

# Stolt Interpolation Basics with MATLAB Algorithms

Josiah W. Smith

**Abstract**—Stolt interpolation, also known as range migration algorithm (RMA) or  $\omega - k$  or  $f - k$  migration, for the rectilinear and cylindrical cases in 2D and 3D using monostatic radar systems for near-field 2D and 3D image reconstruction.

## I. Introduction

This paper is organized as follows. Section II will discuss the application of Stolt interpolation to a 1D-rectilinear SAR scanner using a single monostatic, SISO radar array for 2D image reconstruction in the  $(x, z)$  domains and provide applicable MATLAB algorithms. Section III will extend the application of Stolt interpolation for use on a 2D-rectilinear SAR scanner using a single monostatic, SISO radar array for 3D holographic image reconstruction in the  $(x, y, z)$  domains and provide necessary MATLAB algorithms. Section IV will propose and demonstrate a 1D-cylindrical SAR (CSAR) system and the proper Stolt interpolation to recover a 2D image with relevant MATLAB algorithms. And finally, Section V will discuss the extension of this CSAR system to a 2D-CSAR system for generating high resolution 3D holographic images and provide MATLAB algorithms.

## II. 1D-Rectilinear SISO-SAR System Model for 2D Image Reconstruction by Stolt Interpolation

The one-dimensional rectilinear single-input, single-output synthetic aperture radar (SISO-SAR) system consists of a single transceiver moved to evenly-spaced positions across the horizontal X-domain (can we get a figure?).

The received signal from a monostatic, full-duplex transceiver is of the form given by equation (1).

$$s(x, k) = \iint \frac{p(x, z)}{R^2} e^{j2kR} dx dz \quad (1)$$

$$R = \sqrt{(x - x')^2 + (z - z')^2} \quad (2)$$

Where  $R$  is the distance from the transceiver to the target scene,  $p(x, z)$  is the reflectivity function of the target scene, and  $(x', z')$  is the location of the transceiver.

It has been shown in the literature that  $p(x, z)$  can be recovered via Fourier-based techniques and circular to planar wavefront decomposition as described in equation (3).

$$p(x, z) = IFT_{2D}^{(k_x, k_z)} \left[ Stolt^{(k)} \left( FT_{1D}^{(x)} [s^*(x, z)] \right) \right] \quad (3)$$

Where  $FT$  and  $IFT$  are the forward and backward Fourier transform operators,  $(\cdot)^*$  is the complex conjugate operator, and  $Stolt(\cdot)$  is the Stolt interpolation algorithm.

This is the fundamental equation for rectilinear Stolt interpolation and will be further examined in this paper.

## A. 1D-Rectilinear SISO-SAR 2D Image Reconstruction and Necessity of Stolt Interpolation

The single-input-single-output synthetic aperture radar (SISO-SAR) reconstruction algorithm is described by equations (4),(5),(6). Where equation (5) is the Stolt interpolation step, migrating  $(k_x, k)$  to  $(k_x, k_z)$  by the relation given in equation (7), derived from the dispersion relation of electromagnetic (EM) waves.

$$S(k_x, k) = FT_{(x)}[s^*(x, k)] \quad (4)$$

$$\hat{S}(k_x, k_z) = Stolt^{(k)}[S(k_x, k)] = S(k_x, k) \Big|_{k=\frac{1}{2}\sqrt{k_x^2+k_z^2}} \quad (5)$$

$$p(x, z) = IFT_{(k_x, k_z)}[\hat{S}(k_x, k_z)] \quad (6)$$

$$(2k)^2 = k_x^2 + k_z^2 \quad (7)$$

Without the Stolt interpolation step in equation (5), the 2D image is not resolvable in the  $(x, z)$  plane.

## B. MATLAB Algorithm for 2D-Image Reconstruction using Stolt Interpolation

```

1 function StoltInterpolation2Dk_kz(sxk,f,xStep,
   nSample,nFFTxk,nFFTkz)
2 % 2D Stolt Interpolation from k-kZ
3 % f: Frequency Vector
4 nHorMeasurement: # of Horizontal Measurements
5 xStep: Horizontal Step Size in m
6 nSample: Number of Samples
7 nFFTxk: FFT size for x and k domains
8 nFFTkz: FFT size for z domain
9 % Maintain sxk = s(x,k)
10 size(sxk) == [nHorMeasurement,nSample]
11
12 %% Zeropad sxk
13 sxkPadded = sxk;
14 if (nFFTxk > size(sxk,1))
15     sxkPadded = padarray(sxkPadded,[floor((nFFTxk-
       size(sxk,1))/2) 0],0,'pre');
16     sxkPadded = padarray(sxkPadded,[ceil((nFFTxk-
       size(sxk,1))/2) 0],0,'post');
17 else
18     nFFTxk = size(sxk,1);
19 end
20 %% Obtain SkXk
21 SkXk = fftshift(fft(conj(sxkPadded),nFFTxk,1),1);
22
23 %% Define Some Parameters
24 c = 299792458; % m/s
25 k = reshape(2*pi*f/c,1,[]); % Wavenumber Vector
26
27 kSx = 2*pi/(xStep);
28 kX = reshape(linspace(-kSx/2,kSx/2,nFFTxk),[],1);
29
30 % Consider only visible spectrum of kZ
31 kZU = reshape(linspace(0,2*max(k),nFFTkz),1,[]);
32

```

```

33 KU = 1/2 * sqrt(kX.^2 + kZU.^2);
34
35 %% Interpolate (kX,k) -> (kX,kZ)
36 SkXkZ = zeros(size(KU));
37 for ii = 1:size(KU,1) % Replace with parfor to
    increase speed
38     SkXkZ(ii,:) = interp1(k(:),SkXk(ii,:),KU(ii,:), "
        v5cubic");
39 end
40 % Works with: linear,nearest,next,previous,v5cubic
41 SkXkZ(isnan(SkXkZ)) = 0;
42
43 %% Recovered Reflectivity Function
44 pxz = ifft2(SkXkZ,nFFTxk,nFFtz);

```

### III. 2D-Rectilinear SISO-SAR System Model for 3D Holographic Image Reconstruction with Stolt Interpolation

The system model for 3D image reconstruction is as follows. First, equation (8) is the received signal model.

$$s(y, x, k) = \iiint \frac{p(y, x, z)}{R^2} e^{-j2kR} dx dy dz \quad (8)$$

$$R = \sqrt{(x - x')^2 + (y - y')^2 + (z - z')^2} \quad (9)$$

By similar wavefront decomposition and Fourier analysis, the 3D holographic image reconstruction algorithm can be consolidated by equation (10).

$$p(y, x, z) = IFT_{3D}^{(k_y, k_x, k_z)} \left[ Stolt^{(k)} \left( FT_{2D}^{(y, x)} [s^*(y, x, z)] \right) \right] \quad (10)$$

#### A. 2D-Rectilinear SISO-SAR 3D-Image Reconstruction and Implementation of Stolt Interpolation

Expanding the derivation provided in section II, the single-input-single-output synthetic aperture radar (SISO-SAR) reconstruction algorithm is described by equations (4),(5),(6). Where equation (5) is the Stolt interpolation step, migrating  $(k_x, k)$  to  $(k_x, k_z)$  by the relation given in equation (7), derived from the dispersion relation of electromagnetic (EM) waves.

$$S(k_y, k_x, k) = FT_{(y,x)}[s^*(y, x, k)] \quad (11)$$

$$\begin{aligned} \hat{S}(k_y, k_x, k_z) &= Stolt[S(k_y, k_x, k)] \\ &= S(k_y, k_x, k) \Big|_{k=\frac{1}{2}\sqrt{k_y^2+k_x^2+k_z^2}} \end{aligned} \quad (12)$$

$$p(y, x, z) = IFT_{(k_y, k_x, k_z)}[\hat{S}(k_y, k_x, k_z)] \quad (13)$$

$$(2k)^2 = k_y^2 + k_x^2 + k_z^2 \quad (14)$$

Without the Stolt interpolation step in equation (12), the 3D image is not resolvable in the  $(y, x, z)$  plane.

#### B. MATLAB Algorithm for 3D-Image Reconstruction using Stolt Interpolation

```

1 function StoltInterpolation3Dk_kz(syxx, f, xStep, yStep
    , nSample, nFFTyxk, nFFtz)
2 % 3D Stolt Interpolation from k-kZ
3 % f: Frequency Vector
4 % nHorMeasurement: # of Horizontal Measurements
5 % nVerMeasurement: # of Vertical Measurements
6 % xStep: Horizontal Step Size in m

```

```

7 % yStep: Vertical Step Size in m
8 % nSample: Number of Samples
9 % nFFTyxk: FFT size for x and k domains
10 % nFFtz: FFT size for z domain
11 % Maintain syxx = s(x, k)
12 % size(syxx) == [nVerMeasurement, nHorMeasurement,
    nSample]
13
14 %% Zeropad syxx
15 syxxPadded = syxx;
16 if (nFFTyxk > size(syxx,1))
17     syxxPadded = padarray(syxxPadded, [floor((nFFTyxk
        -size(syxx,1))/2) 0], 0, 'pre');
18     syxxPadded = padarray(syxxPadded, [ceil((nFFTyxk-
        size(syxx,1))/2) 0], 0, 'post');
19 else
20     nFFTyxk = size(syxx,1);
21 end
22 if (nFFTyxk > size(syxx,2))
23     syxxPadded = padarray(syxxPadded, [0 floor((
        nFFTyxk-size(syxx,2))/2)], 0, 'pre');
24     syxxPadded = padarray(syxxPadded, [0 ceil((
        nFFTyxk-size(syxx,2))/2)], 0, 'post');
25 else
26     nFFTyxk = size(syxx,2);
27 end
28
29 %% Obtain SkYkXk
30 SkYkXk = fftshift(fftshift(fft(conj(syxxPadded),
    nFFTyxk,1), nFFTyxk,2),1),2);
31
32 %% Define Some Parameters
33 c = 299792458; % m/s
34 k = reshape(2*pi*f/c,1,1,[]); % Wavenumber Vector
35
36 kSx = 2*pi/(xStep);
37 kX = reshape(linspace(-kSx/2, kSx/2, nFFTyxk),1,[]);
38
39 kSy = 2*pi/(yStep);
40 kY = reshape(linspace(-kSy/2, kSy/2, nFFTyxk),[],1);
41
42 % Consider only visible spectrum of kZ
43 kZU = reshape(linspace(0, 2*max(k), nFFtz),1,1,[]);
44
45 KU = 1/2 * sqrt(kY.^2 + kX.^2 + kZU.^2);
46
47 %% Interpolate (kY,kX,k) -> (kY,kX,kZ)
48 SkYkXkZ = zeros(size(KU));
49 sizeKU2 = size(KU,2); % Necessary for parfor
50 for ii = 1:size(KU,1) % Replace with parfor to
    increase speed
51     for jj = 1:sizeKU2
52         SkYkXkZ(ii,jj,:) = interp1(k(:), squeeze(
            SkYkXk(ii,jj,:)), squeeze(KU(ii,jj,:)), "
                v5cubic");
53     end
54 end
55 % Works with: linear,nearest,next,previous,v5cubic
56 SkYkXkZ(isnan(SkYkXkZ)) = 0;
57
58 %% Recovered Reflectivity Function
59 pyxz = ifftshift(ifftn(SkYkXkZ, [nFFTyxk, nFFTyxk,
    nFFtz]),1);
60 % I am not entirely sure why I have to do an
    ifftshift across the first dimension here

```

### IV. 1D-Cylindrical SISO-CSAR System Model for 2D Image Reconstruction by Stolt Interpolation

Using the same coordinate system described in sections II-III and the basic near-field cylindrical setup, the return signal can be modeled by equation (15).

$$s(\theta, k) = \iint \frac{p(x, z)}{R^2} e^{j2kR} dx dz \quad (15)$$

$$R = \sqrt{(R_0 \cos \theta - x)^2 + (R_0 \sin \theta - z)^2} \quad (16)$$

Where  $R_0$  is the radial distance from the center of the rotator to the radar transceiver.

#### A. 1D-Cylindrical SISO-CSAR 2D-Image Reconstruction and Implementation of Stolt Interpolation

The algorithm for recovering the reflectivity function of the target scene is described by the following derivation.

First, the exponential term in equation (15,  $s(\theta, k)$ ) can be approximated by:

$$e^{j2k\sqrt{(R_0 \cos \theta - x)^2 + (R_0 \sin \theta - z)^2}} \approx \int_{-\pi}^{\pi} e^{j2k(\cos \alpha (R_0 \cos \theta - x) + \sin \alpha (R_0 \sin \theta - z))} d\alpha \quad (17)$$

Substituting into equation (15,  $s(\theta, k)$ ) neglecting amplitude terms.

$$s(\theta, k) = \iint p(x, z) \times \int_{-\pi}^{\pi} e^{j2k(\cos \alpha (R_0 \cos \theta - x) + \sin \alpha (R_0 \sin \theta - z))} d\alpha dx dz \quad (18)$$

$$s(\theta, k) = \int_{-\pi}^{\pi} \iint p(x, z) e^{-j2k(x \cos \alpha + y \sin \alpha)} dx dz \times e^{j2kR_0 \cos(\theta - \alpha)} d\alpha \quad (19)$$

Defining the Fourier relation:

$$p(\alpha, k) \triangleq \iint p(x, z) e^{-j2k(x \cos \alpha + y \sin \alpha)} dx dz \quad (20)$$

Now

$$s(\theta, k) = \int_{-\pi}^{\pi} p(\alpha, k) e^{j2kR_0 \cos(\theta - \alpha)} d\alpha \quad (21)$$

It is obvious that equation (21) represents a circular convolution with respect to  $\theta$ , denoted by  $\otimes_{\theta}$ .

$$h(\theta, k) = e^{j2kR_0 \cos \theta} \quad (22)$$

$$s(\theta, k) = p(\theta, k) \otimes_{\theta} h(\theta, k) \quad (23)$$

Taking the Fourier transform on both sides of equation (23) allows for multiplication in the Fourier domain to replace convolution spatial angular. Note that  $\theta$  and  $\Theta$  are conjugate variables of the Fourier transform.

$$S(\Theta, k) = P(\Theta, k) H(\Theta, k) \quad (24)$$

$$P(\Theta, k) = \frac{S(\Theta, k)}{H(\Theta, k)} \quad (25)$$

$$p(\theta, k) = IFT_{1D}^{(\Theta)} \left[ \frac{S(\Theta, k)}{H(\Theta, k)} \right] \quad (26)$$

By the Fourier relation in equation (20),  $p(x, z)$  can be recovered from  $p(\theta, k)$  by:

$$p(x, z) \triangleq \iint p(\theta, k) e^{-j2k(x \cos \theta + y \sin \theta)} d\theta dk \quad (27)$$

## V. 2D-Cylindrical SISO-CSAR System Model for 3D Holographic Image Reconstruction by Stolt Interpolation

#### A. 2D-Cylindrical SISO-CSAR 3D-Image Reconstruction and Implementation of Stolt Interpolation

The received signal is of the form given by equation (15).

$$s(\theta, y, k) = \iiint \frac{p(x, y, z)}{R^2} e^{-j2kR} dx dy dz \quad (28)$$

$$R = \sqrt{(R_0 \cos \theta - x)^2 + (R_0 \sin \theta - y)^2 + (z' - z)^2} \quad (29)$$

Where  $R_0$  is the scanning radius,  $\theta$  is the scanning angle, and  $y'$  is the scanning height, with the target and scanning domains coinciding.  $p(x, y, z)$  is the complex reflectivity function of the target scene.

Inverting equation (15) to solve for  $p(x, y, z)$  requires decomposing the exponential term representing a spherical-wave into a superposition of plane-wave components using some Fourier-based techniques described in [REF].

Perform a 2D Fourier transform across the  $\theta$  and  $z$  dimensions.

$$S(\Theta, k, k_z) = FT_{2D}^{(\theta, z)}[s(\theta, k, z)] \quad (30)$$

Compute the phase term described in equation (18).

$$k_{\Theta} = \sqrt{4(k_x^2 + k_y^2)R_0^2 - \Theta^2} \quad (31)$$

Multiply the phase term and perform an inverse Fourier transform across the  $\Theta$  domain.

$$\hat{P}(\theta, k, k_z) = IFT_{1D}^{(\Theta)}[S(\Theta, k, k_z) e^{-jk_{\Theta}}] \quad (32)$$

Make the following substitutions:

$$P(k_x, k_y, k_z) = \hat{P}(\theta, k, k_z) \Big|_{\theta = \tan^{-1}(\frac{k_y}{k_x}), k = \frac{1}{2} \sqrt{k_x^2 + k_y^2 + k_z^2}} \quad (33)$$