A beamforming approach to the self-calibration of phased arrays

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Abstract—In this paper, we propose a beamforming method for the calibration of the direction-independent gain of the analog chains of aperture arrays. The gain estimates are obtained by cross-correlating the output voltage of each antenna with a voltage beamformed using the other antennas of the array. When the beamforming weights are equal to the average cross-correlated power, a relation is drawn with the StEFCal algorithm. An example illustrates this approach for few point sources and a 256-element array.

Index Terms—Self-calibration, calibration, beamforming, radio-interferometry, Square Kilometer Array

I. INTRODUCTION

The calibration of the aperture array stations of the Square Kilometer Array (SKA) telescope [1] and of its precursor and pathfinder, the Murchison Widefield Array (MWA) [2] and the Low-Frequency Array (LOFAR) [3], is essential to achieve their expected nominal performance. The increased bandwidth and sensitivity, which has led to large irregular arrays of wideband antennas, inherently renders the station calibration more challenging due to complex coupling effects [4], each element being in a different local environment, and a higher data throughput [5].

The SKA1-low instrument will consist of 512 stations of 256 elements [1] covering the 50-350 MHz band. The analog chain associated to each log-periodic antennas [6] will include low-noise amplifiers and other analog components connected via coaxial cables or optical fibers to remote analog-to-digital converters. The station beamforming will be performed digitally and thus relies on an accurate calibration of the analog chain. Given the limited storage capabilities of the digital backend and the fast time variability of the gains accounting for the propagation of the signals through the analog chain, one currently considers real-time calibrations [5].

Self-calibration [7] is a calibration technique which uses a model of the sky and of the embedded element patterns (EEP) to solve for the gains. Most algorithms implementing self-calibration formulate the problem as a quadratic system of equations based on a correlation matrix containing all the pairs of cross-correlation between the antenna voltages. For a station with many elements, those approaches are computationally expensive since they scale at least quadratically w.r.t. the number of antennas [8], i.e. $O(N^2)$ for $N$ antennas. Recently, [9] proposed a promising technique, called self-holography, which exhibits a linear complexity $O(N)$ and basically consists of cross-correlating each antenna voltage with the signal from a single beam formed by the same station pointing towards strong sources. Overall, beamforming approaches may help to decrease the computational complexity of the calibration since they avoid the construction of the correlation matrix.

This paper presents a method for the gain calibration problem based on cross-correlating each antenna voltage with a voltage beamformed from the other antennas. The difference with [9] is that the beamforming weights, and thus the beams, are allowed to vary for each cross-correlation. This formulation should be well-suited to further accelerations using fast beamforming methods. We will then analyse the convergence and the accuracy of the numerical method for a simple scenario with few point sources and an irregular array similar to the SKA Aperture Array Verification System I (AAVS1) [4].

II. BEAMFORMING-BASED GAIN ESTIMATION

The voltage $v_i(t)$ of antenna $i$ measured at the input port of the receiver can be decomposed into

$$v_i(t) = g_i s_i(t) + n_i(t) \quad (1)$$

where $g_i$ is a complex-valued gain accounting for the $i$th receiving path, $s_i(t)$ is the signal coming from the sky and $n_i(t)$ is the noise produced by the receiving system. The signals are filtered to a bandwidth $B$ and sampled in time at the Nyquist frequency $1/(2B)$ for a duration $T$. The response of the overall system is assumed frequency-flat over the bandwidth $B$.

The calibration technique, developed next, iteratively finds the unknown gains $g_i$ by cross-correlating the voltage of a single element $i$ with a voltage beamformed using all the other elements, as illustrated in Fig. 1. At each iteration, the gains of the beamformed array are assumed perfectly known s.t. only $g_i$ needs to be found. Intuitively, this approximation is relevant when the errors on the gains average out during the beamforming operation.
The cross-correlation product between the voltage $v_i(t)$ of antenna $i$ and the voltage $b_j(t)$ at the output of the array beamformer is measured by

$$R_{bi} = \frac{1}{M} \sum_{m} v_i(t_m) b_j(t_m)^*$$  \hspace{1cm} (2)

where $* \text{ stands for complex conjugate and } M = 2BT \text{ is the number of time samples } t_m = m/(2B). \text{ As illustrated in Fig. 1 for } i = 2, \text{ the beamformed voltage is computed with}$

$$b_j(t) = \hat{w}_i \cdot v(t)$$  \hspace{1cm} (3)

where $\hat{w}_i$ is a normalized vector containing the beamforming weights $\hat{w}_{ij}$ and the antenna voltages have been stacked into the vector $v(t).$

Assuming ergodicity and uncorrelated noises $n_i$, the mean over time of the correlation product $P_{ij}$ is equal to the ensemble average $\langle v_i v_j^* \rangle$ and is given by

$$P_{ij} = k_b \left( g_i T_{ij} g_j^* + T_{ni} \delta_{ij} \right) B$$  \hspace{1cm} (4)

where $k_b$ is the Boltzmann’s constant, $\delta_{ij}$ is the Kronecker’s delta, $T_{ij} = \langle s_i s_j^* \rangle/(k_b B)$ and $T_{ni} = \langle n_i n_j^* \rangle/(k_b B)$ are, respectively, the correlated antenna temperature and the noise temperature. In self-calibration, the correlation product $\langle s_i s_j^* \rangle$ is known and can be evaluated numerically from models of the EEPs and of the source intensities [10].

Using (3) and (4), the correlated power $P_{bi} = \langle v_i b_i^* \rangle$ is expressed as

$$P_{bi} = k_b \left( g_i T_{bi} + \hat{w}_{bi} T_{ni} \right) B$$  \hspace{1cm} (5)

where $T_{bi} = \hat{w}_i^* \cdot (g^* \otimes T_i)$ is the beamformed array temperature with the element-wise product $\otimes$, the gain vector $g$ and the vector $T_i$ standing for the $i$th row of the matrix $T$ which contains all the antenna temperatures $T_{ij}$. Assuming that a guess for $T_{bi}$ has been obtained at the previous iteration and that the signals $s_i$ are ergodic, we can make the approximation $P_{bi} \approx R_{bi}$ and isolate the gain $g_i$ in (5) to have

$$g_i = \frac{R_{bi} - \hat{w}_{bi}^* k_b T_{ni} B}{k_b T_{bi} B}$$  \hspace{1cm} (6)

The gain estimation with (6) can be viewed as two-step method with noise extraction followed by rescaling. It requires the knowledge of the noise temperature $T_{ni}$. When no noise information is available, one can cancel the auto-correlations in (2) by imposing $\hat{w}_{ii} = 0$.

Next, if we insert (2) into (6) and define the vector $\hat{g}$ containing the gain estimates $\hat{g}_i$, the following matrix form is obtained

$$\hat{g} = \left( \sum_{m} v(t_m) \otimes (Wv(t_m))^* \right) \otimes (Mk_b T_i B)$$  \hspace{1cm} (7)

where $\otimes$ denote the element-wise division, $W$ is a matrix whose lines are the weights $\hat{w}_i$ and $T_i$ is a vector containing the array temperatures $T_{bi}$. It is clear that a brute force evaluation of the matrix-vector product in (7) will lead to $O(N^2)$ complexity per time sample. However, it is expected that fast beamforming methods for irregular arrays will help to decrease the computational burden.

In the following example, we will only consider the beamforming weights

$$\hat{w}_i = \frac{g^* \otimes T_i}{\|g^* \otimes T_i\|}$$  \hspace{1cm} (8)

which maximize the average cross-correlated power $P_{bi}$ in (5). It can be proven that the particular choice of weights (8) leads to gain estimates $\hat{g}$ equal to those obtained after a single iteration of the StEFCal algorithm. Indeed, by comparing (7) with the relation (12) in [8] using the same notations, one can easily identify the equivalences $\hat{w}_p = Z_{-p}/\|Z_{-p}\|$ and $k_b T_{bi} B = Z_{-p} \cdot (Z_{-p})^\dagger$.

III. NUMERICAL EXAMPLE

We will now apply the proposed method to calibrate an irregular array of 256 SKALA4 antennas [6] with a random layout similar to the AAVS1 array [4]. The noise temperature of the system is fixed to $T_{ni} = 200$ K for each antenna. As shown in Fig. 2, we have created a sky map composed of 5 unresolved and unpolarized sources of intensities randomly distributed between 0.1 and 1 Jy. Both the signals $s_i(t)$ and $n_{ij}(t)$ are Gaussian random variables filtered to a bandwidth $B = 1$ MHz around the center frequency $f = 148$ MHz. The cross-correlations $R_{bi}$ in (6) are evaluated with an integration time $T = 0.1s$ leading to $M = 2 \times 10^5$ time samples. The
EEP of each antennas, necessary to compute the correlated temperatures $T_{ij}$ in (4), have been simulated with the HARP software [4] and takes into account the mutual coupling effect. The amplitude and phase of the true gains $g_i$ are randomly distributed within the intervals $[0, 2]$ and $[0, \pi]$, respectively. The gain estimates at the initial step are set to $\tilde{g}_i = 1$. The method is stopped after 3 iterations. Each iteration takes a full timeslot. Hence, the duration of the calibration is $0.3s$.

One can observe in Fig. 3 that the beam formed with the weights (8) at the third iteration is producing hot spots in the direction of the point sources. The source close to the horizon is an exception and has been weighted down by the EEPs since the field of view of the SKALA antenna is limited. In Fig. 4, one can see that the relative error, computed with $\epsilon = |\tilde{g}_i - g_i|/|g_i|$, has decreased from $0$ dB for the initial guess to around $-30$ dB after the third iteration. This thus provides a simple numerical validation of the proposed self-calibration technique.

IV. CONCLUSION

We have presented and validated a beamforming method based on the computation of the cross-correlation between each antenna voltage and the voltage formed by a beam using the other antennas of the station. When the beamforming weights match the average antenna cross-correlation, the method is equivalent to a single iteration of StEFCal algorithm. This thus ensures the convergence and the efficiency of the gain estimates for most practical cases.

In the way that fast methods have accelerated matrix-vector products in computational electromagnetics, fast beamforming methods, for instance using the Non-equispaced Fast Fourier Transform [11], will probably help to bring down the computational complexity from $N^2$ for classical approaches to $N \log_2 N$ per time sample.

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