Parameter estimation and imaging of moving targets in bistatic synthetic aperture radar

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Abstract. In high-resolution bistatic synthetic aperture radar (SAR) systems, parameter estimation is essential to moving target imaging quality. However, precise parameters are difficult to obtain without priori information due to the relative along-track and across-track velocities between the moving target and platforms that change with time. A parameter estimation and imaging approach for moving targets is proposed. First, slant range and relative velocities expression are deduced based on the geometry of bistatic SAR model with one stationary configuration. Then, range curvature term are compensated skillfully by fitting the range-compressed curve in two-dimensional time domain, meanwhile, the initial estimated range walk slope can be achieved. Finally, precise Doppler centroid is estimated through searching for the maximum contrast with folding search algorithm, which is giving consideration to both searching precision and computational complexity. Thus, the proposed algorithm provides an effective way for parameter estimation and imaging of moving target without prior information and interpolation operation. Experimental results show the effectiveness of the proposed method. © The Authors. Published by SPIE under a Creative Commons Attribution 3.0 Unported License. Distribution or reproduction of this work in whole or in part requires full attribution of the original publication, including its DOI. [DOI: 10.1117/1.JRS.10.015018]

Keywords: bistatic synthetic aperture radar; parameter estimation; curve fitting; maximum contrast; folding search algorithm.

1 Introduction

In bistatic SAR data acquisition process, beam center offset would reduce signal-to-noise ratio, increase main lobe width and change focusing position, which are caused by motion error of platforms and beam pointing error of antennas. Therefore, accurate Doppler centroid estimation is the premise of high-quality moving target imaging. In general, the Doppler centroid estimation processing includes the base-band Doppler centroid estimation and the Doppler ambiguity number assessing, in which a minor error range is allowed when estimating the base-band Doppler centroid, while the Doppler ambiguity number is required to be completely consistent with the theoretical value.

Presently, several Doppler centroid estimation techniques have been proposed to deal with the Doppler ambiguity problem. By applying an average cross correlation coefficient (ACCC) to signals in range frequency domain, wavelength diversity algorithm (WDA)1 obtains absolute Doppler centroid according to the relation between Doppler centroid and range frequency. Multilook cross correlation (MLCC)2 and multilook beat frequency (MLBF)3 resolve Doppler ambiguity through decomposing the original signal into two subsignals with different carrier frequencies. MLCC achieves Doppler centroid by calculating the phase difference of ACCC between subsignals, while the latter method gets that directly via computing the frequency difference of subsignals. Among the mentioned three algorithms, WDA and MLCC are suitable for low-contrast scene, and MLBF is propitious to high-contrast scene. Multiple PRF4–6 avoids Doppler ambiguity based on a set of PRFs, whose disadvantages lie in the complexity of system design. Therefore, this method is employed mostly in ScanSAR mode.

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Radon transform\textsuperscript{7–9} has been proved to be a high-precision method to estimate Doppler centroid in medium and high-contrast scenes, which give accurate Doppler centroid estimation based on the linear feature of the range-compressed signal. However, the estimation accuracy deteriorates when the range curvature cannot be neglected. Moreover, it is confronted with great computational complexity. In this paper, a new parameter estimation and imaging approach for moving targets is proposed. First, the expression of slant range is derived and approximated based on the bistatic SAR model, from which we know the along-track and across-track velocity cannot be solved without prior information. Then, curve fitting is applied to range-compressed signal in two-dimensional (2-D) time domain. On this basis, precise Doppler center is estimated using high-efficient folding search algorithm. The proposed algorithm does not involve interpolation operation and provides an effective way for parameter estimation and imaging in bistatic SAR.

2 Bistatic Synthetic Aperture Radar with One Stationary Configuration

2.1 Geometry Model

The geometry model of bistatic SAR is shown in Fig. 1. Assume that the transmitter is fixed, and the receiver moves along the $x$-axis direction and works at side looking mode with velocity $v_R$. $R_{T0}$ and $R_{R0}$ are the slant range from the transmitter and receiver to ground moving target at $t_a = 0$, while $R_{T1}$ and $R_{R1}$ stand for the instantaneous slant ranges of the transmitter and receiver from the target. $\theta_T$ and $\theta_R$ are the squint angles from the two platforms to the target, $t_a$ is the slow time. $v_{Ta}$, $v_{Ra}$, $v_{Tr}$, and $v_{Rr}$ denote the relative lateral and radial velocities between the moving target and platforms, respectively. The velocity of the moving target $v$ can be decomposed into $v_{Ta}$ and $v_{Tr}$. Thus, the instantaneous slant range $R(t_a)$ is given by

\[
R(t_a) = \sqrt{(R_{T0} - v_{Tr}t_a)^2 + (v_{Ta}t_a)^2 + 2(R_{T0} - v_{Tr}t_a)(v_{Ta}t_a) \sin \theta_T} \\
+ \sqrt{(R_{R0} - v_{Rr}t_a)^2 + (v_{Ra}t_a)^2}.
\]  

(1)

According to the geometry of bistatic SAR model with one stationary configuration, we have

\[
v / \sin(\pi/2 - \theta_{T1}) = v_{Ta} / \sin(\pi/2 + \theta_{T1} - \theta_1) = \tilde{v}_{Ta} / \sin \theta_1.
\]

Thus, $v_{Ta}$, $v_{Ra}$, $v_{Tr}$, and $v_{Rr}$ is given by

\[
\begin{align*}
  v_{Tr} &= \tilde{v}_{Tr} \cos \epsilon_T = v \sin \theta_1 \cos \epsilon_T / \sin(\pi/2 - \theta_{T1}) \\
  v_{Ta} &= v \sin(\pi/2 + \theta_{T1} - \theta_1) / \sin(\pi/2 - \theta_{T1}) \\
\end{align*}
\]

(2)

Similarly, we have

![Fig. 1 Geometry model.](https://www.spiedigitallibrary.org/journals/Journal-of-Applied-Remote-Sensing on 17 Feb 2022 Terms of Use: https://www.spiedigitallibrary.org/terms-of-use)
\[
\begin{align*}
\begin{cases}
\vRr = v \cos \theta_1 \cos \epsilon_R, \\
\vRa = v \sin \theta_1,
\end{cases}
\end{align*}
\]
(3)
where \(\theta_1\) and \(\theta_2\) represent the angles between \(v_{Ta}\), \(v_{Ra}\), and \(v\), respectively. \(\epsilon_T\) and \(\epsilon_R\) are the grazing angles of transmitter and receiver. Obviously, \(\theta_1\), \(\theta_2\), and \(\theta_{T1}\) satisfies
\[
\begin{align*}
\begin{cases}
\theta_1 + \theta_2 + (\pi/2 - \theta_{T1}) = \pi, \\
\theta_{T1} = \sin^{-1}(\sin \theta_T / \cos \epsilon_T).
\end{cases}
\end{align*}
\]
(4)
From Eqs. (2)–(4), one obtains six equations with eight variables, i.e., \(v_{Ta}, v_{Ra}, v_{Tr}, v_{Ra}, \theta_1, \theta_2, \theta_{T1},\) and \(v\). In order to calculate these variables, range walk and range curvature term are estimated in the following procedures, both of which are the functions of the unknown parameters. In this way, the moving target-related variables can be achieved.

### 2.2 Analysis of Range Migration

Ignoring higher-order terms of phase, the Tailor series expansion coefficients of \(R(t_a)\) at the beam center crossing time is given by\(^{10-11}\)
\[
R(t_a) \approx R_{T0} + R_{R0} - v_{Ta} t_a + v_{Ta} \sin \theta_T - v_{Ra} t_a + \frac{(v_{Ta} \cos \theta_T)^2}{2R_{T0}} + \frac{(v_{Ra} - v_{Ra} t_a)^2}{2R_{R0}}.
\]
(5)
According to the analysis of range migration given in Ref. 12, the Doppler centroid can be easily deduced when the range walk is far greater than the range curvature, and the quadratic term in Eq. (5) can be neglected. That is
\[
|v_{Ta} t_a + v_{Ra} t_a - v_{Ta} \sin \theta_T| \gg \frac{(v_{Ta} \cos \theta_T)^2}{2R_{T0}} + \frac{(v_{Ra} - v_{Ra} t_a)^2}{2R_{R0}}.
\]
However, the above assumption would no longer hold in some cases, e.g., the platform height is low, the beam width is broad, and the receiving platform velocity is high. Thus, the Doppler centroid cannot be estimated accurately using the algorithm in Ref. 12.

### 3 Proposed Parameter Estimation Algorithm

In this section, at first we introduce curve fitting method to achieve the linear and quadratic coefficients of bistatic slant range. Then precise Doppler centroid can be obtained effectively through contrast-based folding search algorithm. Finally, moving target can be well focused using the estimated parameters.

#### 3.1 Curve Fitting Method

Suppose a linear FM signal is transmitted, then the received signal after range compression is given by
\[
s(f_r, t_a) = w_a(t_a) w_r(f_r) \exp \left\{ -j \frac{2\pi(f_0 + f_c)}{c} \left[ \frac{R_{T0} + R_{R0} - v_{Ta} t_a - v_{Ra} t_a + v_{Ta} \sin \theta_T}{2R_{T0}} + \frac{(v_{Ra} - v_{Ra} t_a)^2}{2R_{R0}} \right] \right\},
\]
(6)
where \(w_a(\cdot)\) and \(w_r(\cdot)\) are the range and azimuth envelopes, respectively.

Due to the separated transmitting and receiving antennas, relative along-track and across-track velocities between the moving target and platforms change with time, which lead to the linear and quadratic terms of \(t_a\) in slant range, as is indicated in Eq. (6). Obviously, range curvature compensation is the premise of Doppler centroid estimation, but the relative velocities are unknown and cannot be obtained directly from the echo data with bistatic configuration. Therefore, we could not get range curvature compensation factor without priori
information. Accordingly, the existent of quadratic term makes the Doppler centroid estimation method\textsuperscript{12} not applicable.

In this paper, considering the bending property of the signal in 2-D time domain, the bending degree stands for the quadratic coefficient, range curvature term is compensated by curve fitting method. Meanwhile, the linear coefficient is used for the rough estimation of Doppler centroid. In order to improve the fitting accuracy, the curve is refined before fitting operation. Assuming that the linear and quadratic coefficients achieved by curve fitting method are \( T_1 \) and \( T_2 \), respectively, where \( T_1 \) denotes the Doppler centroid offset and \( T_2 \) represents the defocusing degree. In the process of parameter estimation, range curvature term is first compensated, and then range walk slope is estimated on the basis of this procedure. Theoretically, \( T_1 = \left[ \frac{dR(t_a)}{dt_a} \right]_{t_a = 0} \), \( T_2 = \left[ \frac{d^2 R(t_a)}{dt_a^2} \right]_{t_a = 0} \). Range curvature compensation factor is given by

\[
H_1(f_r, t_a) = \exp \left[ \frac{2\pi}{c} (f_c + f_r) T_2 t_a \right],
\]

\[
H_2 = \exp \left[ \frac{2\pi}{c} (f_c + f_r) T_1 t_a \right].
\]

### 3.2 Definition of Contrast

Assume that the range curvature term has been compensated, then the relationship among the range walk term, the slope \( K \) and the Doppler centroid \( f_{dc} \) can be written as

\[
f_{dc} = -\frac{K}{\lambda} = (v_{Tr} + v_{Rr} - v_{Ta} \sin \theta_T)/\lambda \approx -T_1/\lambda.
\]

Initial Doppler centroid correction factor is given by

\[
C(v, \theta_T) = \sqrt{\frac{1}{N_r} \sum_{n=0}^{N_r-1}[I^2(n) - \frac{1}{N_r} \sum_{n=0}^{N_r-1} I^2(n)]^2},
\]

where \( I(n) \) is defined as

\[
I(n) = \frac{|x(n)|}{\|x\|}, \quad \|x\| = \sum_{n=0}^{N_r-1} |x(n)|.
\]

From the above definition, we know contrast reflects the energy accumulation degree. That is, if the contrast is large, the accumulated energy along azimuth direction gathers near one range cell. Otherwise, small contrast means the energy gathers in several or more range cells. In this case, defocusing phenomena will appear after range compression. As a result, the exact Doppler centroid estimation is equivalent to searching the maximum contrast.

### 3.3 Folding Search Algorithm

The conventional ways for searching for range walk slope are based on one fixed step. These methods may result in many shortcomings. On the one hand, supposing that the initial estimated slope deviation has the same sign as the searching step, the estimated slope error will increase with the searching process, and the optimal slope cannot be obtained. On the other hand, when
the initial slope error is large or the required Doppler centroid accuracy is high, conventional searching methods will face huge computational complexity. In view of the above problems, a folding search algorithm based on maximum contrast is presented in this section, the main steps are as follows:

- Step 1: Initializing step $\Delta K$ and setting $\Delta K_0$ as the termination step in folding search algorithm.
- Step 2: Equalizing range curvature term using the curve fitting method.
- Step 3: Initializing range walk compensation factor with the fitting coefficient, and calculating the initial contrast $C_1$ according to Eq. (10).
- Step 4: Updating the contrast $C_k (k = 2, 3, 4\ldots)$. If $C_k \geq C_{k-1}$, continue this iteration process. Otherwise, turn to step 5.
- Step 5: Outputting the slope and corresponding Doppler centroid, if the step is less than $\Delta K_0$.

Analysis of computational complexity: Assuming that the initial step error is $\Delta E$, the computational complexity of folding search algorithm based on maximum contrast is $\log_2(\Delta E/\Delta K_0)$, while the conventional algorithm based on fixed step is $\Delta E/\Delta K_0$. With the increase of initial step error and enhancement of step accuracy, the computational complexity of the proposed algorithm is significantly reduced compared to that of conventional algorithms.

### 3.4 Moving Target Imaging Procedure

At this stage, moving target can be imaged by means of the estimated linear and quadratic coefficients of bistatic slant range. Assuming that the ultimate range walk slope using folding search algorithm is $T_3$, then the final Doppler centroid correction factor is given by

$$H_3(f_r, t_a) = \exp\left[j\frac{2\pi(f_r + f_c)}{c} T_3 t_a\right].$$  \hspace{1cm} (11)

Multiplying Eq. (6) with Eq. (11), and transforming the results into 2-D frequency domain yields

$$S(f_r, f_a) = \exp\left(-j\left\{\frac{2\pi(f_r + f_c)}{c} (R_{T0} + R_{R0}) \frac{\pi cf_a^2}{2(f_r + f_c) (\frac{(\nu_{\theta_0} t_0 \cos \theta_T^0)^2}{2R_{T0}} + \frac{(V_{\theta_0} t_0 \nu_{\theta_0} t_0)^2}{2R_{R0}})}\right\}\right).$$  \hspace{1cm} (12)

In Eq. (12), as the linear term of $t_a$ mainly affects the envelope position, so it cannot be reflected in phase term. Finally, the azimuth compression factor is given by

$$H_4(f_r, f_a) = \exp\left[-j\frac{\pi cf_a^2}{2(f_r + f_c) T_2}\right].$$  \hspace{1cm} (13)

After transforming the azimuth-compressed signal into 2-D time domain, moving target imaging process with bistatic configuration is accomplished. The final imaging quality depends on the accuracy of fitting coefficients as well as the termination step in folding search algorithm. Figure 2 shows the flow chart of the proposed parameter estimation algorithm.

### 4 Simulation Results

In this section, simulation results of different algorithms are compared to evaluate the effectiveness of the proposed one. Assume four moving targets with the same reflectivity coefficients are located in the scene. The positions of target 1, target 2, target 3, and target 4 are given by $(0, 400)$ m, $(0, -800)$ m, $(62.5, 0)$ m, and $(-375, 0)$ m, respectively. Among the moving targets, target 1 and target 4 have the same velocity 15 m/s and the same angle 12 deg between target velocity
and y-axis, while target 2 and target 3 have the same velocity 17 m/s and the same angle 10 deg between target velocity and y-axis. Other simulation parameters are shown in Table 1.

### 4.1 Simulations for Multiple Moving Targets

In Sec. 3, we apply curve fitting method to the range-compressed signal. In this way, the coefficients that represent the initial range walk slope and range curvature term are obtained. Figure 3(a) shows the range-compressed signal in 2-D time domain. Figure 3(b) denotes the initial range walk slope corrected signal, which indicate that different targets energy are gathered

![Flow chart of the proposed algorithm](https://www.spiedigitallibrary.org/journals/Journal-of-Applied-Remote-Sensing)

**Fig. 2** Flow chart of the proposed algorithm.

| Parameter                        | Value        |
|----------------------------------|--------------|
| Coordinates of transmitter      | (-3, -2, 1) km |
| Coordinates of receiver         | (-15, 0, 1) km |
| Bandwidth                        | 95 MHz       |
| Beam width of transmitter        | 10 deg       |
| Carrier frequency                | 10 GHz       |
| PRF                              | 400 Hz       |
| Synthetic aperture time          | 10.50        |
| Receiver velocity                | 100 m/s      |
| Initial searching precision      | 0.2 m/s      |
| Beam width of receiver           | 4 deg        |
| Terminal searching precision     | 0.02 m/s     |

**Table 1** Simulation parameters.
into different range intervals. As a result, folding search algorithm can be employed to conduct energy accumulation for multiple targets. Here, the slope $K$ of the four targets in Fig. 3(b) are $-177$, $-179.5$, $-183$, and $-191$. Figures 3(c) and 3(d) are the range corrected signal using paper12 algorithm and the proposed algorithm, respectively. In order to evaluate the effectiveness of our method, Figs. 3(e) and 3(f) give the amplified curve of target 3, in which the curve in Fig. 3(e) becomes a straight line, as is shown in Fig. 3(f). This phenomenon results from the fact that the range curvature term has been compensated with the proposed method. Figures 3(g), 3(h), 3(i), and 3(j) show the comparison of the normalized contrast between the folding search algorithm and paper12 algorithm. From the comparison between Figs. 3(i) and 3(j), we know the normalized contrast in Fig. 3(i) focuses on a finite range interval because the range curvature term is ignored with paper12 algorithm, while the proposed algorithm avoids this problem, and the energy gathers more intensively. Obviously, the latter is more beneficial for high-accuracy Doppler centroid estimation.

The estimated Doppler centroid error and computing time using different algorithms are listed in Table 2. By adopting curve fitting method and high-efficient search algorithm, not only the performance but the computation complexity is taken into account in the proposed algorithm even though insufficient prior information is provided, which is consistent with

Fig. 3 Simulations of multiple moving targets: (a) range-compressed data, (b) initial range walk slope corrected data, (c) range corrected data using paper12 algorithm, (d) range corrected data using the proposed algorithm, (e) target 3 without range curvature correction, (f) target 3 after range curvature correction, (g) energy accumulation results using paper12 algorithm, (h) energy accumulation results using the proposed algorithm, (i) amplified target 3 from (g), and (j) amplified target 3 from (h).
the theoretical analysis. On the other hand, velocity errors of different targets can be achieved simultaneously from the derivation in Sec. 2. Parameter estimation error with the proposed algorithm is mainly caused by the approximation of double square-root term equation and the deficiency of compensation factor precision.

### 4.2 The Influence of Parameter Error on Estimation Results

According to the definition of contrast, it is influenced by range walk slope and range curvature term. From Eq. (10), we know these two parameters are composed of the target velocity and the transmitting angle essentially. As a result, the target velocity and the transmitting angle are the direct factor to imaging quality. Based on the parameters of target 3, the errors of linear and quadratic coefficient caused by incorrect target velocity as well as inaccurate transmitting angle are shown in Figs. 4(a) and 4(b), respectively. It is obvious that the range walk slope is more sensitive to the parameter error compared with range curvature term. Therefore, range curvature term can be estimated simply using curve fitting method, while the assessment of range walk slope requires not only curve fitting method but the secondary estimation based on folding search algorithm.

### 4.3 Comparison of Imaging Results

In general, the final purpose of parameter estimation algorithms is to detect and image moving targets. Here, the parameters of target 3 are still employed. Comparison of imaging results using different algorithms is shown in Fig. 5, among which (a), (b), and (c) are obtained by radon transform, paper\textsuperscript{12} method, and the proposed algorithm, respectively. The range interval and step size in radon transform are set to 20 deg and 0.02 deg. Since the previous two algorithms do not give consideration to the influence of range curvature term, precise azimuth compression factor cannot be achieved. Also, defocusing phenomenon will appear after azimuth compression.\textsuperscript{15} Suppose that the azimuth compression function with a certain error is given by

\[
H_\delta(f_r, f_a) = \exp \left[ -j \frac{\delta \pi c f_0^2}{2(f_r + f_c)T_2} \right].
\]  

\[ (14) \]

### Table 2 Comparisons of estimation results.

|                      | Doppler centroid errors of target 1, target 2, target 3, and target 4 (Hz) | Velocity errors of target 1, target 2, target 3, and target 4 (m/s) | Computation time (s) |
|----------------------|--------------------------------------------------------------------------------|-----------------------------------------------------------------|---------------------|
| Radon transform      | 3.60, 7.05, 4.34, 6.88                                                        | 0.5040, 1.2690, 0.7812, 0.9632                                    | 120.80              |
| Paper\textsuperscript{12} method | 2.48, 4.31, 2.82, 4.46                                                       | 0.3472, 0.7758, 0.5076, 0.6244                                    | 61.33               |
| Proposed method      | 0.46, 0.71, 0.42, 0.67                                                        | 0.0644, 0.1278, 0.0756, 0.0938                                    | 28.01               |

**Fig. 4** Simulation results of estimation error caused by velocity and squint angle: (a) range walk slope error and (b) range curvature error.
where the defocusing factor $\delta (0 \leq \delta \leq 1)$ is set to 0.98. With the increase of $\delta$, azimuth focusing performance will be improved. It can be seen from Fig. 5 that even though $\delta$ is as high as 0.98, desired imaging result cannot be achieved. Compared with the previous two algorithms, the focusing accuracy has been significantly improved using the proposed algorithm. The simulation results in this section suggest that the range curvature term is essential when the platform height is low, the beam width is broad, or the receiving platform velocity is high, which are in agreement with the theoretical analysis.

5 Conclusion

For moving targets imaging procedure with bistatic SAR configuration, the relative along-track and across-track velocities between the moving target and platforms are required. Therefore, four independent equations need to be constructed in theory. However, we could not achieve enough related information from the echo data without prior knowledge. In this paper, a parameter estimation and imaging algorithm for moving targets is presented. On the one hand, the expression of slant range and relative velocities are derived based on the geometry of bistatic SAR model with one stationary configuration. On the other hand, range curvature term is compensated with polynomial fitting method, and the precise Doppler centroid is obtained using folding search algorithm. Experimental results show that the proposed algorithm could give consideration to both the performance and computational complexity. In addition, this algorithm can also be employed in bistatic SAR system with arbitrary configuration.

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