

A Generic and Efficient E-field Parallel Imaging Correlator for Next-Generation Radio Telescopes

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ABSTRACT

Abstract here (250 words)

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1 INTRODUCTION

Radio astronomy is entering an era in which interferometers of hundreds to thousands of individual antennas are needed to achieve desired survey speeds. Nowhere is this more apparent than at radio frequencies below 1.4 GHz. The study of the history of hydrogen gas throughout the universe’s evolution is pushing technology development towards arrays of low-cost antennas with large fields of view and densely packed apertures. Similarly, the search for transient objects and regular monitoring of the time-dependent sky is driving instruments in the same direction. A number of new telescopes are under development around the world based on this new paradigm, including the Murchison Widefield Array (MWA, [Tingay et al. 2013](#)), the Precision Array for Probing the Epoch of Reionization (PAPER, [Parsons et al. 2010](#)), the Hydrogen Epoch of Reionization Array (HERA ¹), the Low Frequency ARray (LOFAR, [de Vos et al. 2009](#)), the Canadian Hydrogen Intensity Mapping Experiment (CHIME, [Bandura et al. 2014](#)), the Long Wavelength Array (LWA, [Ellingson et al. 2013](#)), and the low frequency component of the Square Kilometer Array Low Frequency Aperture Array (SKA-Low [Mellema et al. 2013](#)).

This paradigm shift requires a fundamentally new approach to the design of digital correlators ([Lonsdale et al. 2000](#)). Modern correlators calculate the cross-power correlation between all antenna pairs in many narrow frequencies, forming *visibilities*, the traditional fundamental measurement of radio interferometers. The computational requirements for a modern FX correlator scale with the number of antenna pairs, or the square of the number of antennas $\sim N_{\text{ant}}^2$ ([Bunton 2004](#)). For this reason traditional correlators have difficulty scaling to thousands of antennas. As an example, the full HERA correlator for 352 dishes with 200 MHz of bandwidth requires 212 trillion complex multiplies and adds per second (TMACS). Future arrays with thousands of collecting elements will

require orders of magnitude more computation, making the correlator the dominant cost.

For certain classes of radio arrays there is an alternative to the FX correlator that can lower the computational burden by directly performing a spatial fast Fourier transform (FFT) on the electric fields measured by each antenna in the array at each time step, removing the cross-correlation step. This relieves the computational scaling from the harsh N_{ant}^2 to the more gentle envelope of $N_{\text{pix}} \log N_{\text{pix}}$, where N_{pix} is the number of pixels in the Fourier transform (e.g. [Morales 2011](#); [Tegmark & Zaldarriaga 2009](#); [Tegmark & Zaldarriaga 2010](#)). This architecture is often referred to as a “direct imaging” correlator because it eliminates the intermediary cross-correlation data products of the FX and earlier lag correlators, but instead directly forms images from the electric field measurements.

Direct imaging correlators have begun to be explored on deployed arrays including the Basic Element for SKA Training II (BEST-2) array ([Foster et al. 2014](#)), the Omniscope ([Zheng et al. 2014](#)), and an earlier incarnation at higher frequencies with the intent of pulsar timing ([Otoabe et al. 1994](#); [Daishido et al. 2000](#)). However, each of these examples make assumptions about the redundancy of the array layout, and require the collecting elements are identical. On the other hand, the MOFF algorithm achieves the same $N_{\text{pix}} \log N_{\text{pix}}$ computational scaling without placing any restriction on antenna placement, can accommodate non-identical beam patterns, and is a provably optimal mapping ([Morales 2011](#)). This algorithm uses the antenna beam patterns to grid the electric field measurements to a regular grid in the software holography/A-transpose fashion ([Morales & Matejek 2009](#); [Bhatnagar, S. et al. 2008](#); [Tegmark 1997a](#)) before performing the spatial FFT. This process has been shown to theoretically produce a data product identical to images produced from the traditional FX correlator.

Here we present the first software implementation of the MOFF correlator, and announce the public release of the E-field Parallel Imaging Correlator (EPIC) code. We begin with a technical description of the algorithm in §2, then discuss our particular implementa-

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tion in §3. We then verify the output data quality from our code in §4 by presenting simulated images from both the EPIC correlator and comparing to a simulated FX correlator. We also demonstrate the performance with real-world data from the LWA. In §5 we analyze the scaling relationships of the algorithm. We identify specific array design classes where the EPIC correlator is computationally more efficient than the FX algorithm. We conclude and discuss future research prospects in §6.

2 MATHEMATICAL FRAMEWORK

We provide a quick summary of the mathematical equivalence of the MOFF and FX correlators detailed in Morales (2011).

Electric fields from astrophysical sources, $E(\hat{s})$, in the sky coordinate system denoted by unit vector \hat{s} propagate towards the observer as:

$$\tilde{E}(\mathbf{r}) = \int E(\hat{s}) e^{-i2\pi\mathbf{r}\cdot\hat{s}} d^2\hat{s}, \quad (1)$$

where, \mathbf{r} denotes the observer's coordinate system and $\tilde{E}(\mathbf{r})$ is the propagated electric field. Thus the propagated electric field is a linear superposition of the electric fields emanating from astronomical sources with appropriate complex phases. It can also be described as a Fourier transform of the electric fields in the sky coordinates.

An antenna, a , measures a phased sum of these propagated electric fields over its effective collecting area with an additive receiver noise:

$$\tilde{E}_a = \int \tilde{W}_a(\mathbf{r} - \mathbf{r}_a) \tilde{E}(\mathbf{r}) d^2\mathbf{r} + \tilde{n}_a \quad (2)$$

$$= \int \tilde{W}_a(\mathbf{r} - \mathbf{r}_a) \left[\int E(\hat{s}) e^{-i2\pi\mathbf{r}\cdot\hat{s}} d^2\hat{s} \right] d^2\mathbf{r} + \tilde{n}_a \quad (3)$$

$$= \int \delta(\mathbf{r} - \mathbf{r}_a) \left[\int W_a(\hat{s}) E(\hat{s}) e^{-i2\pi\mathbf{r}\cdot\hat{s}} d^2\hat{s} \right] d^2\mathbf{r} + \tilde{n}_a \quad (4)$$

where, $\tilde{W}_a(\mathbf{r})$ is the aperture electric field illumination pattern of the antenna and its Fourier transform, $W_a(\hat{s})$ is the directional antenna voltage response.

Interferometers measure *visibility* – the degree of coherence – between electric fields measured by a pair of antennas (van Cittert 1934; Zernike 1938; Thompson et al. 2001). Visibility, \tilde{V}_p , can be written as:

$$\tilde{V}_p = \langle \tilde{E}_a \tilde{E}_b^* \rangle \quad (5)$$

$$= \int \delta(\mathbf{u} - \mathbf{u}_p) \left[\int e^{-i2\pi\mathbf{u}\cdot\hat{s}} B(\hat{s}) I(\hat{s}) d^2\hat{s} \right] d^2\mathbf{u} + n_p, \quad (6)$$

where, $I(\hat{s})$ is the sky brightness, $B(\hat{s})$ is the antenna power response, \mathbf{u} denotes the coordinates of the uv - measurement plane, $\mathbf{u}_p \equiv \mathbf{r}_a - \mathbf{r}_b$ is the antenna spacing vector corresponding to the antenna pair p , and n_p is the receiver noise. This signifies that the visibility (\tilde{V}_p) measured between a pair of antennas (p) is obtained by the multiplying the sky brightness $I(\hat{s})$ by the antenna power response $B(\hat{s})$ and Fourier transform from the directional coordinates (\hat{s}) to uv coordinates, which are then sampled at the locations of the antenna spacings (or baselines), namely, \mathbf{u}_p using the sampling function $\delta(\mathbf{u} - \mathbf{u}_p)$ and added to the receiver noise n_p . This can be

equivalently re-written as:

$$\tilde{V}_p = \int \delta(\mathbf{u}' - \mathbf{u}_p) \left[\int \tilde{B}(\mathbf{u}' - \mathbf{u}) \times \left[\int e^{-i2\pi\mathbf{u}\cdot\hat{s}} I(\hat{s}) d^2\hat{s} \right] d^2\mathbf{u}' + n_p, \quad (7)$$

where, $\tilde{B}(\mathbf{u})$ denotes the uv -space antenna power response obtained by a Fourier transform of $B(\hat{s})$. Effectively, the multiplication in image space by $B(\hat{s})$ has been replaced by a convolution with $\tilde{B}(\mathbf{u})$ in uv -space. This is the software holographic equivalent of traditional FX correlator output. Following the matrix notation of Morales (2011), the above equation can be expressed as:

$$\mathbf{m}(\mathbf{v}) = \tilde{B}(\mathbf{v}, \mathbf{u}) \mathbf{F}(\mathbf{u}, \hat{s}) I(\hat{s}) + \mathbf{n}(\mathbf{v}), \quad (8)$$

where, the sky brightness $I(\hat{s})$ is Fourier transformed using $\mathbf{F}(\mathbf{u}, \hat{s})$ and the resultant spatial coherence function is weighted and summed using the antenna power response, $\tilde{B}(\mathbf{v}, \mathbf{u})$ in uv -space sampled at the baseline location to obtain the measured visibilities:

$$\mathbf{m}(\mathbf{v}) = \langle \tilde{E}(\mathbf{a}) * \tilde{E}(\mathbf{a}') \rangle_t. \quad (9)$$

$\mathbf{m}(\mathbf{v})$ denotes visibilities measured by cross-correlating measured antenna electric fields over all possible pairs of \mathbf{a} and \mathbf{a}' . It is the same as equation 5 written in matrix notation.

Using the optimal map-making formalism (Tegmark 1997b; Tegmark 1997a), a software holography image is formed using (Morales & Matejek 2009):

$$I'(\hat{s}) = \mathbf{F}^T(\hat{s}, \mathbf{u}) \tilde{B}^T(\mathbf{u}, \mathbf{v}) \mathbf{N}^{-1} \mathbf{m}(\mathbf{v}) \quad (10)$$

where the measured visibilities are weighted by the inverse of the system noise, following a gridding process using the holographic antenna power response as the gridding kernel followed by a Fourier transform to create an image $I'(\hat{s})$ which is the optimal estimate of the true image $I(\hat{s})$.

It may be noted that the antenna power response in uv -plane can be written as:

$$\tilde{B}_p(\mathbf{u}) = \tilde{W}_a(\mathbf{u}) * \tilde{W}_b^*(\mathbf{u}) \quad (11)$$

where, $W_a(\mathbf{u})$ and $W_b(\mathbf{u})$ denote the individual antenna voltage responses. Thus $\tilde{B}_p(\mathbf{u})$ is a cross-correlation of the individual antenna voltage responses. Similarly, it can be shown that

$$\begin{aligned} \langle E(\hat{s}) E^*(\hat{s}') \rangle &= \langle E(\hat{s}) E^*(\hat{s}') \delta(\hat{s} - \hat{s}') \rangle \\ &= \langle |E(\hat{s})|^2 \rangle = I(\hat{s}) \end{aligned} \quad (12)$$

From equations 2, 7, 9, 11 and 12, the optimal image in equation 10 can be expressed equivalently as:

$$I'(\hat{s}) = \left\langle \left| \mathbf{F}^T(\hat{s}, \mathbf{r}) \tilde{\mathbf{W}}^T(\mathbf{r}, \mathbf{a}) \tilde{\mathbf{N}}(\mathbf{a}, \mathbf{a}) \tilde{E}(\mathbf{a}) \right|^2 \right\rangle_t. \quad (13)$$

The term inside the angular brackets before squaring has a very similar form as that in equation 10. It signifies that the measured antenna electric fields are weighted by the antenna noise, weighted and gridded by the antenna aperture kernel, Fourier transformed and finally squared to obtain the same image estimated that would have been obtained using equation 10.

Equations 13 and 10 are identical to each other obtained as a consequence of the multiplication-convolution theorem of Fourier transforms. In other words, equation 10 is obtained by a Fourier transform of the gridded spatial correlation of electric fields measured by antennas while equation 13 is obtained by squaring the Fourier transform of the gridded electric fields measured by the antennas. The time-scale of averaging in equations 5 and 13 is set

by the coherence time-scale of the electric fields or the science requirement (e.g., timescale of variability or transients).

There are some important differences between the two techniques:

(i) The time-averaging cannot be performed on a stochastic measurement but only on its statistical properties. In FX imaging, the visibilities measured between antenna pairs represent spatial correlations which can be time-averaged following which they are gridded and imaged. However, in MOFF imaging both antenna and gridded electric fields are stochastic and therefore have to be imaged before time-averaging.

(ii) In FX imaging, electric fields measured on antennas are not correlated with themselves and hence lack zero spacing measurements. In contrast, in MOFF imaging, since the gridded electric fields are imaged and squared, they contain information from auto-correlated electric fields at zero spacing. Hence, they have to be subtracted from the images.

3 SOFTWARE IMPLEMENTATION

We have implemented the MOFF imaging technique in our “E-field Parallel Imaging Correlator” – a highly parallelized Object Oriented Python package,² now publicly available. Besides implementing the MOFF imaging algorithm it also includes FX imaging using software holography technique and a simulator for generating electric fields from a sky model.

Figure 1 shows the flowchart for MOFF imaging. The propagated electric fields are shown on the left at different time stamps, $t_1 \dots t_M$. At each time stamp, the electric fields measured by antennas are denoted by $E_1(t) \dots E_N(t)$. The F-engine performs a temporal Fourier transform on the electric field time-series to obtain electric field spectra $E_1(f) \dots E_N(f)$ (\vec{E}_a in matrix notation) for each of the antennas. Each of the complex antenna gains are calibrated to correct the corresponding electric field spectra. These calibrated electric fields are gridded using an antenna-based gridding convolution function following which it is spatially Fourier transformed and squared to obtain images for every time stamp. These images are then time-averaged to obtain the accumulated image $I(f)$ ($I(\delta)$ in matrix notation).

Figure 2 shows the flowchart for software holographic imaging from a FX correlator. The antenna-based F-engine is identical to that in the MOFF processing. The electric field spectra each antenna are then cross-multiplied in the X-engine with those from all other antennas to obtain the visibilities $V_{ij}(f)$ ($\mathbf{m}(\mathbf{v})$ in matrix notation). They are calibrated and time-averaged to obtain $\langle V_{ij}(f) \rangle$ which are then gridded and imaged to obtain the image $I(f)$. The $I(f)$ obtained from both techniques are identical as explained in §2.

Here, we discuss the components of these architectures in detail.

3.1 Temporal Fourier transform

This module is common to the MOFF and FX imaging techniques. Time samples of electric fields measured by the antenna and digitized by the A/D converter is Fourier transformed to generate electric field spectra. This step can be parallelized by antennas as shown in Figure 3. The output is then fed to either MOFF and FX imaging pipelines.

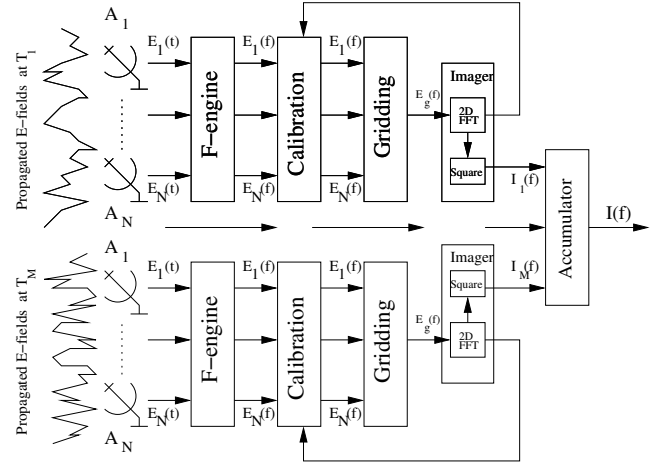


Figure 1. A flowchart of MOFF imaging in EPIC. The propagated electric fields shown on the left are measured as time-series $E_1(t) \dots E_N(t)$ by the antennas which are then Fourier transformed by the F-engine to produce electric field spectra $E_1(f) \dots E_N(f)$. They are calibrated and gridded. The gridded electric fields $E_g(f)$ from each time series are imaged to produce an images $I_1(f) \dots I_N(f)$. These images are time-averaged to obtain the final image $I(f)$.

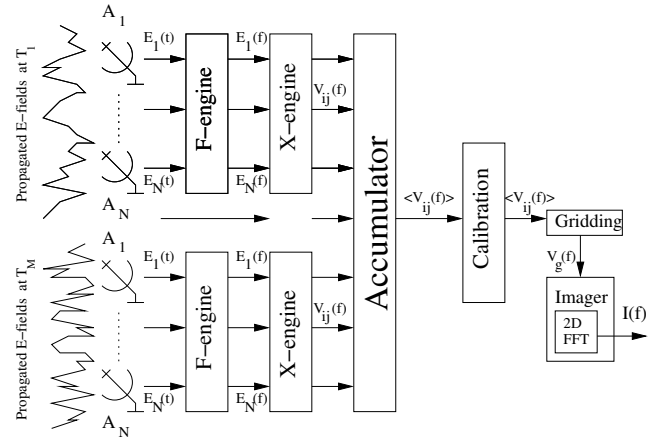


Figure 2. A flowchart of FX imaging in EPIC. The FX process flow shares the F-engine with the MOFF process. Following the F-engine, the electric fields pass through the X-engine to obtain visibilities $V_{ij}(f)$ which are calibrated and time-averaged. Then they are gridded to obtain the gridded visibilities $V_g(f)$ which are then imaged to obtain the image $I(f)$.

3.2 Antenna-to-Grid Mapping

A grid is generated on the coordinate system in which antenna locations are specified with a grid spacing that is at most $\lambda_{\min}/2$ even at the highest frequency to ensure there is no aliasing even from regions of the sky far away from the field of view. The number of locations on the grid is restricted to be a power of 2.

The gridding kernel in the simplest case is given by the antenna aperture illumination function, $B(\mathbf{r} - \mathbf{r}_a)$, which can be specified either by a functional form or as a table of values against locations around the antennas. A nearest neighbor mapping from all antenna footprints to grid locations is created using an efficient k-d tree algorithm (Maneewongvatana & Mount 1999). There is no restric-

² EPIC package can be accessed at <https://github.com/nithyanandan/EPIC>

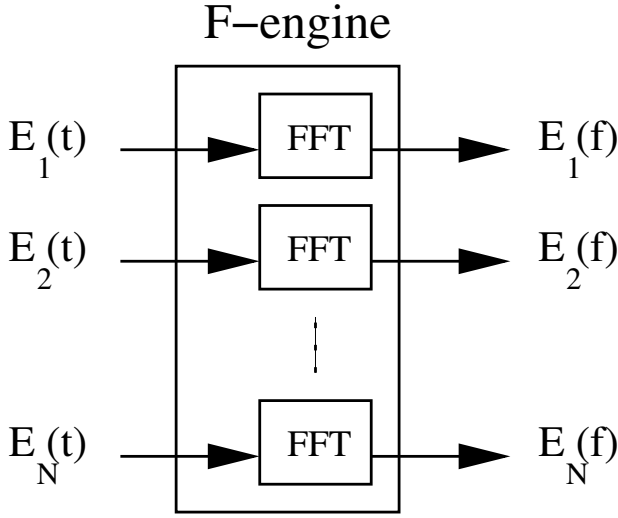


Figure 3. Block diagram of an F-engine. The electric field data streams from antennas are Fourier transformed in parallel to generate electric field spectra.

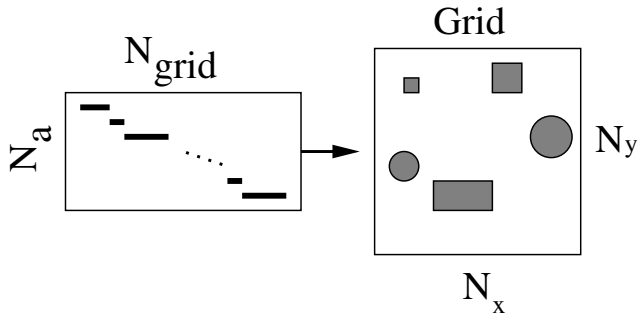


Figure 4. Block diagram of an antenna-to-grid mapping. A sparse block-diagonal matrix of total size $N_{\text{grid}} \times N_a$ is created where each block contains roughly the number of pixels covered by the respective kernel. The antenna aperture illumination kernels do not have to be identical to each other. A discrete set of arbitrarily placed antennas are now placed onto a regular grid.

tion here that the aperture illumination function has to be identical across antennas.

In the most general case, this gridding kernel could contain information on the w -projection effect, and even other time-dependent ionospheric effects. For a stationary antenna array in the absence of any time-dependent effects, this mapping has to be determined only once in the antenna array coordinate frame. The antenna-to-grid mapping matrix, $\mathbf{M}(\mathbf{r}, \mathbf{e})$ is described as a transformation matrix from the space of measured electric fields (\mathbf{e}) to the antenna array grid denoted by \mathbf{r} . Since each antenna occupies a footprint typically the size of its aperture, $\mathbf{M}(\mathbf{r}, \mathbf{e})$ which is generally of size $N_{\text{grid}} \times N_a$, reduces to a sparse block-diagonal matrix with only N_a blocks and roughly N_k non-zero entries per block. Figure 4 illustrates the antenna-to-grid mapping matrix and the grid containing the mapped aperture footprints of the antennas.

3.3 Calibration

Calibration of direct imaging correlator remains an unresolved matter. Contrary to the FX data flow, direct imagers mix the signals from

all antennas before averaging and writing to disk. It is therefore essential to apply gain solutions before the gridding step. Previous efforts have resorted to applying FX-generated calibration solutions (Zheng et al. 2014; Foster et al. 2014), or integrate a dedicated FX correlator which periodically forms the full visibility matrix (Wijnholds & van der Veen 2009; de Vos et al. 2009).

In a companion paper to appear soon, we demonstrate a novel calibration technique (EPICal) which leverages the data products formed by direct imaging correlators to estimate antenna complex gains. This method correlates the antenna electric field signals with an image pixel from the output of the correlator in the feedback calibration fashion outlined in Morales 2011 (illustrated in Figure 1 by the arrow leading from the imager to the calibration block). Furthermore it allows for arbitrarily complex sky models, and following the MOFF algorithm places no restriction on array layout, and accounts for non-identical antenna beam patterns. Because only a single correlation is needed for each antenna, the computation complexity scales only as N_{ant} .

The calibration module included in the EPIC repository allows for application of pre-determined calibration solutions, or can solve for the complex gains using the EPICal algorithm.

3.4 Gridding Convolution

The antenna array aperture illumination over the entire grid, $\tilde{W}(\mathbf{r})$, is obtained by a projection of the individual antenna aperture illuminations:

$$\tilde{W}(\mathbf{r}) = \sum_a \tilde{W}_a(\mathbf{r} - \mathbf{r}_a) \quad (14)$$

$$= \mathbf{M}(\mathbf{r}, \mathbf{e}) \mathcal{I}(\mathbf{e}), \quad (15)$$

where, $\mathcal{I}(\mathbf{e})$ is a row of ones. This is achieved using efficient multiplication with the sparse matrix created in the antenna-to-grid mapping process. Unless $\tilde{W}(\mathbf{r})$ includes time-dependent effects of the ionosphere or the instrument, it needs to be computed just once for the entire observation. However, the gridding of electric fields must be computed at every readout of the electric field spectra. Thus,

$$\tilde{E}(\mathbf{r}) = \mathbf{M}(\mathbf{r}, \mathbf{e}) E(\mathbf{e}), \quad (16)$$

where, $E(\mathbf{e})$ denotes the spectra of measured antenna electric fields.

In practice the grid is set to be a power of 2 such that the grid spacing is at most $\lambda/2$ even at the highest frequency in the observing band.

3.5 Spatial Fourier Transform

Before the spatial Fourier transform, the gridded electric fields are padded with zeros in order to match the grid size and angular size of each image pixel that would have been obtained with software holography of output from an FX correlator.

In MOFF imaging, these are spatially Fourier transformed followed by a squaring operation at every timestamp for every frequency channel. In FX imaging, the spatial Fourier transform is performed only once per integration timescale and does not include a squaring operation.

3.6 Time-averaging

In MOFF imaging, the measured antenna electric fields and the corresponding holographic electric field images are zero-mean stochastic quantities. Hence, they cannot be time-averaged to reduce noise.

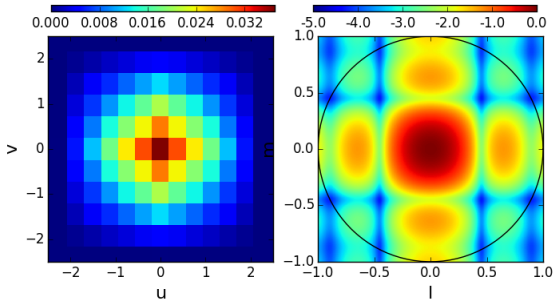


Figure 5. The auto-correlation of weights of a square shaped antenna aperture in the uv plane (left) and the corresponding directional antenna power response on the sky (right) in coordinates specified by direction cosines. The antenna auto-correlation weights are normalized to a sum of unity yielding a peak response of unity in the antenna’s directional power pattern on the sky. The color scale for the directional power pattern is logarithmic. The black circle indicates the sky horizon and values beyond it are not physical and hence ignored.

The statistical quantity stable with time in this case are the square of the holographic electric field images. Thus, squared images have to be formed at every instant of time before averaging as indicated in equation 13.

In contrast, visibilities measured by an antenna are statistically stable within an integration time interval. Hence, they are averaged after calibration as shown in equation 5. It is advantageous to average them in visibilities before imaging because visibilities represent a compact representation of the information in images. Hence it is more computationally efficient to average antenna pair visibilities rather than images. Since this averaging has been performed already on the visibilities over an integration timescale, the imaging step has to be performed only once per integration cycle. FX imaging holds this advantage as long as the square of the number of antennas is smaller than the number of pixels on the image.

3.7 Dealing with antenna auto-correlations

The squaring operation under MOFF imaging in the image plane introduces antenna auto-correlations around the zero spacing in the uv -plane which are absent in traditional visibility-based imaging. In order for a robust comparison, these auto-correlations have to be removed from both the uv and image planes.

The shape and extent of these auto-correlations can be estimated from the antenna aperture illumination pattern which is a quantity already available from the gridding step. Figure 5 shows the estimated weights from antenna auto-correlations in the uv -plane (left) and the corresponding response in the image plane (right). The latter is simply the directional antenna power response.

We inverse Fourier transform the squared images and beams back to the uv plane and subtract the estimated auto-correlation kernel scaled to the peak value centered at the zero spacing pixel. The final averaged image is obtained by Fourier transforming the uv plane data and weights with the auto-correlations subtracted to the image plane. These images are now comparable to those obtained from visibility-based imaging. This step of removing auto-correlations is required to be performed only once per integration timescale and does not add significant cost to the full operation.

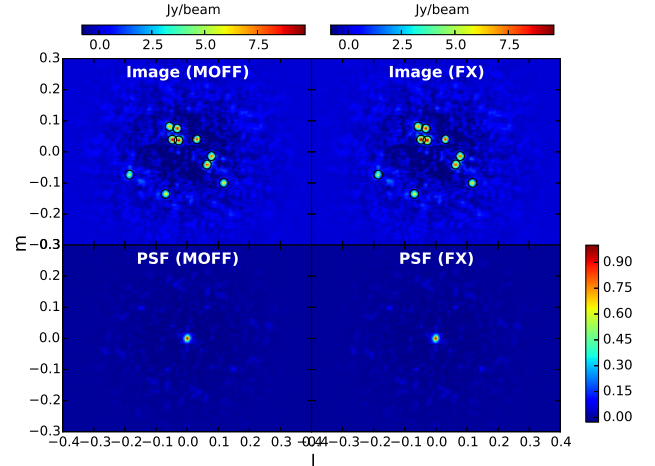


Figure 6. Dirty images (top) and synthesized beams (bottom) obtained from simulated data using antenna-based MOFF algorithm (left) and FX visibility-based software holography (right). The solid black circles in the top panels indicate the simulated source positions. The antenna auto-correlations at zero-spacing have been removed from the MOFF images. The images in either case reconstruct the sources at the right locations with the fluxes expected after multiplication by the antenna power pattern. The synthesized beams from the two algorithms are well matched in size and shape. The overall modulation by the power pattern is seen clearly in both images.

4 VERIFICATION

We use the EPIC simulator to generate electric field samples from a sky model. In our simulations, we use 64 frequency channels each of width $\delta f = 40$ kHz, 10 point sources of flux densities 10 Jy at random locations. The number of timestamps integrated in one integration cycle was kept at eight where each A/D timeseries is $1/\delta f = 25 \mu s$ long. We use the MWA array layout (Beardsley et al. 2012) for demonstration. Only the inner 51 tiles within a square bounding box of 150 m on each side were used. We assumed all tiles are identical and have a square shaped electric field illumination footprint 4.4 m on each side.

Figure 6 shows the dirty images (top) and synthesized beams (bottom) obtained with antenna-based MOFF and FX visibility-based imaging algorithms packaged in EPIC. The antenna auto-correlations that correspond to zero spacing have been removed from the MOFF image and the corresponding synthesized beam. The sky positions of the simulated sources are indicated by solid black circles. The reconstructed sky image has the simulated sources at the expected sky positions in either case. Both algorithms result in images and synthesized beams that are well matched with each other. Their fluxes are modulated by a multiplicative power pattern corresponding to that of a uniform square aperture.

We investigate the two imaging algorithms for differences from the point of view of the quality of their outputs. We begin by comparing the gridded cross-correlation weights in the uv -plane. In MOFF imaging, weights from antenna auto-correlations have been removed as described in §3.7.

Figure 7 shows the cross-correlation weights obtained with MOFF imaging (left) and visibility-based imaging (right). The first notable difference is in the weights around zero spacing. Though both show a dominant void around zero-spacing, the void obtained with MOFF algorithm shows many pixels with non-zero weights. In contrast, the zero-spacing void from traditional imaging consists

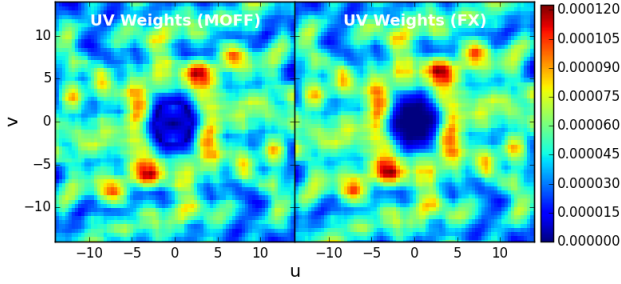


Figure 7. Dirty images (top) and synthesized beams (bottom) obtained from simulated data using antenna-based MOFF algorithm (left) and FX visibility-based software holography (right). The solid black circles in the top panels indicate the simulated source positions. The antenna auto-correlations at zero-spacing have been removed from the MOFF images. The images in either case reconstruct the sources at the right locations with the fluxes expected after multiplication by the antenna power pattern. The synthesized beams from the two algorithms are well matches in size and shape. The overall modulation by the power pattern is seen clearly in both images.

of predominantly zero-valued pixels. The gridding process in the former involves rounding the antenna footprint to the nearest grid pixel. Depending on the exact location of the center of the antenna relative to the grid, the grid pixels that receive contribution from an antenna may be a pixel narrower along one or both axes relative to that from another identical antenna but with a different center location relative to the grid. The resulting auto-correlation footprint will also not necessarily be identical. Hence, using a single expected footprint for subtraction will typically leave some residuals behind as seen in the void region in the weights obtained with MOFF imaging.

These residuals can be mitigated by:

- (i) making the grid spacing finer which makes the rounding error less susceptible to the location of the antenna center relative to the grid, and
- (ii) subtracting each auto-correlation of antenna weights separately by using the shape and extent of the footprint appropriate for that specific antenna aperture.

The latter is the most general solution applicable especially in the case of heterogeneous antenna arrays.

The other notable difference is that weights outside the zero-spacing void in some regions are different from each other at the few percent level though the sum of weights in these regions are identical. For instance, note the difference in weights at $(u, v) \approx (2, -10)$. This is found to arise because with MOFF imaging the antenna locations were rounded to the nearest grid pixel whereas in visibility-based imaging the baseline locations (difference between antenna locations) were rounded to the nearest grid pixel. Since rounding is not a commutative operation (i.e. rounding of the difference value is not necessarily equal to the difference of the rounded values), the gridding operation in the two cases introduces rounding errors in the placement of antenna aperture and cross-correlated aperture weights. Each antenna aperture weight projected to the nearest grid location can be displaced by ~ 1 – 2 pixels. This effect can be mitigated only by making the grid spacing finer.

We study the effect of the differences in gridded weights on

the image plane. The synthesized beam can be written as:

$$B(\hat{s}) = \left| \sum_{\mathbf{r}} \tilde{W}(\mathbf{r}) e^{-i \frac{2\pi}{\lambda} (\mathbf{r} \cdot \hat{s})} \right|^2, \quad (17)$$

$$\text{or, } B(l, m) = \left| \sum_{r_x, r_y} \tilde{W}(r_x, r_y) e^{-i \frac{2\pi}{\lambda} (r_x l + r_y m)} \right|^2 \quad (18)$$

where, $\hat{s} \equiv (l, m)$, $\mathbf{r} \equiv (r_x, r_y)$ and $\tilde{W}(\mathbf{r})$ is given by equation 14. If the errors in the determination of r_x and r_y are denoted by δr_x and δr_y respectively, the error on $B(l, m)$ can be estimated using a simple propagation of errors, where,

$$\frac{\partial B(l, m)}{\partial r_x} = e^{-i \frac{2\pi}{\lambda} \mathbf{r} \cdot \hat{s}} \left[\frac{\partial \tilde{W}(r_x, r_y)}{\partial r_x} - i \frac{2\pi l}{\lambda} \tilde{W}(\mathbf{r}) \right] \quad (19)$$

$$\times \sum_{\mathbf{r}'} \tilde{W}^*(\mathbf{r}') e^{+i \frac{2\pi}{\lambda} \mathbf{r}' \cdot \hat{s}} \quad (20)$$

$$= e^{-i \frac{2\pi}{\lambda} \mathbf{r} \cdot \hat{s}} \left[\frac{\partial \tilde{W}(r_x, r_y)}{\partial r_x} - i \frac{2\pi l}{\lambda} \tilde{W}(\mathbf{r}) \right] W^*(\hat{s}), \quad (21)$$

$$\text{and, } \frac{\partial B(l, m)}{\partial r_y} = e^{-i \frac{2\pi}{\lambda} \mathbf{r} \cdot \hat{s}} \left[\frac{\partial \tilde{W}(r_x, r_y)}{\partial r_y} - i \frac{2\pi m}{\lambda} \tilde{W}(\mathbf{r}) \right] W^*(\hat{s}). \quad (22)$$

It must be noted that although we start with a squaring operation in equation 17, in equations 19 through 22, the summations are performed only when $\mathbf{r} \neq \mathbf{r}'$ and therefore do not include the auto-correlations of the antenna weights. The error expected in $B(l, m)$ is:

$$|\delta B(l, m)|^2 = \sum_{r_x} \left| \frac{\partial B(l, m)}{\partial r_x} \right|^2 \delta r_x^2 + \sum_{r_y} \left| \frac{\partial B(l, m)}{\partial r_y} \right|^2 \delta r_y^2 \quad (23)$$

$$= \delta r_x^2 \sum_{r_x} \left| \frac{\partial \tilde{W}(r_x, r_y)}{\partial r_x} - i \frac{2\pi l}{\lambda} \tilde{W}(\mathbf{r}) \right|^2 |W(\hat{s})|^2 + \delta r_y^2 \sum_{r_y} \left| \frac{\partial \tilde{W}(r_x, r_y)}{\partial r_y} - i \frac{2\pi m}{\lambda} \tilde{W}(\mathbf{r}) \right|^2 |W(\hat{s})|^2 \quad (24)$$

For the current simulations,

$$\delta r_x = \delta r_y \approx 1, \quad (25)$$

$$\tilde{W}(r_x, r_y) = \{0, \text{constant}\}, \quad (26)$$

$$\text{and, } \frac{\partial \tilde{W}(r_x, r_y)}{\partial r_x} = \frac{\partial \tilde{W}(r_x, r_y)}{\partial r_y} = 0. \quad (27)$$

Hence,

$$|\delta B(l, m)|^2 \approx \left(\frac{2\pi}{\lambda} \right)^2 (l^2 + m^2) B(\hat{s}) \sum_{\mathbf{r}} |\tilde{W}(\mathbf{r})|^2, \quad (28)$$

$$\text{or, } |\delta B(l, m)| \approx \left(\frac{2\pi}{\lambda} \right) \sqrt{l^2 + m^2} |W(\hat{s})| \left(\sum_{\mathbf{r}} |\tilde{W}(\mathbf{r})|^2 \right)^{1/2}, \quad (29)$$

where, $B(\hat{s}) \equiv |W(\hat{s})|^2$. Thus, the error on the synthesized beam is proportional to the antenna power pattern and increases linearly outward in radial bins of direction cosines.

Figure 8 shows the difference between the synthesized beams obtained with the two methods. A difference map between the two

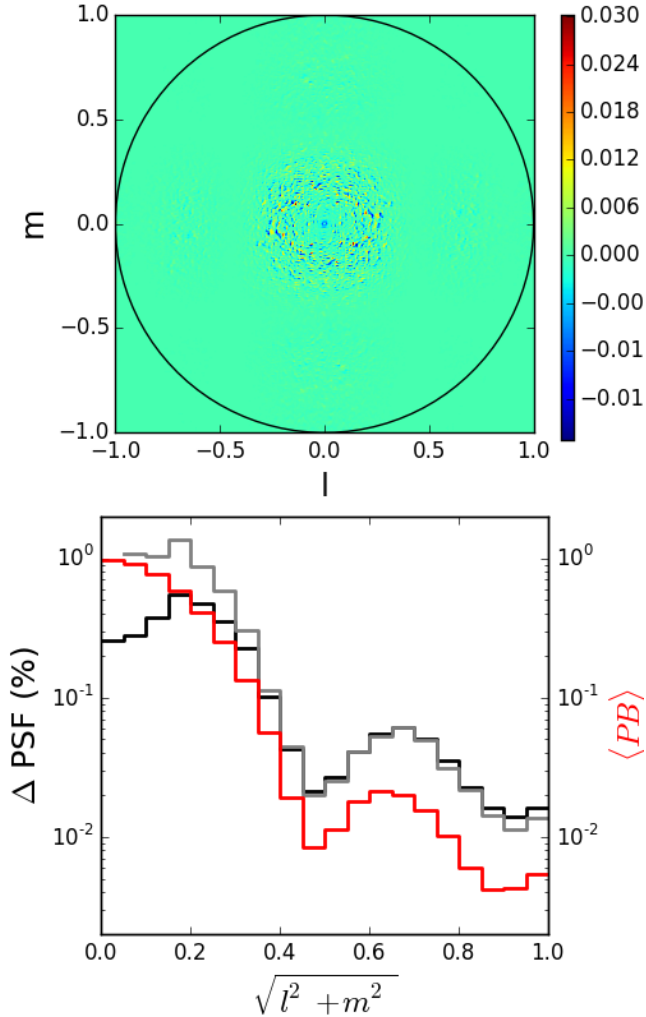


Figure 8. Map of difference between the synthesized beams obtained with the two methods (top) and radial statistics of the synthesized beams and their differences (bottom). The maximum difference is of the order of a few percent. The difference appears to be modulated in amplitude by the power pattern of the antenna.

synthesized beams is shown on top. The amplitude of the difference appears to be modulated by the directional voltage response of the antenna. At the bottom, in radial bins, the rms of the synthesized beam (gray) and the rms of the difference map (black) are plotted in percentage units relative to the peak (to be read using the axis on the left side of the plot). The antenna power pattern (red; to be read using the scale on the right) is plotted for reference.

The synthesized beam rms is proportional to the antenna power pattern as expected from a point spread function of uncorrected for the antenna power pattern. A notable observation is that the rms of the difference of the synthesized beams seems proportional to the directional antenna voltage pattern and also increases gradually with distance from the center as seen from the diverging amplitudes between the black and red curves as predicted in equation 29. The rms of the difference appears to be bounded by the actual rms of the

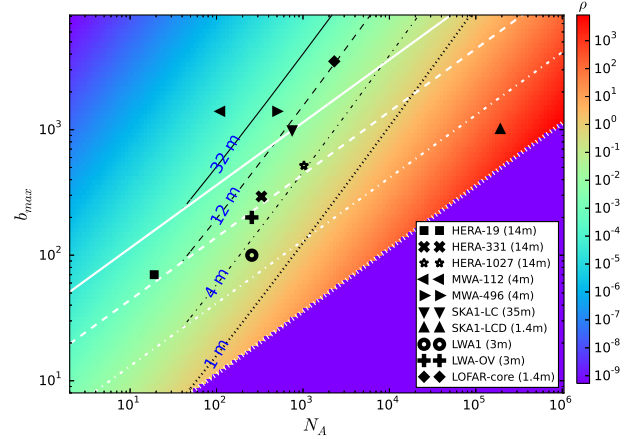


Figure 9. Current and instruments planned for future in parameter space of baseline length and number of antennas with MOFF and FX.

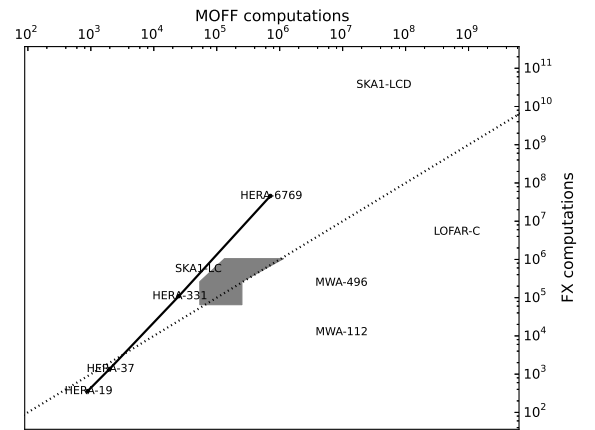


Figure 10. Current and instruments planned for future in parameter space of number of complex multiplies and adds with MOFF and FX.

synthesized beam indicating that these rounding errors in gridding do not affect the statistics of the images or the synthesized beams.

5 ANALYSIS AND FEASIBILITY

5.1 Scaling Relations: MOFF vs. FX

5.2 Scaling Up

5.3 Case Study

6 CONCLUSIONS

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REFERENCES

- Bandura K., et al., 2014, in Society of Photo-Optical Instrumentation Engineers (SPIE) Conference Series. p. 22 ([arXiv:1406.2288](#)), [doi:10.1117/12.2054950](#)
- Beardsley A. P., et al., 2012, *MNRAS*, **425**, 1781
- Bhatnagar, S. Cornwell, T. J. Golap, K. Uson, J. M. 2008, *A&A*, 487, 419
- Bunton J. D., 2004, *Experimental Astronomy*, **17**, 251
- Daishido T., et al., 2000, *Proc. SPIE*, 4015, 73
- Ellingson S. W., et al., 2013, *IEEE Transactions on Antennas and Propagation*, **61**, 2540
- Foster G., Hickish J., Magro A., Price D., Zarb Adami K., 2014, *Monthly Notices of the Royal Astronomical Society*, 439, 3180
- Lonsdale C. J., Doeleman S. S., Cappallo R. J., Hewitt J. N., Whitney A. R., 2000, in Butcher H. R., ed., Society of Photo-Optical Instrumentation Engineers (SPIE) Conference Series Vol. 4015, Radio Telescopes. pp 126–134
- Maneevongvatana S., Mount D. M., 1999, eprint arXiv:cs/9901013,
- Mellema G., et al., 2013, *Experimental Astronomy*, **36**, 235
- Morales M. F., 2011, *PASP*, **123**, 1265
- Morales M. F., Matejek M., 2009, *MNRAS*, **400**, 1814
- Otobe E., et al., 1994, *PASJ*, **46**, 503
- Parsons A. R., et al., 2010, *The Astronomical Journal*, 139, 1468
- Tegmark M., 1997a, *Phys. Rev. D*, **55**, 5895
- Tegmark M., 1997b, *ApJ*, **480**, L87
- Tegmark M., Zaldarriaga M., 2009, *Phys. Rev. D*, **79**, 083530
- Tegmark M., Zaldarriaga M., 2010, *Phys. Rev. D*, **82**, 103501
- Thompson A. R., Moran J. M., Swenson Jr. G. W., 2001, *Interferometry and Synthesis in Radio Astronomy*, 2nd Edition. Wiley
- Tingay S. J., et al., 2013, *PASA - Publications of the Astronomical Society of Australia*, 30
- Wijnholds S., van der Veen A.-J., 2009, *Signal Processing, IEEE Transactions on*, **57**, 3512
- Zernike F., 1938, *Physica*, **5**, 785
- Zheng H., et al., 2014, *MNRAS*, **445**, 1084
- de Vos M., Gunst A., Nijboer R., 2009, *Proceedings of the IEEE*, **97**, 1431
- van Cittert P. H., 1934, *Physica*, **1**, 201

APPENDIX A: SOME EXTRA MATERIAL

If you want to present additional material which would interrupt the flow of the main paper, it can be placed in an Appendix which appears after the list of references.

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