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A STUDY OF FUNDAMENTAL LIMITATIONS TO STATISTICAL DETECTION OF REDSHIFTED HI FROM THE EPOCH OF REIONIZATION

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ABSTRACT

In this paper we explore for the first time the relative magnitudes of three fundamental sources of uncertainty, namely, foreground contamination, thermal noise and sample variance in detecting the HI power spectrum from the Epoch of Reionization (EoR). We derive limits on the sensitivity of a Fourier synthesis telescope to detect EoR based on its array configuration and a statistical representation of images made by the instrument. We use the Murchison Widefield Array (MWA) configuration for our studies. Using a unified framework for estimating signal and noise components in the HI power spectrum, we derive an expression for and estimate the contamination from extragalactic point–like sources in three–dimensional k–space. Sensitivity for EoR HI power spectrum detection is estimated for different observing modes with MWA. With 1000 hours of observing on a single field using the 128–tile MWA, EoR detection is feasible ($S/N > 1$ for $k \lesssim 0.8$ Mpc$^{-1}$). Bandpass shaping and refinements to the EoR window are found to be effective in containing foreground contamination, which makes the instrument tolerant to imaging errors. We find that for a given observing time, observing many independent fields of view does not offer an advantage over a single field observation when thermal noise dominates over other uncertainties in the derived power spectrum.

Keywords: large-scale structure of Universe — methods: statistical — radio continuum: galaxies — radio lines: general — reionization — techniques: interferometric

1. INTRODUCTION

Precise measurements of the cosmic microwave background (CMB) anisotropies have constrained the background cosmology and initial conditions for structure formation. However, understanding the non-linear growth of density perturbations and astrophysical evolution in the Epoch of Reionization (EoR) has been difficult. Even estimates of the epoch as well as the process of reionization in different models of instantaneous reionization, point to a redshift of $z \approx 10.5 - 11$ (Kogut et al. 2003; Jarosik et al. 2011; Komatsu et al. 2011; Larson et al. 2011), which is in conflict with observations of Gunn-Peterson absorption troughs and near-zone transmission towards distant quasars indicating rapid evolution in the ionization fraction as late as $z \approx 6 - 7$ (Becker et al. 2001; Djorgovski et al. 2001; Fan et al. 2002; Mortlock et al. 2011).

Direct observation of redshifted 21 cm spin transition of neutral hydrogen has been identified to be a useful method for detecting structures in cosmological gas at high redshifts (Sunyaev & Zeldovich 1972; Scott & Rees 1990; Madau et al. 1997; Tozzi et al. 2000, Iliev et al. 2002). Tomography of redshifted 21 cm line promises to be a key probe of reionization history (Zaldarriaga et al. 2004). Observing images of the three–dimensional distribution of neutral hydrogen temperature fluctuations in excess relative to the CMB temperature is expected to reveal the epoch as well as the process of reionization in detail; however, Furlanetto & Briggs (2004) point out that such imaging requires the sensitivity of the Square Kilometer Array (SKA). Recently, through the use of simulations, the potential of SKA precursors for direct imag-
ing and detection of ionized regions during late stages of reionization has been demonstrated (Zaroubi et al. 2012; Malloy & Lidz 2013). Numerous first-generation radio telescopes such as the Murchison Widefield Array (MWA; Lonsdale et al. 2009; Tingay et al. 2013), the Low Frequency Array (LOFAR; van Haarlem et al. 2013), and the Precision Array for Probing the Epoch of Reionization (PAPER; Parsons et al. 2010) are becoming operational with enough sensitivity for a statistical detection of the EoR and its cosmological evolution is a first step to understanding structure formation and astrophysics in the EoR.

Power spectrum measurements of the redshifted 21 cm from EoR are difficult for the following reasons. The EoR signal is extremely weak relative to the foreground emission of the Galaxy and extragalactic sources (Bernardi et al. 2008; Ghosh et al. 2012). Considerable effort is required to distinguish their signatures from residual errors even after careful spectral modeling and subtraction of these foregrounds (Di Matteo et al. 2002; Zaldarriaga et al. 2004; Morales & Hewitt 2004) show that the inherent isotropy and symmetry of the EoR signal in frequency and spatial wavenumber (k) space make it distinguishable from sources of contamination which lack such symmetry. But they note that such symmetry considerations provide only an additional tool for separating foreground contamination from the signal, and do not guarantee that foreground contamination will be removed.

An inherent mechanism of foreground contamination via the frequency dependent structure (chromaticity) of the primary and synthesized beams has been pointed out by Bowman et al. (2009) and Morales et al. (2012). The chromatic nature of the primary and synthesized beams carries the transverse structure of contamination due to the residuals of continuum foreground subtraction into the line-of-sight direction. This has been termed mode–mixing. Both analytic calculations of Vedantham et al. (2012) and simulations of Datta et al. (2010) and Trott et al. (2012) have shown that foreground contamination by residuals after source subtraction are predominantly localized to a wedge-like region in k–space. The region excluded by the wedge has been termed the EoR window (Morales et al. 2012; Vedantham et al. 2012). They have also indicated that appropriate choices of bandpass window functions and imaging algorithms can significantly minimize levels of such contamination in specific regions of k–space.

In this paper, we present a unified framework for estimating three fundamental sources of uncertainty, namely, foreground contamination, thermal noise, and sample variance in k–space. We apply this general understanding to the case of MWA using different observing modes. We have also explored the effects of shaping the bandpass window and refining the EoR window. With detailed estimates, we compare the relative magnitudes of different sources of uncertainties and obtain a more complete view of the EoR sensitivity of the 128–tile MWA.

The rest of the paper is organized as follows. 2 provides a quick snapshot of the cosmology that motivates radio observations. 3 sets up the basic radio interferometer measurements of signal and uncertainties. Parameters and notations used are introduced that bridge the radio interferometer measurements and cosmological motivations. 4 introduces the framework upon which we build our understanding and estimates of different sources of uncertainties. Here, we also list some assumptions that have gone into our study. In 5 and 6 we describe the k–space occupancy and estimates of foreground and thermal noise components, respectively, in the power spectrum. 7 provides estimates of the EoR H I power spectrum and sample variance. The detailed interplay between various uncertainties under different observing modes and instrument parameters in determining sensitivity of the instrument for statistical measurements of EoR signatures is discussed in 8. The results are then summarized in 9. In appendices A and B we provide the details behind the derivation and estimation of classical radio source confusion and power spectrum of extragalactic point–like sources in k–space, respectively.

2. BASIC THEORY

Lidz et al. (2008) provide a basis for understanding the power spectrum of the 21 cm brightness temperature (relative to CMB) fluctuations in the limit that the spin temperature, T S, is globally much larger than the CMB temperature, T CMB. Ignoring peculiar velocities, the 21 cm brightness temperature relative to the CMB at spatial position, r, is,

\[ \delta_T(r) = T_0 \left( x_H \right) \left[ 1 + \delta_x(r) \right] \left[ 1 + \delta_\rho(r) \right]. \]  

(1)

Here, T 0 is the 21 cm brightness temperature of a neutral gas element with cosmic mean gas density, at redshift z, observed at frequency f 0 = 1420/ (1 + z) MHz, relative to the CMB temperature at that epoch. T 0 = 28 \left( (1 + z)/10 \right)^{1/2} \mu K for the cosmological parameters we have adopted throughout this paper (symbols have their usual meanings): \( H_0 = 70 \text{ km s}^{-1} \text{ Mpc}^{-1} \), \( \Omega_m = 0.27 \), \( \Omega_\Lambda = 0.73 \), and \( \Omega_K = 1 - \Omega_m - \Omega_\Lambda \). \( \langle x_H \rangle \) is the volume-averaged neutral fraction, \( \delta_x \) is the fractional fluctuation in the neutral fraction, and \( \delta_\rho \) is the fractional gas density fluctuation. The volume-averaged ionization fraction is, \( \langle x_i \rangle = 1 - \langle x_H \rangle \).

The power spectrum of \( \delta_T(r) \) is given in k–space by \( P^\text{HI}(k) \), which is the Fourier transform of \( \langle \delta_T(\mathbf{r}) \delta_T(\mathbf{r} + \Delta \mathbf{r}) \rangle \), and \( \mathbf{K} \) is the Fourier conjugate variable of \( \Delta \mathbf{r} \). Assuming isotropy of neutral hydrogen distribution, \( P^\text{HI}(\mathbf{K}) \) may be described using only the radial coordinate k, as \( P^\text{HI}(k) \). Equivalently, the dimensionless quantity \( \Delta^2_H(k) \) is frequently used to represent power in a logarithmic interval of k, given by (Zaldarriaga et al. 2004; McQuinn et al. 2006; Lidz et al. 2008).

\[ \Delta^2_H(k) = k^3 P^\text{HI}(k)/2\pi^2 T_0^2. \]  

(2)

3. INTERFEROMETER MEASUREMENTS IN THE EOR CONTEXT

\( P^\text{HI}(k) \) is estimated using the image cube, \( I(l,m,f) \), representing the sky brightness distribution in \( (l,m,f) \)–coordinates. In radio interferometry, \( I(l,m,f) \) is obtained by Fourier transforming the visibility measurements, \( V(u,v,f) \), made in \( (u,v,f) \)–coordinates. u and v are baseline lengths in units of wavelength, and l and m denote the direction cosines on the celestial sphere. f denotes the frequency of observation. \( \eta \) represents instrumental delay. \( (l,m,f) \) and \( (u,v,\eta) \) form a Fourier
conjugate pair of variables. We adopt the following convention for Fourier transform:

\[ V(u, v, \eta) = \iint I(l, m, f) e^{-j2\pi(u l + v m + \eta f)} \, dl \, dm \, df, \]

(3)

\[ I(l, m, f) = \iint V(u, v, \eta) e^{j2\pi(u l + v m + \eta f)} \, du \, d\eta, \]

(4)

where \( j = \sqrt{-1}. \)

The sky brightness distribution comprising of the true EoR Hi signal and foregrounds is multiplied by the primary beam power pattern, \( W_{\text{ps}}(l, m, f) \), on the sky. Equivalently, the visibilities of the true EoR Hi signal, \( V_{\text{true}}(u, v, f) \), and foregrounds, \( V_{\text{fg}}(u, v, f) \), are convolved with the spatial frequency response of the primary beam of an individual antenna, \( W_{\text{ps}}(u, v) \), \( W_{\text{ps}}(l, m, f) \) and \( W_{\text{ps}}(u, v) \) form a Fourier transform pair. The convolved visibilities are corrupted by additive thermal noise, \( V_{\text{n}}(u, v, f) \), sampled at the baseline locations given by the sampling function, \( S_{\text{uv}}(u, v) \). The sampling function is the Fourier counterpart of the synthesized beam, \( S_{\text{lm}}(l, m) \). Along the line of sight, the visibilities are modified by the frequency bandpass weights, \( W_{\text{b}}(f) \). Thus, the measured visibilities may be expressed as:

\[ V_{\text{obs}}^{\text{true}}(u, v, f) = \left\{ V_{\text{true}}(u, v, f) + V_{\text{fg}}(u, v, f) \right\} \ast W_{\text{ps}}(u, v) \ast V_{\text{n}}(u, v, f), \]

(5)

where the symbol \( \ast \) denotes convolution. Equation (5) forms the basis of our estimates of signal and various components of uncertainty in the power spectrum. A matrix–based framework is described in [Liu & Tegmark 2011].

In Fourier space, \( V_{\text{obs}}^{\text{true}}(u, v, f) \), is transformed to:

\[ V_{\text{obs}}^{\text{true}}(u, v, \eta) = \int V_{\text{obs}}^{\text{true}}(u, v, f) e^{-j2\pi\eta f} \, df. \]

(6)

This is obtained by Fourier transforming along frequency.

The characteristic size of the spatial frequency response of the tile’s power pattern, \( W_{\text{ps}}^{\text{true}}(u, v) \), is \( A_{\text{w}}/\lambda^{2} \), where \( A_{\text{w}} \) is the effective area of a tile and \( \lambda \) is the observing wavelength. The Fourier response of bandpass window, \( W_{\text{b}}^{\text{f}}(f) \), is \( W_{\eta}^{\text{f}}(\eta) \). The characteristic width of the bandpass response function, \( W_{\eta}^{\text{f}}(\eta) \), is set by the inverse of effective bandwidth, \( B_{\text{eff}} \). \( W_{\text{ps}}^{\text{true}}(u, v, \eta) = W_{\text{ps}}^{\text{true}}(u, v, f) \ast W_{\eta}^{\text{f}}(\eta) \) is the instrumental response in the spatial frequency domain, where \( W_{\text{ps}}^{\text{true}}(u, v, \eta) \) may be interpreted as the spatial frequency response of the tile’s power pattern obtained over a uniform and infinite bandpass window, and \( W_{\text{ps}}^{\text{true}}(u, v, \eta) = W_{\text{ps}}^{\text{true}}(u, v, f) = 0 \). Here, we have assumed that the power pattern of the tile does not vary significantly over the chosen frequency band. \( W_{\text{ps}}^{\text{true}}(u, v, \eta) \) is set such that \( f \int W_{\text{ps}}^{\text{true}} \, du \, d\eta = 1 \), where \( du \, d\eta \approx A_{\text{w}}/(\lambda^{2}B_{\text{eff}}) \). In our adopted Fourier convention, the primary beam, \( W_{\text{ps}}^{\text{true}}(l, m, f) \), and bandpass window function, \( W_{\text{b}}^{\text{f}}(f) \), each have peaks of value unity.

True sky visibilities are uncorrelated between non-identical baseline vectors (spatial frequencies). \( W_{\text{ps}}^{\text{true}}(u, v, \eta) \) is an instrumental function that introduces correlations between true sky visibilities. \( W_{\text{ps}}^{\text{true}}(u, v) \) and \( W_{\eta}^{\text{f}}(\eta) \) are the transverse and line-of-sight components, respectively, of this correlating instrumental function. \( W_{\text{ps}}^{\text{true}}(u, v) \) is a convolution of the electric field distribution over the tile with itself which is, alternatively, the Fourier transform of the power pattern of the tile. If the electric field pattern of the tile was a uniform square, \( W_{\text{ps}}^{\text{true}}(u, v) \) takes the shape of a square pyramid in the \((uv)-plane.\)

We use the array configuration of MWA to estimate the signal and uncertainties. For such an estimation, we gridied the \((uv)-plane with a cell size of \( \lambda/2 \) on each side. This is to avoid aliasing effects in the image up to the spatial scale of the horizon, \( (l, m) \in [-1, +1] \). \( W_{\text{ps}}^{\text{true}}(u, v) \) is obtained on the grid by the convolution described above. We assume each MWA array element to be a \( 4 \times 4 \) array of identical radiators. This results in a discrete form of the square pyramid for \( W_{\text{ps}}^{\text{true}}(u, v) \). Some authors [McQuinn et al. 2006; Bowman et al. 2006, 2007; Beardsev et al. 2013] have assumed that \( W_{\text{ps}}^{\text{true}}(u, v, \eta) \) is a sharply peaked function, and hence approximated it by a delta function. In our study, we use the full functional form. It is important to keep this functional form in order to take into account the multi–baseline mode–mixing, which has been described in [Hazelton et al. 2013]. The shape of \( W_{\text{ps}}^{\text{true}}(u, v) \) use we is also consistent with the baseline power response shown in [Hazelton et al. 2013].

We place the synthesized baselines represented by the sampling function \( S_{\text{uv}}(u, v) \) as a two–dimensional histogram on this grid. Compared to an optimal gridding scheme, the histogram method could place the baselines in the grid with a maximum error of \( \lambda/2 \). This could cause a jitter in the weights on the grid which is most severe on scales comparable to the cell size in the grid (corresponds to spatial scale of horizon or larger). While such a jitter will be important in case of a synthesis imaging procedure, it is not as significant in determining grid weights in our study.

In describing the aforementioned radio interferometer quantities and desired cosmological measurements, we use the following notation interchangeably throughout this paper:

\[ \mathbf{u} \equiv (u, v), \quad \mathbf{l} \equiv (l, m), \quad \mathbf{m} \equiv (u, \eta), \quad \mathbf{f} \equiv (l, f). \]

The spatial wave vectors, \( \mathbf{k} \equiv (k_{x}, k_{y}) \equiv (k_{x}, k_{y}, k_{z}) \), are related to \( \mathbf{m} \) as [Morales & Hewitt 2004]:

\[ \mathbf{k} = \frac{2\pi \mathbf{m}}{D(z)} \equiv \frac{2\pi \mathbf{m}}{H_{0}f_{21}E(z)}, \]

(8)

with \( \Delta D(z) \equiv \frac{c(1+z)^{2}}{H_{0}f_{21}E(z)} \Delta f \),

(9)

where \( H_{0} \) and \( E(z) = \Omega_{m}(1+z)^{3} + \Omega_{\Lambda}(1+z)^{2} + \Omega_{k}^{1/2} \) are standard terms in cosmology and \( D(z) \) is the transverse comoving distance at redshift \( z \) [Hogg 1999]. \( f_{21} \) is the rest frequency of the 21 cm line. In the last equation, \( \Delta D(z) \) may be identified as the line-of-sight comoving width of the observation at redshift \( z \) if \( \Delta f \) is set to the observing bandwidth. The following relations may also
be noted:
\[ k_\perp = |\mathbf{k}| = (k_x^2 + k_y^2)^{1/2} \quad \text{and} \quad k = |\mathbf{K}| = (k_x^2 + k_y^2 + k_z^2)^{1/2}. \]  
(10)

\[ k_\perp \] and \[ k_\parallel \] may be viewed as components of \( \mathbf{K} \) along the transverse and line-of-sight directions respectively.

Figure 1 is a cartoon illustration of different regions of significance in \( k \)-space. The range in \( k_\perp \) is set by the minimum and maximum baseline lengths while that along \( k_\parallel \) is set by the channel resolution and bandwidth. This region is the instrumental window. Residuals from un subtracted foregrounds and the structure of their frequency dependent sidelobes occupy the wedge-shaped region labeled as “foregrounds”. The part of the instrumental window excluding the wedge, called the EoR window, is also shown.

The radio measurements described above are related to the desired cosmological quantities. The power spectrum of EoR HI fluctuations is related to the diagonal of the covariance matrix of true HI visibilities in Fourier space as (Morales & Hewitt 2004; Morales 2005):

\[
P^{\text{HI}}(\mathbf{u}) = \left\langle V_{\text{av},f}^\text{HI, T}(\mathbf{u}) V_{\text{av},f}^\text{HI, T}(\mathbf{u})^* \right\rangle \delta_{ij} = \left\langle V_{\text{av},f}^\text{HI, T}(\mathbf{u})^2 \right\rangle,
\]

where \[ P^{\text{HI}}(\mathbf{u}) = P^{\text{HI}}(\mathbf{K}) \left( \frac{1}{D} \right)^2 \left( \frac{B_{\text{eff}}}{\Delta D} \right). \]

The last equation is obtained using the Fourier conventions and Jacobian in the transformation between quantities in \((k_x, k_y, k_z)\)– and \((u, v, \eta)\)–coordinates.

4. FRAMEWORK

The radio measurements described above are related to the desired cosmological quantities. The power spectrum of EoR HI fluctuations is related to the diagonal of the covariance matrix of true HI visibilities in Fourier space as (Morales & Hewitt 2004; Morales 2005):

\[
P^{\text{HI}}(\mathbf{u}) = \left\langle V_{\text{av},f}^\text{HI, T}(\mathbf{u}) V_{\text{av},f}^\text{HI, T}(\mathbf{u})^* \right\rangle \delta_{ij} = \left\langle V_{\text{av},f}^\text{HI, T}(\mathbf{u})^2 \right\rangle,
\]

where \[ P^{\text{HI}}(\mathbf{u}) = P^{\text{HI}}(\mathbf{K}) \left( \frac{1}{D} \right)^2 \left( \frac{B_{\text{eff}}}{\Delta D} \right). \]

The last equation is obtained using the Fourier conventions and Jacobian in the transformation between quantities in \((k_x, k_y, k_z)\)– and \((u, v, \eta)\)–coordinates.
We only consider point–like extragalactic sources in our analysis and leave the treatment of extended emission from the Galaxy and extragalactic sources to future work. Extragalactic point–like sources above the detection threshold and their sidelobes are assumed to be perfectly subtracted. For simplicity, we assume the flux densities of residual sources are constant with frequency over the band of interest after source subtraction.

The frequency structure of the bandpass could also be affected due to radio frequency interference (RFI). In this paper, we neglect effects of calibration errors and RFI. We also neglect the effects of non-coplanarity of baselines.

### 5. FOREGROUND POWER SPECTRUM

One of the major contaminants in the EoR Hi signal is the synchrotron emission from extragalactic and Galatic foregrounds (Di Matteo et al. 2002; Zaldarriaga et al. 2004). Thus far, neither of these causes of foreground contamination have been accounted for in the estimates of sensitivity of three-dimensional EoR Hi power spectrum in a comprehensive manner. For instance, Beardsley et al. (2013) estimated sensitivity by excluding a wedge shaped region, thereby removing a majority of foreground contamination. But they did not consider possible spillover from this contamination. Does it imply with certainty that foregrounds play no role any more in contaminating the EoR window or in estimates of sensitivity? Detailed estimates of extragalactic foreground contamination as done below in this paper show that contamination is also present in the EoR window outside the wedge-shaped region, even after a perfect subtraction of foregrounds.

#### 5.1. Classical Radio Source Confusion

The foreground contamination in the sought EoR Hi power spectrum is seeded by classical radio source confusion. The cause of classical confusion and the theory behind it have been well studied in literature (Condon 1974; Rohlfis & Wilson 2000). The basic ingredients for estimating classical confusion are radio source count statistics on the sky and instrument parameters such as synthesized beam size.

Assuming sources brighter than five times the classical source confusion noise and their sidelobes have been subtracted perfectly, the source confusion variance, $\sigma^2(C) \equiv P_{fg}(l,m)$, was estimated across the residual image, $\Delta(l,m)$, using the source count statistics provided by Hopkins et al. (2003) at a frequency of 1.4 GHz. $\sigma^2(C)$ increases with the solid angle subtended by an image pixel in $(l,m)$–coordinates, which in turn is a function of the pixel location. Hence, $\sigma^2(C)$ varies with 1.

---

**Table 2**

Parameters used in different case studies of model observations.

<table>
<thead>
<tr>
<th>S. No.</th>
<th>$T_{\text{sys}}$</th>
<th>$N_{\text{fields}}$</th>
<th>$N_{\text{inst}}$</th>
<th>$T_{\text{field}}$</th>
<th>$T_{\text{obs}}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>6 hours</td>
<td>1</td>
<td>166.7</td>
<td>1000 hours</td>
<td>1000 hours</td>
</tr>
<tr>
<td>2</td>
<td>6 hours</td>
<td>20</td>
<td>8.3</td>
<td>50 hours</td>
<td>1000 hours</td>
</tr>
</tbody>
</table>

1. 6 hours synthesis on a single patch of sky repeated about 160 times to get a total observing time of 1000 hours; and,

2. 6 hours synthesis on 20 different patches of sky (different RA with the same declination), each observed about 8 times (also amounting to a total observing time of 1000 hours), where each patch is separated from others by at least one FWHM of the primary beam.

The choice of 6 hours of aperture synthesis was made to confine the observations to within ±3 hours in hour angle on either side of zenith as per the MWA EoR observing plan (Beardsley et al. 2013). The different observing modes used are summarized in Table 2. The first column refers to the number of independent patches of the sky observed, the fourth refers to the number of times each of these fields are observed with their respective times of synthesis, the fifth denotes the total amount of time spent observing each field and the sixth column lists the total observing time used in each case study. We assume $T_{\text{sys}}$ is identical for all patches of sky in these observing modes.

The motivation to explore model observations with multiple fields of view is as follows: if the total time spent observing a single field is divided over multiple fields, an independent measurement of the power spectrum for each of the fields will be obtained. Upon averaging these power spectra, different components of uncertainty, specifically sample variance, will be reduced. However, since the observing time on each field has reduced, the thermal noise component in individual power spectra obtained over each field will be worse than when all the time was spent observing a single field. This is because all visibilities in a single–field observation will be combined coherently before estimating the power spectrum unlike that in a multi–field observation. What are the relative levels of various components of uncertainty in the measured power spectrum? Are there regions in $k$–space where sample variance is the dominant source of uncertainty relative to the thermal noise component? And can sensitivity be improved in these regions by averaging power spectrum measurements from independent fields which reduces sample variance? Should a particular observing mode be preferred over others?

4.3. Assumptions

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19 We model the primary beam power pattern of the MWA tile as a 4x4 array of identical radiators. At 170.7 MHz, the primary beam has a FWHM of $\theta_p \approx 21^\circ$. A strip of 24 hour range in RA at a declination $\delta$ with a declination width $\Delta \delta$ subtends a solid angle $\Delta \Omega = 4 \pi \sin \Delta \delta \cos \delta$. For $\delta = -26.7^\circ$ and $\Delta \delta = \theta_p$, the strip subtends a solid angle of $\approx 2$ sr and corresponds to $\approx 20$ non-overlapping primary beams at the specified FWHM.

---
In obtaining the confusion variance at 170.7 MHz, we used a mean spectral index of \( \alpha = -0.78 \) \cite{ishwarachandra2010}, \( S \propto f^\alpha \). Flux densities are converted to temperature units using \( T = S A_B/(2 k_B) \), where \( k_B \) is the Boltzmann constant. Hereafter, fluctuations in the residual image cube are statistically represented by the classical source confusion variance \( \sigma^2_{\text{C}}(1) \).

In a naturally weighted image made using the 128–tile MWA, a typical pixel close to the zenith subtends a solid angle of \( \Omega \approx 2 \times 10^{-6} \text{ sr} \). Statistically, the flux contained in such pixels have uncertainties whose rms is \( \sigma_C \approx 35 \text{ mJy} \).

Equation \( \ref{eq:15} \) gives details of an iterative procedure used to arrive at the classical source confusion rms \( \sigma_C \). An illustration of the numerical solution for the variation of source confusion variance with solid angle, which varies across the residual image, is also presented.

### 5.2. Foregrounds in k-space

The foreground component of measured visibilities may be written from equation \( \ref{eq:9} \) as:

\[
V_{u,v,f}^{\text{FG}}(u,f) = V_{u,v}^{\text{FG,T}}(u,f) * W_{uv}^{\text{P}}(u) S_{uv}(u) W_{\eta}^{\text{B}}(\eta).
\]

Thus,

\[
V_{u,v,f}^{\text{FG}}(\mathbf{\eta}) = V_{u,v}^{\text{FG,T}}(\mathbf{\eta}) S_{uv}(u) * W_{uv}^{\text{P}}(u) * W_{\eta}^{\text{B}}(\eta),
\]

where the true foreground visibilities, \( V_{u,v}^{\text{FG,T}}(u,f) \), have been convolved with the spatial frequency response of the tile’s power pattern, \( W_{uv}^{\text{P}}(u) \), multiplied by the sampling function in the \((u,v)\)-plane, \( S_{uv}(u) \), multiplied by the frequency bandpass window function, \( W_{\eta}^{\text{B}}(\eta) \), and Fourier transformed along \( f \).

The power spectrum of foregrounds is extracted from the diagonal of the covariance matrix \( \langle V_{u,v}^{\text{FG}}(\mathbf{\eta}) * V_{u,v}^{\text{FG}}(\mathbf{\eta}) \rangle \). After certain algebraic simplifications between quantities in Fourier \((\mathbf{\eta})\)– and real \((\mathbf{l})\)-space, the foreground component of power spectrum may be expressed as:

\[
P_{\text{inst}}^{\text{FG}}(\mathbf{l}) = \int \int P_{\text{inst}}^{\text{FG}}(1) * |S_{\text{inst}}(1)|^2 |W_{\text{inst}}^{\text{P}}(1)|^2 * |W_{\eta}^{\text{B}}(\eta)|^2 * e^{-i \mathbf{\eta} \cdot \frac{\mathbf{x}}{c}} d^2 \mathbf{\eta}.
\]

where \( P_{\text{inst}}^{\text{FG}}(1) = \sigma^2_{\text{C}}(1) \). The classical source confusion variance estimated across the image is weighted by primary beam squared \( |W_{\text{inst}}^{\text{P}}(1)|^2 \), then convolved with the synthesized beam squared \( |S_{\text{inst}}(1)|^2 \), and is finally convolved with the instrumental delay response squared \( |W_{\eta}^{\text{B}}(\eta)|^2 \), which is shifted in \( \eta \) by \( \mathbf{x} \cdot \frac{1}{c} \) as prescribed by the delta function \( \delta \left( \eta + \frac{\mathbf{x} \cdot \mathbf{l}}{c} \right) \). \( \mathbf{x} = \mathbf{\eta} \lambda \) is the baseline vector in units of distance. The details of the derivation of equation \( \ref{eq:15} \) have been laid out in \cite{parsons2012}.

Equation \( \ref{eq:15} \) enables evaluation of foreground contamination in \( k \)-space, which depends on classical source confusion \( \sigma^2_{\text{C}}(1) \), the primary beam pattern of the tile \( W_{\text{inst}}^{\text{P}}(1) \), Fourier response of bandpass weights \( W_{\eta}^{\text{B}}(\eta) \) and synthesized beam \( S_{\text{inst}}(1) \). It expresses the foreground contamination that is in \((u,v,\eta)\)-coordinates (Fourier space) in terms of quantities in \((l,m)\)-coordinates (real space). This expression aids in understanding the mixing aspect of foreground contamination. The delta function indicates that unsubtracted sources at different locations and their sidelobes contribute predominantly to specific values of \( \eta \) in delay space (and corresponding \( k \)-modes) given by \( \eta \approx \frac{\mathbf{x} \cdot \mathbf{l}}{c} = \frac{\mathbf{u} \cdot \mathbf{l}}{f} \). This is consistent with the findings of \cite{vedantham2012}.

The equation naturally leads to a wedge–shaped region, whose boundary is set by the horizon limit \( \left( \text{Parsons et al. 2012} \right) \) in \((l,m)\)-space given by \( \left| \mathbf{l} \right|^2 = l^2 + m^2 = 1 \). The equation in \( k \)-space for the horizon limit (wedge boundary) is:

\[
k_{\parallel} = \frac{H_0 E(z) D_M(z)}{c(1+z)} k_{\perp}.
\]

Using a natural weighted synthesized beam, \( S_{\text{inst}}(1) \), primary beam, \( W_{\text{inst}}^{\text{P}}(1) \), obtained by a phased \( 4 \times 4 \) array of identical radiators, instrumental delay response \( W_{\eta}^{\text{B}}(\eta) \) for different bandpass shapes, and source confusion variance, \( P_{\text{inst}}^{\text{FG}}(1) \), we have estimated foreground contamination of power spectrum in full three–dimensional \( k \)-space by evaluating the integral in equation \( \ref{eq:15} \).

The size of a cell in three–dimensional \( k \)-space, termed voxel, corresponds to \( A_c \) and \( B_{\text{eff}}^2 \) in the transverse and line-of-sight directions respectively. The \( k \)-space volume of a voxel is \( \approx 9.1 \times 10^{-8} \text{ Mpc}^{-3} \).

In order to illustrate the three–dimensional foreground contamination in two–dimensional \((k_{\parallel}, k_{\perp})\)-plane, we azimuthally averaged the foreground contamination using an inverse quadrature weighting as though foregrounds were the only source of uncertainty. Figure \( \ref{fig:2} \) shows the averaged foreground contamination (in units of \( K^2 \)) in \( k \)-space for a 6 hour synthesis, where the effects of band shape, \( W_{\eta}^{\text{B}}(\eta) \), are not applied. The structure of the foreground power spectrum is found to be in agreement with the foreground window illustrated in Figure \( \ref{fig:1} \). The lower and upper bounds on the \( k_{\perp} \)-axis are provided by the minimum (\( \approx 3 \lambda \)) and maximum (\( \approx 1636 \lambda \)) baseline lengths, and those on the \( k_{\parallel} \)-axis are provided by \( \sim 1/B_{\text{eff}}^2 \) (gray horizontal line, \( \lambda_{\text{min}} \approx 0.125 \mu\text{sec} \)) and \( \sim 1/(2 \Delta f) \) (\( \lambda_{\text{max}} \approx 12.5 \mu\text{sec} \)) respectively. The foreground component of the power spectrum occupies a wedge–shaped region in \( k \)-space. The horizon limit (gray line with positive slope) sets the boundary of the wedge. For the parameters listed in Table \( \ref{tab:1} \) the instrumental window is given by \( 0.0014 \text{ Mpc}^{-1} \leq k_{\parallel} \leq 1.47 \text{ Mpc}^{-1} \) and \( 0.048 \text{ Mpc}^{-1} \leq k_{\perp} \leq 4.7 \text{ Mpc}^{-1} \), and the slope of the foreground wedge is 3.18. The maximum foreground contamination is \( \approx 1.5 \times 10^{-9} \text{ K}^2 \) but occurs below the instrumental window. The typical contamination close to \( (k_{\parallel}, k_{\perp}) \approx (0.015, 0.047) \text{ Mpc}^{-1} \) is \( \approx 10^{-10} \text{ K}^2 \). The model eor \( \text{H}1 \) signal strength in this region is \( \approx 3 \times 10^{-10} \text{ K}^2 \). An apparent increase in foreground contamination closer and parallel to the wedge is noted. This is attributed to an increase in \( \sigma^2_{\text{C}}(1) \), which in turn is due to an increase in solid angles as \( \left| \mathbf{l} \right| \) approaches the horizon limit.

If independent power spectra are averaged from \( N_{\text{fields}} \) patches of sky, the foreground contamination in the averaged power spectrum goes as \( \sim 1/\sqrt{N_{\text{fields}}} \).
3.18. The top axis denotes baseline lengths, power is predominantly contained in. The slope of the wedge is into account? overall sensitivity when other uncertainties are also taken dimensional bandpass shaping on foreground contamination in three–

addresses the following questions: what is the effect of window function (Nuttall 1981). Nuttall (2012) have discussed a possibility of using a appropriate choice of bandpass shapes. Vedantham et al.

tamination from foreground emission, which is restricted to the horizon limit k

5.3. Role of Bandpass Shapes in Foreground Contamination

An important consequence of convolution in equation (15) by the term \( W^B_\eta (\eta) \), the response of the bandpass shape in instrumental delay space, is to spill the contamination from foreground emission, which is restricted to the wedge, into the regions beyond. This spillover, therefore, spills even the desirable portions of \( k \)-space, namely, the EoR window. Hence, a simple removal of the wedge shaped region from data analysis does not completely remove all the effects of foreground contamination. The level of this spillover may be controlled by appropriate choice of bandpass shapes. Vedantham et al. (2012) have discussed a possibility of using a Blackman–Nuttall window function (Nuttall 1981).

In the context of bandpass window shaping, this paper addresses the following questions: what is the effect of bandpass shaping on foreground contamination in three–dimensional \( k \)-space? And, what is its significance to overall sensitivity when other uncertainties are also taken into account?

An infinite bandpass will convolve the power spec-

trum of foregrounds by a \emph{delta} function resulting in zero spillover, but is impossible to achieve in practice. A more practical bandshape, such as a rectangular window, will manifest as a \emph{sinc}–shaped response along \( \eta \) (and \( k_{\parallel} \)). A Blackman–Nuttall window is known to have a much reduced sidelobe response (3–4 orders of magnitude) along \( \eta \) but its peak (area under band shape) in \( \eta \)-space is \( \approx 2.76 \) times less than that of a \emph{sinc} function, which implies a loss of sensitivity. The resolution in \( \eta \) is also relatively poorer. However, if the effective bandwidth, \( B_{\text{eff}} \), is made equal to that of a rectangular band shape by extending the standard Blackman–Nuttall window, a significant reduction in sidelobes in the response function along \( \eta \) may be achieved without compromising either sensitivity or resolution, compared to those from a rectangular window. We define the effective bandwidth as:

\[
B_{\text{eff}} = \int_{-B_0/2}^{B_0/2} W^B_f (f) \, df = W^B_f (0),
\]

where the second equality comes from the Fourier transform convention and the limits of the integral form the edges of the band.

Figure 3 shows the rectangular (dotted), standard (solid gray), and extended (solid black) versions of Blackman–Nuttall band shapes. Their effective bandwidths are 8 MHz, 2.9 MHz and 8 MHz respectively. Figure 4 shows using the respective line styles the amplitude of the respective responses, \( W^B_\eta (\eta) \), of the aforementioned band shapes along \( \eta \). As expected, the standard Blackman–Nuttall window has reduced sensitivity visible by its peak and is of a poorer resolution. The extended version, however, is identical in sensitivity and resolution to that of a rectangular window.
Figure 4. The amplitude of the responses of the bandpass window functions of Figure 3 shown versus $\eta$, the instrumental delay. Line styles are identical to those in Figure 3. The responses of rectangular and extended Blackman–Nuttall bandpass windows are almost identical in sensitivity and resolution along $\eta$, while the standard Blackman–Nuttall window is poorer in resolution and $\approx 2.76$ times lesser in sensitivity.

Using $W_B^R(\eta)$ for rectangular and extended Blackman–Nuttall windows described above, the power spectrum of unsubtracted foreground sources was estimated in three-dimensional $k$-space using the integral in equation (15). Once again for purposes of illustration we averaged these power spectra in independent cells azimuthally in $k$-space using inverse quadrature weighting. Figures 5 and 6 show the azimuthally averaged power spectra of foregrounds (in units of K$^2$ Hz$^2$) for rectangular and extended Blackman–Nuttall band shapes respectively, each with an effective bandwidth of 8 MHz. The features already noted in Figure 2 are also noted here. But the bandpass effects were not applied in Figure 2. Due to the presence of the term $W_B^R(\eta)$ in equation (15), the foreground contamination is seen to spill over into the EoR window in both cases. The spillover into the EoR window up to $k_\parallel < 0.1$ Mpc$^{-1}$ is similar in both cases and is $\sim 10^4$ K$^2$ Hz$^2$. A Blackman–Nuttall window is expected to be superior by 7–8 orders of magnitude in reducing the spillover beyond the wedge shaped region relative to a rectangular window because of the term $|W_B^R(\eta)|^2$. Higher levels of foreground contamination are seen due to a rectangular bandpass window in the region $k_\perp \lesssim 0.03$ Mpc$^{-1}, k_\parallel \gtrsim 0.1$ Mpc$^{-1}$ when compared to that due to a Blackman–Nuttall window. In fact, this is confirmed from Figure 7 which compares the foreground power along slices at $k_\perp \simeq 0.01$ Mpc$^{-1}$ shown in Figures 5 and 6 as gray dashed lines. In the range $0.2$ Mpc$^{-1} \lesssim k_\parallel \lesssim 5$ Mpc$^{-1}$, the extended Blackman–Nuttall window produces a foreground contamination spillover $\sim 10^{-6} - 10^{-5}$ K$^2$ Hz$^2$, which is about 7–8 orders of magnitude smaller than that from a rectangular window.

Are there undesirable effects of using an extended Blackman–Nuttall band shape? A wideband observation...
might be analyzed with a sliding window to examine for any change in EoR detection with redshift. Thus, bandwidth is not discarded when an extended Blackman–Nuttall window is deployed. But such a window uses larger total bandwidth (more channels) than a nominal rectangular window to achieve the same effective bandwidth. If there is significant cosmic evolution of the EoR signal within the band, the assumption of statistical stationarity of EoR signal could break down and lead to a dilution of measured signal power.

6. THERMAL NOISE POWER SPECTRUM

Thermal noise component in a sampled visibility measurement in Fourier space, from equations (15) and (16), is:

$$V_{uv}^N(\mathbf{u}) = \int V_{uv,f}^N(\mathbf{u}, f) S_{uv}(\mathbf{u}) W^B(f) e^{-j2\pi uf} df.$$  (18)

The rms of thermal noise in a measured visibility sample in a single frequency channel is given by Morales (2005) and McQuinn et al. (2006):

$$\Delta V_{uv,f}^N(\mathbf{u}, f) = \frac{\lambda^2 T_{sys}}{Ae \sqrt{\Delta f t_{int}}},$$  (19)

where $t_{int}$ is the integration time used to obtain visibility samples.

In a natural weighting scheme, the weight of a certain $(u, v)$-cell is proportional to the number of baselines (or measurements) in that cell. When data measured by baselines inside $(uv)$-cells are averaged, the thermal noise in the averaged cell visibility is inversely proportional to the square root of number of baselines sampling that cell. Thus, the power spectrum uncertainty due to thermal noise in a spatial frequency mode $(\mathbf{u})$ may be written as

$$C^N(\mathbf{u}) = \left( \frac{\lambda^2 T_{sys}}{Ae} \right)^2 \frac{\epsilon B_{eff}}{t_{eff}},$$  (20)

where $\epsilon$ is the factor arising out of bandpass weights and $t_{eff}$ is the effective time of observation for a given mode, which is effectively a product of the number of baselines sampling the mode during aperture synthesis and the integration time used in producing a visibility sample.

Thermal noise power spectrum is estimated as prescribed by Morales (2005) and McQuinn et al. (2006). $\epsilon$ arises in equation (20) from the sum of squares of bandpass weights, $W^B(f)$, while determining the power spectrum. $\epsilon$ is given by:

$$\epsilon = \frac{\sum_{i=1}^{N_{ch}} |W^B(f_i)|^2}{\sum_{i=1}^{N_{ch}} W^B(f_i)}.$$  (21)

where $i$ indexes the frequency channels in the bandpass, $N_{ch}$ is the number of channels in the bandpass and $\sum_{i=1}^{N_{ch}} W^B(f_i) \Delta f = B_{eff}$. The value of $\epsilon$ is 1 and 0.72 respectively for rectangular and extended Blackman–Nuttall window band shapes. It indicates that an extended Blackman–Nuttall window also reduces the thermal noise component in the power spectrum by 28%. As a result, in order to achieve a thermal noise power equal to that from an extended Blackman–Nuttall window using a rectangular window, an observation has to be $\approx 40\%$ longer.

Uncertainty in power spectrum due to thermal noise is very sensitive to observing time. In each $k$-mode, it is inversely proportional to the number of baselines, including redundant ones that sample that $k$-mode during the entire synthesis. It depends on band shape that is parametrized by $\epsilon$. In the context of our model observations listed in Table 2, the thermal noise power spectrum in equation (20) at any $k$-mode may be re-written as:

$$C^N(\mathbf{k}) = \left( \frac{\lambda^2 T_{sys}}{Ae} \right)^2 \frac{\epsilon B_{eff}}{N_{cad} t_{int} N(\mathbf{k}) \sqrt{N_{fields}}},$$  (22)

where $N(\mathbf{k})$ is the number of baselines (redundant ones included) observing the $\mathbf{k}$-mode during a single synthesis observation of duration $t_{syn}$ and $N_{cad} = t_{obs}/(t_{syn} N_{fields})$.

For illustration, this is azimuthally averaged in bins of $k_\parallel = (k_x^2 + k_y^2)^{1/2}$ by adding the thermal noise component of power spectrum contained in cells in this bin, in inverse quadrature, to yield a two-dimensional thermal noise power spectrum, $C^N(k_\parallel, k_\perp)$. In Figures 5 and 9 we show $C^N(k_\parallel, k_\perp)$ for observing modes (1) and (2), respectively, using a rectangular bandpass window. In observing mode (1), the azimuthally averaged thermal noise power spectrum attains minimum ($\approx 640$ K$^2$ Hz$^2$) at $k_\parallel \approx 0.004$ Mpc$^{-1}$ and maximum ($\approx 1.4 \times 10^9$ K$^2$ Hz$^2$) at $k_\perp \approx 1$ Mpc$^{-1}$. The thermal noise power spectra are similar in the two observing modes except that it is higher by a factor of $\approx 4.47$ in the latter as predicted.
by the scaling in equation (22). In the first observing mode, the visibilities are added coherently for a total of 1000 hours. In the second observing mode, the visibilities on each field are added coherently only for 50 hours and power spectra are estimated. These power spectra are then averaged. This increases the thermal noise component in the power spectrum by a factor $\sqrt{20} \approx 4.47$.

Thermal noise power spectrum exhibits cylindrical symmetry about the $k_{||}$ axis. When an extended Blackman–Nuttall window is used, thermal noise component in the power spectrum drops by 28% ($\epsilon = 0.72$) relative to that from a rectangular bandpass window.

The measured $H_i$ component of Fourier space visibilities in equation (32) in a manner similar to equation (14) may be expressed as:

$$V^{HI}_{uv\eta}(\mathbf{u}) = V^{HI,T}_{uv\eta}(\mathbf{u}) S_{uv}(\mathbf{u}) * W_{uv\eta}(\mathbf{u}).$$

(23)

The EoR $H_i$ power spectrum measured by the instrument, $P^{HI}_{\text{inst}}(k)$, is the diagonal of the covariance matrix

$$\left\langle V^{HI}_{uv\eta}(\mathbf{u}_i) V^{HI}_{uv\eta}(\mathbf{u}_j) \right\rangle$$

and is given by (Morales & Hewitt 2003; Morales 2005; McQuinn et al. 2006; Bowman et al. 2006, 2007): 

$$P^{HI}_{\text{inst}}(k) = \iint P^{HI}(\mathbf{u}) \left| W_{uv\eta}(\mathbf{u}) \right|^2 d^2u$$

(24)

Figure 8. Logarithm of thermal power spectrum (in units of $K^2$ Hz$^{-2}$) for a 1000 hour observation with 6 hours synthesis on a single field (first observing mode in Table 2), using a rectangular bandpass window ($\epsilon = 1$). The solid and dotted gray lines are identical to those shown in Figure 4. The grayscale color bar used is in logarithm units. The minimum ($\approx 640 K^2$ Hz$^{-2}$) and maximum ($\approx 1.4 \times 10^9 K^2$ Hz$^{-2}$) are attained at $k_{||} \approx 0.004$ Mpc$^{-1}$ and $k_{\perp} \approx 0.1$ Mpc$^{-1}$ respectively for the natural weighting scheme used. The black vertical segments in the right edge of the image indicate absence of measurements at the corresponding baselines.

Figure 9. Same as Figure 5 but evaluated for 1000 hours with 6 hours synthesis on 20 fields (observing mode (2) in Table 2). In this case, thermal noise power spectrum relative to observing mode (1) (in Figure 8) is worse by a factor $\approx 4.47$ throughout, owing to a decrease in observing time on individual fields as the number of independent fields observed is increased from one to twenty. 

at the sampled baseline locations and $P^{HI}(k)$ is given by equation (12). Although some authors (McQuinn et al. 2006; Bowman et al. 2006, 2007) have approximated $W_{uv\eta}(\mathbf{u})$ by a delta function, it will exhibit some spillover along $\eta$ depending on the bandpass shape $W_{B}(f)$. Hence, we retain the general form of the power spectrum in equation (24) for our work.

In equation (24), we note the convolving effect arising out of instrumental factors, thereby introducing correlations between neighboring spatial frequencies. The observed power spectrum of the signal is a modification of the true EoR $H_i$ power spectrum by instrumental parameters of observation such as primary beam and bandpass shape.

Sample variance is equal to the power spectrum

Jungman et al. 1996; McQuinn et al. 2006. If a number of independent measurements ($N_{\text{fields}}$) of power spectrum are averaged, the sample variance goes as $P^{SV}(k) = P^{HI}_{\text{inst}}(k)/\sqrt{N_{\text{fields}}}$.

$P^{HI}(k)$, in equation (2), represents the variance in $k$-space of the spin temperature fluctuations of $H_i$ relative to the CMB. As already mentioned in §4.3, simulations of Lidz et al. 2008 show that the variance in the ionization field peaks at a value close to 50% ionization. We choose from the family of $P^{HI}(k)$ curves they provide the one parametrized by $((x_1), z) = (0.54, 7.32)$ and use it as the input model in this study. $(x_1)$ and $z$ are the mean volume-averaged ionization fraction and redshift respectively. Redshifted emission from $H_i$ at $z = 7.32$ occurs at 170.7 MHz, which is chosen as our observing frequency. We have obtained the values of power spectrum predicted for $((x_1), z) = (0.54, 7.32)$ from the plots of Lidz et al. 2008 (through private communica-
tion with Adam Lidz). Since we required the predicted values of power spectrum at intermediate values of \( k \) not tabulated, we used a third-order polynomial fit to interpolate the predicted power spectrum to the required values.

Two primary causes contribute to power spectrum of EoR \( \text{H} \text{i} \) fluctuations:

1. the underlying matter density fluctuations, and
2. the ionized bubbles during the reionization process.

The contribution from matter density fluctuations is anisotropic due to redshift–space distortions caused by peculiar velocity effects along the line of sight \( (\text{Barkana & Loeb 2005}) \), whereas, the contribution from ionization fluctuations is isotropic. In our adopted model the ionization fraction is about 50\%, which indicates significant ionization. Hence, the contribution to the EoR can be assumed to be dominated by ionized bubbles rather than due to underlying matter density fluctuations \( (\text{Lidz et al. 2008}) \). Therefore, we neglect anisotropic effects arising out of peculiar velocities in our model power spectrum.

The observed power spectrum is computed from equation (24) using the input model. It is identical for the two observing modes listed in Table 2. The observed sample variance is higher in observing mode (1) relative to observing mode (2) by a factor \( \sqrt{20} \approx 4.47 \).

Figure 10 shows the observed EoR \( \text{H} \text{i} \) power spectrum in \( (k_{\perp}, k_{\|}) \)-plane corresponding to observing mode (1) while employing a rectangular band shape. Bandpass shape causes a convolution with the true power spectrum along \( k_{\|} \), as in the case of foreground power spectrum. Although not shown, as expected the extended Blackman–Nuttall window causes a far lesser spillover of the EoR \( \text{H} \text{i} \) power spectrum relative to the rectangular band shape. For our paper, we term this as “signal spillover”. It is caused by the same reason (bandpass window shape) that causes foreground spillover beyond the horizon limit.

8. EoR SIGNAL DETECTION

We have estimated the EoR \( \text{H} \text{i} \) power spectrum expected to be observed and individual uncertainties in three-dimensional \( k \)-space. The total uncertainty in the power spectrum in three-dimensional \( k \)-space was obtained by summing the component uncertainties,

\[
\Delta P(\mathbf{k}) = P_{\text{net}}(\mathbf{k}) + C^N(\mathbf{k}) + P^\text{SV}(\mathbf{k}).
\]

8.1. Family of EoR Windows

By knowing the occupancy of various uncertainties in \( k \)-space and excluding those regions where uncertainties dominate, the estimates are expected to be relatively free of contamination \( (\text{Morales et al. 2012}) \). Within the instrumental window, results from \( \text{Datta et al. 2010} \), \( \text{Vedantham et al. 2012} \), \( \text{Williams et al. 2012} \) and our study have shown that the wedge-shaped region in the \( k \)-space is contaminated due to unsubtracted foreground sources and their sidelobes relatively more than in other regions in \( k \)-space. Hence, the EoR window for \( \text{H} \text{i} \) power spectrum has been designated as the region in \( k \)-space inside the instrumental window excluding the wedge. The idea of an optimal window is investigated further.
mental delay function $W$ (26). The gray box denotes the instrumental window in $k$-space. The straight dotted line marks the boundary of the wedge-shaped region given by equation (16).

$k_1$ is removed, while $e = 3$ denotes removal of three characteristic widths in addition to the wedge.

eff \eta (u, \nu)$ of convolution due to the instrumental term, $W_{\text{inst}}(u, \nu)$, in equation (21) is not spherically symmetric. During spherical averaging, we ignore this loss of spherical symmetry caused by instrumental distortion. While averaging in shells of $k$, we average the observed signal in these shells and, correspondingly, estimate the uncertainties by adding them in inverse quadrature.

We compare the one-dimensional signal and noise estimates for the cases listed in Table 2 while deploying rectangular and extended Blackman–Nuttall band shapes, for a range of values of $e$.

Figures 12a–12d show in detail the signal and uncertainties expected with the MWA for observing mode (1) listed in Table 2. Figure 12a demonstrates the levels of signal (solid circles) and uncertainty (solid line) expected with either of the bandpass windows employed for the EoR window parameter $e = 0$. Also shown are the individual components of the total uncertainty in different line styles (foregrounds: dot-dashed, thermal noise: dashed, sample variance: dotted). Figure 12b shows the change in the power spectrum of EoR $\text{Hi}$ and that of foregrounds when $e$ is varied ($e = 1, 2, 3$ in red, green and blue respectively) relative to their respective values at $e = 0$ (black). Similarly, Figure 12c shows the change in thermal noise component of power spectrum for different values of $e$, relative to its values at $e = 0$. In other words, Figures 12a and 12c show signal and uncertainty components normalized with respect to themselves obtained at $e = 0$. Hence, the quantities at $e = 0$ are shown for reference in these Figures at a constant value of unity. Figure 12d shows the ratio of signal to uncertainty (S/N) as $e$ is varied. The left sub-panel in each panel corresponds to a rectangular bandpass window while that on the right is obtained with an extended Blackman–Nuttall window.

With observing mode (1) using 128–tile MWA, the signal clearly appears to be detectable (S/N $> 1$) for $k \lesssim 0.8$ Mpc$^{-1}$ (see Figures 12a and 12d). From Figure 12a, with $e = 0$, foreground contamination (dotted line) exceeds thermal noise (dashed line) for $k \lesssim 0.2$ Mpc$^{-1}$ and $k \lesssim 0.1$ Mpc$^{-1}$ while using rectangular and Blackman–Nuttall windows respectively. Beyond this crossover, thermal noise takes over as the dominant source of uncertainty in power spectrum. Foregrounds (dot-dashed line) and sample variance (dotted line) are roughly equal up to this crossover. As $e$ increases, progressively larger regions get excluded from $k$-space. This is clearly visible in Figures 12b–12d especially for $k \lesssim 1$ Mpc$^{-1}$, through a systematic drift of radii of spherical shells (‘+’ symbols) towards higher values as $e$ increases (red to green to blue). In other words, increasing $e$ from 0 to 3 makes it progressively harder to recover scales with 0.06 Mpc$^{-1} \lesssim k \lesssim 0.2$ Mpc$^{-1}$. This results in partial removal of different uncertainties and the signal, besides an inherent decrease in signal strength with increasing $k$. However, the decrease in signal by $\lesssim 1–2$ orders of magnitude (solid red, green and blue curves) is less rapid than that in the foreground contamination (dashed red, green and blue curves), which decreases by $\gtrsim 1–2$ orders of magnitude. This is true for both bandpass shapes but is quite pronounced for the extended Blackman–Nuttall window (see Figure 12b), where the foreground contamination reduces by $\sim 5–10$ orders of magnitude. On the other hand, as $e$ increases, thermal noise component changes at most by a factor of 2 as seen from Figure 12d. Effectively, the thermal noise component is only mildly affected compared to the signal and foregrounds. Regardless of the bandpass window used, the colored curves which are almost coincident in Figure 12d show that there is no improvement in overall sensitivity as $e$ is varied. This is a consequence of the nature of inverse quadrature weighting used in averaging in spherical shells of $k$. However, it is very important to reiterate that the foreground contamination decreases more rapidly than the loss in signal as $e$ is increased. In fact, foregrounds are almost completely removed for $e \gtrsim 1$ while using the Blackman–Nuttall window. Hence, using a combination of extended Blackman–Nuttall band shape with certain members of the family of EoR window ($1 \lesssim e \lesssim 2$) does not appear to improve MWA sensitivity, but offers a significant leverage in reducing foreground contamination from the power spectrum, thereby providing a cleaner EoR window. This could become very significant when imperfect source subtraction (position and calibration errors) and extended emission from extragalactic and Galactic foregrounds are also taken into account. This is due to the 7–8 orders of magnitude of extra tolerance provided by an extended Blackman–
Figure 12. Properties of signal and uncertainty components in power spectrum (in units of $K^2 \text{Hz}^2$) for observing mode (1). These are determined by averaging in shells of $k$ for various values of EoR window parameter, $e$. In each panel, the sub-panel on the left is obtained with a rectangular bandpass shape while that on the right is due to an extended Blackman–Nuttall window. The averaging excludes from the $(k_\perp, k_\parallel)$–plane, the wedge-shaped foreground window and an additional region parametrized by $e$ in equation (26). (a) Expected EoR $\text{H}_i$ signal and uncertainties in the power spectrum. The solid circles and lines denote the signal and the total uncertainty respectively. The latter consists of sample variance (dotted), thermal noise (dashed), and foreground contamination (dot dashed). (b) Signal (solid lines) and foreground contamination (dotted lines) for different values of $e$ ($e = 1, 2, 3$ in red, green and blue respectively) normalized by their respective values at $e = 0$ (black) in Figure 12a. (c) Thermal noise component in power spectrum for different values of $e$ (shown in legend) normalized by its values at $e = 0$. (d) S/N ratios for different values of $e$. In panels b through d, ‘+’ symbols denote the mean radii of spherical shells.
We investigate the sensitivity for observing mode (2), where a total of 1000 hours were divided over 20 patches of sky to obtain 20 independent measurements of power spectrum. This differs from observing mode (1) in that the total observing time is now divided on multiple fields. Figure 13a is the counterpart of Figure 12a and illustrates the signal and different uncertainty components obtained with the MWA in this observing mode. Counterparts to Figures 12b and 12c will be identical and are not shown. Figure 13b shows the ratio of signal to total uncertainty (S/N) for this observing mode.

Since the time on individual fields has reduced by a factor of $N_{\text{fields}} = 20$ leading to a reduction in the number of coherent visibility measurements per field, the thermal noise component in each measurement of the power spectrum has increased by the same factor. The foreground contamination and sample variance in an individual power spectrum measurement remain identical to those in observing mode (1) where only a single field is observed. When independent measurements of power spectra are averaged, all the components of uncertainty in the averaged power spectrum are reduced by a factor 4.47, where thermal noise component has worsened by the same amount. This is evident when Figure 13a is compared with Figure 12a. As a result, the crossover, up to which foreground contamination and sample variance dominate over thermal noise, moves leftward and now occurs at $k \approx 0.1 \text{ Mpc}^{-1}$ for both bandpass windows. This has a two-fold effect on overall sensitivity relative to that in the previous case: (a) the sensitivity in lowest bin of $k$ ($k \approx 0.06 \text{ Mpc}^{-1}$), where sample variance and foreground components dominated over thermal noise component in observing mode (1) has improved to 20, a factor of $\sqrt{N_{\text{fields}}} \approx 4.47$; and, (b) the sensitivity in other bins of $k (k \gtrsim 0.1 \text{ Mpc}^{-1})$ has degraded because the thermal noise component, which was already dominant in this regime, has worsened. The final effect on detectability is that $S/N > 1$ only for $k \lesssim 0.4 \text{ Mpc}^{-1}$.

How does sensitivity in observing mode (2) compare to that in observing mode (1)? Sensitivity is the result of a complex interplay between the relative magnitudes of the desired EoR Hi signal and different uncertainty components in the power spectrum. As far as MWA is concerned, thermal noise component appears to be dominant on scales with $k \gtrsim 0.1–0.2 \text{ Mpc}^{-1}$. Hence, dividing the observing time over multiple fields and averaging the independent measurements of power spectra helps improve the sensitivity by a factor of $\approx 4$ for $k \lesssim 0.1 \text{ Mpc}^{-1}$ but degrades it everywhere else. Consequently, the zone of detectability ($S/N > 1$) in $k$–space becomes narrower from $k \lesssim 0.8 \text{ Mpc}^{-1}$ to $k \lesssim 0.4 \text{ Mpc}^{-1}$. Thus, except for an improvement in sensitivity on the largest scales ($k \lesssim 0.1 \text{ Mpc}^{-1}$), increasing the number of independent fields seems to offer no significant advantage.

As far as effects of bandpass windowing and family of EoR windows are concerned, significant improvement in MWA sensitivity is neither seen with bandpass window shapes nor with the EoR window parameter $e$. However, it is crucial to obtain the cleanest EoR window possible in order to reduce the amount of systematics in the data, which may not be fully understood, such as those arising from unsubtracted foreground residuals. We find that refinements to the EoR window (through parameter $e$), a Blackman–Nuttall bandpass window shape and a combination of both significantly reduce the extragalactic foreground contamination in the measured power spectrum.

9. SUMMARY

The primary goal of this work was to understand and estimate some of the fundamental factors that limit sensitivity of an EoR Hi power spectrum measurement, namely, point–like extragalactic foreground contamination, thermal noise and sample variance of the Hi brightness temperature fluctuations. A secondary goal was to understand how these uncertainties compete with each other on different scales in determining the sensitivity of a radio interferometer array, such as the MWA, towards detection of EoR Hi power spectrum.

An analytic cum–numerical approach was used in representing residual image cubes by assuming a radio source count distribution on the sky. For the 128–tile MWA at 170.7 MHz, when sources above the $5\sigma$ classical source confusion threshold were subtracted, the $1\sigma$ classical source confusion limit near the zenith was found to be $\approx 35 \text{ mJy}$ for a natural weighting scheme.

Unsubtracted foreground sources and their sidelobes contaminate the predicted EoR signal. The frequency dependence of synthesized beam distributes the contamination from sidelobes onto a wedge–shaped region in $k$–space. We have presented a unified framework for signal and noise estimation using which we estimate foreground contamination in three–dimensional $k$–space. This framework also attempts to take into account multi–baseline mode–mixing effects caused by loss of coherence between non–identical baselines inside an independent cell in the spatial frequency domain. Using this framework, we establish an expression for the boundary of the wedge set by the horizon limit.

We show for the first time, quantitatively, how the usage of a finite bandpass spills the contamination from unsubtracted sources and their sidelobes into the EoR window. This spillover decreases by 7–8 orders of magnitude, for instance, in the range $0.2 \text{ Mpc}^{-1} \lesssim k_{\parallel} \lesssim 5 \text{ Mpc}^{-1}$ at $k_{\perp} \approx 0.01 \text{ Mpc}^{-1}$, by switching from a rectangular to an extended Blackman–Nuttall window. We argue this additional tolerance provided by the latter could prove to be of crucial significance in minimizing power spectrum contamination when the impact of imperfect source subtraction (due to position and calibration errors), and extended extragalactic and Galactic foregrounds are also considered. The frequency weighting in an extended Blackman–Nuttall bandpass window also lowers the thermal noise component of the power spectrum by 28% relative to that achievable with a rectangular window. Conversely, in order to achieve the same thermal noise power with both windows, the duration of observing with a rectangular window has to be $\approx 40\%$ longer when compared to an extended Blackman–Nuttall window.

We performed case studies of two different observ-
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The concept of EoR window was probed quantitatively. Foreground contamination can be drastically reduced by using an extended Blackman–Nuttall bandpass window and through refinements ($1 \leq e \leq 2$) to the EoR window.

By modeling in detail the various uncertainties, we have shown the significance of different uncertainties on various scales and their roles in determining overall sensitivity. Observing many independent fields of view worsens the thermal noise and hence degrades the sensitivity for the MWA relative to a single field observation of the same duration. Bandpass window shaping and refinements to the EoR window do not affect sensitivity, but have a significant effect on containing the foreground contamination.

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APPENDIX

A. ESTIMATING CLASSICAL SOURCE CONFUSION NOISE

At any given frequency, the field of view contains many sources. Any given synthesized beam area of an array consists of many unresolved sources along that line of sight. Due to the statistical nature of the distribution of sources, the flux density contained in a synthesized beam area varies across the sky. The classical source confusion is the variation in flux density due to random distribution of unresolved sources across different beam areas on the sky. The theory on confusion is discussed in detail in Condon (1974) and Rohlfs & Wilson (2000). This paper uses the discussion and notations presented in the latter.

The differential number density of sources per unit solid angle with respect to flux density is denoted by $dn/dS$, where $dn$ is the number of sources per steradian in a flux density interval between $S$ and $S + dS$. The variance in confusing source flux density in a given solid angle, $\Omega$, due to the number count distribution of sources is given by,

$$\sigma_C^2 = \Omega \int_{S_{\text{min}}}^{S_C} S^2 \frac{dn}{dS} dS,$$

(A1)

where $S_{\text{min}}$ and $S_C$ are the lower and upper limits on the flux density of radio sources respectively.

Since $\sigma_C^2$ is determined from the entire range of flux densities up to $S_C$, if there are bright sources, $\sigma_C^2$ will be overestimated. Hence, we perform an iterative procedure wherein all foreground sources brighter than $\rho_C \sigma_C$ and their sidelobes are subtracted, thereby updating the upper end of the flux density range, $S_C$. Here, $\rho_C$ acts as the source subtraction threshold factor. This iterative procedure of subtracting foreground sources brighter than $S_C = \rho_C \sigma_C$, and updating $S_C$ and $\sigma_C$, is performed until there are no sources brighter than $\rho_C \sigma_C$. We consider the $\sigma_C^2$ so computed as the “true” classical source confusion variance.

In practice, usually, deconvolution procedures can estimate and subtract foreground sources and their associated sidelobes to a limited extent leaving behind a residual image. With the knowledge of distribution of radio sources encoded in $dn/dS$, a set threshold factor ($\rho_C$), and the solid angle ($\Omega$) corresponding to the angular resolution limit, the depth of source subtraction and corresponding residuals in the residual image can be determined using the iterative procedure described above, and is given by the following equation (Rohlfs & Wilson 2000):

$$\rho_C^2 = \frac{S_C^2}{\Omega \int_{S_{\text{min}}}^{S_C} S^2 \frac{dn}{dS} dS},$$

(A2)

This is obtained by imposing the criterion that the residual image is allowed to have unsubtracted sources of flux densities up to $S_C \leq \rho_C \sigma_C$. We have assumed that the source subtraction threshold, $S_C$, is set only by the classical source confusion noise whereas, in practice, confusion caused by sidelobes in a residual image will also play a role in determining the source subtraction threshold in this iterative procedure.

We use the radio source statistics provided by Hopkins et al. (2003). Their best-fit expression for the source counts is:

$$\log\left(\frac{(dn/dS)}{(S^{-2.5})}\right) = \sum_{i=0}^{6} a_i \left[\log(S/\text{mJy})\right]^i,$$

(A3)

valid at 1400 MHz for $0.05 \text{ mJy} \leq S \leq 1000 \text{ mJy}$, where $a_0 = 0.859$, $a_1 = 0.508$, $a_2 = 0.376$, $a_3 = -0.049$, $a_4 = -0.121$, $a_5 = 0.057$, and $a_6 = -0.008$. $S$ is in units of mJy, and $(dn/dS)/S^{-2.5}$ is in units of $\text{Jy}^{1.5} \text{ sr}^{-1}$. The normalization by $S^{-2.5}$ indicates the expression is relative to a Euclidean universe.

The solid angle at various locations of an image in $(l, m)$–coordinates scales as:

$$\Omega = \frac{\Delta l \Delta m}{\sqrt{1 - l^2 - m^2}},$$

(A4)

where $\Delta l \Delta m$ is the pixel size. This implies, from equation (A2), that confusion variance ($\sigma_C$), and flux density cutoff ($S_C$) are functions of position in the residual image. Estimating confusion noise over a wide range of solid angles requires an extrapolation of the empirical function in equation (A3) on both ends of the flux density range. We have extrapolated at the higher end by a flat function mimicking a Euclidean local universe behavior, and at the lower end by an extension of the same slope found between 2–20 mJy. Figure 14a shows equation (A3) extrapolated at both ends beyond the aforementioned range in flux density (specified by vertical dotted lines). The dashed segments $S < 0.05 \text{ mJy}$ and $2 \text{ mJy} \leq S \leq 20 \text{ mJy}$ have slopes identical to each other.

Under the assumption that the source population remains the same at the relevant frequency (170.7 MHz, for instance) under consideration as at 1400 MHz, and that these sources are unresolved with the MWA, the same
expression for the distribution of source counts can be used once the spectral index is taken into account. There has been conflicting evidence in literature (Randall et al. 2012, and references therein) over the spectral index properties of radio sources at frequencies below 1.4 GHz and whether the spectral index flattens for faint sources at low frequencies. Further, Kellermann (1964) has pointed out that spectral index distribution is not independent of the observing frequency since sources with flatter spectral index are more likely to be observed at higher frequencies and those at lower frequencies tend to have steeper spectral index, and has subsequently provided a spectral index correction that tends to offset this bias. However, for our work, we have adopted a mean spectral index for radio sources as \( \alpha = -0.78 \) (Ishwara-Chandra et al. 2010), where \( S \propto f^\alpha \). Compared to flatter values of spectral index, our adopted value of \( \alpha = -0.78 \) gives us higher values for the classical source confusion at 170.7 MHz, making our sensitivity estimates conservative.

Sources can be subtracted down to various levels of threshold factor, \( \rho_C \). Figure 14b represents a numerical solution for equation (A2) using \( \frac{dn}{dS} \) from equation (A3) for various values of \( \rho_C \), as a function of the beam solid angle, \( \Omega \).

As expected from equation (A2), Figure 14b shows that \( \sigma_C \) and \( S_C \) are non-linear functions of both \( \rho_C \) and \( \Omega \). \( \sigma_C \) depends on the choice of \( \rho_C \): it increases with \( \rho_C \). The dash-dotted line denotes the cutoff \( S_C \) at which on average one source of flux density exceeding the cutoff is expected (Subrahmanyan & Ekers 2002). The gray vertical line at \( \Omega \approx 2 \times 10^{-6} \) sr denotes the typical size of a resolution element for a naturally weighted image obtained with 128–tile MWA.

**B. FOREGROUND CONTAMINATION**

Following the notations established in [3], the true visibilities of foregrounds, \( V_{uvf}^{FG: T}(u,f) \), are modified by the instrument as:

\[
V_{uvf}^{FG: T}(u,f) = \int V_{uvf}^{FG: T}(u,f) S_{uv}(u) W_f^B(f) W_{uv}^P(u) e^{-j2\pi f d} df
\]

\[
= V_{uvf}^{FG: T}(u) S_{uv}(u) W_f^B(\eta) W_{uv}^P(u)
\]

\[
= V_{uvf}^{FG: T}(u) S_{uv}(u) W_{uvf}(u).
\]

(B1)

where the true foreground visibilities, \( V_{uvf}^{FG: T}(u,f) \), have been convolved with the spatial frequency response of the antenna’s power pattern, \( W_{uv}^P(u) \), multiplied by the sampling function, \( S_{uv}(u) \), in the \((u,v)\)-plane, multiplied by the
bandpass window function in frequency, $W^B(f)$, and Fourier transformed along frequency to obtain the measurements in Fourier space ($\mathbf{u}$). $W^B(\eta)$ is the Fourier transform of $W^B(f)$, and $j = \sqrt{-1}$. $W_{uv\eta}(\mathbf{u}) = W^P_{uv\eta}(\mathbf{u}) * W^B(\eta)$, where $W^P_{uv\eta}(\mathbf{u})$ may be interpreted as the spatial frequency response of the antenna’s power pattern over an infinitely uniform bandpass. Assuming changes in the antenna power pattern over the observing band are insignificant, $W^P_{uv}(\mathbf{u}) = W^P_{uv\eta}(\mathbf{u}, \eta = 0)$.

The covariance matrix for the measured foreground visibilities in Fourier space may be written as:

$$C^{FG}(\mathbf{k}_i, \mathbf{k}_j) = \left\langle V^{FG}_{uv\eta}(\mathbf{u}_i) V^{FG}_{uv\eta}(\mathbf{u}_j) \right\rangle$$

$$= \iiint \iiint \frac{V^{FG;T}_{uv\eta}(\mathbf{u}_p) V^{FG;T}_{uv\eta}(\mathbf{u}_q)}{S_{uv}(\mathbf{u}_i) S_{uv}(\mathbf{u}_j) W_{uv\eta}(\mathbf{u}_i - \mathbf{u}_p) W_{uv\eta}(\mathbf{u}_j - \mathbf{u}_q)} d^3\mathbf{u}_p d^3\mathbf{u}_q,$$

(B2)

where $*$ denotes a complex conjugate. But, being an uncorrelated statistical signal, $\left\langle V^{FG;T}_{uv\eta}(\mathbf{u}_p) V^{FG;T}_{uv\eta}(\mathbf{u}_q) \right\rangle = P^{FG}(\mathbf{u}_p) \delta(\mathbf{u}_p - \mathbf{u}_q)$. Hence,

$$C^{FG}(\mathbf{k}_i, \mathbf{k}_j) = \iiint P^{FG}(\mathbf{u}) |S_{uv}(\mathbf{u})|^2 W_{uv\eta}(\mathbf{u}_i - \mathbf{u}_p) W_{uv\eta}(\mathbf{u}_j - \mathbf{u}_q) d^3\mathbf{u}$$

$$= \iiint P^{FG}(\mathbf{u}) |S_{uv}(\mathbf{u})|^2 W^P_{uv}(\mathbf{u}_i - \mathbf{u}) W^P_{uv}(\mathbf{u}_j - \mathbf{u}) W^B(\eta_i - \eta) * W^B(\eta_j - \eta) d^3\mathbf{u},$$

(B3)

The power spectrum is simply the diagonal of the covariance matrix, i.e., when the intensities at locations are compared with themselves: $P^{FG}_{inst}(\mathbf{k}) = C^{FG}(\mathbf{k}_i, \mathbf{k}_j) \delta_{ij}$. Thus,

$$P^{FG}_{inst}(\mathbf{u}) = \iiint P^{FG}(\mathbf{w}) |S_{uv}(\mathbf{u})|^2 |W_{uv\eta}(\mathbf{u} - \mathbf{w})|^2 d^3\mathbf{w}.\quad(B4)$$

An insight into mode–mixing may be obtained if the above expression for power spectrum is re-expressed as a Fourier transform of quantities in the $(l, m, f)$–coordinates as follows:

$$P^{FG}_{inst}(\mathbf{u}) = \iiint P^{FG}_{lmf}(\mathbf{\hat{u}}) * |S_{lm}(\mathbf{\hat{u}})|^2 |W^P_{lm}(\mathbf{\hat{u}})|^2 |W^B(f)|^2 e^{-j2\pi(x_1 f + \eta)} d^3\mathbf{\hat{u}}.\quad(B5)$$

Confusion variance, $P^{FG}_{lmf}(\mathbf{\hat{u}})$, from extragalactic foreground sources will be considered as the cause for foreground contamination in the power spectrum. It may be computed using equations (A2) and (A3). When line-of-sight ($f$ and $\eta$) terms are dropped, the integrand in the above equation is consistent with that in equation (24) of Bowman et al. (2009).

We further assume that $P^{FG}_{lmf}(\mathbf{\hat{u}})$ is independent of frequency, provided the residuals have no spectral variations. Noting that $\mathbf{u} = f \mathbf{x}/c$, the above equation may be re-written as:

$$P^{FG}_{inst}(\mathbf{u}) = \iiint P^{FG}_{lm}(\mathbf{\hat{u}}) * |S_{lm}(\mathbf{\hat{u}})|^2 |W^P_{lm}(\mathbf{\hat{u}})|^2 |W^B(f)|^2 e^{-j2\pi(x_1 f + \eta)} d^2\mathbf{\hat{u}} df$$

$$= \iiint P^{FG}_{lm}(\mathbf{\hat{u}}) * |S_{lm}(\mathbf{\hat{u}})|^2 |W^P_{lm}(\mathbf{\hat{u}})|^2 * |W^B(\eta)|^2 * \delta \left(\frac{x_1}{c} + \frac{\mathbf{\hat{u}}}{\mathbf{\hat{u}}_0}\right) d^2\mathbf{\hat{u}} \quad(B6)$$

where $\mathbf{x}$ is the baseline vector in units of distance. The argument of the $\delta$ function connects transverse spatial structure to that along the line-of-sight. This demonstrates the mode–mixing aspect of contamination from foregrounds in the power spectrum.

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