FOUNDATIONS OF HANKEL TRANSFORM ALGORITHMS

BY

BRUCE W. SUTER

Air Force Institute of Technology, Wright-Patterson Air Force Base, Ohio

Abstract. A brief survey of existing Hankel (Fourier-Bessel) transform algorithms is presented along with a natural way to classify these algorithms. In several cases these algorithms were derived originally by methods that were unnecessarily complicated and not sufficiently general. By using operator notation and Radon transform methods, derivations and generalizations are straightforward. These improvements and generalizations are given at the appropriate places in the discussion.

1.0. Introduction. Assume that \( f(r) \) satisfies the following conditions:

1) \( f(r) \) is defined for \( 0 \leq r < \infty \).

2) \( f(r) \) and \( \frac{df(r)}{dr} \) are piecewise continuous over any finite interval, except at isolated points of measure zero.

3) \( f(r) \) is absolutely convergent,

\[
\int_0^\infty |f(r)|dr < \infty.
\]

Definition. The \( n \)th-order Hankel transform (also called Fourier-Bessel transform) of \( f(r) \) satisfying 1), 2), and 3) is defined [7, 40] as

\[
\hat{f}(u) = 2\pi \int_0^\infty r f(r)J_n(2\pi ru)dr,
\]

where \( J_n \) is a Bessel function of the first kind [45].

The integral transform (1) is expressed in operator notation by

\[
\hat{f}(u) = H_n f(r),
\]

where \( H_n \) represents the \( n \)th-order Hankel transform.

In a variety of applications which exhibit cylindrical symmetry, the Hankel transform of a function must be computed. These include underwater acoustics [17, 18], atomic physics [43], geophysical prospecting [13, 20], laser mode analyses [38], image processing [33], structural dynamics due to earthquakes [28], and speech processing [21]. Thus, the Hankel transform has been applied to problems in diffraction and wave propagation.

A device that will be repeatedly used throughout the paper is the following: By imposing radial symmetry on \( f(x, y) \), then \( f(r) \) can be described as a part of

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separable function \( f(x, y) = f(r) \exp(i n \theta) \). Since the Hankel transform of \( f(r) \) is desired, no generality is lost.

2.0. New classification. It is proposed that Hankel transform algorithms be systematically organized into three different categories: direct, indirect, and miscellaneous methods. Direct methods are based on the definition of the Hankel transform, while indirect methods are based on traversing Radon space while calculating the Hankel transform. Miscellaneous methods are methods that appear to have promise, but require that significant computational problems be solved before they become practical.

Direct methods, the traditional and more established ones, are a class of Hankel transform algorithms directly ascertainable from the definition of the Hankel transform. The direct methods may be organized further into four different classes: quadrature methods, asymptotic algorithms, convolutional algorithms, and simplified two-dimensional fast Fourier transform algorithms (see Table 1).

Indirect methods refer to a class of Hankel transform algorithms based indirectly on the definition of the Hankel transform. Although this class is more recent than the direct method one, several practical algorithms have been developed. The indirect methods are organized further into two different classes: projection-slice algorithms and filtered backprojection algorithms (see Table 1).

All of the classes of Hankel transform algorithms mentioned thus far have been utilized in at least one engineering or scientific application. Two other methods are worthy of being mentioned, although they still require more study before their usefulness can be fully determined. These miscellaneous methods are hybrid methods and orthogonal function methods (see Table 1).

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3.0. Direct methods. In the next four subsections we cover the four classes of direct Hankel transform algorithms listed in the top third of Table 1.

3.1. Quadrature methods. Quadrature methods involve direct numerical integration [15]. An integral is approximated by a linear combination of integrand values:

\[
\int_{a}^{b} g(x)dx \approx \sum_{i=1}^{N} g(x_i)\omega_i, \quad (3)
\]

where \( x_1, x_2, \ldots, x_N \) are \( N \) points, or abscissas, normally chosen in the interval of integration, and the numbers \( \omega_1, \omega_2, \ldots, \omega_N \) are weights assigned to these points.
Many times Gaussian integration is utilized to evaluate the summation over \( i \), since Gaussian formulas are “best” in the sense that they integrate exactly polynomials of as high degree as possible with a formula of the form of (3).

Longman [30, 31] split the Hankel transform (1) into a sum of integrals of the form

\[
\tilde{f}(u) = \sum_{s=0}^{\infty} V_s
\]

and

\[
V_s = 2\pi \int_{j_{n,s-1}}^{j_{n,s}} r f(r) J_n(2\pi ru) dr,
\]

where \( j_{n,0} = 0 \) and \( j_{n,s} \) is the \( s \)th positive zero of the Bessel function \( J_n(2\pi ru) \) [45]. An approximation to (5) in a form similar to (3) is given by the summation

\[
V_s = 2\pi \sum_{j=1}^{N} h_j r_j f(r_j) J_n(2\pi r_j u),
\]

where \( r_j \) is the abscissa and \( h_j \) is the weight. \( V_s \), the terms in the summation over \( s \) in Eq. (4) are often alternating in sign due to the Bessel function’s oscillatory behavior, and the result may be slowly convergent. Numerous techniques exist for transforming a slowly convergent series into one more rapidly convergent. These techniques include Euler transformations, the \( \varepsilon \)-transformation, Aitken’s \( \Delta^2 \) method, and Levin’s \( V \)-transformation [15]. An early review of quadrature methods was provided by Cornille [14].

More recently, Chave [12] presented an iterative technique that utilized a Pade convergence acceleration step together with Patterson’s [36] methods for the optimal addition of points to a quadrature formula. Chave’s [12] technique permits the control of local error, defined as the difference between the result of two successive quadratures, each with a different number of abscissas. In addition, Chave [12] demonstrated that, through the utilization of the Pade approach, it is possible to numerically integrate Hankel transforms that are formally divergent. This algorithm was applied by Chave and Cox [13] to the calculation of the electromagnetic fields associated with the sea floor.

3.2. Asymptotic algorithms. Utilizing the asymptotic expansion for the Bessel function of the first kind [5],

\[
J_n(x) \approx \left[2/(\pi x)\right]^{1/2} \cos[x - (2n + 1)\pi/4],
\]

the Hankel transform (1) becomes

\[
\tilde{f}(u) \approx 2\pi \int_{0}^{\infty} r f(r)[2/\pi(2\pi ru)]^{1/2} \cos[(2\pi ru) - (2n + 1)\pi/4]dr
\]

or equivalently

\[
\tilde{f}(u) \approx (2u^{-1/2}) \int_{0}^{\infty} r^{1/2} f(r) \cos[2\pi ru - (2n + 1)\pi/4]dr.
\]
Extending this integral over \( r \) from \(-\infty\) to \( \infty \) and utilizing the fact that \( \cos \theta = \text{Re}\{\exp(-i\theta)\} \) yields,

\[
\tilde{f}(u) \approx (u^{-1/2}) \text{Re}\{\exp[i(2n + 1)\pi/4]\int_{-\infty}^{\infty} (|r|)^{1/2} f(|r|) \exp(-i2\pi ru)dr\}.
\]

(7)

Utilizing operator notation Eq. (7) becomes

\[
\tilde{f}(u) \approx (u^{-1/2}) \text{Re}\{\exp[i(2n + 1)\pi/4]\mathcal{F}_1[(|r|)^{1/2} f(|r|)]\},
\]

(8)

where \( \mathcal{F}_1 \), the (one-dimensional) Fourier transform in cartesian coordinates, is given by [7, 40]

\[
\mathcal{F}_1(f(x)) = \int_{-\infty}^{\infty} f(x) \exp(-i2\pi xu)dx.
\]

(9)

DiNapoli and Deavenport [17] utilized Eq. (8) to approximate the \( n \)th-order Hankel transform in terms of a fast Fourier transform.

Caution must be exercised when using asymptotic algorithms, since their results are only valid for large values of the output argument. When for a given application this is not a limitation, i.e., only large values of the output argument are desired, then this algorithm can be successfully applied. DiNapoli and Deavenport [17] applied this algorithm to the problem of modeling underwater acoustic propagation loss versus range.

3.3. Convolutional algorithms. Convolutional algorithms have been developed independently by Ghosh [20], Siegman [39], and Talman [43]. Let

\[
H(u) = u\tilde{f}(u),
\]

where \( \tilde{f}(u) \) is the equation for the Hankel transform (1),

\[
H(u) = 2\pi \int_{0}^{\infty} ruf(r)J_n(2\pi ru)dr.
\]

(10)

Assume that \( f(r) = 0 \) for \( r \leq 0 \) and that \( r \) and \( u \) vary in the following manner,

\[
r = r_0 \exp(-n) \quad \text{and} \quad u = u_0 \exp(m).
\]

(11)

Substituting Eq. (11) into Eq. (10) yields

\[
H(m) = 2\pi u_0 \exp(m) \int_{-\infty}^{\infty} [r_0 \exp(-n)]^2 f[r_0 \exp(-n)] J_n[2\pi r_0 u_0 \exp(m - n)]dn.
\]

Let

\[
F(n) = [r_0 \exp(-n)]f[r_0 \exp(-n)]
\]

and

\[
G(m - n) = 2\pi r_0 u_0 \exp(m - n) J_n[2\pi r_0 u_0 \exp(m - n)].
\]

Then \( H(m) \) becomes

\[
H(m) = \int_{-\infty}^{\infty} F(n)G(m - n)dn ,
\]

(12)

since \( f(r) = 0 \) for \( r \leq 0 \). Therefore,

\[
\tilde{f}(u) = H(m)/u = H(m)/[u_0 \exp(m)].
\]

(13)
One approach to the calculation of Eq. (12) is with the Fourier convolution theorem, which states that the convolution of two functions corresponds to the product of their transforms. Hence,

\[ H(m) = F^{-1}_1[F_1F(n)F_1G(n)]. \]

Sheng and Siegman [38] utilized this approach to the problem of modeling high power lasers.

In geophysics, Hankel transforms frequently occur as a set of transforms whose integrands are related algebraically (see, for example, Verma [44]). If each of these Hankel transforms is evaluated using convolutional algorithms, then the resulting set of evaluations will be termed related convolutions [1]. In addition, after the initial direct convolution, subsequent "lagged" convolutions [2] are performed with previously computed and saved function values. Anderson [2] showed that by arranging related and lagged convolutions in a matrix, the number of function calls can be minimized. The resulting algorithm [3] is extensively utilized by geophysicists [4].

3.4. Simplified two-dimensional fast Fourier transform algorithms. The zero-order Hankel transform may be defined as the two-dimensional Fourier transform of a circularly symmetric function, and described by [7]

\[ H_0f(r) = F_2f(r), \]  
where \( r = [x^2 + y^2]^{1/2} \) and \( F_2 \), the two-dimensional Fourier transform in cartesian coordinates, is given by [7, 40]

\[ \hat{f}(u, v) = F_2f(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) \exp[-i2\pi(xu + yv)]dxdy. \]  

Utilizing Eq. (14), Murphy and Gallagher [33] developed a simplified two-dimensional fast Fourier transform algorithm. They did not formulate the problem in a general manner, and extensions to higher orders were not apparent. Murphy and Gallagher [33] utilized this algorithm in image processing applications in which the output data are required to be in the form of a two-dimensional rectangular array for subsequent processing.

Utilizing operator notation, the framework for simplified two-dimensional fast Fourier transform algorithms will be generalized. When \( f(x, y) = f(r) \exp(in\theta) \), the Hankel transform is related to the two-dimensional Fourier transform of \( f(r) \exp(in\theta) \):

\[ F_2f(r) \exp(in\theta) = \exp(in\phi) i^{-n} H_n f(r). \]  

Proof. To see how (16) is obtained it is convenient to express the definition of the two-dimensional Fourier transform in polar coordinates,

\[ F_2f(r, \theta) = \int_{0}^{\infty} \int_{0}^{2\pi} f(r, \theta) \exp[-i2\pi rq \cos(\theta - \phi)] rdrd\theta. \]

Let \( \beta = \theta - \phi \), then

\[ F_2f(r) \exp(in\theta) = \exp(in\phi) \int_{0}^{\infty} df(r) r \int_{0}^{2\pi} \exp[i(n\beta - 2\pi rq \cos \beta)]d\beta. \]
By use of the identity \( J_m(x) = (i^m/2\pi) \int_0^{2\pi} \exp[i(m\alpha - x \cos \alpha)]d\alpha \) [45], it follows that
\[
F_2(f(r) \exp(i\theta)) = \exp(i\phi)n^{-\theta} 2\pi \int_0^\infty r f(r) \tilde{J}_n(2\pi q) dr.
\]
This result in operator notation is
\[
F_2(f(r) \exp(i\theta)) = \exp(i\phi)n^{-\theta} H_n f(r).
\]
QED

It is important to realize that if \( f(r) \) is real, then \( H_n f(r) \) is a real. This suggests that a form simpler than (16) may exist. Since \( \phi \) is a parameter, let \( \phi \) equal zero. When \( \phi \) is set to zero, no generality is lost since the functions \( f(r) \) have radial symmetry. Consequently, if the Hankel transform is known for any angle \( \phi \), then it is known for all \( \phi \).

If \( f(r) \) is real, then the Hankel transform and the two-dimensional Fourier transform are related by
\[
H_n f(r) = \left\{ \begin{array}{ll}
\exp(n\phi)H_n f(r) & \text{if } n \text{ is even} \\
\exp(n\phi)H_n f(r) & \text{if } n \text{ is odd.}
\end{array} \right.
\]

Proof. Since \( f(r) \) is real, then \( H_n f(r) \) is real. Therefore, utilizing Eq. (16) with \( \phi = 0 \) yields,
\[
H_n f(r) = \text{Re}\{i^n F_2(f(r) \exp(\phi))\}.
\]
This equation can be rewritten utilizing the definition of \( i^n \).
\[
H_n f(r) = \left\{ \begin{array}{ll}
\text{Re}\{F_2 f(r) \exp(i\phi)\} & \text{if } n \text{ MOD } 4 = 0 \\
i \text{Im}\{F_2 f(r) \exp(i\phi)\} & \text{if } n \text{ MOD } 4 = 1 \\
-\text{Re}\{F_2 f(r) \exp(i\phi)\} & \text{if } n \text{ MOD } 4 = 2 \\
-i \text{Im}\{F_2 f(r) \exp(i\phi)\} & \text{if } n \text{ MOD } 4 = 3.
\end{array} \right.
\]
The real and imaginary parts of the Fourier transform of \( f(x, y) \) are given by
\[
\text{Re}\{F_2 f(x, y)\} = F_2[f(x, y) + f^*(x, y)]/2
\]
and
\[
\text{Im}\{F_2 f(x, y)\} = F_2[f(x, y) - f^*(x, y)]/2i,
\]
where \( f^*(x, y) \) is the complex conjugate of \( f(x, y) \). Then,
\[
H_n f(r) = \left\{ \begin{array}{ll}
F_2(f(r) \cos(n\theta)) & \text{if } n \text{ MOD } 4 = 0 \\
if F_2(f(r) \sin(n\theta)) & \text{if } n \text{ MOD } 4 = 1 \\
-F_2(f(r) \cos(n\theta)) & \text{if } n \text{ MOD } 4 = 2 \\
-iF_2(f(r) \sin(n\theta)) & \text{if } n \text{ MOD } 4 = 3.
\end{array} \right.
\]
Equivalently,
\[
H_n f(r) = \left\{ \begin{array}{ll}
i^n F_2(f(r) \cos(n\theta)) & \text{if } n \text{ even} \\
if^n F_2(f(r) \sin(n\theta)) & \text{if } n \text{ odd.}
\end{array} \right.
\]
QED

The simplified two-dimensional fast Fourier transform algorithms are placed in perspective through the utilization of operator notation.
4.0. **Indirect methods.** In the next two subsections we cover the two classes of indirect Hankel transform algorithms listed in the middle third of Table 1, and develop extensions and generalizations using Radon transform methods.

4.1. **Projection-slice algorithms.** Through manipulations of the two-dimensional Fourier transform, Kausel and Bouchovalas [27] arrived at the following equations:

\[
\begin{align*}
\cos(n\theta)g(r)dy \\
\end{align*}
\]

and

\[
G(\xi) = i^n \int_{-\infty}^{\infty} f(x) \exp(-i2n\xi x)dx,
\]

where \( \theta = \tan^{-1}(y/x) \), \( r = (x^2 + y^2)^{1/2} \), and \( G(\xi) \) is an approximation to \( H_n g(r) \).

Since Kausel and Bouchovalas [27] did not formulate the problem in a more general manner, the limitations of (18) were not apparent. It will be shown through the utilization of operator notation that Eq. (18) is only correct for Hankel transforms of even order. In spite of these errors, Kausel and Roesset [28] were able to utilize this technique to analyze the dynamics of structures due to earthquakes.

Utilizing operator notation, the framework for projection-slice algorithms will be specified in terms of the projection-slice theorem [16, 24]:

\[
F_2f(x,y) = F_1R_2f(x,y),
\]

where \( R_2 \), the two-dimensional Radon transform, is given by [6, 16, 19, 24, 37]

\[
R_2f(x,y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x,y) \delta(\rho - x \cos \phi - y \sin \phi)dx dy,
\]

and \( \delta \) is the Dirac delta function [7, 29]. Observe that the delta function in the equation above is one-dimensional and thus the integral over the \( xy \) plane reduces to an integral along the line defined by the equation

\[
\rho = x \cos \phi + y \sin \phi.
\]

By substituting Eq. (19) into (17) the connection between the Hankel transform and the projection-slice theorem becomes apparent.

\[
H_n f(r) = \left\{ \begin{array}{ll}
i^n F_1 R_2 f(r) \cos(n\theta), & \text{if } n \text{ is even} \\
i^n F_1 R_2 f(r) \sin(n\theta), & \text{if } n \text{ is odd}
\end{array} \right.
\]

It is important to note that a projection angle \( \phi \) of zero degrees is assumed. Through the utilization of the definitions of the Fourier (15) and Radon (20) transforms, Eq. (21) becomes

\[
\hat{f}(u) = i^n \int_{-\infty}^{\infty} \exp(-i2\pi p u)\hat{f}(p)dp
\]

where

\[
\hat{f}(p) = [\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x,y)\delta(\rho - x \cos \phi - y \sin \phi)dx dy]_{\phi=0} = \int_{-\infty}^{\infty} f(\rho,y)dy
\]
and
\[ f(\rho, y) = \begin{cases} f(r) \cos(n\theta), & \text{if } n \text{ is even} \\ f(r) \sin(n\theta), & \text{if } n \text{ is odd}, \end{cases} \]
\[ r = \left[ \rho^2 + y^2 \right]^{1/2}, \]
and
\[ \theta = \tan^{-1}(y/\rho). \]
Thus, utilizing the insight obtained from using operator notation, we see that Eq. (18) can be corrected by changing \( \cos(n\theta) \) to \( \sin(n\theta) \) for Hankel transforms of odd orders.

By including Tchebycheff transforms, the theoretical framework for projection-slice algorithms becomes richer. Tchebycheff transforms are related to Radon transforms by the following equation:
\[ R_2 f(r) \exp(in\theta) = \exp(in\phi) T_n f(r), \tag{24} \]
where \( T_n \), the \( n \)-th-order Tchebycheff transform, is given by
\[ T_n f(r) = 2 \int_0^\infty f(r) T_n(\rho/r) dr / [1 - (\rho/r)^2]^{1/2}, \tag{25} \]
where \( T_n \) is a Tchebycheff polynomial, Type I [5].

**Proof.** To see how (24) is obtained it is convenient to express the definition of the Radon transform (20) in polar coordinates,
\[ R_2 f(r) \exp(in\theta) = \int_0^\infty \int_0^{2\pi} f(r) \exp(in\theta) \delta[\rho - r \cos(\theta - \phi)] r dr d\theta. \]
If the change of variable \( \beta = \theta - \phi \) is made, then
\[ R_2 f(r) \exp(in\theta) = \exp(in\phi) \int_0^\infty d\rho f(r) r \int_0^{2\pi} \exp(in\beta) \delta(\rho - r \cos \beta) d\beta. \]
The integral over \( \beta \) can be evaluated using the fact that
\[ \int a(x) \delta(h(x)) dx = \sum_i a(x_i) \left| dh/dx \right|_{x=x_i}, \]
where \( x_i \)'s are zeroes of multiplicity one of \( h(x) \) within the limits of integration of \( x \) [26].
The zeroes of the argument of the delta function are
\[ \beta = \pm \cos^{-1}(\rho/r), \quad r \geq \rho. \]
By using this result the integral over \( \beta \) yields
\[ \int_0^{2\pi} \exp(in\beta) \delta(\rho - r \cos \beta) d\beta = 2 \cos[n \cos^{-1}(\rho/r)] / [r(1 - (\rho/r)^2)^{1/2}], \quad r \geq \rho. \]
Since the Tchebycheff polynomial is related to the cosine function by \( T_n(x) = \cos[n \cos^{-1}(x)] \), then
\[ \int_0^{2\pi} \exp(in\beta) \delta(\rho - r \cos \beta) d\beta = 2T_n(\rho/r) / [r(1 - (\rho/r)^2)^{1/2}], \quad r \geq \rho. \]
Hence,
\[ R_2 f(r) \exp(i \theta) = 2 \exp(i \phi) \int_{|p|}^{\infty} f(r) T_n(\rho/r) dr / [1 - (\rho/r)^2]^{1/2}. \]
This result in operator notation is
\[ R_2 f(r) \exp(i \theta) = \exp(i \phi) T_n f(r). \]
QED

Tchebycheff transforms are related to Hankel transforms by the following equation:
\[ H_n f(r) = \int_{n \Phi} T_n f(r). \] (26)

**Proof.** From (16),
\[ \exp(i \phi) i^{-n} H_n f(r) = F_2 f(r) \exp(i \theta), \]
and the projection-slice theorem for the two-dimensional Fourier transform (19),
\[ \exp(i \phi) i^{-n} H_n f(r) = F_1 R_2 f(r) \exp(i \theta). \]
Utilizing the relationship between the Radon transform and the Tchebycheff transform (24) results in the following equation:
\[ \exp(i \phi) i^{-n} H_n f(r) = \exp(i \phi) F_1 T_n f(r). \]
Hence, this equation in operator notation is given by
\[ H_n f(r) = \int_{n \Phi} T_n f(r). \]
QED

The utilization of (26) to develop projection-slice algorithms of Hankel transforms of order \( n \) is presented in Suter [41, 42]. For zero order, Eq. (26) can be rewritten as
\[ H_0 f(r) = \int_{n \Phi} A f(r), \] (27)
where \( A \), the Abel transform, is given by
\[ A f(r) = 2 \int_{|p|}^{\infty} f(r) dr / [1 - (\rho/r)^2]^{1/2}. \]
Equation (27) forms the basis for the Hankel transform algorithms of Mook [32] and Hansen [23].

Substituting the definitions of the Fourier and Tchebycheff transforms in Eq. (26) yields
\[ H_n f(r) = \int_{-\infty}^{\infty} \hat{f}(\rho) \exp(-i2\pi \rho u) d\rho \] (28)
where
\[ \hat{f}(\rho) = 2 \int_{|p|}^{\infty} f(r) T_n(\rho/r) dr / [1 - (\rho/r)^2]^{1/2}. \] (29)
Assume \( f(r) = 0 \) for \( r > R \). Then, Eq. (29) can be rewritten as
\[ \hat{f}(\rho) = 2 \int_{|p|}^{R} f(r) T_n(\rho/r) dr / [1 - (\rho/r)^2]^{1/2}. \] (30)
The change of variables $z^2 = r^2 - \rho^2$ yields

$$\tilde{f}(\rho) = 2 \int_{0}^{1} dz f[(z^2 + \rho^2)^{1/2}] T_n[\rho/(z^2 + \rho^2)^{1/2}].$$

(31)

Although Oppenheim, Frisk, and Martinez [34, 35] derived Eqs. (28) and (31), their primary interest appears to be order zero. In a companion paper to Oppenheim, Frisk, and Martinez [35], Frisk, Oppenheim, and Martinez [18] applied this algorithm to the problem of modeling ocean acoustic propagation. In support of this algorithm, Frisk, Oppenheim, and Martinez [18] developed a new measurement technique of the plane wave reflection coefficient.

4.2. Filtered backprojection algorithms. Through the use of Bessel function identities, Candel [8, 10] and Gopalan and Chen [22] obtained the following equations:

$$\phi(r) = \int_{-\infty}^{\infty} \exp(i2\pi r) f(\xi)d\xi$$

(32)

and

$$F(r) = \int_{0}^{\pi} \phi(r \sin \theta) \exp(-in\theta)d\theta$$

(33)

where $F(r)$ is an approximation to $H_n f(\xi)$.

Candel [8] and Gopalan and Chen [22] were concerned with zero order only, while Candel [10] was directed towards the $n$th-order Hankel transform. Since Candel [8, 10] and Gopalan and Chen [22] did not formulate the problem in a more general way, the limitations of (32) and (33) were not apparent. Moreover, flaws in the derivation of the equations were compensated in the actual implementation by symmetry arguments. It will be shown that the factor $\xi$ in (32) should be $|\xi|$. In addition, a factor of $i^{-n}$ is missing in (33). In spite of these errors, Gopalan [21] successfully utilized Eqs. (32) and (33) in the area of speech processing.

The following result is a generalization of the work of Candel [8, 10] and Gopalan and Chen [22]. Utilizing operator notation the framework for the filtered backprojection will be specified in terms of the relationship between the Hankel transform and the two-dimensional Fourier transform.

$$F_2^{-1} \tilde{f}(r, \theta) = \exp(-in\phi) i^n H_n^{-1} \tilde{f}(r),$$

(34)

where

$$\tilde{f}(r, \theta) = \tilde{f}(r) \exp(i\theta).$$

(35)

This equation is obtained by taking the inverse of both sides of the equation previously given as (16). Since the Hankel transform has a Fourier kernel [40], i.e., $H_n f(r) = H_n^{-1} f(r)$, then Eq. (34) becomes

$$F_2^{-1} \tilde{f}(r, \theta) = \exp(-in\phi) i^n H_n \tilde{f}(r).$$

(36)

Given a function $\tilde{f}(r, \theta)$, then its inverse two-dimensional Fourier transform can be described in terms of a filtered backprojection (see for example Deans [16]). It is given by

$$f(x, y) = F_2^{-1} \tilde{f}(r, \theta) = BF_1^{-1}[|r| \tilde{f}(r, \theta)],$$

(37)
where $B$, the backprojection operator, is defined by

$$B g(x \cos \theta + y \sin \theta, \theta) = \int_0^\pi g(x \cos \theta + y \sin \theta, \theta) d\theta. \quad (38)$$

Let $x = k \cos \phi$ and $y = k \sin \phi$. Then Eqs. (37) and (38) become

$$f(k, \phi) = F^{-1}_2 \hat{f}(r, \theta) = BF^{-1}_1 \|r|f(r, \theta)\|, \quad (39)$$

where

$$B g(k \cos(\phi - \theta), \theta) = \int_0^\pi g(k \cos(\phi - \theta), \theta) d\theta. \quad (40)$$

Substituting Eq. (39) into (36) yields

$$BF^{-1}_1 \|r|f(r, \theta)\| = \exp(-in\phi) H_n \hat{f}(r). \quad (41)$$

Since $\phi$ is a parameter, let $\phi$ equal zero. When $\phi$ is set equal to zero, no generality is lost since the functions $\hat{f}(r)$ have radial symmetry. Consequently, if the Hankel transform is known for any angle $\phi$, then it is known for all $\phi$. Hence, Eqs. (40) and (41) become

$$H_n \hat{f}(r) = i^{-n} BF^{-1}_1 \|r|f(r, \theta)\|, \quad (42)$$

where

$$B g(k \cos \theta, \theta) = \int_0^\pi g(k \cos \theta, \theta) d\theta. \quad (43)$$

It is important to realize that the inverse Fourier transform is performed along the $r$ direction. Recall that $\hat{f}(r, \theta)$ was assumed to have the form $\hat{f}(r, \theta) = \hat{f}(r) \exp(in\theta)$. Then utilizing the definition of the inverse Fourier transform together with Eqs. (42) and (43) yields the following pair of integral equations

$$H_n \hat{f}(r) = i^{-n} \int_0^\pi h(k \cos \theta) \exp(in\theta) d\theta, \quad (44)$$

where

$$h(k) = \int_{-\infty}^{\infty} \|r|\hat{f}(r) \exp(i2\pi rk) dr. \quad (45)$$

Let $\alpha = \theta - \pi/2$. Since the resulting integrand is an even function, the interval of integration can be halved. Hence, Eq. (44) becomes

$$H_n \hat{f}(r) = 2 \int_0^{\pi/2} h(k \sin \alpha) \exp(in\alpha) d\alpha. \quad (46)$$

Through the utilization of Bessel function identities, Higgins and Munson [25] derived Eqs. (45) and (46), which define their $n$th-order Hankel transform algorithm.

### 5.0. Miscellaneous methods

In the next two subsections we cover the two classes of miscellaneous Hankel transform algorithms listed in the bottom third of Table 1.

#### 5.1. Hybrid methods

Hybrid methods refer to a class of Hankel transform algorithms based on the use of a slow, accurate algorithm for small values of the output variable and the use of a fast, approximate algorithm for large values of the output variable. Both the slow, accurate algorithm and the fast, approximate algorithm can be either direct or indirect method.
Candel [9] developed a working algorithm utilizing a combination of backprojection methods and asymptotic methods together with an ad hoc technique for switching between the slow and the fast algorithms; however, the accuracy of Candel’s results was data dependent. No practical hybrid method algorithms can be expected until the switching point between the slow and the fast algorithms can be chosen to maximize accuracy.

5.2. Orthogonal function methods. Orthogonal function methods refer to a class of Hankel transform algorithms based on expanding a function in terms of an orthogonal set of functions whose members have known Hankel transforms. Cavanaugh and Cook [11] developed a working algorithm using Gaussian-Laguerre polynomials. A disadvantage is that a large number of terms are required for convergence. These algorithms could be improved by selecting the orthogonal functions to minimize the number of terms required for convergence.

Summary. A new classification scheme for Hankel transform algorithms was presented. This classification is significant because: first, it specifies a way to systematically characterize Hankel transform algorithms; and, second, it provides a formalism for extending and generalizing previously stated results. Accordingly, extensions or generalizations were provided for the simplified 2-D FFT algorithms, the projection-slice algorithms, and the filtered backprojection algorithms.

References