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A Discrete Fourier Transform for Virtual Memory Machines

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Summary

The discrete Fourier transform (DFT) is a pervasive tool in scientific and engineering calculation. It lies at the heart of many methods for solving partial differential equations, data analysis, statistical calculations, and of course Fourier Analysis (see ref 1). More than one hundred papers on fast algorithms have been published! So, why another? Most DFT algorithms assume use of high speed computer memory that is local and contiguous; others assume the data is stored in blocks of external – remote – memory. In contrast, the efficient DFT algorithm provided in this paper assumes neither case; instead, it works efficiently on computers with virtual or cache memory where the location of the data is mapped automatically onto a small high speed memory by the computer's operating system. Important features of the algorithm are:

- no details of page mechanism to be used
- the code should be compact and clear
- no auxiliary storage should be required
- efficient execution when the data resides entirely in high speed memory.

Additionally, a matrix theory of the DFT is given. This theory provides the basis of the "fast" calculation of the transforms and the proofs give the insight to efficient organization of the calculation under different assumptions.

Implementing DFT's on Virtual Memory Machines

The DFT is most directly expressed as the product of a matrix with a vector. If $(x_1, x_2, \dots, x_n)^T$ denotes the data to be transformed, $\omega_n = \exp(2\pi i/n)$, and

$$\mathcal{W}_{n} = \begin{bmatrix} 1 & 1 & 1 & \cdots & 1 \\ 1 & \omega_{n}^{-1} & \omega_{n}^{-2} & \cdots & \omega_{n}^{1-n} \\ 1 & \omega_{n}^{-2} & \omega_{n}^{-4} & \cdots & \omega_{n}^{2 \cdot (1-n)} \\ \vdots & & & & & & \\ 1 & \omega_{n}^{1-n} & \cdots & \omega_{n}^{(1-n) \cdot (1-n)} \end{bmatrix}$$
(1)

then $a = w_n$ x is the discrete Fourier transform of x. If a super bar is used to denote complex conjugate, $\frac{1}{n} \overline{w_n}^{-1} \cdot x$ is the inverse discrete Fourier transform of x. w_n is called the discrete Fourier operator.

Appendix B provides decompositions of the matrix w_n as a product of sparse matrices. These decompositions provide a framework for implementing versions of fast DFTs. The results are not new, but the details of the proofs provide computational insight.

DFT algorithms to evaluate (1) are based on writing W_n as a product of sparse matrices which are then sequentially applied to x (ref 2, ref 3, ref 4.). The decomposition is not unique, so we may choose one that has properties that we find important.

The salient feature of computers with hierarchies of memory is that transfers of data between different levels is by blocks rather than individual words. Therefore, more complete use of data in a block yields a premium in inter-memory traffic. The features that we wish to emphasize, then, flow from a high locality of data. One simple approach is to access data in sequentially increasing order and to use every datum as much as possible before a block is discarded. Unfortunately, the latter is difficult to achieve without detailed knowledge of block sizes, but, careful use of the former yields almost all of the efficiency gain.

The following discussion assumes that the calculation overwrites the data to be transformed and the vector length is a power of two. Appendix B decomposes the DFT into a two phases. One scrambles the data; the other converts the data into its Fourier coefficients.

In the scrambling phase, usually the binary reversal of an index is calculated directly and the two elements of the vector are exchanged. A number and its binary reversal usually occur in different parts of the vector. The binary reversal of two successive integers are in different halves of the vector; even integers have binary reversals in the lower half, odd integers have their binary reversal in the upper half. So potentially almost as many block transfers occur as half the number of elements in the vector. Direct application of the unscrambling matrix decomposition requires one sequential pass through the data for each factor. Thus, total data motion is considerably reduced. When the vectors are long and real memory is small, the overhead from data motion is reduced considerably.

During the other phase of the DFT, the computation can be organized in several ways, one is to apply a factor of the decomposition to the same element in each block. This minimizes the computation of trigonometric function values (which are mostly calculated by recurrence relations), but does not address the data in a strictly sequential fashion. This organization is advantageous for machines with vector processing capabilities. Applying a factor of the decomposition on a complete block by complete block basis leads to strictly sequential access of data at the cost of repeated calculation of trigonometric function values. However, the decrease in overhead from data motion can often more than compensate this cost. This organization is also more suitable for machines with multiple processors than the one above.

The algorithm given in Appendix A is based upon the factorization in equation (15b). The transformation phase is based upon base four transforms rather than purely base two transforms because there is an increase in computational efficiency without complication to the coding of the algorithm.

References

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APPENDIX A - Fortran Code

SUBROUTINE VMDFT(N, A, IER)

•	SUBROUTINE VMDFT(N, A, IER)	
C	VIRTUAL MEMORY DISCRETE FOURIER TRANSFORM	I
C	THIS ROUTINE EVALUATES	I
С		Ι
С	N-1 JK	I
.C	X(J) = SUM (A(K)W), J = 0(1)N-1	I
C	K=0	I
C	NURDE N - EVE /2+DI+I/N\ I - CORE/ 1\ N IC & DOMED OF 2	I I
C C	WHERE $W = EXP(2*Pi*i/N)$, $I = SQRT(-1)$, N IS A POWER OF 2	<u> </u>
C	FORMAL PARAMETERS:	I
C		Ī
C	N INTEGER, IN. LENGTH OF A	I
С		I
С	A COMPLEX VECTOR OF LENGTH N. AT ENTRY, A CONTAINS THE	I
С	DATA TO BE TRANSFORMED. AT EXIT, IT CONTAINS THE	I
С		Ι
C		I
C		I
C C		I I
C		Ī
C		
C	CACM CATOGORY: J1	I
C		-
C C	KEYWORDS: FOURIER TRANSFORM, VIRTUAL MEMORY, BINARY REVERSAL	_
С	THE ALGORITHM IS A VARIATION OF THE COOLEY-TUKEY ALGORITHM.	I
С		Ι
C	-,	I
С	-,	I
C		I I
C C		I
C	.,	I
C	o,	I
Ċ		I
С		I
	MUST BE AT LEAST 4 DATA PAGES PLUS ROOM FOR THE CODE FOR	
С	MOST DE AT BEAST 4 DATA TAGES TESS MOST TOX THE CODE TOX	Ι
C C		I
C C	IMPROVEMENT TO SHOW 9) THE CODE IS CLEAR AND COMPACT	
C C C	IMPROVEMENT TO SHOW 9) THE CODE IS CLEAR AND COMPACT 10) THE COST IN CPU TIME OVER THE MOST EFFICIENT DFT ALGORITH	I I I
C C C	IMPROVEMENT TO SHOW 9) THE CODE IS CLEAR AND COMPACT 10) THE COST IN CPU TIME OVER THE MOST EFFICIENT DFT ALGORITH IS SMALL	I I
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C C C C	IMPROVEMENT TO SHOW 9) THE CODE IS CLEAR AND COMPACT 10) THE COST IN CPU TIME OVER THE MOST EFFICIENT DFT ALGORITH IS SMALL USAGE:	I I I - I
C C C C C C C C	IMPROVEMENT TO SHOW 9) THE CODE IS CLEAR AND COMPACT 10) THE COST IN CPU TIME OVER THE MOST EFFICIENT DFT ALGORITH IS SMALL USAGE:	I I I
C C C C C C C C C	IMPROVEMENT TO SHOW 9) THE CODE IS CLEAR AND COMPACT 10) THE COST IN CPU TIME OVER THE MOST EFFICIENT DFT ALGORITH IS SMALL USAGE: TO EVALUATE THE FOURIER TRANSFORM OF THE N = 2**M DATA STORED	I I I - I
C C C C C C C C	IMPROVEMENT TO SHOW 9) THE CODE IS CLEAR AND COMPACT 10) THE COST IN CPU TIME OVER THE MOST EFFICIENT DFT ALGORITH IS SMALL USAGE: TO EVALUATE THE FOURIER TRANSFORM OF THE N = 2**M DATA STORED IN THE COMPLEX VECTOR A:	I I I - I I I
C C C C C C C C C C C	IMPROVEMENT TO SHOW 9) THE CODE IS CLEAR AND COMPACT 10) THE COST IN CPU TIME OVER THE MOST EFFICIENT DFT ALGORITH IS SMALL USAGE: TO EVALUATE THE FOURIER TRANSFORM OF THE N = 2**M DATA STORED IN THE COMPLEX VECTOR A:	I I I I I I I I

c	PARA	METER DECLARATIONS			
-	INTEGER N, IER COMPLEX A(N)				
c	INTE	RNAL DECLARATIONS			
	COMPLEX AO, A1, A2, REAL C1, C2, C3, DC REAL S1, S2, S3, TH INTEGER K, KB, KK, INTEGER KO, K1, K1M				
C	CON	STANTS			
Č	DATA ZERO /0.0/, ON DATA FOUR /4.0/, HA RAD = FOUR * ATAN(O	E /1.0/, TWO /2.0/ LF /0.5/, QRTR /0.25/, THRHLF /1.5/ NE)			
С	STAGE 0 - TEST INPUT AND INITIALIZE I				
C	IER = 1 NL = N IF (NL .LE. 0) IER = 2	RETURN			
C	M = 0	>>> COMPUTE LOG BASE 2 OF N			
10	MM = 1 IF (MM .GE. NL) MM = MM + MM M = M + 1 GOTO 10	GOTO 20			
20	IF (MM .GT. NL) IER = 0 MP = M / 2	RETURN			

```
C-----
        STAGE 1 - SCRAMBLING THE DATA
C----
   BINARY REVERSAL BY SEQUENTIAL EXCHANGE OF SYMMETRICALLY LOCATED
С
    BITS REQURIED M/2 SEQUENTIAL PASSES OVER THE DATA AND HAS MUCH
C
C
    LOWER OVERHEAD FOR MACHINES WITH PAGED MEMORY THAN USING A
   BINARY REVERSAL COUNTER
С
C-----
    THE CODE IS A VERY SLIGHTLY MODIFIED VERSION OF THE SUBROUTINE I
С
С
    "UNSBIN" GIVEN IN
    POLGE, R. J., B. K. BHAGAVAN, AND J. M. CARSWELL, "FAST
С
                                                     I
    COMPUTATIONAL ALGORITHMS FOR BIT REVERSAL, " IEEE TRANSACTIONS
С
   ON COMPUTERS, C-23(1), PP. 1-9.
    KSP = 1
    KSH = NL
C----->>> DO (M/2) PASSES OVER THE DATA
    DO 140 KP = 1, MP
        KSPP = KSP + 1
         KSPM = KSP - 1
         KS = KSH
         KSH = KSH / 2
         KST = KSH - KSP
        KSP = KSP + KSP
C---->>STEP THROUGH THE FILE SEGMENTS
        DO 130 KO = KSPP, NL, KS
           K1MX = K0 + KST
C---->>>STEP THROUGH BLOCKS IN EACH FILE SEGMENT
             DO 120 K1 = K0, K1MX, KSP
             K2MX = K1 + KSPM
C---->>>SWAP TWO BLOCKS OF DATA
                  DO 110 K2 = K1, K2MX
                      K3 = K2 + KST
                      S = A(K2)
                      A(K2) = A(K3)
                      A(K3) = S
                  CONTINUE
110
120
              CONTINUE
130
         CONTINUE
140 CONTINUE
```

C		
c	STAGE 2 - THE TRANSFORMATIONS	_ I _
c c c	RADIX 4 + 2 FAST FOURIER TRANSFORM ROUTINE USING A COOLEY-TUKEY LIKE ALGORITHM FOR BINARY REVERSED DATA. THE ORGANIZATION USED REQUIRES THE MINIMUM NUMBER OF SEQUENTIAL PASSES THROUGH THE DATA AND ALLOWS THE SINE-COSINE VALUES TO BE GENERATED IN SEQUENTIAL ORDER AS WELL.	I I I I
C C C C C	THE CODE IS ADAPTED FROM THE ALGOL PROCEDURE "REVFFT4" FROM CACM ALGORITHM 345, "ALGORITHM 345, AN ALGOL CONVOLUTION PROCEDURE BASED ON THE FAST FOURIER TRANSFORM," BY RICHARD C, SINGLETON.]]]
С	IF M IS ODD THEN DO A RADIX 2 TRANSFORM	_
210	<pre>KSP = 1 IF (2 * MP .EQ. M) GOTO 215 DO 210 K0 = 1, NL, 2</pre>	
_	NOW DO M/2 RADIX 4 TRANSFORMS	
215	IF (MP .LT. 1) RETURN DO 250 KP = 1, MP RAD = QRTR * RAD	
C C C	NOTE: SIN-COS REFERENCES CAN BE REPLACED BY INLINE APPROXIMATIONS	
C	DC = TWO * SIN(RAD) ** 2 DS = SIN(TWO * RAD) KST = 4 * KSP	

```
C-----BLOCK BEING TRANSFORMED
           DO 240 KB = 1, NL, KST
                C1 = ONE
                 S1 = ZERO
                 -----TRANSFORM THE BLOCK
                 DO 230 KK = 1, KSP
                      K = KK - 1
                      K0 = KB + K
                      K1 = K0 + KSP
                      K2 = K1 + KSP
                      K3 = K2 + KSP
                      A0 = A(K0)
                      A2 = A(K1)
                      A1 = A(K2)
                      A3 = A(K3)
                      IF ( K .EQ. 0 )
                                         GOTO 225
С
                                        ------
С
                                       ADVANCE TRIG FUNCTION VALUES
С
                      C2 = C1 - (DC * C1 + DS * S1)
                      S1 = S1 + (DS * C1 - DC * S1)
С
С
                                THE FOLLOWING 3 STATEMENTS COMPENSATE
С
                               FOR TRUNCATION ERROR. IF ROUNDED
С
                                ARITHMETIC IS USED, REPLACE THEM WITH
С
                                          C1 = C2
С
                     C1 = THRHLF - HALF * (C2 * C2 + S1 * S1)
                     S1 = C1 * S1
                     C1 = C1 * C2
                     C2 = C1 * C1 - S1 * S1
                     S2 = TWO * S1 * C1
                     C3 = C2 * C1 - S2 * S1
                     S3 = C2 * S1 + S2 * C1
C---->>>APPLY RADIX 4 TRANSFORM
                     A2 = A2 * CMPLX(C2, S2)
                     A1 = A1 * CMPLX(C1, S1)
                     A3 = A3 * CMPLX(C3, S3)
225
                     S = (A0 + A2)
                     T = (A1 + A3)
                     A(K0) = S + T
                     A(K2) = S - T
                     S = (A0 - A2)
                     T = (A1 - A3)
                     TR = REAL(T)
                     TI = AIMAG(T)
                     T = CMPLX(-TI, TR)
                     A(K1) = S + T
                     A(K3) = S - T
230
                 CONTINUE
240
          CONTINUE
          KSP = KST
250
     CONTINUE
                           RETURN
```

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END

APPENDIX B - ALGEBRAIC THEORY

DFT Identities

This appendix gives the mathematical background and required identities for the DFT viewed as a matrix operator. We begin with some definitions. The results have been previously published, but the development is more complete than published results and the actual development is more directly useful in teasing out details for actual implementation.

A **permutation matrix** is any matrix which can be obtained from an identity matrix by permuting its rows or columns. Let the set of numbers $\{c(j)\}$, $0 \le j \le p-1$, be a permutation of the integers $0 \dots p-1$, and let $\{r(j)\}$ be the inverse permutation:

$$r(c(j)) = c(r(j)) = j$$

The permutation matrix P, of order p, is defined by

$$P_{i,k} = \delta(j,r(k)) = \delta(c(j),k), 0 \le j, k \le p-1.$$

The functions r(j) and c(j) are called the **row** and **column** functions, respectively. Thus the matrix has just one non-zero element (equal to unity) in each row and column. The unit element in row j is c(j), and the unit element in column k is in row r(k).

The following properties of permutation matrices follow directly from the definition:

- (1) The transpose is the inverse: $P^T = P^{-1}$
- (2) If P_1 and P_2 are permutation matrices of the same order with row functions r_1 and r_2 and columns c_1 and c_2 , then their product $P = P_1P_2$ is a permutation matrix with row and column functions given by the transitive formulas

$$\mathbf{r}(\mathbf{j}) = \mathbf{r}_1[\mathbf{r}_2(\mathbf{j})]$$

$$c(j) = c_2[c_1(j)]$$

(3) IF X is an arbitrary matrix, the effect of premultiplying by a permutation matrix is to permute the rows, while post multiplying permutes the columns:

Y = PX

Z = XP

Then

Let

 $Y_{j\,k} = X_{c(j),\,k}$

 $Z_{i,k} = X_{i,r(k)}$

In particular, if

 $U = XP^T$

then

 $U_{ik} = X_{i, c(k)}$

That is, premultiplying by P produces a permutation of the rows, postmultiplying by the transpose P^{T} produces the *same* permutation of the columns.

Two important permutation matrices are the square deal $S_{p,q}$ and its transpose (and inverse) the perfect shuffle, $T_{q,p}$, each of order pq, defined by the the row and column functions

$$c_S(j + lq) = jp + l = r_T(j + lq)$$

 $r_S(jp + l) = j + lq = c_T(jp + l)$

with $0 \le j \le q-1$, $0 \le k \le p-1$.

Premultiplying by S represents dealing a deck of pq cards to p players, while premultiplying by T represents shuffling (merging) p decks of q cards each. The terminology is actually somewhat ambiguous, since the roles are reversed in postmultiplication. Note the identities

$$\begin{split} S_{q, p} &= T_{p, q} = S_{p, q}^{T} = S_{p, q}^{-1} \\ S_{p, p} &= T_{p, p} \\ S_{p, 1} &= S_{1, p} = T_{p, 1} = T_{p, 1} = I_{p} \end{split}$$

The **tensor product** of two matrices, A and B, denoted $A \otimes B$, is the partitioned block matrix obtained by multiplying each element of A by the matrix B

$$A \otimes B = (A_{l, m} B)$$

more specifically, if A is of order p and B is of order q, and $C = A \otimes B$, then C is of order pq, with elements given by

$$C_{j+lq, k+mq} = A_{l, m} B_{j, k}$$
, $0 \le j, k \le q-1, 0 \le l, m \le p-1$.

The following properties follow directly from the definition:

Associative Law:

$$(A \otimes B) \otimes C = A \otimes (B \otimes C) = A \otimes B \otimes C \tag{2}$$

Similarity Law: If A is of order p and B is of order q, then

$$B \otimes A = T_{p,q}(A \otimes B) S_{p,q}$$
(3)

Transpose:
$$(A \otimes B)^T = A^T \otimes B^T$$
 (4)

Inverse:
$$(A \otimes B)^{-1} = A^{-1} \otimes B^{-1}$$
 (5)

Identity:
$$A \otimes I_1 = I_1 \otimes A = A$$

$$I_p \otimes I_q = I_{pq}$$
 (6)

Multiplicative Law:
$$(AB)\otimes(CD) = (A\otimes C)(B\otimes D)$$
 (7)

Distributive Law:
$$I \otimes \prod_{j} A_{j} = \prod_{j} (I \otimes A_{j}), \left(\prod_{j} A_{j}\right) \otimes I = \prod_{j} (A_{j} \otimes I)$$
 (8)

where the Aj are all of the same order, I is an identity matrix of arbitrary order, and \prod_{i} denotes the

ordinary product of a finite number of factors.

Basic Factor Theorem If P is of order p and Q is of order Q. Then

$$P \otimes Q = (PI_p) \otimes (I_q Q)$$

$$= (P \otimes I_q)(I_p \otimes Q)$$

$$= (I_p P) \otimes (QI_q)$$

$$= (I_p \otimes Q)(P \otimes I_q)$$
(9)

Thus, the tensor product can be expressed as an ordinary (commutative) product), each factor being a tensor product with an identity matrix.

If P and Q are permutation matrices, then so is their tensor product $A = P \otimes Q$, and its column function is

$$c_A(j + lq) = c_O(j) + q c_P(l), 0 \le j < q, 0 \le l < p$$

and its row function is

$$r_A(k + mq) = r_Q(k) + qr_P(m), 0 \le k < q, 0 \le m < p$$

since

$$A_{j+lq, k+mq} = P_{l, m}Q_{j, k}$$

The tensor product has a simple representation when one factor is a permutation matrix and the other is an identity matrix. Let P be a permutation matrix of order n with column function c(j), and let I_s be the identity matrix of order s. The matrices

$$Q = P \otimes I_s$$

$$R = I_s \otimes P$$

are permutation matrices with column functions

$$c_{Q}(js+1) = sc(j) + 1$$

$$c_{R} = \ln + c(j)$$

$$0 \le j < n$$

$$0 \le 1 < s$$

So, if P is regarded as the operator that shuffles (permutes) a deck of cards, then the effect of Q is to perform the same permutation on a deck of "packets", each packet containing s cards that are stuck together. Thus, for example, the matrices

$$S_{p,q} \otimes I_{s}$$

$$T_{p,q} \otimes I_{s}$$
 (10)

can be described as the "dirty deal" and the "sticky shuffle" respectively. On the other hand, the effect of R is to shuffle's decks in parallel and then stack them together. Combining these gives the formula for the triple product:

Let $B = I_p \otimes P_q \otimes I_r$ where P_q is a permutation matrix of order q with column function $c_p(j)$. Then B is a permutation matrix of order pqr with column function $c_p(j)$ given by

$$c_B(\alpha qr + \beta r + \gamma) = \alpha qr + rc_P(\beta) + \gamma$$
 for $0 \le \alpha < p$, $0 \le \beta < q$, $0 \le \gamma < r$

More generally, if A, B, C are arbitrary matrices of order p, q, r respectively, then the elements of the triple product

$$D = A \otimes B \otimes C$$

are given by

$$D_{iqr + kr + m, jqr + lr + n} = A_{i, j}B_{k, l}C_{m, n}$$

 $0 \le i, j < p; 0 \le k, l < q; 0 \le m, n < r$

Even Order DFT

The reduction of the Fourier matrix of even order M = 2q to a product of simpler matrices depends upon the simple observation

$$\begin{array}{c} \omega_{2q}^{2jk} = \omega_q^{jk} \\ \omega_{2q}^{(2j+1)k} = \omega_q^{jk} \omega_{2q}^{k} \end{array} \begin{array}{c} 0 \leq j < q \\ 0 \leq k < 2q \end{array}$$

That is, the even numbered rows of W_{2q} contain the elements of W_q , while the elements in the odd numbered rows are products of elements from both W_q and W_{2q} . This suggests permuting the rows of W_{2q} to bring the even numbered rows together in one block and the odd numbered rows into another. This can be done by using $S_{2,q}$. Define $V = S_{2,q} W_{2q}$. The elements of V are

$$V_{j,k} = \omega_q^{jk}$$

$$V_{j+q,k} = \omega_q^{jk} \omega_{2q}^{k}$$

for $0 \le j$, k < q. The columns can be partitioned by noting

$$\omega_q^{j(k+q)} = \omega_q^{jk}$$

$$\omega_{2q}^{(k+q)} = -\omega_{2q}^{k}$$

for $0 \le j$, k < q. Thus the elements of V are

$$V_{j,k} = \omega_q^{jk}, V_{j,k+q} = \omega_q^{jk}$$

$$V_{j+q,k} = \omega_q^{jk} \omega_{2q}^{k}, V_{j+q,k+q} = -\omega_q^{jk} \omega_{2q}^{k}$$

for $0 \le j$, k < q. Define

Then

$$V = \begin{bmatrix} w_{q} & w_{q} \\ w_{q}D_{q}(\omega_{2q}) & -w_{q}D_{q}(\omega_{2q}) \end{bmatrix}$$
$$= \begin{bmatrix} w_{q} & 0 \\ 0 & w_{q} \end{bmatrix} \begin{bmatrix} I_{q} & 0 \\ 0 & D_{q}(\omega_{2q}) \end{bmatrix} \begin{bmatrix} I_{q} & I_{q} \\ I_{q} & -I_{q} \end{bmatrix}$$

which using the results from the previous section can be written compactly as

$$W_{2q} = T_{2,q} \left(I_2 \otimes w_q \right) \Delta_{2,q} \left(w_2 \otimes I_q \right) \tag{11}$$

Where

$$\Delta_{2, q} = \begin{bmatrix} I_q & 0 \\ 0 & D_q(\omega_{2q}) \end{bmatrix}$$
 and $W_2 = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$

The Fundamental Factor Theorem

From the above, the DFT of M = 2q points can be written as a product of sparse matrices. If the integer q is, itself, a power of two, then the above can be iterated to yield the complete Cooley-Tukey algorithm. However, the process generalizes to general factors.

Let

$$\Delta_{p,q} = \begin{bmatrix} I_q & & & & & \\ & D_q(\omega_{pq}) & & & & \\ & & D_q(\omega_{pq}^2) & & & \\ & & & \ddots & & \\ & & & & D_q(\omega_{pq}^{p-1}) \end{bmatrix}$$

Note that

$$\Delta_{\mathbf{q}, \mathbf{p}} = \mathbf{T}_{\mathbf{p}, \mathbf{q}} \, \Delta_{\mathbf{p}, \mathbf{q}} \, \mathbf{S}_{\mathbf{p}, \mathbf{q}} \tag{12}$$

and reduces to the identity if either subscript is unity

$$\Delta_{p, 1} = \Delta_{1, p} = I_p$$

Fundamental Factor Theorem

$$W_{pq} = T_{p, q} (I_p \otimes W_q) \Delta_{p, q} (W_p \otimes I_q)$$
(13a)

alternatively,

$$W_{pq} = (W_q \otimes I_p) \Delta_{q, p} (I_q \otimes W_p) T_{p, q}$$
(13b)

Proof: We first prove (13a). By analogy with the even order case. Start by applying the square deal to w

$$V = S_{p,q} W_{pq}$$

The elements of V are

$$V_{j \; + \; lq, \; k \; + \; mq} = \sum_{r \; = \; 0}^{pq \; - \; l} \; S_{p,q_{j \; + \; lq, \; r}} \; \mathcal{W}_{r, \; k \; + \; mq}, \; 0 \leq j, \; k \leq q, \; 0 \leq l, \; m < p.$$

But

$$S_{p,q_{j+1}q,n} = \delta[n, c_{S}(j+1q)]$$

= $\delta[n, jp + 1]$

so that

$$V_{j+lq, k+mq} = W_{jp+l, k+mq}$$

= $\omega_{pq}^{(jp+l)(k+mq)}$

but recalling that

$$\omega_{pq} = e^{2\pi i/pq},$$

the identities

$$\omega_{pq}^p = \omega_q$$
 $\omega_{pq}^q = \omega_q$
 $\omega_{pq}^{pq} = 1$

give

$$V_{j+lq, k+mq} = \omega_q^{jk} \omega_{pq}^{lk} \omega_p^{lm}$$

Thus, V is a partitioned block matrix, with (l, m) denoting the row and column position of each block and (j, k) denoting the local row and column position of each element within the block $V = (V_{l,m})$, $0 \le l$, m < q. Hence,

$$V_{l, m} = \mathcal{W}_q D_q (\omega_{pq}^l) \omega_p^{lm} I_q$$

Since W_q is a common factor of every block and $D_q(\omega_{pq}^{-1})$ is a common factor of every block in row 1, V can be factored:

$$\mathbf{V} = \begin{bmatrix} w_q & & & \\ & w_q & & O \\ & & w_q & \\ & O & \ddots & \\ & & & w_q' \end{bmatrix} \begin{bmatrix} \mathbf{I}_q & & & & \\ & \mathbf{D}_q(\omega_{pq}) & & O \\ & & \mathbf{D}_q(\omega_{pq}^2) & \\ & O & & \ddots & \\ & & & \mathbf{D}_q(\omega_{pq}^{p-1}) \end{bmatrix} (w_p \otimes \mathbf{I}_q)$$

that is,

$$V = (I_p \otimes \mathcal{W}_q) \Delta_{p, q} (\mathcal{W}_p \otimes I_q)$$

Since, $V = S_{p,q} \mathcal{W}_{pq}$ and $T_{p,q}$ is the inverse of $S_{p,q}$, the result follows immediately.

(13b) follows by writing

$$\begin{split} \mathcal{W}_{pq} &= T_{p, q} \left(I_{p} \otimes \mathcal{W}_{q} \right) S_{p, q} T_{p, q} \Delta_{p, q} S_{p, q} T_{p, q} \left(\mathcal{W}_{p} \otimes I_{q} \right) S_{p, q} T_{p, q} \\ &= \left(T_{p, q} \left(I_{p} \otimes \mathcal{W}_{q} \right) S_{p, q} \right) \left(T_{p, q} \Delta_{p, q} S_{p, q} \right) \left(T_{p, q} \left(\mathcal{W}_{p} \otimes I_{q} \right) S_{p, q} \right) T_{p, q} \end{split}$$

and then applying (3) and (12).

Complete Factor Theorem

Clearly the fundamental factor theorem can be iterated if either of the integers p, q is not prime. For example, take $p = p_2$, $q = p_1$, then

$$\mathcal{W}_{p_2p_1} = T_{p_2,\;p_1}\;\left(I_{p_2}\!\!\otimes\!\mathcal{W}_{p_1}\right)\Delta_{p_2,\;p_1}\left(\mathcal{W}_{p_2}\!\!\otimes\!I_{p_1}\right)$$

Now take $p = p_3$, $q = p_2p_1$ in the fundamental theorem giving

$$\mathcal{W}_{p_{3}p_{2}p_{1}} = T_{p_{3},\; p_{2}p_{1}}\;\left(I_{p_{3}} \otimes \mathcal{W}_{p_{2}p_{1}}\right) \Delta_{p_{3},\; p_{2}p_{1}}\left(\mathcal{W}_{p_{3}} \otimes I_{p_{2}p_{1}}\right)$$

Inserting the expression given above for $W_{p_2p_1}$ and using the distributive law,

$$\mathcal{W}_{p_{3}p_{2}p_{1}} = T_{p_{3},\ p_{2}p_{1}}\ \left(I_{p_{3}} \otimes T_{p_{2},\ p_{1}}\right)\left(I_{p_{3}p_{2}} \otimes \mathcal{W}_{p_{1}}\right)\left(I_{p_{3}} \otimes \Delta_{p_{2},\ p_{1}}\right)\left(I_{p_{3}} \otimes \mathcal{W}_{p_{2}} \otimes I_{p_{1}}\right)\Delta_{p_{3},\ p_{2}p_{1}}\left(\mathcal{W}_{p_{3}} \otimes I_{p_{2}p_{1}}\right)$$

To continue the process, only a proper notation is necessary. To this end, consider an indefinite sequence of positive integers, $\{p_j\}$, each $p_j > 1$. Define arrays r_j and $q_{n,j}$ by

$$r_{i+1} = 1$$

$$r_{i+1} = p_i r_i, j \ge 1$$

so that

$$r_{j} = \prod_{k=1}^{j-1} p_{k}, j \ge 2$$

$$q_{n,n} = 1, n \ge 0$$

$$q_{n,j} = p_{j+1}q_{n,j+1}, 0 \le j < n$$

so that

$$q_{n, j} = \prod_{k=j+1}^{n} p_k, n \ge 1$$

Further, define

$$M_n = \prod_{k=1}^n p_k, \, n \ge 1$$

so that

$$\boldsymbol{r}_{n+1} = \boldsymbol{q}_{n,\,0} = \boldsymbol{M}_n$$
 , $n \geq 1$

Now note that

$$\begin{aligned} q_{n+1, j} &= p_{n+1} q_n, j, 0 \le j \le n \\ p_j q_{n, j} r_j &= M_n, 1 \le j \le n \end{aligned}$$

Then the procedure outlined above yields, by induction,

The Complete Factorization Theorem (CFT) Let

$$B_{M_{1}} = B_{p_{1}} = I_{p_{1}}$$

$$B_{M_{n+1}} = T_{p_{n+1}, I_{n+1}} (I_{p_{n+1}} \otimes B_{M_{n}}), n \ge 1$$
(14)

then

$$\mathcal{W}_{\mathsf{M}_{\mathsf{n}}} = \mathsf{B}_{\mathsf{M}_{\mathsf{n}}} \prod_{\mathsf{j}=1}^{\mathsf{n}} \big(\mathsf{I}_{\mathsf{q}_{\mathsf{n},\mathsf{j}}} \otimes \Delta_{\mathsf{p}_{\mathsf{j}},\mathsf{r}_{\mathsf{j}}} \big) \big(\mathsf{I}_{\mathsf{q}_{\mathsf{n},\mathsf{j}}} \otimes \mathcal{W}_{\mathsf{p}_{\mathsf{j}}} \otimes \mathsf{I}_{\mathsf{r}_{\mathsf{j}}} \big) \tag{15a}$$

alternatively,

$$\mathcal{W}_{\mathsf{M}_{\mathsf{n}}} = \prod_{j=n}^{1} \left(I_{\mathsf{q}_{\mathsf{n},j}} \otimes \mathcal{W}_{\mathsf{p}_{j}} \otimes I_{\mathsf{r}_{j}} \right) \left(I_{\mathsf{q}_{\mathsf{n},j}} \otimes \Delta_{\mathsf{p}_{j}, \, \mathsf{r}_{j}} \right) C_{\mathsf{M}_{\mathsf{n}}} \tag{15b}$$

where

$$C_{M_1} = C_{p_1} = I_{p_1}$$

$$C_{M_{n+1}} = (I_{p_{n+1}} \otimes C_{M_n}) T_{r_{n+1}, p_{n+1}}, n \ge 1$$

Notice that B_{M_n} is a permutation matrix of order M_n . Also notice that the notation is not complete because the matrix B_{M_n} depends on the **ordered** set of integers $p_1, p_2, ..., p_n$, and not simply on their product M_n . In fact, by induction,

$$B_{\mathbf{M}_{\mathbf{n}}} = \prod_{j=n}^{1} \left(\mathbf{I}_{\mathbf{q}_{\mathbf{p},j}} \otimes \mathbf{T}_{\mathbf{p}_{\mathbf{p}}, \mathbf{r}_{j}} \right) \tag{16}$$

Since matrix multiplication is not commutative, any permutation of the order of the p_j would change the matrix B_{M_n} . Also, the arrays r and q are defined with respect to the **ordered** sequence of p's.

Proof: First we prove (15a). First note that $r_1 = q_{n, n} = 1$, $\Delta_{p_1, r_1} = I_p$ so that the theorem is true for n = 1, 2, 3. Assume it is true for n. Then

$$\mathcal{W}_{M_{n+1}} = \mathcal{W}_{p_{n+1}M_n} = \mathcal{W}_{p_{n+1}r_{n+1}}$$

Setting $p = p_{n+1}$, $p = r_{n+1} = M_n$ in the fundamental factor theorem gives

$$\mathcal{W}_{M_{n+1}} = T_{p_{n+1}, f_{n+1}} (I_{p_{n+1}} \otimes \mathcal{W}_{M_n}) \Delta_{p_{n+1}f_{n+1}} (\mathcal{W}_{p_{n+1}} \otimes I_{f_{n+1}})$$

By the induction hypothesis, W_{M_n} is given by the theorem. Hence, by the distributive law,

$$I_{p_{n+1}} \otimes \mathcal{W}_{M_n} = (I_{p_{n+1}} \otimes B_{M_n}) \prod_{j=1}^{n} (I_{p_{n+1} q_{n,j}} \otimes \Delta_{p_j, r_j}) (I_{p_{n+1} q_{n,j}} \otimes \mathcal{W}_{p_j} \otimes I_{r_j})$$

But $p_{n+1}q_{n,j} = q_{n+1,j}$. Hence,

$$\mathcal{W}_{M_{n+1}} = T_{p_{n+1}, \, r_{n+1}} (I_{p_{n+1}} \otimes B_{M_n}) \prod_{j=1}^{n} \big(I_{q_{n+1,j}} \otimes \Delta_{p_j, \, r_j} \big) \big(I_{q_{n+1,j}} \otimes \, \mathcal{W}_{p_j} \otimes I_{r_j} \big) \Delta_{p_{n+1}, \, r_{n+1}} \big(\, \mathcal{W}_{p_{n+1}} \otimes I_{r_{n+1}} \big)$$

The last two factors are simply those for j = n + 1. Therefore, the (15a) is proved. Equation (15b) can be proved similarly.

Computing the DFT using the CFT

To compute the DFT of a vector, two basic matrix-vector algorithms are needed.

$$I. \vec{b} = (I_q \otimes \mathcal{W}_p \otimes I_r) \vec{a}$$

II.
$$\vec{c} = (I_q \otimes \Delta_{p, r}) \vec{b}$$

The definitions yield:

$$\vec{b}_{kpr+lr+m} = \sum_{\alpha=0}^{p-1} \omega_p^{l\alpha} a_{kpr+\alpha r+m}$$

$$\vec{c}_{kpr + lr + m} = \omega_{pr}^{lm} \vec{b}_{kpr + lr + m}$$

for $0 \le k < q$, $0 \le l < p$, $0 \le m < r$, and, for any β , $\omega_B = \exp(2\pi i / \beta)$. In the binary case,

$$\begin{array}{ll} I\colon & b_{\alpha} = a_{\alpha} + a_{\alpha+r} \\ & b_{\beta} = a_{\beta-r} - a_{\beta} \end{array} \qquad \qquad \begin{array}{ll} \text{for } a = 2kr + m, \, b = (2k+1)r + m \\ \text{II:} & c_{\alpha} = b_{\alpha} \\ & c_{\beta} = \omega_{pr}^{m} \, b_{\beta} \end{array}$$

Of course, there remains the the final multiplication of a vector by the permutation matrix B_{M_n} . This requires an expression for its column function. To obtain this, we briefly digress into number theory.

Theorem If j is any integer in the range $0 \le j < M_n$ then j has the unique representations

$$j = \sum_{k=1}^{n} \alpha_k r_k$$

(2)
$$j = \sum_{k=1}^{n} \beta_{k} q_{n, k}$$

with $0 \le \alpha_k$, $\beta_k < p_k$ where the a and β are given by:

(1)
$$R_1 = \left[\frac{j}{p_1}\right], \alpha_1 = j - p_1 R_1$$
 for $k = 2, \dots, n$, $R_k = \left[\frac{R_{k-1}}{p_k}\right], \alpha_k = R_{k-1} - p_k R_k$

(2)
$$Q_{1} = \left[\frac{j}{p_{n}}\right], \beta_{n} = j - p_{n}Q_{n}$$

$$\begin{cases} \text{for } k = n, \dots, 2, \ Q_{k-1} = \left[\frac{Q_{k}}{p_{k-1}}\right], \beta_{k-1} = Q_{k} - p_{k-1}Q_{k-1} \end{cases}$$

where [x] is the greatest integer less than or equal to x.

Proof: of (1). The definition of [x] gives

$$\frac{j}{p_1} - 1 < R_k \le \frac{j}{p_1}$$

giving

$$0 \le \alpha_1 < p_1$$

and

$$\frac{R_{k-1}}{p_k} - 1 < R_k \le \frac{R_{k-1}}{p_k}$$

gives

$$0 \le \alpha_k < p_k$$

Next,

$$R_1 \le \frac{j}{p_1} < \frac{M_n}{p_1} = q_{n, 1}$$

and, by induction, $R_k < q_{n, k}$ Hence, $R_n < 1$, that is, $R_n = 0$. Thus,

$$\sum_{k=1}^{n} \alpha_k r_k = \alpha_1 r_1 + \sum_{k=2}^{n} \alpha_k r_k$$

$$= j - p_1 R_1 + \sum_{k=2}^{n} (R_{k-1} - p_k R_k) r_k$$

$$= j - p_1 R_1 + \sum_{k=2}^{n} (R_{k-1} r_k - R_k r_{k+1})$$

$$= j - p_1 R_1 + R_1 r_2 - R_n r_{n+1}$$

but
$$r_1 = p_1$$
, $R_n = 0$, so $\sum_{k=1}^{n} \alpha_k r_k = j$.

(2) is proved similarly.

The Scrambling Matrix The column function for the scrambling matrix, B_{M_n} , can now be

given. For $0 \le j < M_n$, we have $j = \sum_{k=1}^n \, \beta_k q_{n, k}.$ The column function is

$$C_{B_{M_n}}(j) = \sum_{k=1}^n \beta_k r_k$$

This is easily proved by induction on n.

For n = 1, k = 1, $q_{1, 1} = r_1 = 1$ and $j = \beta_1 = c(j)$ which is correct since $B_{p_1} = I_{p_1}$, the identity matrix.

Assume the result is true for n. By the recursive definition,

$$B_{M_{n+1}} = T_{p_{n+1}, \, r_{n+1}} (I_{p_{n+1}} \otimes B_{M_n})$$

and for $0 \le j < M_{n+1}$ use the representation $j = \sum_{k=1}^{n+1} \beta_k q_{n+1,k}$. Since $q_{n+1,k} = p_{n+1} q_{n,k}$ for

 $1 \le k \le n$, we can write $j = \sum_{k=1}^{n} \beta_k q_{n, k} p_{n+1} + \beta_{n+1}$, or $j = l p_{n+1} + m$ with

$$1 = \sum_{k=1}^{n} \beta_{k} q_{n, k}, m = \beta_{n+1}$$

so that

$$0 \le 1 < M_n = r_{m+1}, 0 \le m < p_{n+1}.$$

By definition of the perfect shuffle,

$$1 = \sum_{k=1}^{n} \beta_{k} q_{n, k}, m = \beta_{n+1}$$

Let $D = I_{p_{n+1}} \otimes B_{M_n}.$ The column function for D is

$$c_{D}(1 + mM_{n}) = mM_{n} + c_{B_{n}}(1) = \beta_{n+1} r_{n+1} + c_{B_{n}}(1).$$

By the inductive hypothesis

$$c_{B_n}(1) = \sum_{k=1}^n \beta_k r_k$$

Hence,

$$c_{B_{n+1}}(j) = c_{D}[c_{T}(j)] = \beta_{n+1}r_{n+1} + \sum_{k=1}^{n} \beta_{k}r_{k} = \sum_{k=1}^{n+1} \beta_{k}r_{k}$$

Calculation of $c_{B_{M_n}}(j)$ is easily implemented. If

$$j = \sum_{l=1}^{n} \beta_l q_{n, l}$$

then

$$k \equiv c(j) = \sum_{l=1}^{n} \beta_{l} r_{l}$$

Define

$$k_m = \sum_{l=1}^{m} \beta_l r_l$$
, $1 \le m \le n$, $k_0 = 0$

then

$$k_{m} = k_{m-1} + \beta_{m} r_{m}, 1 \le m \le n, k_{n} = k$$

Set j = 0

Alternatively, define $J_n = \sum_{l=m}^n \beta_l q_{l,n}$, $i \le m \le n$, so that $J_1 = j$ and $J_m = J_{m+1} + \beta_m q_{n,n}$, $1 \le m < n$.

Thus, set K = 0

$$\begin{array}{l} \text{for } j_n := 0 \text{ to } q_{n,\; n-1} - 1 \text{ by } q_{n,\; n} \text{ do} \\ \text{for } j_{n-1} := j_n \text{ to } q_{n,\; n-2} - 1 \text{ by } q_{n,\; n-1} \text{ do} \\ \vdots \\ \text{for } j_m := j_{m+1} \text{ to } q_{n,\; m-1} - 1 \text{ by } q_{n,\; m} \text{ do} \\ \vdots \\ \text{for } j_1 := j_2 \text{ to } M - 1 \text{ by } q_{n,\; 1} \text{ do} \\ \text{begin} \\ c(j_1) := k; \\ k := k+1; \\ \text{end}; \end{array}$$

For the binary case, $p_j = 2$, $r_j = 2^{j-1}$, $q_{n,j} = 2^{n-j}$. So if $j = \sum_{k=1}^{n} \beta_k 2^{n-k}$, $0 \le \beta_k \le 1$, then

$$c_B(j) = \sum_{k=1}^{n} \beta_k 2^{k-1}, 0 \le \beta_k \le 1$$

is the column function, which can be written

$$c_B(j) = \sum_{k=1}^n \beta_{n+1-k} 2^{n-k}$$

so B is symmetric and hence $B^{-1} = B$. Thus when j is represented as an n-bit binary number, c(j) is obtained from j by simply reversing the order of its bits. For this reason, B is also called the **bit-reversal** matrix. In this case, a simpler notation can be used for the perfect shuffle and square deal matrices. Define

$$T_{2^n} \equiv T_{2, 2^{n-1}}$$

 $S_{2^n} \equiv S_{2, 2^{n-1}}$ $n \ge 1, T_1 = S_1 = I_1$

Then with $j = \sum_{k=1}^{n} \beta_k 2^{n-k}$, the column functions for S and T are

$$c_{T}(j) = \sum_{k=1}^{n} \gamma_{k} 2^{n-k}$$

$$c_{S}(j) = \sum_{k=1}^{n} \sigma_{k} 2^{n+k}$$

where $\gamma_1=\beta_n$, $\gamma_k=\beta_{k-1}$, k>1, $\gamma_1=\beta_n$, $\sigma_n=\beta_1$, $\sigma_k=\beta_{k+1}$, k< n. Thus writing j in binary notation: $j=\beta_1\beta_2...\beta_{n-1}\beta_n$ gives

$$c_T(j) = \beta_n \beta_1 \beta_2 \cdots \beta_{n-1}$$

$$c_{S}(j) = \beta_{2}\beta_{3} \cdot \cdot \cdot \beta_{n}\beta_{1}$$

So $c_T(j)$ = is obtained by shifting the bits in j one place to the right, end around, and $c_S(j)$ is

obtained by shifting the bits one place to the left, end around.

We now introduce one more permutation matrix, the exchange matrix X_{2^n} , defined by its column function $c_x(j)$:

$$j = \sum_{k=1}^{n} \beta_k 2^{n-k}$$

$$c_X(j) = \sum_{k=1}^{n} \chi_k 2^{n-k}, \chi_1 = \beta_n, \chi_n = \beta_1, \chi_k = \beta_k, 1 < k < n$$

Thus, the action of X is to exchange the first and last bits of j. So

$$c_{X}(2k+1) = 2k + \frac{M}{2}$$

$$c_{X}(2k + \frac{M}{2}) = (2k+1) \int_{0}^{\infty} 0 \le k \le \frac{M}{4}$$

$$c_{X}(j) = j \text{ for all other values of } j$$
(19)

Thus, if

$$\vec{b} = X \vec{a}$$

then \vec{b} is obtained from \vec{a} simply by moving the appropriate element $\pm (M/2 - 1)$. Note that X is symmetric. In other words, the odd numbered components in the first half of \vec{a} are exchanged with the even components of the second half of \vec{a} . Directly from the definitions we have

$$B_2 = X_2 = T_2 = S_2 = I$$

 $B_4 = X_4 = T_4 = S_4$
 $B_8 = X_8$

$$T_{2^{n}} = X_{2^{n}} (I_{2} \otimes T_{2^{n-1}}) = (T_{2^{n-1}} \otimes I_{2}) X_{2^{n}}$$

$$S_{2^{n}} = (I_{2} \otimes S_{2^{n-1}}) X_{2^{n}} = X_{2^{n}} (S_{2^{n-1}} \otimes I_{2})$$

$$B_{2^{n}} = T_{2^{n}} (I_{2} \otimes B_{2^{n-1}}) = (B_{2^{n-1}} \otimes I_{2}) T_{2^{n}}$$

$$B_{2^{n}} = T_{2^{n}} (I_{2} \otimes B_{2^{n-1}}) = (B_{2^{n-1}} \otimes I_{2}) T_{2^{n}}$$

$$= (I_{2} \otimes B_{2^{n-1}}) S_{2^{n}} = S_{2^{n}} (B_{2^{n-1}} \otimes I_{2})$$

$$= (I_{2} \otimes B_{2^{n-2}} \otimes I_{2}) X_{2^{n}} = X_{2^{n}} (I_{2} \otimes B_{2^{n-2}} \otimes I_{2})$$

$$(20)$$

By induction, these yield,

$$T_{2^{n}} = \prod_{j=0}^{n-2} (I_{2^{j}} \otimes X_{2^{j-n}}) = \prod_{j=n-2}^{0} (X_{2^{n-j}} \otimes I_{2^{j}})$$

$$S_{2^{n}} = \prod_{j=0}^{n-2} (X_{2^{n-j}} \otimes I_{2^{j}}) = \prod_{j=n-2}^{0} (I_{2^{j}} \otimes X_{2^{n-j}})$$

$$(21)$$

and

$$B_{2^{n}} = \prod_{j=0}^{\left[\frac{n}{2}\right]-1} (I_{2^{j}} \otimes X_{2^{n-2j}} \otimes I_{2^{j}})$$

$$= \prod_{j=\left[\frac{n}{2}\right]-1}^{0} (I_{2^{j}} \otimes X_{2^{n-2j}} \otimes I_{2^{j}}), n \ge 2$$
(22)

So scrambling, shuffling, and dealing can be done purely in terms of exchange matrices.

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