

# DEM-BASED SAR PIXEL AREA ESTIMATION FOR ENHANCED GEOCODING REFINEMENT AND RADIOMETRIC NORMALIZATION

Othmar Frey<sup>(1)</sup>, Maurizio Santoro<sup>(2)</sup>, Charles L. Werner<sup>(2)</sup>, and Urs Wegmuller<sup>(2)</sup>

<sup>(1)</sup>*Gamma Remote Sensing AG, Switzerland / Earth Observation & Remote Sensing, ETH Zurich*

<sup>(2)</sup>*Gamma Remote Sensing AG, Switzerland*

*E-mail: frey@gamma-rs.ch*

## ABSTRACT

Precise terrain-corrected georeferencing of SAR images and derived products in range-Doppler coordinates is important with respect to several aspects, such as data interpretation, combination with other geodata products, and transformation of, e.g., terrain heights into SAR geometry as used in DInSAR applications. For georeferencing a look-up table is calculated and refined based on a coregistration of the actual SAR image to a simulated SAR image. The impact of using two different implementations of such a simulator of topography-induced radar brightness, an approach based on angular relationships and a pixel-area based method are discussed in this paper. It is found that the pixel-area-based method leads to considerable improvements with regard to the robustness of georeferencing and also with regard to radiometric normalization in layover-affected areas.

**Index Terms**— Geocoding, radiometric calibration, terrain-based radiometric normalization

## 1. INTRODUCTION

Precise terrain-corrected georeferencing of SAR images and derived products in range-Doppler coordinates is relevant for data interpretation and for the combination with other geodata products. In addition, the transformation of data from map geometry to range-Doppler geometry is also very important, e.g., the transformation of terrain heights into SAR geometry as used in DInSAR applications. A method for automated terrain-corrected SAR geocoding with a refinement step using a simulation of topography-induced radiometric features was described in [1]. The quality of the correlation-based geocoding look-up table refinement is strongly dependent on the availability of distinctive common features in both the real SAR intensity image and its simulated counterpart. This SAR image intensity simulation makes use of the angular relationship between the surface normal of the local terrain patch, and the range-azimuth geometry, respectively. This approach and similar ones based on angular relationships were also assessed in [2,3]. This type of algorithm yields a realistic simulation of the terrain-induced variation of the backscattering coefficient

except for areas with strong foreshortening and layover regions, which notably are the most distinctive terrain-induced radiometric features.

A much better simulation of backscatter in layover regions is possible following ideas presented by Small *et al.* [4–7], recently summarized in [8]. We implemented such a DEM-based method for realistic SAR pixel area estimation and, in this paper, we assess the improvement achieved in both the geocoding refinement and the radiometric normalization.

## 2. METHODS

### 2.1. Simulation of Illuminated Area

The radar brightness  $\beta^0$  is defined as the average radar cross section per unit image area in range/azimuth coordinates. To obtain sensor-independent, comparable measurements, the SAR image is commonly calibrated to  $\sigma^0$  backscatter coefficients, defined as the average radar cross section per unit ground area, or to  $\gamma^0$  backscatter coefficients, defined as the average radar cross section per unit area obtained by projecting the ground area into the plane perpendicular to line-of-sight. Thus, in practice, the  $\sigma^0$  and the  $\gamma^0$  average backscatter coefficients are obtained by relating the radar brightness  $\beta^0$  to the respective reference areas  $A_{\beta^0}$ ,  $A_{\sigma^0}$ , and  $A_{\gamma^0}$ :

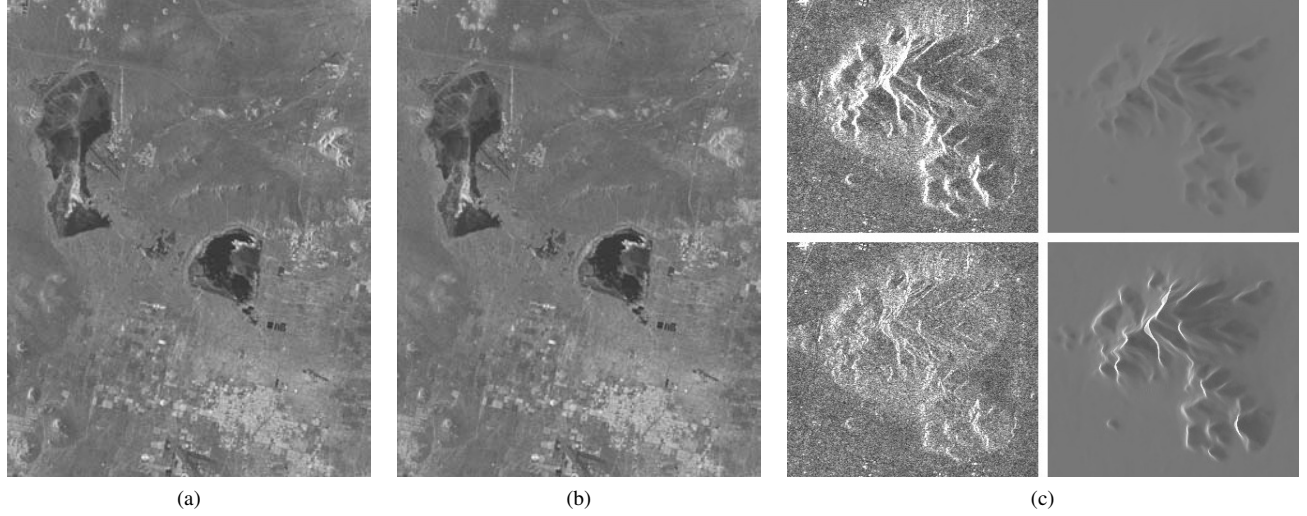
$$\sigma^0 = \beta^0 \frac{A_{\beta^0}}{A_{\sigma^0}} \quad (1)$$

$$\gamma^0 = \beta^0 \frac{A_{\beta^0}}{A_{\gamma^0}} \quad (2)$$

For the standard ellipsoid-based products the following simple trigonometric relationships involving only the incidence angle  $\theta_i$  are valid:

$$A_{\sigma^0_{ell}} = \frac{A_{\beta^0}}{\sin(\theta_i)} \quad (3)$$

$$A_{\gamma^0_{ell}} = A_{\sigma^0} \cos(\theta_i) = \frac{A_{\beta^0}}{\tan(\theta_i)} \quad (4)$$



**Fig. 1:** ERS-2 multilook intensity image of the Mojave Desert before (a) and after (b) radiometric normalization using the pixel-area-based simulated SAR image. Fig. (c) shows, clockwise and starting from top left, 1) a zoom image of a detail from the ERS-2 multilook intensity image of the Mojave Desert (see (a)), 2) the corresponding  $\cos(\psi)$ -based image simulation, 3) the pixel-area-based image simulation, and 4) the pixel-area-based  $\sigma_{pa}^0$  backscatter coefficients.

offset poly. coeff	$a_0$	$a_1(x)$	$a_2(x^2)$	$a_0$	$a_1(x)$	$a_2(x^2)$
range. :	2.41045	1.20996e-04	5.53648e-04	3.07041	-2.50452e-04	2.95323e-04
range err.:	1.10811e-02	5.78788e-06	5.00273e-06	3.45840e-03	1.79884e-06	1.59742e-06
azimuth:	1.59825	-4.74088e-04	-3.61166e-04	-0.00481	-2.18779e-05	3.80772e-06
azimuth err.:	1.74349e-02	9.10663e-06	7.87128e-06	2.05678e-03	1.06981e-06	9.50022e-07
model fit std. dev.	$\sigma$			$\sigma$		
range:	0.2349			0.0982		
azimuth:	0.3696			0.0584		

(a)

(b)

**Table 1:** Estimated georeferencing offset polynomial coefficients and corresponding errors as well as the standard deviation of the model fit in range and azimuth for the Mojave Desert data set using as a reference: (a) the  $\cos(\psi)$ -based simulated image and (b) the pixel-area-based simulated image.

Obviously, this approach yields inadequate estimates of the true ground area or the projected area in cases of rugged topography. In the following, two terrain-dependent methods to estimate the  $A_{\sigma^0}$  and  $A_{\gamma^0}$  reference areas are highlighted:

### 2.1.1. Projection Cosine Approach

Ulander [9] proposed a projection cosine method, which was applied for automated terrain-corrected SAR geocoding by Wegmuller [1]:

$$A_{\sigma_{cos}^0} = \frac{A_{\beta^0}}{\cos(\psi)} \quad (5)$$

$\psi$  denotes the angle between the image plane normal and the surface normal, i.e.,  $\psi$  relates the unit image area to the unit ground area. The respective  $A_{\gamma_{cos}^0}$  reference area in the plane perpendicular to line-of-sight is calculated via the cosine of

the local incidence angle  $\theta_l$

$$A_{\gamma_{cos}^0} = A_{\sigma_{cos}^0} \cos(\theta_l). \quad (6)$$

### 2.1.2. Pixel Area Integration Method

In reality, many DEM pixels may contribute to a single range/azimuth coordinate. This is particularly the case in layover-affected areas. Therefore, a more realistic simulation of the topography-induced variation of radar brightness is obtained by integrating all DEM-facets  $dA_{DEM}$  that contribute to a specific SAR pixel at range/azimuth coordinates  $(r, a)$ , i.e., that belong to the illuminated area, henceforth termed pixel area  $A_{pa}$ . This idea was brought forward by Small *et al.* [4]. In our implementation, the complete DEM-surface covered by the SAR image is divided into small triangular surface patches  $dA_{DEM}$  which are then “distributed” into

range/azimuth “buckets”, the SAR pixels, according to the associated range/azimuth value as obtained from a geocoding look-up table. Note, that the size of the triangular surface patches  $dA_{DEM}$  varies throughout the DEM. After having worked sequentially through all the rows and columns (i, j) of the DEM, we end up with the sum of all contributing facets in each bucket, i.e. the total illuminated topographic area, or *pixel area*,  $A_{\sigma_{pa}^0}$  for each pixel:

$$A_{\sigma_{pa}^0}(r, a) = \sum_{i,j \in \mathcal{A}} dA_{DEM}(i, j), \quad (7)$$

where  $\mathcal{A} := \{i, j \mid \rho(i, j) = r, \eta(i, j) = a\}$ ;  $\rho$  and  $\eta$  are the range and azimuth values at DEM position (i, j) as obtained by bilinear interpolation of the look-up table and rounding to the next integer value. Similarly, the  $A_{\gamma_{pa}^0}$  reference area is obtained via

$$A_{\gamma_{pa}^0}(r, a) = \sum_{i,j \in \mathcal{A}} [dA_{DEM}(i, j) \cdot \cos(\theta_l(i, j))] . \quad (8)$$

## 2.2. Geocoding Refinement

The SAR geocoding method described in [1] makes use of the satellite orbit, a digital terrain model, and the SAR imaging parameters to calculate the corresponding slant-range and along-track position for each grid point of the digital elevation model. These initial slant-range and along-track positions are stored in a look-up table. As a next step, a simulated SAR intensity image is calculated. Then, offsets between the simulated image and the real SAR intensity image are determined and used to refine the look-up table. Eventually, the SAR image is transformed in one resampling step based on the refined look-up table. We compare the quality and robustness of this geocoding approach for the two different image simulation approaches, 1) the projection cosine-based approach and 2) the pixel-area-based approach.

## 2.3. Radiometric Normalization

The terrain-corrected normalized backscatter coefficients  $\sigma_{cos}^0$  and  $\sigma_{pa}^0$  are obtained by inserting the reference areas  $A_{\sigma_{cos}^0}$  from (5) and  $A_{\sigma_{pa}^0}$  from (7), respectively, into (1). Similarly, the  $\gamma_{cos}^0$  and  $\gamma_{pa}^0$  backscatter coefficients are obtained using (6) and (8) in (2). In the following, the quality of the  $\cos(\psi)$ -based and the pixel-area-based radiometric normalization is assessed.

# 3. RESULTS

## 3.1. Geocoding Refinement

To demonstrate the increased robustness with respect to geocoding refinement provided by the new pixel-area-based SAR simulation a subset of an ERS-2 data set over the Mojave Desert in California is used. The area is mostly flat

only showing a very limited number of distinct topographic features (see Fig.1) The second scene, also taken by ERS-2, shows an alpine area in southeastern Switzerland on the border to Italy (see Fig. 2). For the Mojave data set the SRTM 1 arcsecond digital surface model is used. For the second example the Swiss DHM25 digital elevation model is used, complemented by SRTM 3 arcsecond DSM data covering the regions outside Switzerland.

Table 1 contains the coefficients of the quadratic polynomials that describe the range- and azimuth-varying geocoding offsets between the actual SAR image and the simulated image. Two different simulation methods, the  $\cos(\psi)$ -based, and the pixel-area-based simulation of topography-induced radar brightness were employed. The resulting coefficients are found in Table 1 (a) and (b), respectively. In addition to the polynomial coefficients, also their respective errors, as well as the standard deviations of the model fit are given.

## 3.2. Radiometric Normalization

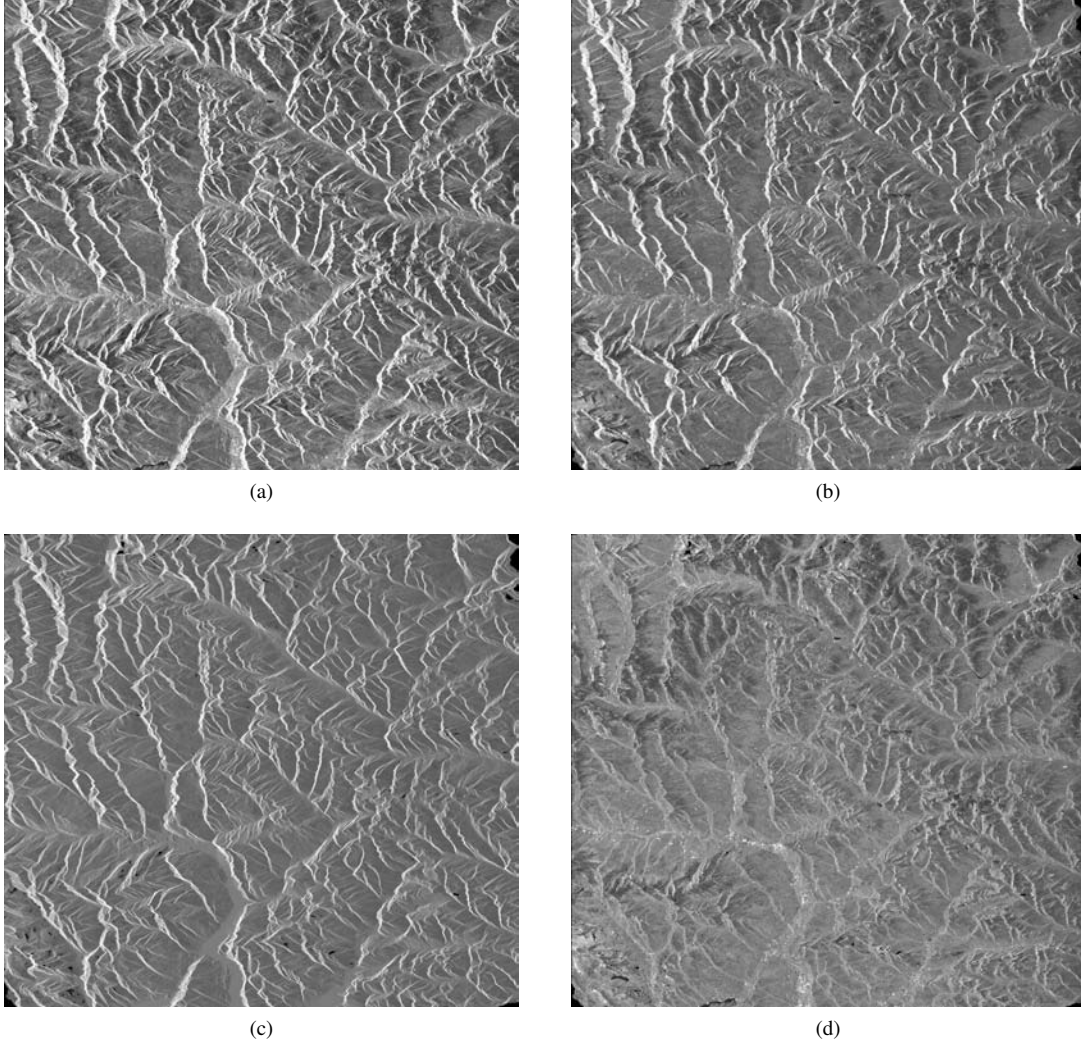
In Fig. 3 the performance of the three different radiometric calibration procedures, 1) ellipsoid-based, 2)  $\cos(\psi)$ -based, and, 3) pixel-area-based normalization, is evaluated, quantitatively, for an ERS-2 scene of an alpine region in southeastern Switzerland. The backscatter values are subdivided into classes according to the corresponding local incidence angles. Each box plot represents the median, the 25%, and the 75% percentiles of that particular class, the width of the classes being 5 degree. The whiskers attached to the boxes indicate the 5% and the 95% percentiles. The plots in the first row of Fig. 3 show the distribution of  $\sigma^0$  values employing, from left to right, (a) the ellipsoid-based, (b) the  $\cos(\psi)$ -based, and, (c) the pixel-area-based radiometric normalization. In the second row, the respective data evaluation is given only for the nonlayover regions of the same SAR scene. The histogram in Fig. 3g contains the relative frequency of occurrence of local incidence angles throughout the SAR image for the two cases 1) with, and 2) without inclusion of layover-affected areas. Finally, the distributions of the  $\gamma_{pa}^0$  backscatter values through pixel-area-based normalization with and without layover regions are shown in Figs. 3h & 3i, respectively.

# 4. DISCUSSION

## 4.1. Geocoding Refinement

The small patch of the ERS-2 scene depicting the Mojave Desert (see Fig. 1) exhibits very few topography-induced radiometric features only. The importance for accurate modeling of these features based on integration of all illuminated DEM surface patches is reflected in the much-improved estimation errors—by nearly one order of magnitude—when using the pixel-area-based image simulation as a reference.

In this particular and, notably, rather extreme case, it can even be observed that the resulting coefficients of the off-

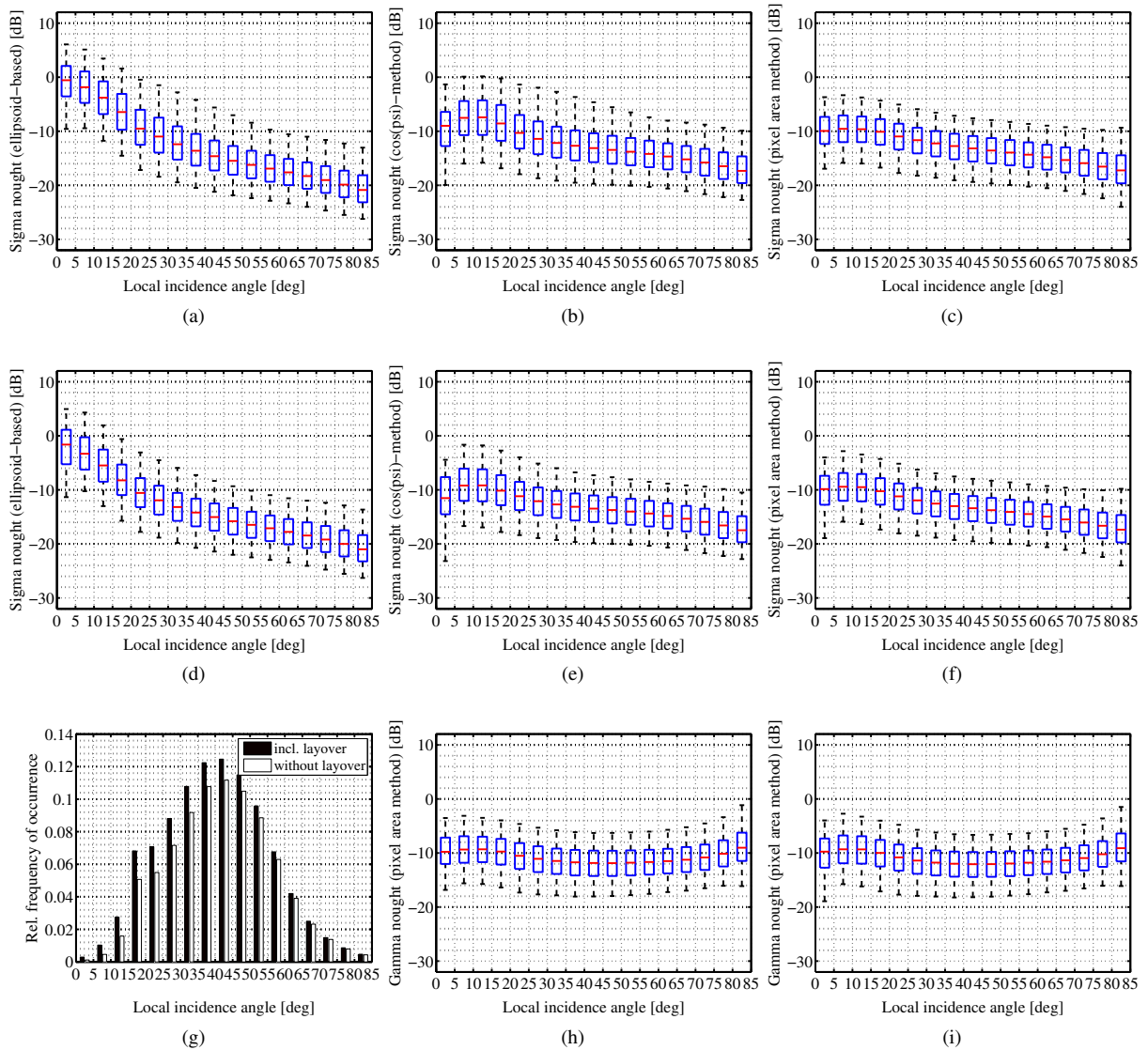


**Fig. 2:** (a) ERS-2 MLI  $\sigma^0$  backscatter map of an alpine area in southeastern Switzerland using ellipsoid-based calibration. (b) ERS-2 MLI  $\sigma_{cos}^0$  values after radiometric normalization using the DEM-based  $\cos(\psi)$  correction approach. (c) Simulated SAR image based on integration of illuminated DEM-pixels per range/azimuth coordinate. (d) ERS-2 MLI ( $\sigma_{pa}^0$ ) after radiometric normalization using the pixel-area-based simulated SAR image.

set polynomials are quite different for the two simulations compared: in azimuth, the constant coefficients differ by more than 1.5 samples, and in range the difference is about 0.6 samples. Judging from the lower estimation errors and the improved standard deviation of the model fit the pixel-area-based geocoding solution is more robust in this case where only very few terrain features are present in a SAR image, whereas the  $\cos(\psi)$ -based method leads to a biased estimate. This finding is also supported by the fact that if a larger excerpt of the same frame including more topography is geocoded the estimated coefficients tend towards the solution obtained for the pixel-area-based simulation with only the small patch of data.

#### 4.2. Radiometric Normalization

A visual inspection of the radiometrically normalized backscatter values (see Fig. 2) indicates most strikingly the enhanced radiometric calibration obtained with the pixel-area-based technique. This visual impression is quantitatively underlined by a comparison of the box plots of  $\sigma^0$  values, shown in Fig. 3: while the median backscatter value ranges from 0 dB, for very small local incidence angles, to -21 dB, for large local incidence angles in the case of standard ellipsoid based normalization, the range of median  $\sigma^0$  values is reduced to the interval  $[-7.5, -17.5]$  dB for the  $\cos(\psi)$ -based method, and reduced to  $[-9.5, -17.5]$  dB for the pixel-area-based method. In the latter case, the outer quantiles, which mark the range of values where 50% and 90% of the data in a



**Fig. 3:** The first row contains box plots (5%, 25%, 50% (median), 75%, and 95% percentiles) according to classes of local incidence angles for the backscatter values of the ERS-2 scene of the alpine region in southeastern Switzerland shown in Fig. 2, including areas affected by layover: (a) ellipsoid-based  $\sigma^0$ , (b)  $\cos(\psi)$ -based  $\sigma_{cos}^0$ , and (c) pixel-area-based  $\sigma_{pa}^0$ . In the second row, the respective box plots are shown for the same scene, but the layover areas were masked out, beforehand: (d) ellipsoid-based  $\sigma^0$ , (e)  $\cos(\psi)$ -based  $\sigma_{cos}^0$ , (f) pixel-area-based  $\sigma_{pa}^0$ . The last row contains (g) histograms of relative frequencies of occurrence of SAR pixels according to local incidence angles with and without layover regions, as well as the pixel-area-based  $\gamma_{pa}^0$  with (h), and without (i) layover-affected areas.

particular bin are contained, even span a considerably smaller range ( $\approx 5$  dB and  $\approx 12$ -15 dB, respectively) than in the case of the  $\cos(\psi)$ -based normalization ( $\approx 5$ -7 dB and  $\approx 12$ -20 dB, respectively). The improvement is particularly notable for local incidence values from 0 to 50 degree. The enhancement is less pronounced if only nonlayover pixels are considered (second row of Fig. 3). If the data are calibrated to  $\gamma^0$  backscatter coefficients the median values even stay within  $[-9, -12]$  dB.

The bottom line of this analysis is that incorporation of a digital elevation model into the radiometric normalization is indispensable in mountainous areas and that a pixel-area-based normalization is required to obtain an adequate radiometric normalization of layover-affected pixels. It has to be noted that the resolution and the quality of the DEM used for the calculation of the pixel area both have a great influence on the quality of the geocoding refinement and, consequently,

also on the quality of the radiometric normalization.

## 5. CONCLUSION

The advantages of a pixel-area-based simulation of topography-induced variation of radar brightness in SAR images were highlighted using ERS-2 SAR data and digital elevation models. The added value of this pixel-area-based method, compared to conventional methods based on angular relationships, is twofold: 1) a more accurate geocoding refinement is obtained since the distinct radiometric features which are introduced by layover areas, and which are crucial to obtain a good correlation between the SAR image and the simulated image in the geocoding process, are now modeled realistically, limited only by the resolution and accuracy of the digital elevation model. 2) A second advantage of the pixel-area-based normalization method compared to the  $\cos(\psi)$ -based approach lies in the much-improved backscatter normalization in *layover areas*, thereby leading to a smoother appearance of the backscatter map in SAR scenes with distinct topography. Although the resolution remains low in these areas the improved radiometric normalization aids in detecting features of interest that would otherwise be hidden by topography-induced, high backscatter values.

## 6. ACKNOWLEDGMENT

ERS-2 data © ESA. SRTM DEM © USGS. DHM25 © 2003 swisstopo.

## 7. REFERENCES

- [1] U. Wegmuller, "Automated terrain corrected SAR geocoding," in *Proc. IEEE Geosci. Remote Sens. Symp.*, vol. 3, 1999, pp. 1712–1714.
- [2] A. Löw and W. Mauser, "Generation of geometrically and radiometrically terrain corrected SAR image products," *Remote Sensing of Environment*, vol. 106, no. 3, pp. 337–349, feb 2007.
- [3] O. Frey, E. Meier, and D. Nüesch, "Processing SAR data of rugged terrain by time-domain back-projection," in *SPIE Vol. 5980: SAR Image Analysis, Modeling, and Techniques X*, 2005.
- [4] D. Small, M. Jehle, E. Meier, and D. Nüesch, "Radiometric terrain correction incorporating local antenna gain," in *Proc. 5th EUSAR*, Ulm, Germany, 2004.
- [5] D. Small, N. Miranda, and E. Meier, "Local incidence angle considered harmful," in *Proc. CEOS SAR Cal/Val Workshop*, Pasadena, CA, Nov. 2009.
- [6] D. Small, N. Miranda, and E. Meier, "A revised radiometric normalisation standard for SAR," in *Proc. IEEE Geosci. Remote Sens. Symp.*, vol. 4, July 2009, pp. 566–569.
- [7] D. Small, N. Miranda, L. Zuberbühler, A. Schubert, and E. Meier, "Terrain-corrected gamma: Improved thematic land-cover retrieval for SAR with robust radiometric terrain correction," in *Proc. ESA Living Planet Symp., ESA SP-686*, Bergen, Norway, July 2010.
- [8] D. Small, "Flattening gamma: Radiometric terrain correction for SAR imagery," *IEEE Trans. Geosci. Remote Sens.*, vol. 49, no. 8, pp. 3081–3093, 2011.
- [9] L. M. H. Ulander, "Radiometric slope correction of synthetic-aperture radar images," *IEEE Transactions on Geoscience and Remote Sensing*, vol. 34, no. 5, pp. 1115–1122, 1996.