Geometrical Considerations for Terrain Mapping in SBR Applications

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SUMMARY & CONCLUSIONS

Knowledge aided signal processing calls for the use of existing information such as digitized image maps to further suppress clutter in a typical radar detection operation. In the context of space based radar (SBR) applications, given the SBR parameters and the look angles the first problem is to translate these parameters into a data point on a map. This is realized by converting these parameters into conventional latitude and longitude and then using the corresponding image intensity to estimate the clutter covariance matrix. Using this approach, it is possible to pre-compute the desired weight vector for a location of interest on earth that the SBR is assigned to visit. As a result, using the measured data together with the weight vector, it is possible to test for target detection hypothesis accurately at the time of visit by the SBR.

1. INTRODUCTION

An interesting problem in knowledge aided Space Time Adaptive Processing (KA-STAP) is how to make use of traditional image maps to cancel out clutter and interference components while looking for a target. This approach has applications in any `search and detect’ operation including airborne and space based radar (SBR). In what follows we shall illustrate our approach for the SBR applications. The problem in this context is to translate the local geometry to the SBR platform so that the map intensities can be translated into meaningful clutter covariance estimates.

In its general setting two points with respective latitudes and longitudes \((\alpha_1, \beta_1)\) and \((\alpha_2, \beta_2)\) are chosen on a map, the first indicating the location of the transmitter/receiver such as an SBR and the second indicating the point of interest (see Fig. 1). In addition, given the SBR height \(H\) and its orbit inclination angle \(\eta\) at the equator, the first step is to determine all other desired parameters such as SBR elevation angle \(\theta_e\), SBR array azimuth angle \(\theta_{az}\), the true ground range \(R\), grazing angle \(\psi\), the array footprint on earth, range-foldover points \(R_1, R_2, \ldots\) on the ground, and finally the correction factor on the azimuth due to the earth’s rotation (crab angle) (see Fig. 2 and Fig. 3) [1, 2, 3].

Fig. 1: SBR and point of interest co-ordinates

Fig. 2: SBR-Earth Geometry
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Standard Form 298 (Rev. 8-98)  
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Fig. 3: SBR 2-D Array Gain Pattern and Range Foldovers

The SBR Parameters are listed below (see [4] for derivations):

\[ R = R_e \cos^{-1}(\sin \alpha \sin \beta + \cos \alpha \cos \beta \cos(\beta_2 - \beta_1)) \]  
(1)

\[ \alpha = \theta_e = R / R_e \]  
(2)

\[ R_i = \sqrt{R_e^2 + (R_e + H)^2 - 2R_e(R_e + H) \cos(R/R_e)} \]  
(3)

\[ \sin(\pi / 2 + \psi) = \frac{\sin \theta_e}{R_e} = \frac{\sin \theta_{EL}}{R_e} \]  
(4)

\[ \theta_{EL} = \sin^{-1}\left( \frac{R_e \sin(R/R_e)}{R_e} \right) \]  
(5)

\[ \sin \gamma = \cos \alpha \sin(\beta_2 - \beta_1) / \sin \alpha \]  
(6)

\[ \theta_{AZ} = \gamma + \delta - \pi, \]  
(7)

where

\[ \delta = \sin^{-1}\left( \frac{\cos \eta}{\cos \alpha} \right). \]  
(8)

These formulas give the SBR parameters in terms of \((\alpha, \beta), (\alpha', \beta'),\) and the SBR height \(H.\) Using these parameters the SBR array pattern on the ground can be plotted as shown in Fig. 3. Clearly when pointed at \((\alpha, \beta),\) the SBR receives data from all points that are illuminated in Fig. 3, and hence it is necessary to make angular increments both in the elevation and azimuth directions to access the neighbouring points. As a result, it is necessary to develop the inverse formulas when the parameters are appropriately incremented. Such is the case when computing the clutter covariance matrix \( \mathbf{R}_c \) using an image map.

In general the clutter covariance matrix has the form [1, 2, 3],

\[ \mathbf{R}_c = \sum_k P_k G(\theta_k) s(\theta_k, \omega_k) \mathbf{s}^H(\theta_k, \omega_k) \]  
(9)

where \(P_k\) is the intensity of the \(k^{th}\) clutter patch, \(G(\theta_k)\) the corresponding array gain, and \(s(\theta_k, \omega_k)\) is the space-time steering vector associated with the \(k^{th}\) patch. Here \(\theta_k\) refers to both azimuth angle \(\theta_{AZ_k}\) and the elevation angle \(\theta_{EL_k}\) of the point of interest, and in practice are incremented to cover all guard cells used to estimate \( \mathbf{R}_c \) in (9). In what follows inverse formulas for latitude and longitude, given \(\theta_{EL}\) and \(\theta_{AZ}\), are developed.

2. AZIMUTH INCREMENT

From Fig. 4, we have

\[ \theta_{AZ}' = \theta_{AZ} + \Delta \theta_{AZ}, \]  
(10)

where \(\theta_{AZ}\) and \(\theta_{AZ}'\) refer to the azimuth angles corresponding to the point of interest before and after the increment, respectively.

Note that in this case the range remains the same and there is a change in the azimuth angle only. Thus, (7) can be used to compute the new \(\gamma\).

Fig. 4: Azimuth increment

Fig. 5: SBR-point of interest spherical geometry
In the spherical triangle $ABC$ (Fig. 5), two sides and an included angle $\gamma$ are known. This gives [5]

$$\cos c = \sin \alpha_i \cos \alpha + \cos \alpha_i \sin \alpha \cos \gamma$$  \hspace{1cm} (11)

and hence the desired latitude of the new point is given by

$$\alpha_i = \pi / 2 - c.$$  \hspace{1cm} (12)

Once $\alpha_i$ is known, $\beta_i$ can be obtained from the relation [5]

$$\frac{\sin(\beta_i - \beta)}{\sin \alpha} = \frac{\sin \gamma}{\sin c}.$$  \hspace{1cm} (13)

3. ELEVATION INCREMENT

In this case the new elevation angle corresponding to the point of interest is given by (see Fig. 6)

$$\theta_{\text{el}}' = \theta_{\text{el}} + \Delta \theta_{\text{el}},$$  \hspace{1cm} (14)

where $\theta_{\text{el}}$ refers to the elevation angle before the increment.

![Fig. 6: Elevation increment](image)

With the incremented value of the elevation angle, Eq. (4) can be used to determine the grazing angle $\psi$. Thus,

$$\psi = \cos^{-1} \left( \frac{\sin \theta_{\text{el}}}{1 + H / R_e} \right)$$  \hspace{1cm} (15)

and the relation

$$\theta_e = \pi / 2 - \theta_{\text{el}} - \psi$$  \hspace{1cm} (16)

can be used to determine the new range $R$. This gives

$$R = R_e \theta_e = R_e (\pi / 2 - \theta_{\text{el}} - \psi).$$  \hspace{1cm} (17)

Note that $\theta_{\text{el}}$ and $\gamma$ remain unchanged and (11) – (13) can be used to compute $(\alpha_i, \beta_i)$. Knowing the latitude and longitude, the power level can be extracted from the map. This procedure is repeated to cover all guard cells of interest in (9).

Another refinement in modelling the clutter covariance matrix is to introduce the effect of range foldover ambiguities into (9). Range foldover occurs when clutter returns from previously transmitted pulses are coherently combined with the mainbeam at the point of interest, as shown in Fig. 7. The associated range points corresponding to these returns constitute the range ambiguities for the point of interest.

4. RANGE FOLDOVER

The ground plane resolution of the SBR is limited by the pulse frequency and the grazing angle $\psi$. This causes range ambiguities spaced at the limiting resolution $\delta_r$, given by

$$\delta_r = \frac{c T}{2} \sec \psi.$$  \hspace{1cm} (18)

From (18) and for short range regions where the grazing angle $\psi$ is closer to $\pi / 2$, the range resolution is very poor since $\delta_r \to \infty$. For long ranges, the resolution approaches its limiting value $c T / 2$ as $\psi \to 0$.

![Fig. 7: Range Foldover](image)

If $R_e$ represents the slant range at the end of one pulse, then $R_e + c T / 2$ is the new slant range at the end of the next pulse. Let $R + \Delta R$ represent the new range corresponding to the second pulse shadow on the ground. Using Fig. 2, we have

$$(R_e + c T / 2)^2 =$$

$$R_e^2 + (R_e + H)^2 - 2 R_e (R_e + H) \cos \left( \frac{R + \Delta R}{R_e} \right)$$  \hspace{1cm} (19)

or

$$\Delta R = R_e \cos^{-1} \left( \frac{R_e^2 + (R_e + H)^2 - (R_e + c T / 2)^2}{2 R_e (R_e + H)} \right) - R.$$  \hspace{1cm} (20)

Note that $\Delta R$ is a decreasing function of $R$. Moreover when $R$ is relatively small, the distance between the pulse shadows on the ground is large and it decreases as $R$ increases. However for large values of range, $\Delta R$ remains constant at its limiting value $c T / 2$. As a result and as the range increases,
this has the additional effect of packing several range ambiguities within a mainbeam footprint. If we assume \( \Delta R \) to be constant within a mainbeam footprint, then the number of range ambiguities within the mainbeam is given by

\[
N_a = (R_f - R_H)/\Delta R.
\]  

(21)

Otherwise, starting at the heel of the mainbeam located at range \( R_H \), \( \Delta R \), \( \Delta R_1 \), \ldots, are sequentially calculated with the mainbeam footprint until the toe is reached. Thus, the number of range ambiguities \( N_a \) within a mainbeam satisfies the equation

\[
\Delta R_k + \Delta R_{k+1} + \ldots + \Delta R_{k+m} = R_f - R_H.
\]  

(22)

Each range ambiguity contribution to the clutter covariance matrix has the same form as in (9). As a result, the clutter covariance matrix that accounts for range ambiguities has the following form

\[
R_c = \sum_{n=0}^{N_a} \sum_{k} P_{km} G(\theta_{km}) s(\theta_{km}, \omega_{km}) s^H(\theta_{km}, \omega_{km}).
\]  

(23)

Here \( P_{km} \) corresponds to the intensity of the \( k^{th} \) patch from the \( m^{th} \) range foldover. For each range ambiguity in (23), the azimuth increment method described earlier can be applied to compute its clutter covariance matrix contribution.

Finally, the clutter Doppler \( \omega_{km} \) in (23) can be modelled with and without taking earth’s rotation into account. The effect of earth’s rotation tends to modify the clutter Doppler in terms of a crab angle \( \phi_c \) and a crab magnitude \( \rho_c \) affecting its phase and magnitude [2, 3]. With \( T \) representing the transmit pulse period and \( V_p \) the satellite velocity, it can be shown that the clutter Doppler with earth’s rotation for the \( k^{th} \) patch is given by [2, 3]

\[
\omega_{km} = \frac{2V_p T}{\lambda/2} \rho_c \sin \theta_{km} \cos(\theta_{km} + \phi_c).
\]  

(24)

5. TERRAIN MAPPING

Using the method described above to update the latitude and longitude, terrain data from any stand alone image map can be used to custom compute the covariance matrix. For illustration purposes, we have used image intensity data from a world map at http://www.cs.brown.edu/exploratories.

Fig. 8 shows the SBR coverage for two regions, (i) the eastern United States (\( \alpha_i = 34^\circ \), \( \beta_i = -100^\circ \)) and (ii) the Middle East. To compute the clutter covariance matrix for each region, the array patterns and the range foldovers are projected on the surface of the earth using (10)-(17). In addition, the range foldovers are included in this case since we are using a pulse repetition frequency of 500 Hz. For an SBR at a height of 506 km, we have \( R_{max} = 2400 \) km and \( N_a \) is equal to 6 or 7 depending on the target range. By carrying out the latitude/longitude – azimuth/elevation conversion, the terrain of interest can be superimposed on an image map to obtain the true terrain image intensities. These intensity values \( P_{km} \) together with the terrain geometry described earlier can be used to estimate the clutter covariance matrix \( R_c \) for the point of interest with elevation and azimuth angles \( \theta_{km} \) and \( \omega_{km} \). Knowing the SBR array/pulse configuration, if \( s(\theta_{km}, \omega_{km}) \) represents the space-time steering vector for the target location, then

\[
w = R_c^{-1} s(\theta_{km}, \omega_{km})
\]  

(25)

represents the desired weight vector that can be applied to a data sample to determine the presence or absence of a target of interest.

Fig. 9 shows the performance of the optimum weight vector in terms of its ability to perform clutter suppression with and without the presence of range foldover using the Signal to Interference plus Noise (SINR) measure, given by

\[
\text{SINR} = s^H(\theta_{km}, \omega_{km}) R_c^{-1} s(\theta_{km}, \omega_{km}).
\]  

(26)

Their combined effect results in appreciable performance degradation as shown in Fig. 9. In this context, the presented approach gives a tool to evaluate the performance degradation at various points on earth. A preliminary simulation shows the clutter suppression for different points of interest across the United States.
6. EFFECTS OF TERRAIN ON CLUTTER SIMULATION

As previously described, the intensities obtained from an image map can be used when estimating the clutter covariance matrix, $R_c$. To test these effects, multiple simulations were run at different points over the United States within the three regions of interest as shown in Fig. 11. First, the target range and azimuth angle were held constant at 900 km and 88° and the location of the SBR was moved among the three longitudes (-125°, -110°, -95°) and four latitudes (36°, 38°, 40°, 42°). By making use of (23)-(26) the clutter suppression performance at each of these locations can be studied using the output SINR. The results are shown in terms of the Minimum Detectable Velocity (MDV) measure for ground moving target detection. From Fig. 10 MDV is a measure of the ability to detect slow moving targets in the presence of clutter and noise. From there, clearly a sharp clutter ridge provides superior performance. For the sake of comparison, we have set the threshold for MDV at 5 dB below the noise floor as shown in Figs. 9-10. Notice that when earth’s rotation is present MDV is undefined as shown in Fig. 9, indicating the a severe degradation in performance.

The average MDV values for the three regions are shown in Fig. 11. The MDV variations at different latitudes within each region are shown in Fig. 12. Notice that the Midwest region gave an average MDV of 23.2749 m/s whereas the West region gave an average MDV of 24.3179 m/s (See Fig. 11). The East region MDV was also slightly higher than the Midwest. The Midwest is generally flatter and more uniform than the other regions so one would expect lower MDV’s due to the “homogeneous” nature of the background terrain.

Similarly, we repeated the above simulations while keeping the latitude constant at 40° and adjusting the azimuth around 90° (See Fig. 13). In this case, MDV increase above the other regions for the Midwest at larger azimuth angles. This could be due to terrain variations such as the southern end of the Rocky Mountains and the Gulf of Mexico. Yet as before, MDV values for the Midwest are lower when the beam covered the plains.
7. CONCLUSIONS

This paper discusses methods for incorporating existing digital maps using SBR-earth geometry to generate realistic clutter covariance nulling schemes for ground moving target detection applications. SBR-earth geometry parameters need to be updated into the latitude-longitude scale to determine the actual image intensities. This approach allows pre-computed optimum weight vectors to be determined that can be used on a single snapshot of measurement to determine the presence or absence of a target. Performance results using this approach across the United States terrain are reported in terms of Minimum Detectable Velocity (MDV) and indicate the clutter suppression efficiency.

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REFERENCES


BIOGRAPHIES

Ke Yong Li is a staff engineer with C & P Technologies, Inc., New Jersey specializing in radar signal processing. His research interest includes Space Time Adaptive Processing (STAP), Optimum Trans-Receiver design technology and space based radar signal processing.

Stephen V. Mangiat is an engineering student-intern of C & P Technologies, Inc. He will receive the bachelors’ degree in electrical and computer engineering from Carnegie Mellon University in June 2005. He has developed Java Applets to illustrate blind echo cancellation algorithms and graphical user interface application for Space Based Radar clutter simulation.

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