



Fast Fourier Transforms

Book by: C. Sidney Burrus

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DIF Radix-2 FFT Algorithm

Below is the Fortran code for a Decimation-in-Frequency, Radix-2, three butterfly Cooley-Tukey FFT followed by a bit-reversing unscrambler.

```
C  A COOLEY-TUKEY RADIX 2, DIF  FFT PROGRAM
C  THREE-BF, MULT BY 1 AND J ARE REMOVED
C  COMPLEX INPUT DATA IN ARRAYS X AND Y
C  TABLE LOOK-UP OF W VALUES
C  C. S. BURRUS, RICE UNIVERSITY, SEPT 1983
```

```
C-----
```

```
      SUBROUTINE FFT (X,Y,N,M,WR,WI)
```

```
      REAL X(1), Y(1), WR(1), WI(1)
```

```
C-----MAIN FFT LOOPS-----
```

```
C
```

```
      N2 = N
```

```
      DO 10 K = 1, M
```

```
          N1 = N2
```

```
          N2 = N2/2
```

```
          JT = N2/2 + 1
```

```
          DO 1 I = 1, N, N1
```

```
              L = I + N2
```

```
              T = X(I) - X(L)
```

```
              X(I) = X(I) + X(L)
```

```
              X(L) = T
```

```
              T = Y(I) - Y(L)
```

```
              Y(I) = Y(I) + Y(L)
```

```
              Y(L) = T
```

```
          1      CONTINUE
```

```
          IF (K.EQ.M) GOTO 10
```

```
          IE = N/N1
```

```
          IA = 1
```

```
          DO 20 J = 2, N2
```

```
              IA = IA + IE
```

```
          IF (J.EQ.JT) GOTO 50
```

```
          C = WR(IA)
```

```
          S = WI(IA)
```

```
          DO 30 I = J, N, N1
```

```
              L = I + N2
```

```
              T = X(I) - X(L)
```

```
              X(I) = X(I) + X(L)
```

```
              TY = Y(I) - Y(L)
```

```
              Y(I) = Y(I) + Y(L)
```

```
              X(L) = C*T + S*TY
```

This page is in 2 books:

- **Fast Fourier Transforms**

- **Authors:** C. Sidney Burrus
- **Revised:** Nov 18, 2012
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- **Fast Fourier Transforms (6x9 Version)**

- **Authors:** C. Sidney Burrus
- **Revised:** Aug 16, 2012
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principle, size 1). This might naturally suggest a recursive implementation in which the tree is traversed “depth-first” as in [Figure](#)(right) and the algorithm of [Pre](#)—one size $n/2$ transform is solved completely before processing the other one, and so on. However, most traditional FFT implementations are non-recursive (with rare exceptions [\[link\]](#)) and traverse the tree “breadth-first” [\[link\]](#) as in [Figure](#)(left)—in the radix-2 example, they would perform n (trivial) size-1 transforms, then $n/2$ combinations into size-2 transforms, then $n/4$ combinations into size-4 transforms, and so on, thus making $\log_2 n$ passes over the whole array. In contrast, as we discuss in “[Discussion](#)”, FFTW employs an explicitly recursive strategy that encompasses **both** depth-first and breadth-first styles, favoring the former since it has some theoretical and practical advantages as discussed in “[FFTs and the Memory Hierarchy](#)”.

```

Y[0, ...,  $n - 1$ ]  $\leftarrow$  recfft2( $n$ , X,  $\iota$ ):
IF  $n=1$  THEN
    Y[0]  $\leftarrow$  X[0]
ELSE
    Y[0, ...,  $n/2 - 1$ ]  $\leftarrow$  recfft2( $n/2$ , X,  $2\iota$ )
    Y[ $n/2$ , ...,  $n - 1$ ]  $\leftarrow$  recfft2( $n/2$ , X +  $\iota$ ,  $2\iota$ )
    FOR  $k_1 = 0$  TO  $(n/2) - 1$  DO
         $t \leftarrow$  Y[ $k_1$ ]
        Y[ $k_1$ ]  $\leftarrow$   $t + \omega_n^{k_1}$  Y[ $k_1 + n/2$ ]
        Y[ $k_1 + n/2$ ]  $\leftarrow$   $t - \omega_n^{k_1}$  Y[ $k_1 + n/2$ ]
    END FOR
END IF

```

NOT_CONVERTED_YET: caption

A depth-first recursive radix-2 DIT Cooley-Tukey FFT to compute a DFT of a power-of-two size $n = 2^m$. The input is an array **X** of length n with stride ι (i.e., the inputs are **X**[$\ell\iota$] for $\ell = 0, \dots, n-1$) and the output is an array **Y** of length n (with stride 1), containing the DFT of **X** [Equation 1]. **X** + ι denotes the array beginning with **X**[ι]. This algorithm operates out-of-place, produces in-order output, and does not require a separate bit-reversal stage.

A second ordering distinction lies in how the digit-reversal is performed. The classic approach is a single, separate digit-reversal pass following or preceding the arithmetic computations; this approach is so common and so deeply embedded into FFT lore that many practitioners find it difficult to imagine an FFT without an explicit bit-reversal stage. Although this pass requires only $O(n)$ time [\[link\]](#), it can still be non-negligible, especially if the data is out-of-cache; moreover, it neglects the possibility that data reordering during the transform may improve memory locality. Perhaps the oldest alternative is the Stockham **auto-sort** FFT [\[link\]](#), [\[link\]](#), which transforms back and forth between two arrays with each butterfly, transposing one digit each time, and was popular to improve contiguity of access for vector computers [\[link\]](#). Alternatively, an explicitly recursive style, as in FFTW, performs the digit-reversal implicitly at the “leaves” of its computation when operating out-of-place (see section “[Discussion](#)”). A simple example of this style, which computes in-order output using an out-of-place radix-2 FFT without explicit bit-reversal, is shown in the algorithm of [Pre](#) [corresponding to [Figure](#)(right)]. To operate in-place with $O(1)$ scratch storage, one can interleave small matrix transpositions with the butterflies [\[link\]](#), [\[link\]](#), [\[link\]](#), [\[link\]](#), and a related strategy in FFTW [\[link\]](#) is briefly described by

$N = 2^p 3^q 5^r$. Set $NI = 2^p$, $IP = p$. We first compute the required rotation and set up the table of twiddle factors:

```

COMPLEX TRIGS(NI)
DEL = 4.0*ASIN(1.0)/FLOAT(NI)
IROT = MOD((N/NI),NI)
KK = 0
DO 10 K = 1, NI
  ANGLE = FLOAT(KK)*DEL
  TRIGS(K) = CMPLX(COS(ANGLE),SIN(ANGLE))
  KK = KK + IROT
  IF (KK.GT.NI) KK = KK - NI
10 CONTINUE

```

The first $(p+1)/2$ radix-2 passes are then performed by the following code:

```

COMPLEX X(N), W, Z
NH = N/2
INC = N/NI
DO 50 L = 1, (IP+1)/2
  LA = 2**(L-1)
  JA = 0
  JB = NH/LA
  KK = 1
  DO 40 K = 0, JB-1, INC
    W = TRIGS(KK)
    DO 30 J = K+1, N, N/LA
      IA = JA + J
      IB = JB + J
      DO 20 I = 1, INC
        Z = W*(X(IA) - X(IB))
        X(IA) = X(IA) + X(IB)
        X(IB) = Z
        IA = IA + NI
        IF (IA.GT.N) IA = IA - N
        IB = IB + NI
        IF (IB.GT.N) IB = IB - N
      20 CONTINUE
    30 CONTINUE
    KK = KK + LA
  40 CONTINUE
50 CONTINUE

```

The details of the indexing may be understood by comparing this code with Table 1 ($N = 40$, $NI = 8$). The three outer loops are very similar to the three loops of the code presented in [18], which performed the first half of a self-sorting in-place radix-2 algorithm. In the present case these loops set up base addresses in the first column of Table 1. The advantage of the Puritanian map is that the entries in the first column

A GENERALIZED PRIME FACTOR FFT ALGORITHM FOR ANY $N = 2^p 3^q 5^r$

CLIVE TEMPERTON†

Abstract. Prime factor fast Fourier transform (FFT) algorithms have two important advantages: they can be simultaneously self-sorting and in-place, and they have a lower operation count than conventional FFT algorithms. The major disadvantage of the prime factor FFT has been that it was only applicable to a limited set of values of the transform length N . This paper presents a generalized prime factor FFT, which is applicable for any $N = 2^p 3^q 5^r$, while maintaining both the self-sorting in-place capability and the lower operation count. Timing experiments on the Cray Y-MP demonstrate the advantages of the new algorithm.

Key words. fast Fourier transform (FFT), prime factor algorithm (PFA), self-sorting FFT, in-place FFT

AMS(MOS) subject classification. 65T05

1. Introduction. Fast Fourier transform (FFT) algorithms can be defined whenever the transform length N can be factorized as $N = N_1 N_2 \cdots N_k$, where the factors N_i are integers. Though there are many variants of these algorithms, they fall into two basic categories: those based on the prime factor algorithm (PFA) of Good [5], which are only applicable if the factors N_i are mutually prime, and those descended from the algorithm of Cooley and Tukey [3], for which there is no such restriction (indeed the most familiar case is $N_i = 2$ for all i).

The prime factor algorithms have two important advantages. For a given value of N , the operation count is lower than that for the corresponding Cooley-Tukey algorithm. Moreover, the PFA can be made both self-sorting (input and output both

Fig. 3. Minimization of number of operations.

COOLEY *et al.*: FAST FOURIER TRANSFORM

```

SUBROUTINE FFT(A,M)
COMPLEX A(1024),U,W,T
N= 2**M
NV2 = N/2
NM1 = N-1
J=1
DO 7 I=1,NM1
  IF(I.GE.J) GO TO 5
  T = A(J)
  A(J) = A(I)
  A(I) = T
5 K=NV2
6 IF(K.GE.J) GO TO 7
  J = J-K
  K=K/2
  GO TO 6
7 J= J+K
  PI = 3.14159265358979
  DO 20 L=1,M
    LE = 2**L
    LE1 = LE/2
    U = (1.0,0.)
    W=CMPLX(COS(PI/LE1),SIN(PI/LE1))
    DO 10 J=1,LE1
      DO 10 I=J,N,LE
        IP = I+LE1
        T=A(IP)*U
        A(IP)=A(I)-T
10 A(I)=A(I)+T
20 U=U*W
    RETURN
  END

```

optimally programmed FORTRAN.

Fig. 6 shows a slightly different

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algorithm. This one requires the indices j and i to be mutually prime and uses a different set of one-dimensional indices j and i for the (j_1, j_0) and (n_1, n_0) , respectively. The result is a procedure quite different from the one above, except that the phase factor is calculated in the calculation of $A_1(j_0, n_0)$. This calculation is in combination with the previous one, in ample, to introduce an odd factor. The result is a power of two. Then one can use a factor algorithm using the power of two to calculate the r -point subseries.

The time required for computation of the fast Fourier transform is compared with a conventional program and by the use of the fast method illustrated in Fig. 7. It is seen how to arrive at values of N which are not feasible by conventional methods. The time required for the fast methods shows that the advances permit the measurement of the increasing rates, the demands

The Fast Fourier Transform and Its Applications

JAMES W. COOLEY, PETER A. W. LEWIS, AND PETER D. WELCH

Abstract—The advent of the fast Fourier transform method has greatly extended our ability to implement Fourier methods on digital computers. A description of the algorithm and its programming is given here and followed by a theorem relating its operands, the finite sample sequences, to the continuous functions they often are intended to approximate. An analysis of the error due to discrete sampling over finite ranges is given in terms of aliasing. Procedures for

It is most likely that with the relatively small values of N used in preelectronic computer days, the former methods were easier to use and took fewer operations. Consequently, the methods requiring $N \log N$ operations were neglected. With the arrival of electronic computers capable of doing calculations of Fourier transforms with


```
procedure FFT (A, n, w)
```

```
  # Preconditions:
```

```
  #   A is a Vector of length n;
```

```
  #   n is a power of 2;
```

```
  #   w is a primitive n-th root of unity.
```

```
  #
```

```
  # The Vector A represents the polynomial
```

```
  #    $a(z) = A[1] + A[2]*z + \dots + A[n]*z^{(n-1)}$  .
```

```
  #
```

```
  # The value returned is a Vector of the values
```

```
  #   [ a(1), a(w), a(w^2), ... , a(w^(n-1)) ]
```

```
  # computed via a recursive FFT algorithm.
```

```
  if n = 1 then
```

```
    return A
```

```
  else
```

```
    Aeven <-- Vector( [A[1], A[3], ..., A[n-1]] )
```

```
    Aodd  <-- Vector( [A[2], A[4], ..., A[n]] )
```

```
    Veven <-- FFT( Aeven, n/2, w^2 )
```

```
    Vodd  <-- FFT( Aodd, n/2, w^2 )
```

```
    V <-- Vector(n)  # Define a Vector of length n
```

```
    for i from 1 to n/2 do
```

```
      V[i] <-- Veven[i] + w^(i-1)*Vodd[i]
```

```
      V[n/2 + i] <-- Veven[i] - w^(i-1)*Vodd[i]
```

```
    end do
```

```
    return V
```

```
  end if
```

```
end procedure
```

Vector radix fast Fourier transform

4 Author(s)

D. Harris ; J. McClellan ; D. Chan ; H. Schuessler [View All Authors](#)

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Abstract

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Abstract:

A new radix-2 two-dimensional direct FFT developed by Rivard is generalized in this paper to include arbitrary radices and non-square arrays. It is shown that the radix-4 version of this algorithm may require significantly fewer computations than conventional row-column transform methods. Also, the new algorithm eliminates the matrix transpose operation normally required when the array must reside on a bulk storage device. It requires the same number of passes over the array on bulk storage as efficient matrix transpose routines, but produces the transform in bit-reversed order. An additional pass over the data is necessary to sort the array if normal ordering is desired.

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Authors



References



Citations



Keywords



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[Proceedings] ICASSP 91: 1991 International Conference on Acoustics, Speech, and Signal Processing
Published: 1991

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in each step is proportional to n .

Code 1.1 (recursive radix 2 DIT FFT) *Pseudo code for a recursive procedure of the (radix 2) DIT FFT algorithm, is must be +1 (forward transform) or -1 (backward transform):*

```
procedure rec_fft_dit2(a[], n, x[], is)
// complex a[0..n-1] input
// complex x[0..n-1] result
{
    complex b[0..n/2-1], c[0..n/2-1]    // workspace
    complex s[0..n/2-1], t[0..n/2-1]    // workspace
    if n == 1 then // end of recursion
    {
        x[0] := a[0]
        return
    }
    nh := n/2
    for k:=0 to nh-1 // copy to workspace
    {
        s[k] := a[2*k]    // even indexed elements
        t[k] := a[2*k+1]  // odd indexed elements
    }

    // recursion: call two half-length FFTs:
    rec_fft_dit2(s[],nh,b[],is)
    rec_fft_dit2(t[],nh,c[],is)
    fourier_shift(c[],nh,is*1/2)
    for k:=0 to nh-1 // copy back from workspace
    {
        x[k]      := b[k] + c[k];
        x[k+nh]   := b[k] - c[k];
    }
}
```

[source file: recfftdit2.spr]

The data length n must be a power of 2. The result is in $x[]$. Note that normalization (i.e. multiplication of each element of $x[]$ by $1/\sqrt{n}$) is not included here.

[FXT: recursive_dit2_fft in slow/recfft2.cc] The procedure uses the subroutine

Code 1.2 (Fourier shift) For each element in $c[0..n-1]$ replace $c[k]$ by $c[k]$ times $e^{v2\pi i k/n}$. Used with $v = \pm 1/2$ for the Fourier transform.

```
procedure fourier_shift(c[], n, v)
{
  for k:=0 to n-1
  {
    c[k] := c[k] * exp(v*2.0*PI*I*k/n)
  }
}
```

cf. [FXT: fourier_shift in fft/fouriershift.cc]

The recursive FFT-procedure involves $n \log_2(n)$ function calls, which can be avoided by rewriting it in a non-recursive way. One can even do all operations *in place*, no temporary workspace is needed at all. The price is the necessity of an additional data reordering: The procedure `revbin_permute(a[], n)` rearranges the array $a[]$ in a way that each element a_x is swapped with $a_{\bar{x}}$, where \bar{x} is obtained from x by reversing its binary digits. This is discussed in section 8.1.

Code 1.3 (radix 2 DIT FFT, localized) Pseudo code for a non-recursive procedure of the (radix 2) DIT algorithm, is must be -1 or +1:

```
procedure fft_dit2_localized(a[], ldn, is)
// complex a[0..2*ldn-1] input, result
{
  n := 2*ldn // length of a[] is a power of 2
  revbin_permute(a[], n)
  for ldm:=1 to ldn // log_2(n) iterations
  {
    m := 2*ldm
    mh := m/2
    for r:=0 to n-m step m // n/m iterations
    {
      for j:=0 to mh-1 // m/2 iterations
      {
        e := exp(is*2*PI*I*j/m) // log_2(n)*n/m*m/2 = log_2(n)*n/2 computations
        u := a[r+j]
        v := a[r+j+mh] * e
        a[r+j] := u + v
        a[r+j+mh] := u - v
      }
    }
  }
}
```

[source file: fftdit2localized.spr]

[FXT: dit2_fft_localized in fft/fftdit2.cc]

This version of a non-recursive FFT procedure already avoids the calling overhead and it works in place. It works as given, but is a bit wasteful. The (expensive!) computation $e := \exp(is*2\pi i*j/m)$ is done $n/2 \cdot \log_2(n)$ times. To reduce the number of trigonometric computations, one can simply swap the two inner loops, leading to the first ‘real world’ FFT procedure presented here:

Code 1.4 (radix 2 DIT FFT) Pseudo code for a non-recursive procedure of the (radix 2) DIT algorithm, is must be -1 or +1:

```
procedure fft_dit2(a[], ldn, is)
// complex a[0..2*ldn-1] input, result
```



```

{
  n := 2**ldn
  revbin_permute(a[],n)
  for ldm:=1 to ldn // log_2(n) iterations
  {
    m := 2**ldm
    mh := m/2
    for j:=0 to mh-1 // m/2 iterations
    {
      e := exp(is*2*PI*I*j/m) // 1 + 2 + ... + n/8 + n/4 + n/2 = n-1 computations
      for r:=0 to n-m step m
      {
        u := a[r+j]
        v := a[r+j+mh] * e
        a[r+j] := u + v
        a[r+j+mh] := u - v
      }
    }
  }
}

```

[source file: `fftdit2.spr`]

[FXT: `dit2_fft` in `fft/fftdit2.cc`]

Swapping the two inner loops reduces the number of trigonometric (`exp()`) computations to n but leads to a feature that many FFT implementations share: Memory access is highly nonlocal. For each recursion stage (value of ldm) the array is traversed mh times with n/m accesses in strides of mh . As mh is a power of 2 this can (on computers that use memory cache) have a very negative performance impact for large values of n . On a computer where the CPU clock (366MHz, AMD K6/2) is 5.5 times faster than the memory clock (66MHz, EDO-RAM) I found that indeed for small n the localized FFT is slower by a factor of about 0.66, but for large n the same ratio is in favour of the ‘naive’ procedure!

It is a good idea to extract the $ldm==1$ stage of the outermost loop, this avoids complex multiplications with the trivial factors $1 + 0i$: Replace

```

for ldm:=1 to ldn
{
  for r:=0 to n-1 step 2
  {
    {a[r], a[r+1]} := {a[r]+a[r+1], a[r]-a[r+1]}
  }
  for ldm:=2 to ldn
  {

```

$z^{(2j+\delta)n/2} = e^{\pm i\pi\delta}$ is equal to plus/minus 1 for $\delta = 0/1$ (k even/odd), respectively.

The last two equations are, more compactly written, the

Idea 1.2 (radix 2 DIF step) *Radix 2 decimation in frequency step for the FFT:*

$$\mathcal{F}[a]^{(even)} \stackrel{n/2}{=} \mathcal{F}\left[a^{(left)} + a^{(right)}\right] \quad (1.36)$$

$$\mathcal{F}[a]^{(odd)} \stackrel{n/2}{=} \mathcal{F}\left[S^{1/2}\left(a^{(left)} - a^{(right)}\right)\right] \quad (1.37)$$

Code 1.5 (recursive radix 2 DIF FFT) *Pseudo code for a recursive procedure of the (radix 2) decimation in frequency FFT algorithm, is must be +1 (forward transform) or -1 (backward transform):*

```
procedure rec_fft_dif2(a[], n, x[], is)
// complex a[0..n-1] input
// complex x[0..n-1] result
{
  complex b[0..n/2-1], c[0..n/2-1] // workspace
  complex s[0..n/2-1], t[0..n/2-1] // workspace
  if n == 1 then
  {
    x[0] := a[0]
    return
  }

  nh := n/2
  for k:=0 to nh-1
  {
    s[k] := a[k] // 'left' elements
    t[k] := a[k+nh] // 'right' elements
  }

  for k:=0 to nh-1
  {
    {s[k], t[k]} := {(s[k]+t[k]), (s[k]-t[k])}
  }

  fourier_shift(t[],nh,is*0.5)
  rec_fft_dif2(s[],nh,b[],is)
  rec_fft_dif2(t[],nh,c[],is)

  j := 0
  for k:=0 to nh-1
  {
    x[j] := b[k]
    x[j+1] := c[k]
    j := j+2
  }
}
```

[source file: recfft2dif2.spr]

The data length n must be a power of 2. The result is in $x[]$.

Algorithms for programmers

ideas and source code

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This document¹ was L^AT_EX'd at September 26, 2002

Algorithm 7.2 The iterative radix-2 DIT FFT algorithm in pseudo-code.

begin

$PairsInGroup := N/2$

Begin with $N/2$ butterflies in one group

$NumOfGroups := 1$

Same twiddle factor is employed in a group

$Distance := N/2$

while $NumOfGroups < N$ **do**

for $K := 0$ **to** $NumOfGroups - 1$ **do**

Combine pairs in each group

$JFirst := 2 * K * PairsInGroup$

$JLast := JFirst + PairsInGroup - 1$

$Jtwiddle := K$

Access consecutive $w[m]$

$W := w[Jtwiddle]$

Assume $w[m] = \omega_N^\ell$, m bit-reverses ℓ

for $J := JFirst$ **to** $JLast$ **do**

$Temp := W * a[J + Distance]$

$a[J + Distance] := a[J] - Temp$

$a[J] := a[J] + Temp$

end for

end for

$PairsInGroup := PairsInGroup/2$

$NumOfGroups := NumOfGroups * 2$

$Distance := Distance/2$

end while

end

with the understanding that on input, $a[i_4 i_3 i_2 i_1 i_0]$ contains $x_{i_4 i_3 i_2 i_1 i_0}$, and on output, $a[i_4 i_3 i_2 i_1 i_0]$ contains the bit-reversed $X_{i_0 i_1 i_2 i_3 i_4}$. Refer to Figure 7.6 for the decimal subscripts of all 32 bit-reversed output elements $X_{i_0 i_1 i_2 i_3 i_4}$.

COMPUTATIONAL MATHEMATICS SERIES

The background of the cover features a complex diagram of Fast Fourier Transform (FFT) butterfly networks. It consists of multiple stages of interconnected nodes, represented by small squares with circles inside, and lines with arrows indicating the flow of data. The diagram is overlaid with the title text.

INSIDE the FFT BLACK BOX

**Serial and Parallel Fast
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