A New Metric for Estimating the Disparity of Antenna Patterns in Synthetic Aperture Imaging Radiometry

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Abstract— The Soil Moisture and Ocean Salinity (SMOS) satellite has provided for about 10 years systematic passive L-band measurements from space. For several months, phase 0 studies are conducted by the French space agency for a second generation High Resolution (HR) follow-on mission. This contribution is making the connection between this SMOShr project and the SMOS mission by revisiting the following problematic: the impact of the disparity of the antenna patterns on the reconstruction floor error observed in the retrieved brightness temperatures. This impact is revisited in light of the progress made since that time and a new metric is introduced for estimating the disparity between antenna patterns. It would be helpful for the design of future missions based on imaging by aperture synthesis with interferometric arrays comprising a large number of antennas.

Index Terms—Imaging radiometry, aperture synthesis, interferometric array, antenna pattern.

I. INTRODUCTION

The Soil Moisture and Ocean Salinity (SMOS) space mission [1] was launched in November 2009 by ESA and CNES and for almost a decade this 1st generation satellite has provided accurate radiometric brightness temperature maps with a spatial resolution of ~ 50 km at L-band. These maps have been used for retrieving surface soil moisture (SM) [2] as well as ocean salinity (OS) [3]. In addition, these brightness temperatures are operationally monitored at the European Centre for Medium-Range Weather Forecasts (ECMWF) [4].

For about two years, phase 0 studies are conducted by the French space agency for a 2nd generation High Resolution (HR) follow-on mission [5]. The goal of this SMOShr project is to ensure the continuity of L-band measurements while increasing the spatial resolution to ~ 10 km, without degrading the radiometric sensitivity and keeping the revisit time of 3 days unchanged. This would require a typical antenna size of ~ 18 m. Taking into account the difficulty of deployment of a real aperture of this size in space [6] and the successful

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alternative approach used for SMOS [7], the 2nd generation SMOS*hr* will also perform aperture synthesis [8] with three times more antennas than SMOS has [9].

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This work is making the connection between the two generations, the present SMOS and the future SMOShr, by revisiting a problematic that was discovered too late in the SMOS project for being solved in due time. This issue concerns the impact of the disparity of the antenna patterns [10] on the reconstruction floor error observed in the retrieved brightness temperatures [11]. The observing equation of an imaging radiometer by aperture synthesis is recalled in section II. The impact of the disparity of the antenna patterns is revisited in section III in light of the progress made since the time this issue was observed with SMOS. A new metric is introduced in section IV for estimating the disparity between antenna patterns. It is illustrated with the antenna patterns of SMOS measured prior launch in an anechoic chamber [12]. Finally, it is shown how this new metric would have been useful for detecting those antennas in the interferometric array with a pattern outside a confidence region and which contribute too much to the reconstruction floor error and therefore to the degradation of the mission performances.

II. OBSERVING EQUATION OF INTERFEROMETRIC ARRAYS

While classical radiometers measure the power collected by a highly directive antenna, interferometric measurements with an aperture synthesis radiometer are obtained by crosscorrelating the signals collected by pairs of non-directive antennas having overlapping fields of view. As a consequence, if total power radiometers provide direct measurements of the brightness temperature in the main beam direction of the antenna, aperture synthesis ones required inversion of the socalled complex visibilities [13] with the aid of the computer in order to retrieve the brightness temperature distribution in front of the instrument [14]. The collection of antennas, also referred as the interferometric array in aperture synthesis, as well as their characteristics, play a key role in this inversion process and therefore in the performances of the instrument.

The relationship between the complex visibilities and the brightness temperature of the scene under observation has been revisited in order to take into account the mutual coupling of close antennas [13]. Without polarimetric considerations [15], the complex visibility V_{pq} for a pair of antennas \mathcal{A}_p and \mathcal{A}_q

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is given by:

$$V_{pq} = \frac{1}{\sqrt{\Omega_p \Omega_q}} \int_{\xi^2 + \eta^2 \le 1} F_p(\xi, \eta) \overline{F_q}(\xi, \eta) \left(T_b(\xi, \eta) - T_{rec} \right) \\ \times \widetilde{r}_{pq}(t) \frac{e^{-2j\pi \left(u_{pq}\xi + v_{pq}\eta + w_{pq}\sqrt{1 - \xi^2 - \eta^2} \right)}}{\sqrt{1 - \xi^2 - \eta^2}} \, \mathrm{d}\xi \mathrm{d}\eta, \tag{1}$$

where (u_{pq}, v_{pq}, w_{pq}) are the components of the baseline vector \mathbf{b}_{pq} normalized to the central wavelength of observation λ , $\xi = \sin \theta \cos \phi$ and $\eta = \sin \theta \sin \phi$ are direction cosines in the reference frame of the arrays, θ and ϕ are the classical spherical co-ordinates (the colatitude and the azimuth, respectively), T_{rec} is the physical temperature of the receivers (assumed to be the same for all the receivers), F_p and F_q are the normalized voltage patterns of the two antennas with equivalent solid angles Ω_p and Ω_q (the overbar indicates the complex conjugate), \tilde{r}_{pq} is the fringe-washing function which accounts for spatial decorrelation effects and finally $t = (u_{pq}\xi + v_{pq}\eta + w_{pq}\sqrt{1 - \xi^2 - \eta^2})/f$ is the spatial delay with $f = c/\lambda$ the central frequency of observation.

Referring back to (1), after discretization of the integral over an appropriate sampling grid in the direction cosines domain, the relationship between the complex visibilities Vand the brightness temperature distribution T_b of the scene under observation can be written in the linear form:

$$V = \mathbf{G}T,\tag{2}$$

where **G** is the modeling operator of the instrument. The inverse problem aims at retrieving $T = T_b$ (the constant T_{rec} is cancel out from the visibilities with the aid of the response to a flat target [16], whatever the method used for the inversion [17]). As this inverse problem is ill-posed, it has to be regularized in order to provide a unique and stable solution for T. Among the regularization methods, the minimum-norm solution is widely used in imaging radiometry [18]:

$$T_r = \min_{T} ||T||^2 \text{ s.t. } \mathbf{G}T = V.$$
 (3)

This map is obtained through the computation of the pseudoinverse \mathbf{G}^+ of \mathbf{G} :

$$T_r = \mathbf{G}^+ V, \tag{4}$$

where \mathbf{G}^+ is computed with the aid of a truncated singular values decomposition of \mathbf{G} which discards the smallest singular values prior the inversion. In order to filter out the Gibbs effects due to the sharp frequency cut-off associated to the limited extent of the experimental frequency coverage H of the instrument, T_r has to be damped by an appropriate windowing function: $\mathcal{T}_r = W \star T_r$ [19]. This map has to be compared to $\mathcal{T}_w = W \star T_b$ (which is the "ideal" temperature map to be reconstructed and apodized with the same window W) and not to T_b (which is not at the same spatial resolution).

Early in the project of the SMOS mission, numerical simulations have been conducted to analyze the performances of the inversion process. A systematic error has been observed in the retrieved brightness temperature maps when comparing T_r to T_w , although no modeling errors were introduced in **G** and no noise was added to V. It has been shown that

this reconstruction floor error cannot be attributed to the high frequencies components of the scene T_b since they are filtered out by the instrument itself. Two origins have been investigated [20]: an instrument dependent component which has revealed the key role played by the voltage patterns of the antennas [21], and a scene dependent component which has shown the influence of the aliased regions of the synthesized field of view [22]. If various approaches have been successfully proposed and implemented for reducing the effect of the latter component [23], nothing has been done for the former one [24].

III. INFLUENCE OF ANTENNA PATTERNS

The influence of the antennas and especially that of the disparity of the voltage patterns F_p and F_q from one baseline b_{pq} to the other has been analyzed in [20] with the aid of numerical simulations conducted within the frame of the SMOS project. However, the antenna voltage patterns introduced in the modeling matrix **G** were derived from a very simple model [25] and were not the radiation patterns of the 69 antennas of SMOS which were not available at that time [12]. As there are evidences that models of radiation patterns do not reproduce measurements made on antennas in an anechoic chamber, this study is briefly revisited hereafter.

Shown on Figure 1 are the distributions of the singular values of the operator **G** with the current modeling of SMOS instrument which includes those antenna patterns measured prior launch [12]. As the average pattern $\langle F \rangle$ plays a key role in the analysis, the same distribution is shown for a virtual instrument like SMOS but equipped with equal antennas sharing the same voltage pattern $\langle F \rangle$. In both cases, there are clearly two groups of singular values with a very well-determined gap in between: this situation has been already explained [26]. Although here the two modeling operators are sharing exactly the same first group of singular values, the second groups with the smallest ones are very different. Owing to the role played by these smallest singular values in the pseudo-inversion of **G**, only those lying in the first group



Fig. 1. Distributions of the singular values of the modeling operator **G** in X polarization. Two groups of singular values separated by a well-determined gap are observed. However, in the ideal case (green, same voltage pattern for each antenna), this gap is wider than in the present case of SMOS (red, 69 different antenna patterns). The same behavior is observed in Y polarization.



Fig. 2. Typical error maps $T_r - T_w$ of simulations performed in X polarization with the current modeling of SMOS (left, 69 different antenna patterns) and with that of a virtual instrument like SMOS (right, same voltage pattern for each antenna). With the disparity of the 69 antenna patterns of SMOS, the bias and the standard deviation of the error maps in that part of the field of view free from Earth aliases are respectively about 0.7 K and 1.8 K. On the contrary, when the disparity is reduced to zero they are respectively less than 0.01 K and 0.05 K. The same behavior is observed in Y polarization.

are kept in the computation of \mathbf{G}^+ with the aid of the socalled truncated singular value decomposition. As illustrated in Figure 2 with error maps $\mathcal{T}_r - \mathcal{T}_w$ corresponding to these two cases, the wider this gap between the two groups of singular values, the smaller is the reconstruction floor error. On this example, with the disparity of the 69 different antenna patterns F of SMOS, the bias and the standard deviation of the error maps are respectively about 0.7 K and 1.8 K whereas they are respectively less than 0.01 K and 0.05 K when the disparity is reduced to zero with equal antennas sharing the same voltage pattern $\langle F \rangle$.

Between these two extreme cases, intermediate situations can be studied by substituting $\langle F \rangle + (F - \langle F \rangle) / \alpha$ for each F as it is a convenient way to play with the disparity between the radiation patterns in the modeling of the instrument. The case of the present SMOS with 69 different antenna patterns is obtained with $\alpha = 1$. That of the previous instrument equipped with identical antennas sharing the same voltage pattern $\langle F \rangle$ is achieved when $\alpha \to +\infty$. For any $\alpha > 1$, the disparity between the 69 patterns is α times smaller than that of SMOS. On the contrary, for any $\alpha < 1$, it is $1/\alpha$ times larger. Thanks to this approach it is possible to study the impact of this disparity onto the inversion process. This is what is shown on Figure 3 where the bias and the standard deviation of reconstruction error maps $T_r - T_w$ are reported as a function of the parameter α . As suggested by the two cases reported in Figure 2, $\alpha = 1$ and $\alpha \to +\infty$, the reconstruction floor error is clearly a decreasing function of the parameter α : the larger the disparity between the 69 different patterns, the larger the bias and the standard deviation in the error maps.

The parameter α has been introduced for controlling the disparity between SMOS antenna patterns. Although it has been a useful approach for illustrating the impact of this disparity onto the reconstruction floor error, it is suffering from some drawbacks. The main negative point of this approach lies in its inability to quantify the disparity between the antenna patterns. It can only compare it to the one observed in the 69 antenna patterns of SMOS. Moreover, it cannot distinguish



Fig. 3. Bias (bullets) and standard deviation (bars) of the floor error as a function of the factor α used for changing the disparity between the 69 radiation patterns in the modeling of an instrument like SMOS in X polarization. For $\alpha = 1$ the situation is that of SMOS (red) while for $\alpha \rightarrow +\infty$ it is that of an instrument like SMOS (green) but with the same voltage pattern for each antenna (these two cases are those illustrated in Figures 1 and 2). The reconstruction floor error is a decreasing function of α and when the disparity between the 69 radiation patterns is attenuated by a factor larger than 10 it becomes negligible. The same behavior is observed in Y polarization.

the patterns which are strong contributors to the reconstruction floor error from those contributing a little.

IV. COMPARISON OF ANTENNA PATTERNS

There has been many attempts to define objective quantitative metrics of comparison between radiation patterns that can be used to assess the accuracy, the sensitivity and also the repeatability associated with the manufacturing of antennas [27]. None is perfect because they are all suffering from drawbacks. Many of these figures of merit are only suited for the comparison of power patterns because they are not sensitive to the phase pattern. Moreover, none of them has been specially designed for large data sets [28] like those involved in imaging radiometry by aperture synthesis (SMOS is equipped with 69 antennas [7], preliminary design studies of SMOShr 4

are using 231 antennas [9]). There is a broad consensus that such a comparison is not an easy task. As a consequence, accounting for the remark drawn at the end of the last section, there is a need for such a metric for comparing large number of antennas and for having an objective way to reject those not satisfying specifications or quality requirements.

Parametric approaches have already been used in the attempt to self-characterize antenna patterns with few parameters [29]. However, they have never proven to be well adapted to experimental radiation patterns which are not rigorously reproduced by such models with too few parameters. As an example, when focusing on the main beam of the power patterns of the 69 antennas of SMOS, one cannot see any outliers in the distribution of the Half-Power Beam Width (HPBW) nor in that of the First Null Beam Width (FNBW). Nevertheless, a simple view at the 69 amplitude and phase patterns shows large discrepancies from one antenna to another. This example makes it clear that devil is in the details, details which are not seen nor reproduced by parametric approaches. As a consequence, this new metric should rely only on measurements and not on any radiation pattern modeling.

Coming back to the observing equation (1), the equivalent solid angle Ω of any antenna is given by:

$$\Omega = \iint_{\xi^2 + \eta^2 \le 1} \frac{|F(\xi, \eta)|^2}{\sqrt{1 - \xi^2 - \eta^2}} \,\mathrm{d}\xi \mathrm{d}\eta.$$
(5)

As this definition involves the power pattern $|F|^2$ (see chap. 2 in [30]), Ω is a real-valued quantity although the voltage pattern F may be a complex-valued one. Keeping in mind the writing of (1), this definition suggests to introduce a similar quantity but for a pair of antennas \mathcal{A}_p and \mathcal{A}_q so that both antenna voltage patterns F_p and F_q contribute simultaneously to a unique complex value:

$$\Omega_{pq} \equiv \int_{\xi^2 + \eta^2 \le 1} \frac{F_p(\xi, \eta) \overline{F}_q(\xi, \eta)}{\sqrt{1 - \xi^2 - \eta^2}} \, \mathrm{d}\xi \mathrm{d}\eta, \tag{6}$$

which carries information on the amplitude and on the phase of the two voltage patterns. This new quantity can be named an "equivalent cross solid angle". It satisfies the conjugation relation $\Omega_{qp} = \overline{\Omega}_{pq}$ and it reduces to the standard equivalent solid angles $\Omega_p = \Omega_{pp}$ or $\Omega_q = \Omega_{qq}$ when p = q. As $\Omega_{pq}/\sqrt{\Omega_p \Omega_q}$ is a complex sesquilinear form, an inner product between two voltage patterns can be defined accordingly:

$$\langle F_p \mid F_q \rangle \equiv \frac{1}{\sqrt{\Omega_p \Omega_q}} \int_{\xi^2 + \eta^2 \le 1} \frac{F_p(\xi, \eta) \overline{F}_q(\xi, \eta)}{\sqrt{1 - \xi^2 - \eta^2}} \, \mathrm{d}\xi \mathrm{d}\eta. \tag{7}$$

When $F_p = F_q = F$ we therefore have $\langle F | F \rangle \equiv ||F||^2 = 1$: the norm of any voltage pattern is equal to one according to this inner product. Although this remark is purely anecdotal because voltage patterns have nothing to do with them, this definition reminds that of the inner product between Chebyshev polynomials of the first kind [31]. Actually, this definition is very similar to that of the envelope correlation ρ_e defined by equation (1) in [32]. However, as $\rho_e \equiv |\langle F_p | F_q \rangle|^2$ is a real-valued quantity, the phase information transported by ρ_e is very limited compared to that carried by $\langle F_p | F_q \rangle$. As the idea is not to compare all the antennas of an interferometric array to each other, but only to compare them to a given reference antenna \mathcal{A}_{ref} in order to have a estimate of the distance between any voltage pattern F and the voltage pattern F_{ref} of the reference antenna, the following quantity is useful:

$$\langle F_{ref} \mid F \rangle \equiv \frac{1}{\sqrt{\Omega_{ref}\Omega}} \int_{\xi^2 + \eta^2 \le 1} \frac{F_{ref}(\xi,\eta)\overline{F}(\xi,\eta)}{\sqrt{1 - \xi^2 - \eta^2}} \, \mathrm{d}\xi \mathrm{d}\eta. \quad (8)$$

The reference antenna A_{ref} may be, for example, a specification given to a manufacturer or the output model of a simulation. In this contribution, the voltage pattern F_{ref} is the average voltage pattern $\langle F \rangle$ of the 69 radiation patterns F of SMOS that have been measured with accuracy [33] prior launch in an anechoic chamber [12]. The distribution of the corresponding inner products is therefore an indicator of the disparity of the 69 different patterns with respect to the average one: they are shown on Figure 4 in X and Y polarizations. If most patterns are confined inside the 3σ / 99.73% confidence ellipses, many of them are outside the $1 \sigma / 68.27\%$ ones [34]. The closer to $||F_{ref}||^2 \equiv 1$ is the complex value $\langle F_{ref} | F \rangle$, the closer this F is to F_{ref} . On the contrary, the farthest pattern from the reference pattern is that one whose $\langle F_{ref} | F \rangle$ is the farthest from 1. These patterns are shown in Figure 5 in X polarization: with a scalar product equal to 1.0013 + 0.0002i, the closest one is that of antenna C20 (according to the naming of the antennas of SMOS found in Figure 1 of [35]). On the contrary, the farthest one is that of antenna AB03 with 0.9918 - 0.0886i. The qualitative comparizon of the amplitude and phase patterns of these antennas is not in contradiction with what is deduced from this new metric. Similar results are obtained in Y polarization: the closest pattern is that of antenna B06 (1.0012 + 0.0007i) whereas the farthest one is that of antenna CA03 (0.9918 - 0.0982i).

Clearly identifiable on Figure 4 are the voltage patterns with their inner products outside the $1\,\sigma$ / 68.27% confidence



Fig. 4. Inner products (dots) between the voltage patterns F of the 69 antennas of SMOS and a reference voltage pattern F_{ref} (here equal to the average of the 69 radiation patterns) in X (red) and Y (blue) polarizations. The $1 \sigma / 68.27\%$ and the $3 \sigma / 99.73\%$ confidence ellipses of the complex-valued distributions $\langle F_{ref} | F \rangle$ are shown with dotted-lines and dashed-lines. The closer to $1 = ||F_{ref}||^2$ is a dot, the closer the corresponding F is to F_{ref} . Attention should be paid on the different scales along both axis.



Fig. 5. Amplitude (bottom) and phase (top) patterns of three typical voltage patterns out of the 69 antennas of SMOS in X polarization: average pattern F_{ref} (center), patterns F closest to (left, antenna C20) and farthest from (right, antenna AB03) the average one according to the distance from the complex-valued scalar products $\langle F_{ref} | F \rangle$ to $||F_{ref}||^2 = 1$, as illustrated in Figure 4. Here $\langle F_{ref} | F_{C20} \rangle \simeq 1.0013 + 0.0002i$ and $\langle F_{ref} | F_{AB03} \rangle \simeq 0.9918 - 0.0886i$, so that these distances are respectively about 0.0013 for antenna C20 and 0.0890 for AB03 (whether by coincidence or not, that is a ratio close to 69).

ellipses: 15 in X polarization and 17 in Y polarization, out of 69, are concerned by this issue. The corresponding antennas are localized on SMOS instrument in Figure 6. Among the 22 unique antennas concerned, it should be noted that 10 are outside the confidence ellipse in both X and Y polarizations. If we were at the time of accepting or rejecting antennas coming out from an assembly chain, this new metric would suggest to



Fig. 6. Localization of the 15 antennas of SMOS in X polarization (red) and the 17 ones in Y polarization (blue) which have their voltage pattern F with an inner product $\langle F_{ref} | F \rangle$ outside the $1 \sigma / 68.27\%$ confidence ellipse, as shown on Figure 4. Among the 22 unique antennas concerned by this issue, 10 are outside the confidence ellipse in both polarizations.

pay attention to these ones and to take appropriate decisions. As this time is over, the present study can only be completed with simulations in order to check the impact of these patterns on the reconstruction floor error by reducing artificially their distance to the average pattern with the aid of the parameter α introduced in the previous section. Consequently, to be clear, as SMOS is now in orbit and visibilities have to be inverted with the same antenna patterns as those at the time they were measured, the modifications made hereafter are purely virtual, illustrative and in no way an improvement that could be brought to SMOS now.

Referring back to the simulations of Figure 2, with the disparity of the 69 antenna patterns of SMOS, the bias and the standard deviation of the reconstruction error map are about 0.7 K and 1.8 K. Shown on Figure 7 is the reconstruction error map obtained with the same modeling of SMOS but with those 10 antenna patterns identified outside the 1σ / 68.27% confidence ellipses modified appropriately with $\alpha = 2$. With this virtual modeling the two components of the floor error are reduced down to 0.5 K and 1.3 K. This is a clear illustration of the impact of the disparity, and specially here on the main contributing antennas, onto the reconstruction floor error. Of course, the reduction is not as important as it would be if all the antennas have benefited from this improvement, as illustrated in Figure 3 where the two components of the floor error are equal to 0.3 K and 0.9 K for $\alpha = 2$. However, this virtual



Fig. 7. Error map $\mathcal{T}_r - \mathcal{T}_w$ of simulations performed in X polarization with the current modeling of SMOS equipped with 69 different antenna patterns but with 10 of them identified outside the $1 \sigma / 68.27\%$ confidence ellipses (and only those ones) modified appropriately with $\alpha = 2$. The bias and the standard deviation in that part of the field of view free from Earth aliases are about 0.5 K and 1.3 K. This map has to be compared to the map shown on the left hand of Figure 2 without any modification of these patterns. The same behavior is observed in Y polarization.

modification of only 10 out of the 69 patterns made it possible to make half the distance between the present situation of SMOS and a virtual one where the disparity of all 69 patterns would be twice smaller.

As shown on Figure 8 the previous virtual reduction is fully consistent with the decrease of the floor error as the area of the 1σ / 68.27% confidence ellipses decreases. It should be pointed out that here as only 10 patterns have been modified, this virtual case cannot be plotted on Figure 3 where the floor error is reported only when all 69 patterns have been modified. Hence this Figure 8 which does not presuppose the number of modified patterns nor how they were modified, contrary to Figure 3. Finally, a connection between Figures 4 and 8 can be made through the comparison between the distances from the scalar products $\langle F_{ref} | F \rangle$ to $||F_{ref}||^2 = 1$ and the radii of the disks equivalent to the 1σ / 68.27% confidence ellipses, for example. As a numerical illustration, this radius is equal to 0.0068 for this virtual case shown on Figure 7 and reported in orange on Figure 8. This value can be compared to the distances calculated in Figure 5 for the closest (0.0013) and the farthest (0.0890) antenna patterns.

V. CONCLUSION

A new metric has been introduced for estimating the disparity between voltage patterns of large data sets of antennas, like those involved in imaging radiometry by aperture synthesis. It is derived from the definition of the equivalent solid angle and inspired from the observing equation of the interferometry in aperture synthesis. Contrary to the well-known equivalent solid angle which is a real-valued quantity of a single antenna, this equivalent cross solid angle is a complex-valued quantity which concerns two antennas. Unlike many other figures of merit, it carries information on the amplitude and on the phase of the two voltage patterns. It can be useful for assessing the accuracy with respect to a given numerical model as well as for



Fig. 8. Same as Figure 3 but here the floor error is reported as a function of the area of the 1σ / 68.27% confidence ellipse in X polarization. In addition to the situation of the current SMOS (red) and to that of an ideal instrument with the same voltage pattern for each antenna (green), the case of an instrument like SMOS (orange) but with 10 antenna patterns identified outside the 1σ / 68.27% confidence ellipses modified appropriately with $\alpha = 2$ is reported for comparison (this case is illustrated in Figure 7). The same behavior is observed in Y polarization.

evaluating the repeatability associated with the manufacturing of many antennas.

This work is making the connection between the present SMOS mission and the future SMOS*hr* one.

Within the frame of the SMOS mission, the impact of the dissimilarity of the antenna patterns on the reconstruction error observed in the retrieved brightness temperatures has been revisited in light of the progress made since the preliminary design of the instrument. The average voltage pattern has proven to play a particular role in the floor level of the reconstruction error. The usefulness of this new metric has been illustrated with the 69 voltage patterns of SMOS that have been measured prior launch. It has been shown how it is easy to identify those antennas that are the most contributors to the reconstruction floor error because their patterns are too different from the average pattern with respect to this new metric. If this metric had been available at the time these antennas have been built and assembled, it would have been possible to detect them and to conduct specific studies to investigate solutions for reducing the dissimilarity of their patterns and to explore the possibility of manufacturing others with different requirements connected to the target level for the reconstruction floor error in the retrieved brightness temperatures.

Within the frame of the SMOS*hr* project, a research and technology study is currently conducted at industrial level on the design of the future antennas for this new generation of imaging radiometer by aperture synthesis with high resolution performances. Thanks to the lessons learned from SMOS, finite element simulations of the radiation patterns of the 231 antennas are carried on with the aid of the full-wave solver of CST Studio Suite [36], instead of using parametric models that have proven their total inefficiency. As co-design is the new paradigm [37], time consuming end-to-end simulations with engineers, scientists and end-users actively involved are playing an ongoing role for assessing the sensitivity and the

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robustness of mission performances to driving parameters, to instrument errors and noises as well as to data processing. This new metric is therefore particularly useful for this purpose, as soon as antenna patterns are all the more concerned in the reduction of the reconstruction floor error as they are many.

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