

SOIL MOISTURE EVALUATION USING MULTI-TEMPORAL SYNTHETIC APERTURE RADAR (SAR) IN SEMIARID RANGELAND

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ABSTRACT

There have been several efforts to utilize satellite-based Synthetic Aperture Radar (SAR) measurements to determine surface soil moisture (to 5 cm) conditions of rangeland regions. The results have been mixed since the relation between the SAR signal and surface soil moisture is confounded by variations in topographic features, surface roughness and vegetation density. We designed an experiment to investigate the sensitivity of C-band SAR backscatter (σ^0) to surface soil moisture (θ_s) in a semi-arid rangeland and to test a data-fusion approach based on both optical (Landsat TM) and radar (ERS-2 SAR) measurements to improve regional estimates of surface soil moisture content. The data-fusion approach (from Sano, 1997) utilized the difference between dry- and wet-season SAR σ^0 to normalize roughness effects, and utilized surface reflectance in optical wavelengths to account for differences in vegetation density. We focused the study on three flat, uniformly-vegetated sites of known surface roughness, monitored variations in surface soil moisture, vegetation density and SAR signal over time, and obtained eight optical/SAR image pairs throughout the dry and wet seasons. For

these sparsely vegetated sites during this dry year (1997), we found that the SAR signal was not significantly attenuated by sparse green vegetation cover (green leaf area index < 0.35) and dense standing brown vegetation cover (brown leaf area index up to 1.5). Consequently, the optical data was not required for this application, and the approach could be implemented by simply taking the difference between the dry- and wet-season SAR σ° ($\sigma^{\circ} - \sigma^{\circ}_{\text{dry}}$). For a data set of eight dates at three study sites, we confirmed that the relation between ERS-2 C-band SAR σ° and ϵ_s was weak ($r^2=0.27$); yet for the same data set, that the relation between $\sigma^{\circ} - \sigma^{\circ}_{\text{dry}}$ and ϵ_s was strong and significant ($r^2=0.93$). This study also raised two concerns: 1) the overall sensitivity of SAR σ° to ϵ_s was relatively low, and 2) the approach required a high level of accuracy in the estimate of green leaf area level that may not be obtainable with standard optical remote sensing algorithms. In any case, the positive results from this study should encourage the use of a multi-temporal SAR and optical/SAR fusion for monitoring semi-arid range conditions, and improving management of scarce resources.

1. INTRODUCTION

Knowledge of distributed surface soil moisture content ($\sim 5\text{cm}$ depth) is important for many hydrologic applications including mapping rainfall events, monitoring differential drying patterns, and assessing water availability for plant growth. Surface soil moisture can also be used to parameterize soil water simulation models that estimate soil moisture content with depth in the plant rooting zone (e.g., Hymer et al., 2000). Though the demand for distributed surface soil moisture information is high, the means for obtaining such information are few. Conventional measurement techniques (e.g., gravimetric and time-domain reflectometry (TDR)) are generally point-based, and require on-site operators and tedious post-processing. Such sensor attributes are not conducive to measurement of regional surface soil moisture conditions on a frequent basis.

There is some evidence that satellite-based Synthetic Aperture Radar (SAR) sensors could

provide a regional assessment of surface volumetric soil moisture content (ϵ_s) (Engman and Chauhan, 1995). Theoretically, SAR backscatter (σ^0) detected by orbiting satellite-based sensors is directly related to the target dielectric constant (\hat{a}'), where \hat{a}' is the real part of a complex parameter that describes the electrical properties of a medium relative to the dielectric constant of “free space”. For water, $\hat{a}' \approx 80$; for dry soil, $\hat{a}' \approx 2$. Consequently, an increase in soil moisture content changes \hat{a}' markedly, and results in a strong sensitivity of the SAR signal to ϵ_s . In practice, σ^0 is also highly influenced by topographic features, vegetation density, and variations in small-scale surface roughness. For an agricultural test site, Wang et al. (1986) reported that L-band σ^0 was strongly dependent on ϵ_s and surface roughness, and to a lesser extent, on vegetation cover. And in most cases, the range of σ^0 response to variations in surface soil moisture is equal to the range of σ^0 response to variations in surface roughness of natural surfaces. Thus, it is a difficult task to convert single-channel SAR images directly into maps of regional ϵ_s for heterogeneous terrain.

Most reviews on this topic suggest that surface soil moisture could be retrieved with sufficient accuracy from a system providing two or more polarizations of a single frequency such as L-band SAR (e.g., Ulaby et al., 1996). Unfortunately, all currently orbiting satellite SAR sensors are single-polarization, single-frequency systems. Alternatively, several studies have suggested that the accuracy of SAR-based ϵ_s estimates could be improved by combining data from optical sensors (e.g., surface reflectance and temperature) to discriminate the SAR signal response to vegetation (Moran et al., 1997).

In this project, we designed an experiment to study the link between SAR backscatter and surface soil moisture using the European Remote Sensing (ERS-2) SAR sensor [C-band (5.35 GHz), VV polarization and 23E incidence angle]. To minimize the influence of other conditions, we focused our study on flat, uniformly-vegetated sites, and monitored the variations in surface soil moisture and vegetation cover over time. By choosing flat sites, we avoided the effects of topography; and by monitoring the sites over time (rather than multiple sites over space), we minimized the influence of variations in small-scale roughness conditions. Furthermore, by measuring vegetation density on a monthly basis at each site, we were able to quantify changes in vegetation that might influence SAR σ^0 .

In addition, we ordered an image from a satellite-based optical sensor concurrent with each

SAR scene acquisition. Images were obtained from the Landsat Thematic Mapper (TM) sensor which measures surface reflected radiance in six wavelengths (from 0.45 to 2.35 μm) and measures surface temperature in a single spectral waveband covering 10.42 to 11.66 μm . Due to orbital constraints, the Landsat TM scenes were generally acquired within 0-7 days of the ERS-2 SAR acquisition.

The objectives of this work were to:

- a) investigate the sensitivity of C-band SAR σ° to surface soil moisture content (to 5 cm) in a semi-arid rangeland with sparse vegetation cover; and
- b) test an approach based on both optical (Landsat TM) and radar (ERS-2 SAR) measurements to improve regional estimates of surface soil moisture content .

2. APPROACH

The basic approach for the use of SAR/optical synergism for estimation of surface soil moisture content was developed by Sano (1997). He proposed a semi-empirical approach which accounted for both soil roughness and vegetation effects on the SAR signal, and greatly improved the relation between SAR backscatter and surface soil moisture content in a semiarid region. In this approach, the effects of soil roughness were taken into account by taking the difference between the SAR backscatter from a given image and the backscatter from a “dry season” image ($\sigma^{\circ} - \sigma^{\circ}_{\text{dry}}$). The vegetation influence was corrected by using an empirical relationship between $\sigma^{\circ} - \sigma^{\circ}_{\text{dry}}$ and green leaf area index (GLAI), where GLAI is green leaf area divided by corresponding ground area. GLAI was derived using an operational approach based on a radiation model and measured reflectance in the red and near-infrared spectrum (Qi et al., 2000).

This approach is illustrated hypothetically in Figure 1. Sano (1997) found that the vertical distance between a given point and the line defining the $(\sigma^{\circ} - \sigma^{\circ}_{\text{dry}})/\text{GLAI}$ relation was independent of surface roughness and vegetation density, and directly related to the surface soil moisture content of the site. This relation was found to be linear for low values of GLAI ($\text{GLAI} < 0.6$) as shown by Ulaby et al. (1984) in their Figures 10 and 11. Furthermore, a given relation, as illustrated in Figure 1, would be valid only for a single SAR configuration (e.g., sensor polarization and frequency) and would need to be

adjusted for the influence of topography on local incidence angle. In Figure 1, though the values of δ° - δ°_{dry} for points A-C increase from A (2 dB) to B (5 dB) to C (6 dB), the surface soil moisture conditions are related to the length of the vertical arrow; thus, the surface soil moisture content of B is greatest and A least, with C intermediate. Sano et al. (1998) reported that this approach worked well for estimating surface soil moisture conditions for sparsely-vegetated (GLAI<1), semi-arid sites in Arizona. However, they admitted that validation of this approach was questionable because the small number of soil moisture samples (three per site) didn't properly characterize each site.

3. EXPERIMENT

Three sites were chosen in the Upper San Pedro River Basin (USPB) in southeast Arizona for investigation of the SAR/optical approach for monitoring surface soil moisture content. The sites were characterized by level terrain and uniform vegetation cover (over a 300 x 300 m area), and were named by the dominant vegetation type: Tobosa, Sacaton and Creosote (Figure 2). The Tobosa site was located in a swale which supports a mix of tobosa grass (*Hilaria mutica*) and Creosote (*Larrea tridentata*) shrubs. The Sacaton site was dominated by big sacaton (*Sporobolus wrightii*) with some tobosa grass. The Creosote site was on a flat mesa and is characterized by scattered creosote shrubs (*Larrea tridentata*) with very few grasses or annual forbs. The soils in this region are a very gravelly sandy loam.

In this study, we requested 10 ERS-2 SAR scenes covering our study sites during late 1996 and throughout 1997. The dates of these overpasses were selected to correspond closely with the dates of overpasses of the Landsat-5 satellite (Table 1). In southeast Arizona, the majority of the precipitation is in late July and August, resulting in maximum vegetation greenness in August and September. Thus, the experiment was designed to continue through October 1997, when the vegetation had passed maximum density and was beginning to senesce.

Field Measurements

During each ERS-2 overpass, we visited all three sites and made 49 gravimetric measurements of soil moisture content to 5cm depth over a target area of 90 x 90 m centered within the larger uniform area of 300 x 300 m. These measurements were made within \pm one hour of the ERS-2 overpass time at approximately 11:00am MST. Gravimetric soil moisture measurements were converted to volumetric soil moisture using an estimate of the soil bulk density based on measurements made by Sano (1997).

We also measured vegetation cover, biomass, height, and leaf area every three weeks to record seasonal changes. At each site, destructive samples of green and brown vegetation were made within seven systematically-selected one-m² quadrats within the 90 x 90 m target area. Based on these samples, green and brown leaf area index (GLAI and BLAI, respectively) were measured in the laboratory using a LICOR LI3000 leaf area meter. Just prior to destructive sampling, the total plant area index (TPAI) was measured *in situ* using a LICOR LAI2000 plant canopy analyzer. A distinction is made here between GLAI, BLAI and TPAI: GLAI is an index of only green vegetation, BLAI is an index of only dead, standing vegetation, and TPAI is an index of all standing live and dead vegetation. Note that the approach developed by Sano (1997) is based on the SAR signal attenuation by GLAI (Figure 1). For this study during the period from January through July, GLAI was near zero; that is, BLAI = TPAI and both were composed largely of standing, dry grasses and desiccated creosote shrubs.

Surface soil roughness was measured at the Sacaton and Creosote sites using a roughness meter consisting of a 1-m row of 100 pins spaced at 1 cm intervals (Simanton et al., 1978). For each site, a total of 50 measurements were made at approximately 15 m intervals. At the Tobosa and Sacaton site, all 50 measurements were taken in bare interspace areas because there was no significant difference between the roughness measurements made under the shrub and in the bare interspace areas. At the Creosote site, 25 measurements were taken in bare interspace areas and 25 measurements were taken directly underneath creosote cover to account for discrete roughness changes between the measurement points. Individual lines were digitized using a Geographic Information System software package (Arc/Info) to derive the root mean squared (RMS) height, termed the roughness value. Data presented in Table 2 represent the average roughness value (cm) for the Sacaton and Creosote sites

(data for the Tobosa site were not included since it was not covered by the imagery).

During each TM overpass, we visited the sites and made measurements of surface reflectance and temperature over large areas for comparison with the TM measurements. We also deployed a solar radiometer and arranged for the launch of a radiosonde balloon to measure atmospheric conditions for eventual atmospheric correction of the TM image to obtain estimates of surface reflectance and temperature. On several occasions, we deployed a radiometer aboard a small aircraft to measure surface reflectance and temperature at fine resolution (1-2 m) to provide us with local estimates of vegetation vigor for validation of our satellite-based analysis (Goodrich et al., this issue).

Image Acquisition and Processing

A temporal series of ERS-2 SAR images was obtained covering our study area in southeast Arizona (Figure 3). We used the standard, calibrated amplitude images provided by the European Space Agency (ESA) processed as 3-looks (reduced noise) with ground resolution of 12.5 x 12.5 m. The digital numbers (dn) were converted to a backscattering coefficient expressed in dB using the SAR calibration coefficient (Tom Lukowski, Canadian Centre for Remote Sensing, personal communication). Values of C-band σ^0 within a 7 by 7 pixel window were averaged to one value for each of the three 90 x 90 m field targets. This averaging process minimized the effects of the characteristic SAR image “speckle” due to multiple within-pixel scatterers. It is notable in Figure 3 that the image coverage shifted to the west after the January scene, and to our misfortune, the images no longer included the Tobosa site. Furthermore, on some dates (scenes acquired in April and September), the scene was shifted north such that the Sacaton site was not included.

Landsat TM images were acquired within 0-7 days of the ERS-2 SAR images covering the same general area (Figure 3). Green leaf area index was derived from these images using the modeling approach proposed by Qi et al. (2000). In this approach, the temporal set of Landsat TM images was used as an input to invert a bi-directional reflectance distribution function (BRDF) model and make estimates of GLAI. Next, each Landsat TM image was converted to a normalized-difference vegetation index (NDVI),

$$NDVI = (\tilde{n}_{NIR} - \tilde{n}_{Red}) / (\tilde{n}_{NIR} + \tilde{n}_{Red}), \quad (1)$$

where \tilde{n}_{NIR} and \tilde{n}_{Red} are near-infrared and red reflectance factors derived from TM digital number using measurements of atmospheric conditions and earth-sun geometry (Moran et al., 1995). A regression process and fuzzy neural system were used to find the best-fit equation (linear, exponential, power and polynomial) between image-derived GLAI and NDVI. The purpose of this step was to reduce the computing time and to remove some outliers due to model convergence failure or external noise. For this region, the optimal equation for converting NDVI to GLAI was determined to be

$$GLAI = 18.999NDVI^3 - 15.240NDVI^2 + 6.124NDVI - 0.3520 \quad (2)$$

Qi et al. (2000) validated this approach with ground-based measurements and reported that GLAI measurements could be obtained with high accuracy. For the three experimental sites in this study, values of TM-derived GLAI were extracted from a 3 by 3 pixel window covering 90 x 90 m within each site, and averaged to one value per site per date.

4. RESULTS

Due to favorable weather conditions, we obtained a wide range of surface soil moisture conditions for our study (see Figure 4 and notes in Table 3). An unusual January snowstorm, followed by several days of hot sunny weather, resulted in regional soil moisture conditions near field capacity. February rains resulted in moist, variable soil conditions. A drying trend continued through March and April; in June and July, soil conditions at all sites were very dry. Summer rainstorms preceded the August and September overpasses resulting in moist soil conditions, and in October, soil conditions were again dry.

During the July SAR overpass, the soil moisture conditions at all sites were extremely dry, and the late summer greenup of the vegetation had not yet occurred. Consequently, we designated it as the “dry” scene and subtracted the July SAR backscatter (σ_{dry}°) from the backscatter measured on the

January, March, April, June, August and September dates to account for the contribution of surface roughness to the SAR signal.

For the dry period from January through July, the TPAI estimated *in situ* with the LAI2000 plant canopy analyzer compared relatively well with the BLAI of the sampled vegetation measured in the laboratory with the LI3100 leaf area meter (Figure 5). It appears that TPAI was underestimated with the LAI2000 for high TPAI values due to the clumped vegetation (see Nouvellon et al., this issue), and that BLAI was underestimated with the LI3000 for low BLAI because scattered shrubs were not sampled and measured. Measurements with the LAI2000 showed that the TPAI of the Sacaton site was significantly greater than that of the Creosote and Tobosa sites, with average values for June of 1.2, 0.39 and 0.35, respectively. Since there was no greenup of vegetation during this period, there was no significant change in site TPAI. In summary, for the period from January through July, the TPAI was primarily a measure of dry, standing biomass which differed from site to site, but not from date to date.

We investigated the influence of standing dry biomass on the SAR σ° by deriving a relation between $\sigma^\circ - \sigma^\circ_{\text{dry}}$ and TPAI for the June overpass at the Creosote and Sacaton sites (recall that due to an orbital shift, the Tobosa site was not included in images after January). Based on these two points, there was no significant variation in $\sigma^\circ - \sigma^\circ_{\text{dry}}$ associated with the measured variation in TPAI at the two sites. Our conclusion was that the differences in standing brown vegetation biomass at these two sites were accounted for in the roughness correction. Consequently, for the dry season, the values of $\sigma^\circ - \sigma^\circ_{\text{dry}}$ would be directly related to surface soil moisture conditions at the site, with no need for a vegetation-related correction.

The amount of green leaf area during the period of August through October was relatively small for these three vegetation types during 1997 (Figure 6). In September, the GLAI values for all three sites measured in the laboratory using the LI3000 were less than 0.2. GLAI values derived from the Landsat TM images were slightly larger, ranging from 0.25 for the Tobosa site to 0.45 for the Sacaton grass. The small discrepancy between the LI3000- and TM-derived GLAI could be due to the heterogeneity of the site (where the TM pixel represented a 30 x 30 m areal extent and the ground samples were taken from multiple 2 x 2 m plots) or to the errors inherent in the derivation of Eq. (2), as

discussed by Qi et al. (2000). Assuming that the laboratory GLAI measurements were the more accurate of the two GLAI estimates, then according to the relation presented in Figure 1, the values of $\sigma^{\circ}-\sigma^{\circ}_{\text{dry}}$ for this season did not require a correction for vegetation attenuation of the SAR signal. Thus, for these biomes during the 1997 growing season, values of $\sigma^{\circ}-\sigma^{\circ}_{\text{dry}}$ would be directly related to surface volumetric soil moisture. This premise was supported by the following results.

As expected, results showed that measurements of C-band SAR backscatter (σ° , dB) were poorly correlated ($r^2=0.27$) with surface soil moisture content ($\hat{\epsilon}_s$) at the three field sites (Figure 7). Theoretically, this poor correlation was due to differences in surface roughness at the three sites (Table 2). Comparing the Creosote and Sacaton sites, the greater root mean squared (RMS) roughness resulted in a generally higher SAR σ° for the Creosote site than for the Sacaton site (Figure 7). The data were subsequently corrected for differences in surface roughness and standing brown vegetation biomass by taking the difference between the SAR backscatter from a given image and the backscatter from the dry season image (July). This correction resulted in a good correlation ($r^2=0.93$) between $\sigma^{\circ}-\sigma^{\circ}_{\text{dry}}$ and $\hat{\epsilon}_s$ (Figure 8). On the other hand, there was a general insensitivity of SAR $\sigma^{\circ}-\sigma^{\circ}_{\text{dry}}$ to low soil moisture conditions. That is, for $\hat{\epsilon}_s$ from 0.07 to 0.20, $\sigma^{\circ}-\sigma^{\circ}_{\text{dry}}$ varied by less than 1 dB, which is close to the stated accuracy of the ERS-2 SAR sensor calibration (- 1 dB).

Using the relation between $\sigma^{\circ}-\sigma^{\circ}_{\text{dry}}$ and $\hat{\epsilon}_s$ presented in Figure 8, regional maps of surface volumetric soil moisture were obtained from the January and March SAR images. The maps of these two dates show a good contrast between regional surface soil moisture conditions (Figure 9). On 12 January, due to recent snowmelt, the soil conditions at all three sites and throughout the region were near field capacity. On 23 March, due to an exceptionally dry two-month period, the regional soil conditions were very dry. The scattered anomalously high soil moisture values (blue colors) in the March image (Figure 9) are an artifact due to the errors associated with image-to-image registration for SAR scenes of mountainous terrain. This error was found to be as large as 200 m in the mountainous regions of the image.

5. DISCUSSION AND CONCLUDING REMARKS

Field validations, by their nature, are generally at the mercy of circumstances beyond the control of the experimenters. In this study, we were quite fortunate with atmospheric and surface soil moisture conditions; we were less fortunate with plant conditions and image acquisition. For a variety of reasons, the plant greenup was minimal and measured GLAI was lower than the threshold for σ° attenuation by vegetation set by Sano (1997) in Figure 1. Thus, we were able to confirm the relation between $\sigma^\circ - \sigma^\circ_{\text{dry}}$ and ϵ_s for GLAI values less than 0.35, but we were unable to test the relation between GLAI and $\sigma^\circ - \sigma^\circ_{\text{dry}}$ for $\text{GLAI} > 0.35$.

Image acquisition was also problematic. Due to an unexpected orbital shift in the ERS-2 satellite, one of our field sites (Tobosa) was excluded from seven of eight SAR images. Furthermore, one of the Landsat-5 TM and two ERS-2 SAR images that were ordered were not acquired by the satellite sensor. This is not an uncommon experience in such field validation experiments (Moran, 1994). Finally, the σ° of the SAR scene acquired on 16 February was substantially lower than that of all other scenes. According to our ground-based measurements of meteorological conditions, vegetation cover, roughness and surface soil moisture, it was unreasonable for the February σ° to be 2-3 dB lower than σ° measured in July. The February scene was consequently not included in the analysis, and is currently being investigated with assistance from the European Space Agency to explain this anomaly.

Despite these setbacks, we were able to provide preliminary validation of an operational approach for using SAR and optical data for monitoring vegetation growth and surface soil moisture conditions. The general approach requires a temporal set of SAR images during the “dry” and “wet” seasons, and intermittent optical data at times of vegetation greenup. The preliminary results were encouraging. As expected, the correction for surface roughness conditions resulted in a substantial improvement in the correlation between the SAR signal and ϵ_s . We confirmed that for GLAI values less than 0.35, it was unnecessary to correct for attenuation of the SAR signal by vegetation. Furthermore, it appeared that differences in standing brown vegetation ($\text{TPAI} < 1.5$) did not affect the SAR backscatter signal after the correction for roughness was applied. The latter result is significant for two reasons. First, it means that accurate estimates of surface soil moisture may be possible without a-priori information about standing dry biomass and dry litter. Second, it gives support to the use of an

optical/SAR approach for mapping surface soil moisture because the optical data is sensitive to changes in GLAI rather than TPAI.

Results from this analysis also raised some concerns. First, the difference between measured and modeled GLAI was relatively small, but not negligible (root mean squared error 0.13). A high level of accuracy (-0.05) in the estimate of GLAI is required for this approach since a minimal change in GLAI will result in a large change in $\delta^\circ - \delta^\circ_{\text{dry}}$ (see Figure 1 and work by Troufleau et al., 1997). To achieve such accuracy, it may be necessary to assimilate remotely-sensed information into a plant growth model to determine GLAI (e.g., Nouvellon et al., 1998). Second, despite the good relation between $\delta^\circ - \delta^\circ_{\text{dry}}$ and ϵ_s (Figure 8), the overall sensitivity of the SAR signal to changes in surface soil moisture was low. For the Sacaton site, a change in surface soil moisture of 25% resulted in a change in $\delta^\circ - \delta^\circ_{\text{dry}}$ of only 3 dB. This is notable since 25% is the maximum soil moisture range expected for sandy loam soils in Arizona's semiarid rangeland.

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7. REFERENCES

Engman, E.T. and N. Chauhan (1995) Status of microwave soil moisture measurements with remote sensing, Rem. Sens. Env. 51:189-198.

- Goodrich, D.C., A. Chehbouni, B. Goff, B. MacNish, T. Maddock, S. Moran, et al. (2000) An overview of the activities of the semi-arid land-surface-atmosphere (SALSA) program, Ag. For. Met. (this issue).
- Hymer, D.C., M.S. Moran and T.O. Keefer (2000) Soil moisture evaluation using a hydrologic model and calibrated sensor network, Soil Sci. Soc. Am. J. 64:319-326.
- Moran, M.S. (1994) Irrigation management in Arizona using satellites and airplanes, Irrig. Sci. 15:35-44.
- Moran, M.S., R.D. Jackson, T.R. Clarke, J. Qi, F. Cabot, K.J. Thome and B.L. Markham (1995) Reflectance factor retrieval from landsat TM and SPOT HRV data for bright and dark targets, Rem. Sens. Environ. 52:218-230.
- Moran, M.S., A. Vidal, D. Troufleau, Y. Inoue, J. Qi, T.R. Clarke, P.J. Pinter Jr., T.A. Mitchell and C.M.U. Neale (1997) Combining multi-frequency microwave and optical data for farm management, Rem. Sens. Environ. 61:96-109.
- Nouvellon, Y., A. Bégué, M.S. Moran, D. Lo Seen, S. Rambal, D. Luquet, A.G. Chehbouni and Y. Inoue (2000) PAR extinction in shortgrass-ecosystems: Effects of clumping, sky conditions and soil albedo. Ag. For. Meteorol. (this issue).
- Nouvellon, Y., D. Lo Seen, A. Bégué, S. Rambal, M.S. Moran, J. Qi, A. Chehbouni and Y. Kerr (1998) Combining remote sensing and plant growth modeling to describe the carbon and water budget of semi-arid grasslands, IGARSS, 6-10 July, Seattle, Washington, p. 2393-2398.
- Qi, J., Y.H. Kerr, M.S. Moran, M. Weltz, A.R. Huete, S. Sorooshian and R. Bryant (2000) Leaf area index estimates using remotely sensed data and BRDF models in a semi-arid region, Rem. Sens. Env. (In press).
- Sano, E.E. (1997) Sensitivity analysis of C- and Ku-band synthetic aperture radar data to soil moisture content in a semiarid regions, Ph.D. Dissertation, Univ. of Arizona, Dept. of Soil, Water and Environmental Sci., Tucson, Arizona, 122p.
- Sano, E.E., J. Qi, A.R. Huete and M.S. Moran (1998) The use of SAR/TM synergy for estimating soil moisture content over a semi-arid rangeland, Proc. 2nd Latino-American Seminar on Radar Remote Sensing INPE/ESA, Santos, SP, 11-12 September, ESA-sp-434.

- Simanton, J.R., Dixon, R.M., and McGowan, I. (1978) A microroughness meter for evaluating rainwater infiltration. In Hydrology and Water Resources in Arizona and the Southwest, American Water Resources Association/Arizona-Nevada Academy of Science.
- Troufleau, D., A. Vidal, A. Beaudoin, M.S. Moran, M.A. Weltz, D.C. Goodrich, J. Washburne and A.F. Rahman (1997) Optical-microwave synergy for estimating surface sensible heat flux over a semi-arid rangeland, Rem. Sens. Rev. 15:113-132.
- Ulaby, F.T., C.T. Allen, G. Eger III and E. Kanemasu (1984) Relating the microwave backscattering coefficient to leaf area index, Rem. Sens. Env. 14:113-133.
- Ulaby, F.T., P.C. Dubois and J. Van Zyl (1996) Radar mapping of surface soil moisture, J. Hydrol. 184:57-84.
- Wang, J., E. Engman, J. Shiue, M. Ruzek and C. Steinmeier (1986) The SIR-B observations of microwave backscatter dependence on soil moisture, surface roughness and vegetation cover, IEEE Trans. Geosci. Rem. Sens. GE-24:510-516.

Figure 1. A graphic illustration of the SAR/optical approach for evaluating surface soil moisture developed by Sano (1997). The vertical distance of points A-C from the solid line is related directly to soil moisture content.

Figure 2. Photographs of the three experimental sites in the Upper San Pedro Basin (USPB) in southeast Arizona, named by the dominant vegetation type: Tobosa (*Hilaria mutica*), Sacaton (*Sporobolus wrightii*), and Creosote (*Larrea tridentata*).

Figure 3. Subscenes extracted from three of the eight ERS-2 C-band SAR scenes (left, grey scale) and Landsat TM scenes (right, false color) covering the study area.

Figure 4. Range of surface volumetric soil moisture conditions over the temporal and spatial range of this study.

Figure 5. Comparison of the total plant area index (TPAI) estimated *in situ* with the LAI2000 plant canopy analyzer and the brown leaf area index (BLAI) measured in the laboratory with the LI3100 leaf area meter, for all sites and all dates. The solid line is the best-fit regression line with a correlation coefficient $r^2 = 0.86$. The mean absolute difference (MAD) between the two measurements was 0.2.

Figure 6. Green leaf area index (GLAI) in September 1997 at the three field sites (Creosote, Sacaton and Tobosa) measured in the laboratory using the LI3000, and derived from the Landsat TM images.

Figure 7. The relation between C-band SAR backscatter and surface (5cm) soil moisture content for three sites [labeled S (Sacaton), C (Creosote) and T (Tobosa)] and six dates in January, March, April, June, August and September.

Figure 8. Same as Figure 7, except the SAR backscatter was normalized for differences in surface roughness by subtracting the July SAR backscatter signal from the backscatter signal of the other dates.

The solid line is the best-fit regression line with a correlation coefficient $r^2 = 0.93$.

Figure 9. Regional maps of surface volumetric soil moisture based on ERS-2 C-band SAR images and the relation between $\sigma^\circ - \sigma^\circ_{\text{dry}}$ and ϵ_s presented in Figure 8. The maps of these two dates show a good contrast between regional soil moisture conditions in winter and spring 1997. The solid red colors in the lower left and right corners are areas that weren't covered by the SAR image.

Table 1. Dates of acquisitions of Landsat TM and ERS-2 SAR image pairs and dates of collection of supporting data at the Sacaton (S), Tobosa (T) and Creosote (C) sites.

Landsat TM	ERS-2 SAR	Vegetation Sampling	Soil Moisture
11/12/96	*11/3/96		
1/15/97	1/12/97		1/12/97
†2/16/97	2/16/97		2/16/97
3/20/97	3/23/97	S: 3/20/97 T: 4/3/97 C: 4/8/97	3/23/97
4/21/97	4/27/97	T: 4/14/97 S: 4/21/97 C: 5/5/97 T: 5/12/97	4/27/97
6/8/97	6/1/97	S: 6/2/97 C: 6/9/97 T: 6/16/97	6/1/97
7/10/97	7/6/97	C: 7/14/97 S: 7/21/97 T: 7/28/97	7/6/97
*8/11/97	8/10/97	C: 8/4/97 S: 8/9/97 T: 8/19/97	8/10/97

9/12/97	9/14/97	C: 8/26/97 S: 9/2/97 T: 9/8/97	9/14/97
10/14/97	*10/19/97	C: 10/6/97 S:10/14/97 T: 10/20/97	10/19/97
* Ordered but not acquired † Field sites occluded by clouds			

Table 2. Average surface soil roughness values for the Sacaton and Creosote sites. Separate measurements in the interspace and under creosote cover were taken to account for discrete roughness changes between measurement points.

Site	Roughness Value (cm)
Sacaton	0.53 ± 0.15
Creosote (interspace)	0.55 ± 0.23
Creosote (under shrub)	1.68 ± 0.56

Table 3. Surface (5 cm) volumetric soil moisture content (%) in 1997 at three sites in the Upper San Pedro River Basin, Arizona.

Date	Volumetric Soil Moisture (m ³ /m ³)			Notes
	Sacaton	Creosote	Tobosa	
1/12/97	28.2±5.1	9.0±1.4	19.3±6.1	It snowed in early January, and the snow had melted by the time of the ERS-2 overpass. Thus, the soil conditions were near field-capacity throughout the region.
2/16/97	13.2±5.7	3.4±0.9	8.0±3.8	It rained several times in February, resulting in moist soil conditions.
3/23/97	7.3±4.3	1.1±0.8	3.7±1.1	There was minimal rain in March and April, and the soil conditions were moderately dry.
4/27/97	5.9±2.2	1.2±0.8	3.1±1.5	
6/1/97	3.5±1.3	0.9±0.2	3.3±1.1	During the hot months of June and July, the soil at all sites was very dry, the grass was brown, and no annuals were present.
7/6/97	3.1±1.1	0.7±0.2	2.2±0.6	
8/10/97	27.6±7.3	8.7±1.7	13.1±7.5	A large storm preceded the August overpass, resulting in high soil moisture conditions in combination with peak vegetation greenness.
9/14/97	13.8±5.0	3.8±1.1	7.0±2.6	A small storm preceded the September overpass.
10/19/97	5.8±2.6	1.2±0.3	3.4±1.5	The soil conditions were dry and the vegetation was senescent.