

integer arithmetic modulo some large prime $N+1$, and the N th root of 1 by the modulo arithmetic equivalent. Strictly speaking, these are not *Fourier* transforms at all, but the properties are quite similar and computational speed can be far superior. On the other hand, their use is somewhat restricted to quantities like correlations and convolutions since the transform itself is not easily interpretable as a “frequency” spectrum.

CITED REFERENCES AND FURTHER READING:

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12.3 FFT of Real Functions, Sine and Cosine Transforms

It happens frequently that the data whose FFT is desired consist of real-valued samples f_j , $j = 0 \dots N - 1$. To use `four1`, we put these into a complex array with all imaginary parts set to zero. The resulting transform F_n , $n = 0 \dots N - 1$ satisfies $F_{N-n}^* = F_n$. Since this complex-valued array has real values for F_0 and $F_{N/2}$, and $(N/2) - 1$ other independent values $F_1 \dots F_{N/2-1}$, it has the same $2(N/2 - 1) + 2 = N$ “degrees of freedom” as the original, real data set. However, the use of the full complex FFT algorithm for real data is inefficient, both in execution time and in storage required. You would think that there is a better way.

There are *two* better ways. The first is “mass production”: Pack two separate real functions into the input array in such a way that their individual transforms can be separated from the result. This is implemented in the program `twofft` below. This may remind you of a one-cent sale, at which you are coerced to purchase two of an item when you only need one. However, remember that for correlations and convolutions the Fourier transforms of two functions are involved, and this is a handy way to do them both at once. The second method is to pack the real input array cleverly, without extra zeros, into a complex array of half its length. One then performs a complex FFT on this shorter length; the trick is then to get the required answer out of the result. This is done in the program `realft` below.

Transform of Two Real Functions Simultaneously

First we show how to exploit the symmetry of the transform F_n to handle two real functions at once: Since the input data f_j are real, the components of the discrete Fourier transform satisfy

$$F_{N-n} = (F_n)^* \quad (12.3.1)$$

where the asterisk denotes complex conjugation. By the same token, the discrete Fourier transform of a purely imaginary set of g_j 's has the opposite symmetry.

$$G_{N-n} = -(G_n)^* \quad (12.3.2)$$

Therefore we can take the discrete Fourier transform of two real functions each of length N simultaneously by packing the two data arrays as the real and imaginary parts, respectively, of the complex input array of `four1`. Then the resulting transform array can be unpacked into two complex arrays with the aid of the two symmetries. Routine `twofft` works out these ideas.

```

SUBROUTINE twofft(data1,data2,fft1,fft2,n)
  INTEGER n
  REAL data1(n),data2(n)
  COMPLEX fft1(n),fft2(n)
C  USES four1
    Given two real input arrays data1(1:n) and data2(1:n), this routine calls four1
    and returns two complex output arrays, fft1(1:n) and fft2(1:n), each of complex length n
    (i.e., real length 2*n), which contain the discrete Fourier transforms of the respective data
    arrays. n MUST be an integer power of 2.
  INTEGER j,n2
  COMPLEX h1,h2,c1,c2
  c1=cplx(0.5,0.0)
  c2=cplx(0.0,-0.5)
  do 11 j=1,n
    fft1(j)=cplx(data1(j),data2(j))      Pack the two real arrays into one complex
  enddo 11                                array.
  call four1(fft1,n,1)                   Transform the complex array.
  fft2(1)=cplx(aimag(fft1(1)),0.0)
  fft1(1)=cplx(real(fft1(1)),0.0)
  n2=n+2
  do 12 j=2,n/2+1
    h1=c1*(fft1(j)+conjg(fft1(n2-j)))    Use symmetries to separate the two trans-
    h2=c2*(fft1(j)-conjg(fft1(n2-j)))    forms.
    fft1(j)=h1                            Ship them out in two complex arrays.
    fft1(n2-j)=conjg(h1)
    fft2(j)=h2
    fft2(n2-j)=conjg(h2)
  enddo 12
  return
END

```

What about the reverse process? Suppose you have two complex transform arrays, each of which has the symmetry (12.3.1), so that you know that the inverses of both transforms are real functions. Can you invert both in a single FFT? This is even easier than the other direction. Use the fact that the FFT is linear and form the sum of the first transform plus i times the second. Invert using `four1` with

$\text{isign} = -1$. The real and imaginary parts of the resulting complex array are the two desired real functions.

FFT of Single Real Function

To implement the second method, which allows us to perform the FFT of a *single* real function without redundancy, we split the data set in half, thereby forming two real arrays of half the size. We can apply the program above to these two, but of course the result will not be the transform of the original data. It will be a schizophrenic combination of two transforms, each of which has half of the information we need. Fortunately, this schizophrenia is treatable. It works like this:

The right way to split the original data is to take the even-numbered f_j as one data set, and the odd-numbered f_j as the other. The beauty of this is that we can take the original real array and treat it as a complex array h_j of half the length. The first data set is the real part of this array, and the second is the imaginary part, as prescribed for `twofft`. No repacking is required. In other words $h_j = f_{2j} + if_{2j+1}$, $j = 0, \dots, N/2 - 1$. We submit this to `four1`, and it will return a complex array $H_n = F_n^e + iF_n^o$, $n = 0, \dots, N/2 - 1$ with

$$\begin{aligned} F_n^e &= \sum_{k=0}^{N/2-1} f_{2k} e^{2\pi i k n / (N/2)} \\ F_n^o &= \sum_{k=0}^{N/2-1} f_{2k+1} e^{2\pi i k n / (N/2)} \end{aligned} \quad (12.3.3)$$

The discussion of program `twofft` tells you how to separate the two transforms F_n^e and F_n^o out of H_n . How do you work them into the transform F_n of the original data set f_j ? Simply glance back at equation (12.2.3):

$$F_n = F_n^e + e^{2\pi i n / N} F_n^o \quad n = 0, \dots, N - 1 \quad (12.3.4)$$

Expressed directly in terms of the transform H_n of our real (masquerading as complex) data set, the result is

$$F_n = \frac{1}{2}(H_n + H_{N/2-n}^*) - \frac{i}{2}(H_n - H_{N/2-n}^*) e^{2\pi i n / N} \quad n = 0, \dots, N - 1 \quad (12.3.5)$$

A few remarks:

- Since $F_{N-n}^* = F_n$ there is no point in saving the entire spectrum. The positive frequency half is sufficient and can be stored in the same array as the original data. The operation can, in fact, be done in place.
- Even so, we need values H_n , $n = 0, \dots, N/2$ whereas `four1` returns only the values $n = 0, \dots, N/2 - 1$. Symmetry to the rescue, $H_{N/2} = H_0$.
- The values F_0 and $F_{N/2}$ are real and independent. In order to actually get the entire F_n in the original array space, it is convenient to return $F_{N/2}$ as the imaginary part of F_0 .

- Despite its complicated form, the process above is invertible. First peel $F_{N/2}$ out of F_0 . Then construct

$$\begin{aligned} F_n^e &= \frac{1}{2}(F_n + F_{N/2-n}^*) \\ F_n^o &= \frac{1}{2}e^{-2\pi in/N}(F_n - F_{N/2-n}^*) \end{aligned} \quad n = 0, \dots, N/2 - 1 \quad (12.3.6)$$

and use `four1` to find the inverse transform of $H_n = F_n^{(1)} + iF_n^{(2)}$. Surprisingly, the actual algebraic steps are virtually identical to those of the forward transform.

Here is a representation of what we have said:

```

SUBROUTINE realft(data,n,isign)
INTEGER isign,n
REAL data(n)
C USES four1
  Calculates the Fourier transform of a set of n real-valued data points. Replaces this data
  (which is stored in array data(1:n)) by the positive frequency half of its complex Fourier
  transform. The real-valued first and last components of the complex transform are returned
  as elements data(1) and data(2), respectively. n must be a power of 2. This routine
  also calculates the inverse transform of a complex data array if it is the transform of real
  data. (Result in this case must be multiplied by 2/n.)
INTEGER i,i1,i2,i3,i4,n2p3
REAL c1,c2,h1i,h1r,h2i,h2r,wis,wrs
DOUBLE PRECISION theta,wi,wpi,wpr,
* wr,wtemp
theta=3.141592653589793d0/dble(n/2)
c1=0.5
if (isign.eq.1) then
  c2=-0.5
  call four1(data,n/2,+1)
else
  c2=0.5
  theta=-theta
endif
wpr=-2.0d0*sin(0.5d0*theta)**2
wpi=sin(theta)
wr=1.0d0+wpr
wi=wpi
n2p3=n+3
do 11 i=2,n/4
  i1=2*i-1
  i2=i1+1
  i3=n2p3-i2
  i4=i3+1
  wrs=sngl(wr)
  wis=sngl(wi)
  h1r=c1*(data(i1)+data(i3))
  h1i=c1*(data(i2)-data(i4))
  h2r=-c2*(data(i2)+data(i4))
  h2i=c2*(data(i1)-data(i3))
  data(i1)=h1r+wrs*h2r-wis*h2i
  data(i2)=h1i+wrs*h2i+wis*h2r
  data(i3)=h1r-wrs*h2r+wis*h2i
  data(i4)=-h1i+wrs*h2i+wis*h2r
  wtemp=wr
  wr=wr*wpr-wi*wpi+wr
  wi=wi*wpr+wtemp*wpi+wi
enddo 11

```

Double precision for the trigonometric recurrences. Initialize the recurrence.

The forward transform is here.

Otherwise set up for an inverse transform.

Case i=1 done separately below.

The two separate transforms are separated out of data.

Here they are recombined to form the true transform of the original real data.

The recurrence.

```

if (isign.eq.1) then
  h1r=data(1)
  data(1)=h1r+data(2)
  data(2)=h1r-data(2)
else
  h1r=data(1)
  data(1)=c1*(h1r+data(2))
  data(2)=c1*(h1r-data(2))
  call four1(data,n/2,-1)
endif
return
END

```

Squeeze the first and last data together to get them all within the original array.

This is the inverse transform for the case isign=-1.

Fast Sine and Cosine Transforms

Among their other uses, the Fourier transforms of functions can be used to solve differential equations (see §19.4). The most common boundary conditions for the solutions are 1) they have the value zero at the boundaries, or 2) their derivatives are zero at the boundaries. In the first instance, the natural transform to use is the *sine* transform, given by

$$F_k = \sum_{j=1}^{N-1} f_j \sin(\pi j k / N) \quad \text{sine transform} \quad (12.3.7)$$

where f_j , $j = 0, \dots, N-1$ is the data array, and $f_0 \equiv 0$.

At first blush this appears to be simply the imaginary part of the discrete Fourier transform. However, the argument of the sine differs by a factor of two from the value that would make this so. The sine transform uses *sines only* as a complete set of functions in the interval from 0 to 2π , and, as we shall see, the cosine transform uses *cosines only*. By contrast, the normal FFT uses both sines and cosines, but only half as many of each. (See Figure 12.3.1.)

The expression (12.3.7) can be “force-fit” into a form that allows its calculation via the FFT. The idea is to extend the given function rightward past its last tabulated value. We extend the data to twice their length in such a way as to make them an *odd* function about $j = N$, with $f_N = 0$,

$$f_{2N-j} \equiv -f_j \quad j = 0, \dots, N-1 \quad (12.3.8)$$

Consider the FFT of this extended function:

$$F_k = \sum_{j=0}^{2N-1} f_j e^{2\pi i j k / (2N)} \quad (12.3.9)$$

The half of this sum from $j = N$ to $j = 2N-1$ can be rewritten with the substitution $j' = 2N-j$

$$\begin{aligned} \sum_{j=N}^{2N-1} f_j e^{2\pi i j k / (2N)} &= \sum_{j'=1}^N f_{2N-j'} e^{2\pi i (2N-j') k / (2N)} \\ &= - \sum_{j'=0}^{N-1} f_{j'} e^{-2\pi i j' k / (2N)} \end{aligned} \quad (12.3.10)$$

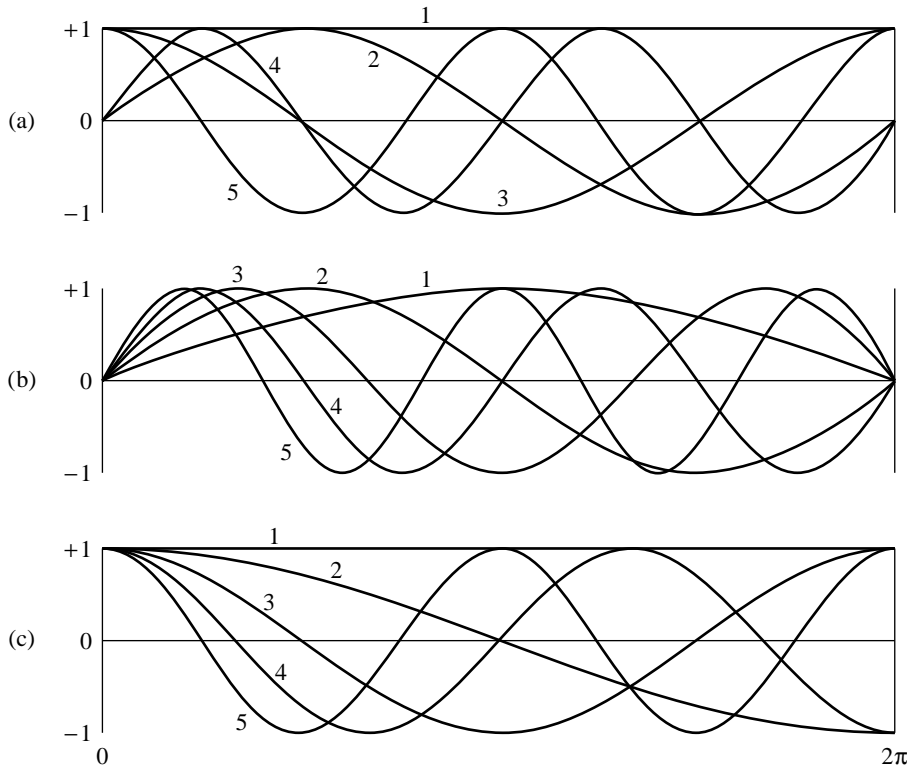


Figure 12.3.1. Basis functions used by the Fourier transform (a), sine transform (b), and cosine transform (c), are plotted. The first five basis functions are shown in each case. (For the Fourier transform, the real and imaginary parts of the basis functions are both shown.) While some basis functions occur in more than one transform, the basis sets are distinct. For example, the sine transform functions labeled (1), (3), (5) are not present in the Fourier basis. Any of the three sets can expand any function in the interval shown; however, the sine or cosine transform best expands functions matching the boundary conditions of the respective basis functions, namely zero function values for sine, zero derivatives for cosine.

so that

$$\begin{aligned}
 F_k &= \sum_{j=0}^{N-1} f_j \left[e^{2\pi i j k / (2N)} - e^{-2\pi i j k / (2N)} \right] \\
 &= 2i \sum_{j=0}^{N-1} f_j \sin(\pi j k / N)
 \end{aligned}
 \tag{12.3.11}$$

Thus, up to a factor $2i$ we get the sine transform from the FFT of the extended function.

This method introduces a factor of two inefficiency into the computation by extending the data. This inefficiency shows up in the FFT output, which has zeros for the real part of every element of the transform. For a one-dimensional problem, the factor of two may be bearable, especially in view of the simplicity of the method. When we work with partial differential equations in two or three dimensions, though, the factor becomes four or eight, so efforts to eliminate the inefficiency are well rewarded.

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From the original real data array f_j we will construct an auxiliary array y_j and apply to it the routine `realft`. The output will then be used to construct the desired transform. For the sine transform of data f_j , $j = 1, \dots, N-1$, the auxiliary array is

$$y_0 = 0$$

$$y_j = \sin(j\pi/N)(f_j + f_{N-j}) + \frac{1}{2}(f_j - f_{N-j}) \quad j = 1, \dots, N-1 \quad (12.3.12)$$

This array is of the same dimension as the original. Notice that the first term is symmetric about $j = N/2$ and the second is antisymmetric. Consequently, when `realft` is applied to y_j , the result has real parts R_k and imaginary parts I_k given by

$$R_k = \sum_{j=0}^{N-1} y_j \cos(2\pi jk/N)$$

$$= \sum_{j=1}^{N-1} (f_j + f_{N-j}) \sin(j\pi/N) \cos(2\pi jk/N)$$

$$= \sum_{j=0}^{N-1} 2f_j \sin(j\pi/N) \cos(2\pi jk/N)$$

$$= \sum_{j=0}^{N-1} f_j \left[\sin \frac{(2k+1)j\pi}{N} - \sin \frac{(2k-1)j\pi}{N} \right]$$

$$= F_{2k+1} - F_{2k-1} \quad (12.3.13)$$

$$I_k = \sum_{j=0}^{N-1} y_j \sin(2\pi jk/N)$$

$$= \sum_{j=1}^{N-1} (f_j - f_{N-j}) \frac{1}{2} \sin(2\pi jk/N)$$

$$= \sum_{j=0}^{N-1} f_j \sin(2\pi jk/N)$$

$$= F_{2k} \quad (12.3.14)$$

Therefore F_k can be determined as follows:

$$F_{2k} = I_k \quad F_{2k+1} = F_{2k-1} + R_k \quad k = 0, \dots, (N/2 - 1) \quad (12.3.15)$$

The even terms of F_k are thus determined very directly. The odd terms require a recursion, the starting point of which follows from setting $k = 0$ in equation (12.3.15) and using $F_1 = -F_{-1}$:

$$F_1 = \frac{1}{2}R_0 \quad (12.3.16)$$

The implementing program is

```

SUBROUTINE sinft(y,n)
INTEGER n
REAL y(n)
C  USES realft
    Calculates the sine transform of a set of n real-valued data points stored in array y(1:n).
    The number n must be a power of 2. On exit y is replaced by its transform. This program,
    without changes, also calculates the inverse sine transform, but in this case the output array
    should be multiplied by 2/n.
INTEGER j
REAL sum,y1,y2
DOUBLE PRECISION theta,wi,wpi,wpr,
*   wr,wtemp      Double precision in the trigonometric recurrences.
theta=3.141592653589793d0/dble(n) Initialize the recurrence.
wr=1.0d0
wi=0.0d0
wpr=-2.0d0*sin(0.5d0*theta)**2
wpi=sin(theta)
y(1)=0.0
do 11 j=1,n/2
    wtemp=wr
    wr=wr*wpr-wi*wpi+wr      Calculate the sine for the auxiliary array.
    wi=wi*wpr+wtemp*wpi+wi  The cosine is needed to continue the recurrence.
    y1=wi*(y(j+1)+y(n-j+1)) Construct the auxiliary array.
    y2=0.5*(y(j+1)-y(n-j+1))
    y(j+1)=y1+y2            Terms j and N - j are related.
    y(n-j+1)=y1-y2
enddo 11
call realft(y,n,+1)        Transform the auxiliary array.
sum=0.0
y(1)=0.5*y(1)              Initialize the sum used for odd terms below.
y(2)=0.0
do 12 j=1,n-1,2
    sum=sum+y(j)
    y(j)=y(j+1)
    y(j+1)=sum              Even terms in the transform are determined directly.
                              Odd terms are determined by this running sum.
enddo 12
return
END

```

The sine transform, curiously, is its own inverse. If you apply it twice, you get the original data, but multiplied by a factor of $N/2$.

The other common boundary condition for differential equations is that the derivative of the function is zero at the boundary. In this case the natural transform is the *cosine* transform. There are several possible ways of defining the transform. Each can be thought of as resulting from a different way of extending a given array to create an even array of double the length, and/or from whether the extended array contains $2N - 1$, $2N$, or some other number of points. In practice, only two of the numerous possibilities are useful so we will restrict ourselves to just these two.

The first form of the cosine transform uses $N + 1$ data points:

$$F_k = \frac{1}{2}[f_0 + (-1)^k f_N] + \sum_{j=1}^{N-1} f_j \cos(\pi j k / N) \quad (12.3.17)$$

It results from extending the given array to an even array about $j = N$, with

$$f_{2N-j} = f_j, \quad j = 0, \dots, N - 1 \quad (12.3.18)$$

If you substitute this extended array into equation (12.3.9), and follow steps analogous to those leading up to equation (12.3.11), you will find that the Fourier transform is just twice the cosine transform (12.3.17). Another way of thinking about the formula (12.3.17) is to notice that it is the Chebyshev Gauss-Lobatto quadrature formula (see §4.5), often used in Clenshaw-Curtis adaptive quadrature (see §5.9, equation 5.9.4).

Once again the transform can be computed without the factor of two inefficiency. In this case the auxiliary function is

$$y_j = \frac{1}{2}(f_j + f_{N-j}) - \sin(j\pi/N)(f_j - f_{N-j}) \quad j = 0, \dots, N-1 \quad (12.3.19)$$

Instead of equation (12.3.15), `realft` now gives

$$F_{2k} = R_k \quad F_{2k+1} = F_{2k-1} + I_k \quad k = 0, \dots, (N/2 - 1) \quad (12.3.20)$$

The starting value for the recursion for odd k in this case is

$$F_1 = \frac{1}{2}(f_0 - f_N) + \sum_{j=1}^{N-1} f_j \cos(j\pi/N) \quad (12.3.21)$$

This sum does not appear naturally among the R_k and I_k , and so we accumulate it during the generation of the array y_j .

Once again this transform is its own inverse, and so the following routine works for both directions of the transformation. Note that although this form of the cosine transform has $N + 1$ input and output values, it passes an array only of length N to `realft`.

```

SUBROUTINE cosft1(y,n)
  INTEGER n
  REAL y(n+1)
C  USES realft
  Calculates the cosine transform of a set y(1:n+1) of real-valued data points. The trans-
  formed data replace the original data in array y. n must be a power of 2. This program,
  without changes, also calculates the inverse cosine transform, but in this case the output
  array should be multiplied by 2/n.
  INTEGER j
  REAL sum,y1,y2
  DOUBLE PRECISION theta,wi,wpi,wpr,wr,wtemp      For trig. recurrences.
  theta=3.141592653589793d0/n                    Initialize the recurrence.
  wr=1.0d0
  wi=0.0d0
  wpr=-2.0d0*sin(0.5d0*theta)**2
  wpi=sin(theta)
  sum=0.5*(y(1)-y(n+1))
  y(1)=0.5*(y(1)+y(n+1))
  do 11 j=1,n/2-1                               j=n/2 unnecessary since y(n/2+1) unchanged.
    wtemp=wr
    wr=wr*wpr-wi*wpi+wr                         Carry out the recurrence.
    wi=wi*wpr+wtemp*wpi+wi
    y1=0.5*(y(j+1)+y(n-j+1))                   Calculate the auxiliary function.
    y2=(y(j+1)-y(n-j+1))
    y(j+1)=y1-wi*y2                             The values for j and N - j are related.
    y(n-j+1)=y1+wi*y2
    sum=sum+wr*y2
  enddo 11                                     Carry along this sum for later use in unfolding the
                                              transform.

```

```

call realft(y,n,+1)           Calculate the transform of the auxiliary function.
y(n+1)=y(2)                  sum is the value of  $F_1$  in equation (12.3.21).
y(2)=sum
do 12 j=4,n,2
  sum=sum+y(j)                Equation (12.3.20).
  y(j)=sum
enddo 12
return
END

```

The second important form of the cosine transform is defined by

$$F_k = \sum_{j=0}^{N-1} f_j \cos \frac{\pi k(j + \frac{1}{2})}{N} \quad (12.3.22)$$

with inverse

$$f_j = \frac{2}{N} \sum_{k=0}^{N-1} F_k \cos \frac{\pi k(j + \frac{1}{2})}{N} \quad (12.3.23)$$

Here the prime on the summation symbol means that the term for $k = 0$ has a coefficient of $\frac{1}{2}$ in front. This form arises by extending the given data, defined for $j = 0, \dots, N-1$, to $j = N, \dots, 2N-1$ in such a way that it is even about the point $N - \frac{1}{2}$ and periodic. (It is therefore also even about $j = -\frac{1}{2}$.) The form (12.3.23) is related to Gauss-Chebyshev quadrature (see equation 4.5.19), to Chebyshev approximation (§5.8, equation 5.8.7), and Clenshaw-Curtis quadrature (§5.9).

This form of the cosine transform is useful when solving differential equations on “staggered” grids, where the variables are centered midway between mesh points. It is also the standard form in the field of data compression and image processing.

The auxiliary function used in this case is similar to equation (12.3.19):

$$y_j = \frac{1}{2}(f_j + f_{N-j-1}) - \sin \frac{\pi(j + \frac{1}{2})}{N}(f_j - f_{N-j-1}) \quad j = 0, \dots, N-1 \quad (12.3.24)$$

Carrying out the steps similar to those used to get from (12.3.12) to (12.3.15), we find

$$F_{2k} = \cos \frac{\pi k}{N} R_k - \sin \frac{\pi k}{N} I_k \quad (12.3.25)$$

$$F_{2k-1} = \sin \frac{\pi k}{N} R_k + \cos \frac{\pi k}{N} I_k + F_{2k+1} \quad (12.3.26)$$

Note that equation (12.3.26) gives

$$F_{N-1} = \frac{1}{2} R_{N/2} \quad (12.3.27)$$

Thus the even components are found directly from (12.3.25), while the odd components are found by recursing (12.3.26) down from $k = N/2 - 1$, using (12.3.27) to start.

Since the transform is not self-inverting, we have to reverse the above steps to find the inverse. Here is the routine:

```

SUBROUTINE cosft2(y,n,isign)
INTEGER isign,n
REAL y(n)

```

C *USES realft*

Calculates the “staggered” cosine transform of a set $y(1:n)$ of real-valued data points. The transformed data replace the original data in array y . n must be a power of 2. Set $isign$ to $+1$ for a transform, and to -1 for an inverse transform. For an inverse transform, the output array should be multiplied by $2/n$.

```

INTEGER i
REAL sum,sum1,y1,y2,ytemp
DOUBLE PRECISION theta,wi,w1,wpi,wpr,wr,wr1,wtemp,PI
  Double precision for the trigonometric recurrences.
PARAMETER (PI=3.141592653589793d0)
theta=0.5d0*PI/n           Initialize the recurrences.
wr=1.0d0
wi=0.0d0
wr1=cos(theta)
w1=sin(theta)
wpr=-2.0d0*w1**2
wpi=sin(2.d0*theta)
if(isign.eq.1)then         Forward transform.
  do 11 i=1,n/2
    y1=0.5*(y(i)+y(n-i+1))   Calculate the auxiliary function.
    y2=w1*(y(i)-y(n-i+1))
    y(i)=y1+y2
    y(n-i+1)=y1-y2
    wtemp=wr1                Carry out the recurrence.
    wr1=wr1*wpr-w1*wpi+wr1
    w1=w1*wpr+wtemp*wpi+w1
  enddo 11
  call realft(y,n,1)         Calculate the transform of the auxiliary function.
  do 12 i=3,n,2              Even terms.
    wtemp=wr
    wr=wr*wpr-w1*wpi+wr
    w1=w1*wpr+wtemp*wpi+w1
    y1=y(i)*wr-y(i+1)*w1
    y2=y(i+1)*wr+y(i)*w1
    y(i)=y1
    y(i+1)=y2
  enddo 12
  sum=0.5*y(2)              Initialize recurrence for odd terms with  $\frac{1}{2}R_{N/2}$ .
  do 13 i=n,2,-2            Carry out recurrence for odd terms.
    sum1=sum
    sum=sum+y(i)
    y(i)=sum1
  enddo 13
else if(isign.eq.-1)then    Inverse transform.
  ytemp=y(n)
  do 14 i=n,4,-2           Form difference of odd terms.
    y(i)=y(i-2)-y(i)
  enddo 14
  y(2)=2.0*ytemp
  do 15 i=3,n,2            Calculate  $R_k$  and  $I_k$ .
    wtemp=wr
    wr=wr*wpr-w1*wpi+wr
    w1=w1*wpr+wtemp*wpi+w1
    y1=y(i)*wr+y(i+1)*w1
    y2=y(i+1)*wr-y(i)*w1
    y(i)=y1
    y(i+1)=y2
  enddo 15
  call realft(y,n,-1)
  do 16 i=1,n/2            Invert auxiliary array.
    y1=y(i)+y(n-i+1)

```

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```

    y2=(0.5/wi1)*(y(i)-y(n-i+1))
    y(i)=0.5*(y1+y2)
    y(n-i+1)=0.5*(y1-y2)
    wtemp=wr1
    wr1=wr1*wpr-wi1*wpi+wr1
    wi1=wi1*wpr+wtemp*wpi+wi1
  enddo 16
endif
return
END

```

An alternative way of implementing this algorithm is to form an auxiliary function by copying the even elements of f_j into the first $N/2$ locations, and the odd elements into the next $N/2$ elements in reverse order. However, it is not easy to implement the alternative algorithm without a temporary storage array and we prefer the above in-place algorithm.

Finally, we mention that there exist fast cosine transforms for small N that do not rely on an auxiliary function or use an FFT routine. Instead, they carry out the transform directly, often coded in hardware for fixed N of small dimension [1].

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12.4 FFT in Two or More Dimensions

Given a complex function $h(k_1, k_2)$ defined over the two-dimensional grid $0 \leq k_1 \leq N_1 - 1$, $0 \leq k_2 \leq N_2 - 1$, we can define its two-dimensional discrete Fourier transform as a complex function $H(n_1, n_2)$, defined over the same grid,

$$H(n_1, n_2) \equiv \sum_{k_2=0}^{N_2-1} \sum_{k_1=0}^{N_1-1} \exp(2\pi i k_2 n_2 / N_2) \exp(2\pi i k_1 n_1 / N_1) h(k_1, k_2) \quad (12.4.1)$$

By pulling the “subscripts 2” exponential outside of the sum over k_1 , or by reversing the order of summation and pulling the “subscripts 1” outside of the sum over k_2 ,