived from the symmetries of the sine and cosine functions. In a relatively short section of Runge and König^[8] it was shown how one could use the *periodicity* of the sine-cosine functions to obtain a 2N-point Fourier analysis from two N-point analyses with only slightly more than N operations. Going the other way, if the series to be transformed is of length N and N is a power of 2, the series can be split into log₂ N subseries and this doubling algorithm can be applied to compute the finite Fourier transform in log₂ N doublings. The number of computations in the resulting successive doubling algorithm is therefore proportional to $N \log_2 N$ rather than N^2 . The use of symmetries only reduces the proportionality factor while the successive doubling algorithm replaces N^2 by $N \log N$. This distinction

was not important for the values of N used in the days of Runge and König. However, when the advent of computing machinery made calculations with large N possible, and the N log N methods should have been thoroughly exploited, they were apparently overlooked, even though they had been published by well-read and well-referenced authors.

The fast Fourier transform algorithm of Cooley and Tukey^[1] is more general in that it is applicable when N is composite and not necessarily a power of 2. Thus, if two factors of N are used, so that $N=r\cdot s$, the data is, in effect, put in an r-column, s-row rectangular array, and a two-dimensional transform is performed with a phase-shifting operation intervening between the transformations in the two dimensions. This results in N(r+s) operations instead of N^2 . By selecting N to be highly composite, substantial savings result. For the very favorable situation when N is equal to a power of 2, the Cooley-Tukey method is essentially the successive doubling algorithm mentioned above and takes $N \log_2 N$ operations.

The 23-year hiatus in the use of the algorithm seemed quite remarkable, and prompted us to inquire of Prof. L. H. Thomas at the IBM Watson Scientific Computing Laboratory, New York, N. Y., as to whether he was familiar with the successive doubling algorithm for computing Fourier series, and knew of any occasions when it had been used. It turned out that Prof. Thomas had spent three months in 1948 doing calculations of Fourier series on a tabulating machine, using what he referred to as the "Stumpff method of subseries." The algorithm described by Thomas^[10] was thought at first to be essentially the same as the fast Fourier transform algorithm of Cooley and Tukey since it also achieved its economy by performing onedimensional Fourier analysis by doing multidimensional Fourier analysis. However, the algorithms are different for the following reaons: 1) in the Thomas algorithm the factors of N must be mutually prime; 2) in the Thomas algorithm the calculation is precisely multidimensional Fourier analysis with no intervening phase shifts or "twiddle factors" as they have been called; and 3) the correspondences between the one-dimensional index and the multidimensional indexes in the two algorithms are quite different. The Thomas or "prime factor" algorithm is described in detail and compared with the fast Fourier transform algorithm in the next section.1 It can be extremely useful when used in combination with the fast Fourier transform algorithm.

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¹ Actually, Stumpff^[9] gave only a doubling and a tripling algorithm and suggested (see Stumpff^[9], p. 442, line 11) that the reader generalize to obtain the method for factors of N other than 2 or 3. Thomas made a further assumption (assuming that the index called r by Stumpff was equal to N/s where s=2 or 3) which led to his algorithm. Without this assumption, Stumpff's description leads to the Cooley-Tukey algorithm.

Several other calculations have been reported in the literature and in private communications which use one or the other of the two algorithms.

Another line of development has since led to the Thomas algorithm is its full generality. This comes from work in the analysis and design of experiments. Let $A(k_0, k_1, \dots, k_{m-1})$ be, for example, a crop yield when a level k_i of treatment i, which may be an amount of fertilizer, is used. Yates^[11] considered the case where $k_i=0$ or 1, meaning treatment i is or is not used. This yields $N=2^m$ values of crop yields and, to get all possible differences between all possible averages, one would, in principle, have to compute N linear combinations of all of the A's. This would require N^2 operations. Yates devised a scheme whereby one computed a new array of N sums and differences of pairs of the A's. The process was repeated on the new array with pairs selected in a different order. This was done $m = \log_2 N$ times, meaning he did the calculation in $N \log_2 N$ operations instead of N^2 .

Good^[4] noted that the Yates method could be regarded as m-dimensional Fourier analysis with only two points in each direction and that the procedure could be generalized to one for an arbitrary number of points in each direction. Then Good showed that if N is composite, with mutually prime factors, i.e., $N = r_1, r_2, \dots, r_m$, one could do a onedimensional Fourier analysis of N points by doing m-dimensional Fourier analysis on an *m*-dimensional, $r_1 \times r_2 \times \cdots$, $\times r_m$, array of points. With these ideas put together and developed, Good's paper contains the full generalization of the Thomas prime factor algorithm.

THE PRIME FACTOR ALGORITHM

As mentioned in the previous section, the algorithm used by Thomas and described later by Good has been mistakenly said to be equivalent to the fast Fourier transform algorithm of Cooley and Tukey. It is important to distinguish between these two algorithms since each has its particular advantages which can be exploited in appropriate circumstances.

The differences will be illustrated by considering the calculation of a Fourier series using two factors of N. The Fourier series is

$$X(j) = \sum_{n=0}^{N-1} A(n) W_{N^{jn}}$$
 (1)

where $W_N = e^{2\pi i/N}$. Consider first the fast Fourier transform algorithm. We assume $N = r \cdot s$, and define a one-toone mapping between the integers j, $0 \le j < N$, and the pairs of integers (j_1, j_0) , $0 \le j_0 < r$, $0 \le j_1 < s$, by the relation

$$j = j_1 r + j_0. (2)$$

Similarly, we let

$$n = n_1 s + n_0, \tag{3}$$

where

$$0 \le n < N$$
, $0 \le n_0 < s$, $0 \le n_1 < r$.

This enables us to refer to A(n) and X(j) as though they respectively. Substituting (6) and (8) and using (7) gives

TABLE I CORRESPONDENCE BETWEEN ONE- AND TWO-DIMENSIONAL INDEXING IN THE ARBITRARY FACTOR ALGORITHM FOR THE CASE r = 8, s = 3, AND N = 24

			n = sr	$n_1 + n_0 =$	$=3n_1+$	n_0			
				n_1					
		0	1	2	3	4	5	6	7
	0	0	3	6	9	12	15	18	21
n_0	1	1	4	7	10	13	16	19	22
	2	2	5	8	11	14	17	20	23
			j=r	$j_1 + j_0 =$	$=8j_1+j_1$	0			
				$\overset{j_0}{2}$			_		_
		0	1	2	3	4	5	6	7
	0	0	1	2	3	4	5	6	7
j_1	1	8	9	10	11	12	13	14	15
	2	16	17	18	19	20	21	22	23

were two-dimensional arrays and permits us to do the Fourier analysis in two steps:

$$A_1(j_0, n_0) = \sum_{n_1=0}^{r-1} A(n_1, n_0) W_r^{j_0 n_1}$$
 (4)

$$X(j_1, j_0) = \sum_{n_0=0}^{s-1} A_1(j_0, n_0) W_s^{j_1 n_0} W_N^{j_0 n_0}.$$
 (5)

Table I shows where A(n) and X(j) are placed in the twodimensional arrays indexed by (n_1, n_0) and (j_1, j_0) , respectively, for r=8 and s=3. For this case, (4) consists of three eight-term Fourier series, one for each row of the n table. Then if j_0 is taken to be the column index of the results, $A_1(j_0, n_0)$, (5) describes eight Fourier series of three terms each on the columns of the array of $A_1(j_0, n_0)W_N^{j_0n_0}$. The factor W^{jon_0} , referred to as the "twiddle factor" by Gentleman and Sande^[3], is usually combined with either the $W_r^{j_0n_1}$ factor in (4) or the $W_s^{j_1n_0}$ factor in (5).

For the Thomas prime factor algorithm, one must require that r and s be mutually prime. In this case, different mappings of the one-dimensional arrays into two-dimensional arrays are used. These are also one-to-one mappings and are defined as follows. Let

$$n \equiv rn_0 + sn_1 \pmod{N} \qquad (0 \le n < N) \tag{6}$$

and

$$j_0 \equiv j \pmod{r} \qquad (0 \le j_0 < r)$$

$$j_1 \equiv j \pmod{s} \qquad (0 \le j_1 < s). \tag{7}$$

Then the expression of j, in terms of j_0 and j_1 , is a solution of the "Chinese remainder problem" and is given by

$$j \equiv s \cdot s_r j_0 + r \cdot r_s j_1 \pmod{N} \qquad (0 \le j < N) \tag{8}$$

where s_r and r_s are solutions of

$$s \cdot s_r \equiv 1 \pmod{r}$$
 $s_r < r$
 $r \cdot r_s \equiv 1 \pmod{s}$ $r_s < s$

TABLE II

CORRESPONDENCE BETWEEN ONE- AND TWO-DIMENSIONAL INDEXING IN THE PRIME FACTOR ALGORITHM FOR THE CASE r=8, s=3, and N=24

	n≡	$rn_0 + s$	$sn_1 \equiv 8n$	$n_0 + 3n_1$	(mod :	24, 0≤	n < N)		
		0	1	$\stackrel{n_1}{2}$	3	4	5	6	7
	0	0	3	6	9	12	15	18	21
n_0	1	8	11	14	17	20	23	2	5
•	2	16	19	22	1	4	7	10	13
	$j \equiv s$	$s_r j_0 + i$	$r \cdot r_s j_1 =$	$=9j_0+1$	6j ₁ (m	od 24,	0≤ <i>j</i> < <i>i</i>	V)	
	$j \equiv s$	$s_r j_0 + i$	$r \cdot r_s j_1 = 1$	$=9j_0+1$ j_0 2	6j ₁ (m	od 24.	0≤ <i>j</i> < <i>i</i>	V) 6	7
	$j \equiv s$								7
j_1		0	1	<i>j</i> ₀ 2	3	4	5	6	7 15 7

$$W_{N}^{jn} = W_{N}^{jn_0r}W_{N}^{jn_1s} = W_{s}^{jn_0}W_{r}^{jn_1} = W_{s}^{j_1n_0}W_{r}^{j_0n_1}$$

which enables one to write the Fourier series (1) in the form

$$A_1(j_0, n_0) = \sum_{n_1=0}^{r-1} A(n_1, n_0) W_r^{j_0 n_1}$$
 (9)

$$X(j_1, j_0) = \sum_{n_0=0}^{s-1} A_1(j_0, n_0) W_s^{j_1 n_0}.$$
 (10)

As in the fast Fourier transform algorithm, this is a two-dimensional Fourier transform. The essential difference is that the "twiddle factor" $W_N^{jon_0}$ does not appear in (10) and the correspondence between one- and two-dimensional indexing is different. The presence of the "twiddle factor"

does not introduce any more computation, but it does increase programming complexity slightly. To illustrate better how the indexing in the two algorithms differs, the mappings of n and j for the Thomas prime factor algorithm are given in Table II for comparison with the indexing described in Table I.

The prime factor algorithm can be programmed very easily in a source language like FORTRAN and, therefore, can be used efficiently with a subroutine designed for a number of terms equal to a power of two. For example, if r is a power of two and s is any odd number, the subseries (9) can be computed by the power of two subroutine.

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Limits on Bandlimited Signals

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Abstract—The problem of estimating the errors arising in a variety of numerical operations involving bandlimited functions is considered. The errors are viewed as responses of suitably created systems, and the analysis is based on the evaluation of the maximum response of these systems in terms of the energy or power of their input. The investigation includes deterministic and random signals, and it is extended to two-dimensional functions and Hankel transforms. Finally, the results are related to the uncertainty principle in one and two variables.

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I. Introduction

N MANY AREAS of engineering and applied sciences, a recurring task is the simplification of various operations by suitable approximations: derivatives by differences, integrals by sums, infinite series by finite truncations, and others. Such approximations are meaningful only if the resulting errors are small in some sense, and a basic problem in numerical analysis is to establish realistic bounds to these errors. With few exceptions, this problem is difficult and in most cases it is avoided by using, in truncations, for example, a "sufficiently" large number of terms, or, in quadratures, "small enough" intervals. An interval is small,