# MULTI-REDSHIFT LIMITS ON THE 21CM POWER SPECTRUM FROM PAPER

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#### ABSTRACT

We present new observations from the Donald C. Backer Precision Array for Probing the Epoch of Reionization (PAPER) telescope which place new deeper limits on the HI power spectrum over the redshift range of 7.5 < z < 10.5, extending previously published single redshift results to cover the full range accessible to the instrument. The epoch of reionization power spectrum is expected to evolve strongly with redshift, and it is this variation with cosmic history that will allow us to begin to place constraints on the physics of reionization. The primary obstacle to this goal are bright foregrounds and the attendant systematics associated with high dynamic range measurements. Here we present an integration across a full season of data and find systematics just beginning to dominate, though well below any levels previously reported. This 500 hour integration is the longest such yet recorded and tests foreground removal to a theoretical dynamic range of 10,000 to 1. We use filtering techniques that take advantage of the large instrumental bandwidth to make a high dynamic range subtraction of foreground power by averaging together many low sensitivity measurements. . Power spectra at different points across the redshift range reveal the variable efficacy of the foreground subtraction. Noise limited measurements of  $\Delta^2$  at  $k=0.2 \mathrm{hMpc}^{-1}$  and z=7.55 reach as low as  $(48 \mathrm{mK})^2$   $(1\sigma)$ . Most spectra demonstrate a robust foreground removal of foreground signals to the thermal noise limit, leaving systematics on the scale of the noise. To better understand the source of these residual systematics we examine different estimates of the noise in the data and compare with theoretical predictions. We find that in the case of the highest redshift power spectrum, foregrounds begin to dominate over the noise, reflecting the increased difficulty in filtering foregrounds near the edge of the instrumental band.

Subject headings: reionization

# 1. INTRODUCTION

The Epoch of Reionization, when the first luminous objects ionized the pervasive cosmological hydrogen, is predicted to be observable in highly redshifted 21 cm radiation. The Donald C. Backer Precision Array for Probing the Epoch of Reionization (PAPER, Parsons et al. (2010))<sup>13</sup> is a low frequency radio interferometer experiment dedicated to opening this window on the universe. Challenges include foregrounds which are brighter by several orders of magnitude and long integration times necessitated by the limited collecting areas of first-generation instruments. Direct observation of hy-

drogen before and during re-ionization is predicted to deliver a wealth of cosmological and astrophysical data, including the nature of the first stellar objects and the timing and rate of galaxy formation. Reviews on the physics of reionization as well as theoretical expectations on the nature of foregrounds may be be found in Furlanetto et al. (2006); Morales & Wyithe (2010); Pritchard & Loeb (2012).

Other telescopes seeking to measure this signal include the Giant Metre-wave Radio Telescope (GMRT; Paciga et al. (2013)), the Low Frequency Array (LOFAR<sup>14</sup>; Yatawatta et al. (2013)) and the Murchison Widefield Array (MWA<sup>15</sup>; Bowman et al. (2013) and Tingay et al. (2013)).

PAPER is located in the Karoo desert at the site of the South African portion of the future Square Kilometer Array<sup>16</sup> and has doubled in size on a yearly basis since 2009; science-grade observations have been made with each stage of the build-out.

Here we report on deep integrations made with a 32 element array in 2011, first described in Parsons et al. (2014), hereafter P14. Our data reduction pipeline was described in detail in P14, where the same methods were used to give the deepest limits yet on the HI power spectrum in the presence of bright foregrounds at redshift

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7.68. However, the Epoch of Reionization signal is expected to evolve strongly with redshift. In fact, it is this signature variation which will distinguish it from foregrounds and it is this variation with cosmic history that will allow us to begin to place constraints on the physics of reionization (Pritchard & Loeb 2008; Pober et al. 2014). Therefore, while a detection of the 21 cm signal at even a single frequency would be a tremendous breakthrough, analysis techniques must be developed to capitalize on the wide bandwidths of the current generation of high-redshift 21cm telescopes. Using the same data set as P14, this paper presents improved upper limits on the HI power spectrum over the redshift range 10.7 > z > 7.2. In Section 2 we summarize the observations, in Section 3 review the reduction methodology, we present the new upper limits in Section 4, and in Section 5 we offer conclusions and discussion of future work.

#### 2. OBSERVATIONS

The work here follows the same basic procedure and uses the same underlying data set as P14. Here we provide a guick summary and refer the reader to P14 for a more in-depth discussion. A general overview of the PA-PER system can be found in Parsons et al. (2010), calibration of the primary beam in Pober et al. (2012), and imaging results in Jacobs et al. (2011, 2013) and Stefan et al. (2013). Sensitivity analysis described in Parsons et al. (2012a) revealed that for the low gain elements employed by PAPER, a highly redundant "grid" type arrangement offers a significant sensitivity boost extending the performance of the compact designs motivated by Morales & Hewitt (2004): Bowman et al. (2006): Lidz et al. (2008). In most interferometers the locations of the antennas are optimized such that each baseline samples a different Fourier mode of the sky; this is the ideal case for reconstructing images where each mode contains different information. For a power spectrum measurement the key metric is sensitivity per mode, rather than number of modes. A grid configuration allows many samples of each cosmological mode, to be averaged to a high sensitivity before being combined with other Fourier modes. The PAPER South Africa 32 antenna deployment (PSA32) was arranged in a  $4\times8$  grid with a column spacing of 30m and a row spacing of 4m. In our analysis, as in P14, we include only the three shortest types of spacings where the reionization power is expected to be brightest. This selection includes those between adjacent columns and within at least one row of each other, a selection containing 70  $\sim$ 30m-long baselines. We will use these baselines to make a one dimensional estimate -spectral line of sight only— of the HI power spectrum.

Observations spanning the band between 100 to 200-MHz (13.1 > z > 6.1) were recorded at a resolution of 48kHz and 10.7s beginning Dec 7, 2011 and ending March 19, 2012 (with some down-time for maintenance) giving a total of 513 hours over 92 nights. Within this set we included observations in the LST range 1h - 9h where the sky dominated system temperature is at a minimum. Note that this LST range is slightly shorter than in P14, which extended to LST of 12 hours. These last three hours were found to contribute minimally to increasing sensitivity while being dominated by bright galactic foreground emission and so have been excluded here.

#### 3. REDUCTION

Here we summarize our data reduction steps; for more details see Section 3 of P14. In summary, we use 70 nearly identical baselines to make a 1D estimate along the spectral or line-of-sight direction of the reionization era HI power spectrum. All processing, save calibration, and the final cross-multiplication step treats each baseline as independent. Foregrounds and interference are removed on a per-baseline basis with no a-priori sky model using signal processing techniques and a physical model of the array. In the final cross-multiplication step, the last layer of systematics is estimated and removed by projecting non-physical correlations between baselines.

# 3.1. Delay/Fringe-Rate Filtering: Averaging and Foreground Removal

In several stages throughout the analysis process we take a 2D Fourier transform of the visibility spectra  $V(\nu,t)$  into "delay/fringe rate" space where delay is the Fourier dual to frequency and similarly fringe-rate for time. In this space, smooth spectrum sources are physically localized to delays shorter than the light travel time length of the baseline and fringe rates shorter than the sidereal rotation rate of the tip of the east-west component of the baseline vector. Sources at the horizon, in the direction of the baseline vector, have the longest delays, while fringe rates are highest where the celestial equator crosses the horizon.

In this Fourier space, sources are highly localized with deviations from a flat spectrum manifesting as a slight dispersion. The spectrum sampling function, which is uneven due to flagging of interference takes the form of a convolution by a point-spread-function (PSF) in the same way an imperfect sampling of the *uv* plane gives rise to the angular PSF of an interferometer. If enough data is missing this PSF can cause smooth-spectrum sources to leak beyond the "horizon", the light travel time limit. To account for this, we use a CLEAN like, iterative, peakfinder and subtraction algorithm which is limited to finding peaks within the physically allowable ranges of delay and/or fringe-rate. In this case, the 1D "CLEAN" beam is the Fourier transform of the spectral or time sampling function (Parsons & Backer 2009).

The data analysis pipeline essentially consists of iterative application of the delay or fringe-rate transform process, with an ever tightening allowable number of modes, interleaved with stages of averaging (time, frequency, night), before finally computing a power spectrum. This final step takes advantage of the redundant baselines to make an unbiased power spectrum estimate by cross-multiplying identical baselines and then averaging the power spectrum modes. By not combining baselines until the last step and by assuming that the line of sight transform is well approximated by the delay transform we dramatically simplify the analysis procedure to a series of signal processing steps. It also obviates the need for precision sky and instrumental beam models which are required by imaging arrays.

This isolation of foregrounds to a region below a line-of-sight  $k_{\parallel}$  mode that increases with baseline length is also the much discussed "wedge" (Liu et al. 2014a,b; Thyagarajan et al. 2013; Pober et al. 2013a; Trott et al. 2012; Morales et al. 2012; Parsons et al. 2012b; Vedantham

et al. 2012; Datta et al. 2010; Parsons & Backer 2009). As we only have three baseline types of nearly the same length the wedge manifests as a single value of  $k \sim k_{\parallel}$ , below which foregrounds are expected to dominate. The details of each application of the delay/fringe-rate transform will be laid out in the following sections as we provide a brief walk-through of the processing pipeline.

## 3.2. Selection of Redshift Bins

The redshift bins for which we have computed power spectra –shown in Figure 1– have been selected from the available bandwidth using two criteria: minimizing covariance between redshifts and avoiding missing spectral data. We minimize covariance between adjacent redshift bins by limiting overlap of each bin to the spectrum at the outer half of the bin channel range which, as will be described in section 3.6, is significantly down-weighted by the use of a Blackman-Harris window. Avoiding missing data is important because the power spectrum method, developed in P14, which leverages the redundant baselines to estimate non-sky covariance and project out these contaminated modes is particularly sensitive to missing data -the inverse of the covariance is not definedso we also select only redshift ranges with channels that have no missing data over the entire sidereal period (see the dotted line in Figure 1). With these constraints we arrive at the redshifts 10.3, 8.5,7.9 and 7.45. For the purposes of comparison with P14 we also include the redshift 7.68 bin. Dillon et al. (2013) describes a pseudo-inverse method for handling the covariance introduced by missing data. Though not implemented here, this method is in the process of being adapted for use in highly redundant analysis (Liu et al. 2014a,b).

The spectral width of the redshift bins is dictated by two competing needs. First, sensitivity and foreground reduction both benefit from wider bandwidths. On the other hand, power spectrum measurements which trace the evolution with redshift require smaller bandwidths. Often, the evolution scale is taken to be on order  $\Delta z \approx 0.5$ , which translates to a bandwidth limit  $B < 0.5(f^2/1421)$  MHz. To balance these two competing constraints we choose a single spectral bandwidth of 20MHz, weighted by a Blackman-Harris window for an effective width of 10MHz, or  $\Delta z = 0.5$  at z = 7.45 ranging up to a  $\Delta z = 0.86$  at redshift 10.3.

### 3.3. Initial Averaging

First, the raw data are down-selected to just the 70  $\sim$ 30m long baselines described in Section 2. The visibilities are then compressed in the frequency and time directions by removing delay modes and highest frequency fringe rates corresponding to a 300m baseline (the longest baseline in the array). This filtering<sup>17</sup> is done in tandem with a radio frequency interference (RFI) flagging step, using the residuals which have had bright sky-like signals removed to flag  $4\sigma$  deviations before feeding the flags back into another iteration of the compression step. This mitigates the effects of bright, narrow band, interference being scattered into higher delay modes (where only the reionization signal is expected to be found) and results in a time and frequency bin of length 40s and

width 492.61kHz. This process reduces the data volume by a factor of  $\sim 20$ , or roughly an order of magnitude improvement on traditional time and frequency averaging which in this array would be limited to 100kHz and 10s to avoid averaging away fringes.

## 3.4. Calibration

We model the gain as a per-antenna amplitude and a phase slope -physically a single time delay- and single real, low order polynomial passband for all antennas. Because the array samples correlations redundantly, the relative calibration between antenna is numerically overdetermined and tractable as a linear algebra problem (Liu et al. 2010). As described in P14, we compute the ratio between redundant baselines, fit for a gain and phase slope and then algebraically solve for a per-antenna solution. Using this method we have avoided calibrating each channel independently to preserve as much frequency variation as possible. These solutions vary little over the three month observing period, exhibiting less than 1\% r.m.s. variation. A single solution derived for the Dec 7 data set is used for the entire observing run. Time and frequency variation of redundant solutions is explored in general in Zheng et al. (2014).

Relative calibration allows the gains of different antennas to be calibrated against one another. However, to obtain a correctly normalized power spectrum, it is necessary to set a flux scale for which we need a flux measurement of a calibration standard. For this we form a beamform on a known, bright calibration source. By itself, the redundant calibration does not contain enough phase information to phase coherently to a sky location, there remain two free phase parameters which cannot be solved by redundancy alone. We find these by fitting a model of Pictor A, Fornax A, and the Crab Nebula during a time when the sky is dominated by these three sources while marginalizing over the  $unknown^{18}$  apparent flux ratio between the three sources. With the delays in place we are now able form a beam on Pictor A and (for each channel) set the overall amplitude to the calibration value of 382 (f/150MHz)<sup>-0.77</sup>Jy found in Jacobs et al. (2013).

## 3.5. Foreground Filtering and Night Averaging

Foregrounds are filtered from the calibrated data by removing all bright delay components with light travel times less then the baseline length. Where during the previous compression step a liberal horizon of 300m (1800ns, much longer than the 30 meter baselines under study) was used to calculate the window size, we now choose a window corresponding to the 30m baselines under study with an extra buffer of 15ns to provide a slight buffer against the 1/B~12ns resolution of the delay spectrum. The broadband delay spectrum model is iteratively built then subtracted from the data leaving residuals which we carry into the next stage. Next, a four hour long running mean is subtracted. This removes excess correlation due to cross-talk in the analog signal chain. The residuals are then flagged once more for RFI

<sup>&</sup>lt;sup>18</sup> Unknown in the sense of a joint uncertainty in source flux and primary beam pattern. Though the fluxes of these sources is in some cases fairly well constrained, fluxes at 150MHz are still fairly uncertain as is the PAPER beam in those directions.

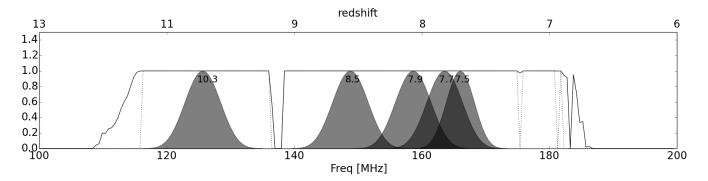


Figure 1. The average amount of data remaining after interference flagging over the 3 month period between Dec 2011 and March 2012 (black line) is quite high. Redshift bins (in grey, redshift center indicated with text label) are chosen to include spectral channels with uniform weight, i.e. no missing channels while maximizing coverage over the band. Redshift 7.68 is included for comparison with P14. Channels with no missing data are indicated by the dotted line, visible at the edge of flagged channels. Each redshift bin is 20 MHz wide, but weighted by a Blackman-Harris window function which heavily down-weights the outer 10MHz for a Noise Equivalent Bandwidth of 10MHz. The interference is almost exclusively dominated by two features: ORBCOMM satellites at 137MHz and an unidentified intermittent line emitter at 175MHz. The roll off at 115MHz is due the rising noise at low frequencies being incorrectly flagged as interference.

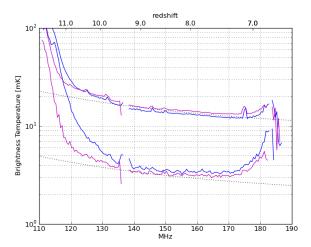


Figure 2. Root mean square (rms) noise before and after integrating to 13 minute LST bins (top and bottom sets of curves) indicating the rising significance of foregrounds as we take the last deep integration step. Noise is estimated by differencing adjacent frequencies (magenta) and between redundant baselines (blue), compared with an estimate of the theoretical noise level with a  $T_{sys}$ of 550K (dotted). The top three lines show noise after filtering foregrounds and binning into 40s long sidereal bins. At this noise level the frequency and baseline differences are roughly similar. both demonstrate the same small bumps of increased noise due to interference flagging and are consistent with noise over much of the band. The bottom three lines show the noise level after integrating up to the maximum fringe rate of 776s. The step between the top and bottom sets is the last coherent integration with noise largely decreasing by the expected factor  $\sqrt{776/40}$  except in the difference between baselines (blue) which demonstrates a clear excess at all frequencies, particularly above redshift 10. With this last large jump in sensitivity we are now seeing the slight dominance of baseline covariance over the spectral derivative rms.

before the 92 nights of data are averaged into 40 second long local sidereal time (LST) bins, which as PAPER is a drift-scanning instrument, are equivalent to bins in Right Ascension (Declination is fixed at -30°). During averaging we found that some LST bins were dominated by a small number of exceedingly bright samples lying well outside the rest of the gaussian distributed data. To compensate we filter the 10% brightest samples in each bin. The source of these outliers is not known, a likely possibility is an instability in the analog signal chain stim-

ulated by weather or bright interference, a circumstance that has since been observed in later seasons.

Though the frequency and repeated nightly observations have been averaged to their maximum extent, at 40s integrations the time axis has yet to be fully exploited. Sky-like fringes on a 30m baseline rotate much slower than 40 seconds. Performing a final fringe-rate filter, limiting to fringe-rates expected on a 30m baseline (down from 300m in the last iteration), we arrive at a data-set averaged to 789s, the maximum possible while still maintaining fringe coherence. The root mean square of the residual signal (seen in Figure 2) at the end of this process is close to the 3mK level expected given the total integration time and system temperature.

## 3.6. Power Spectrum

The output of the above steps is eight sidereal hours of calibrated, foreground filtered visibility data averaged over 92 nights. The power spectrum is estimated in the delay spectrum of a 10MHz bandwidth range centered on the redshift of interest. To preserve the isolation of any foregrounds which remain, we increase the spectral range by 5MHz on each side and multiply by a Blackman-Harris window thus providing a much higher dynamic range delay spectrum point spread function.

This leaves us with 40 delay samples on each of 70 baselines which are divided into three redundant groups. Within these groups we cross correlate delay spectra between different redundant baselines. The cross multiplication of the same delay modes between different redundant baselines provides an unbiased estimate of the power spectrum. These "sky-like" correlations should be identical between all redundant baselines to within the level of the noise, while all other cross multiplications between delay modes should not be correlated between different baselines. In practice these non-sky-like modes do occasionally have significant power which leaks into the correlations which sample the power spectrum. These are removed by iteratively dividing the covariance into a model of systematics and a model of sky-like emission and then projecting out large residual modes. This is done by dividing the baselines into different groups such that all cross-multiplications are done without introducing noise bias. For more see Appendix C of P14.

The residual elements of the correlation matrix corresponding to cross-multiplication of matching delay bins between different baselines are all un-biased samples of the power spectrum. To estimate the final power spectrum and its uncertainty we compute the mean and variance of many random randomly-selected subsamples, sampling the dimensions: sidereal time, redundant baseline pair, and delay sign<sup>19</sup>.

#### 4. RESULTS

## 4.1. Foreground Filtering and Noise Levels

The root-mean-square of the filtered visibilities measures how well we've removed foregrounds. In Figure 2 we examine  $T_{rms}$  as calculated by differencing between adjacent channels and between redundant baselines. We see that in the last stage of coherent integration, the step between 40 second to 789 second integrations, the noise level decreases by the expected factor of  $\sqrt{789/40}$ . At both stages the noise level found by both methods is roughly consistent with the theoretical level, only deviating significantly at the edges where the effective signal to noise due to the decreasing passband gain, Blackman Harris window, and uneven sampling (see Fig 1) drops precipitously.

The theoretical visibility r.m.s. noise level is a straightforward calculation, but given the method employed here, should be given due description. The noise level in a long integration (Eq. 2) is caused by the temperature of the sky (mostly due to galactic emission) which varies with LST and by the receiver noise, both on the scale of a few 100K. The noise of any long integration is also dependent on the number of observations in each LST. The number of points in each bin is mostly determined by the observing schedule, which for PAPER is sunrise to sunset, but also by any flagging of data or observing outage (only 92 of the 103 days in the observing period have recorded data). Unfortunately, technical problems limited the retention of the exact cumulative count of points averaged into each LST bin during the averaging process. Here we have chosen to use the first order estimate based on the observing schedule to estimate the number of observations in each LST. Some bins are observed only a few times, while a slim range in the LST range 6 - 9 hours is observed on every night. Over a number of LST bins  $N_{\rm lst}$  the average effective time per bin  $N_{eff}$  is given by

$$\frac{1}{\sqrt{N_{eff}}} = \left\langle \frac{1}{\sqrt{N(lst)}} \right\rangle_{lst} \tag{1}$$

On average each bin is in this dataset is effectively measured 39.6 times. Given this accounting of the number of samples N(lst), we estimate the r.m.s. temperature of the averaged data set to be

$$T_{rms} = \frac{1}{N_{\text{lsts}}} \sum_{\text{lst}} \frac{T_{sky}(lst) + T_{rcvr}}{\sqrt{2BtN(lst)}} \approx \frac{(\langle T_{sky} \rangle + T_{rcvr})}{\sqrt{2BtN_{eff}}}$$
(2)

<sup>19</sup> As visibilities are complex, both the positive and negative delays carry information. Physically the two signs correspond to the two halves of the sky.

where B and t are the 496kHz channel width and lst bin length, respectively and the approximation indicates that we have assumed average values for the number of samples and the time dependent sky component of the system temperature. In P14 the total system temperature  $(T_{sky} + T_{rcvr})$  was estimated to be 550K using 20 nights of the 92 night set which covered LSTs 3-9. Using the global sky model (de Oliveira-Costa et al. 2008) and a model of the PAPER primary beam, we estimate the mean  $T_{sky}$  during that 20 night period to be 250K, implying a receiver temperature of 300K. The mean temperature over the full 92 night period is only 4K higher; the estimate of 550K for the system temperature is a good estimate for the full observing period.  $T_{rms}$  using 550K and an effective sample count of 39.6 is plotted as a dotted line in Figure 2 for before and after fringe rate filtering (t=40s and 789s)

The final fringe-rate step is the last coherent average of the processing routine and represents the last large gain in sensitivity; all following averaging steps are done incoherently on the square of the visibilities where sensitivity is gained at a much slower rate. For this reason the noise curves of the fringe-rate averaged data set (bottom set of curves of Figure 2) merit closer inspection. At this sensitivity level we see a notable excess of power in the r.m.s. difference between baselines. Though the cause of this excess is unknown, it is suggestive of the excess covariance between redundant baselines which we remove in section 3.6, a hypothesis also supported by the high levels of residual foregrounds in the redshift 10.5 bin where the baseline difference r.m.s. is highest.

### 4.2. Power Spectra

In Figure 3 we show the power spectra at different redshift bins –showing both the spherically averaged power spectrum P(k) and the volume weighted  $\Delta^2(k) \equiv \frac{k^3}{2\pi^2} P(k)$ . The P(k) spectrum averages the three baseline types but preserves the positive and negative delays; it is essentially just the average delay spectrum scaled to temperature and cosmological units. Note that due to the choice of window function, the adjacent k bins correlate at the 50% level. In Figure 5 we see different k mode slices as a function of redshift, all plotted with  $1\sigma$  error bars derived from the bootstrap process and compared with the theoretical noise level. Table 1 we list the data plotted in these slices.

The theoretical noise power spectrum (dashed line in Figures 3 and 4 and grey region in Figure 5) is estimated using the method described in Pober et al. (2013b), assuming a system temperature of 550K and the observing scheme described in Section 2. In all figures the noise levels are plotted for comparison with the power spectrum values rather than the error bars, i.e. 68% of uncorrelated, noise-dominated points should lie below the line or within the grey region. Points where the error bar does not cross zero are inconsistent with noise at >68% these are listed as "detections" in Table 1.

If our measurements were completely noise dominated, 68% of the data points ought to include zero within their error bars. In Table 1, we divide the points into "Detections" (power spectrum values smaller than  $1\sigma$  errors), and "Upper limits" (measured values larger). With only one exception, this scheme yields the same results

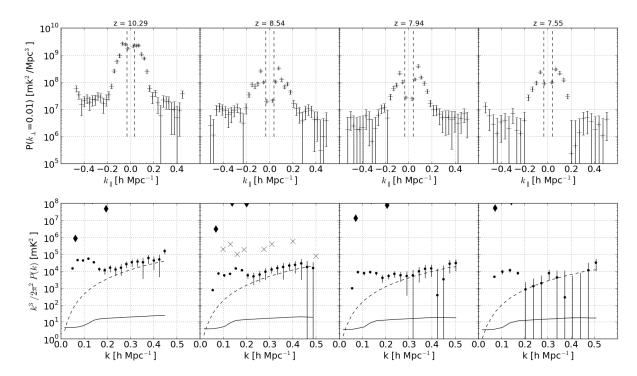


Figure 3. PAPER power spectra at four redshifts. On the top,  $P_k$  spectra (a simple units change from the raw delay spectrum) provide a useful diagnostic on foreground rejection, while on the bottom we plot in  $\Delta^2$  cosmological units with 1  $\sigma$  error bars. Redshift increases left to right. All spectra have an effective bandwidth of  $\pm 10 \text{MHz}$  covering the redshift span  $\Delta z \approx 1421 \text{MHz}$  ( $B/f^2$ ) which ranges from 0.8 at z=10.3 to 0.45 at z=7.4. The noise curve (dashes) is calculated using the method described in Pober et al. (2013b) and indicates the  $1\sigma$  confidence bounds on data points consisting purely of noise; 65% of uncorrelated noise like data points will lie below the curve. However, note the caveat that due to the weighting of the delay transform, adjacent k bins are 50% correlated. See Pober et al. (2014, 2013b) for a discussion of the approximations made in those calculations. The black line is a fiducial model at 50% ionization (Lidz et al. 2008). GMRT points from Paciga et al. (2013) indicated with 'x's and MWA points (also using 32 antenna) are black diamonds Dillon et al. (2014).

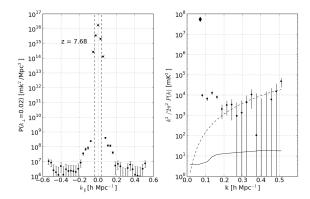


Figure 4. The redshift 7.68 bin has been reprocessed for comparison with the P14 result. See Figure 3 for a detailed description of curves. Compare with Figure 6 of P14. Note that the LST range processed here is 3 hours less than P14 for a small reduction in sensitivity but large reduction in foreground. The data shown here was processed exactly as the data in Figure 3, the foregrounds have been filtered. However the power spectrum shown in P14 added back in the narrow band delay spectrum points (the central five bins inside the horizon which is indicated by dashed lines) for comparison purposes we have added back in these points from that paper. They are marked with '+'s.

whether we use the theoretical errors or the bootstrap errors. Only a minority,  $\sim 20\%$ , of points are consistent with noise at the 1  $\sigma$  level, and considering all the data points together, the k modes above 0.2 in Table 1 are inconsistent with zero at 6.6  $\sigma$ . This quantifies the positive bias that is visually apparent in Figures 3 and 5.

Most of the power spectrum points have an excess

power at about the  $2\sigma$  level. As these measurements have a sensitivity level which is still two orders of magnitude away from theoretical reionization levels we consider the likelihood that these detections are evidence of reionization to be small. Far more likely is that we are limited by a residual foreground or systematic arising from approximations in the data analysis. The statistical significance of the residual  $-2\sigma$  after averaging over the entire season— is low enough that it is not possible to break it down along likely axes like sidereal time or time of day to probe its origin. Such studies must await higher sensitivity measurements with more antennas. However, there are several additional features that seem to suggest an excess originating in foregrounds rather than a HI signal

The predicted signature of the HI signal during reionization is one of a relatively short burst of power at small k values as the ionized bubbles are briefly at their largest, before the signal from the ever increasing ionized medium drops sharply (Pritchard & Loeb 2008). In contrast, this observed residual is relatively constant with redshift while rising at both ends of the band, mimicking the RMS noise curve (Figure 2). This suggests that the accuracy of the foreground filter decreases towards the edges of the band. Recall that the wide-band foreground model we have subtracted was built by weighting the entire band by a Blackman-Harris window function which provides a much higher isolation of foregrounds inside the horizon. The tradeoff that is made is that data in the outer parts of the spectrum are heavily down weighted. For example data at 126MHz are down-weighted  $\sim$ 75% compared to 150Mhz. This amounts to making an assumption about the spectral smoothness of the data, namely that 126MHz can be mostly modeled by extrapolation from 150MHz. The precision of this subtraction can be judged by examining the foreground residuals inside the horizon in the P(k) power spectra plotted in the upper part of Figure 3<sup>20</sup>. The relative depth of the redshift 10.5 P(k) foreground residual is substantially shallower then at 7.55 and is generally correlated with the amount of residual found at higher ks.

The residual level is also illustrated in the k slices of Figure 5. The shortest k mode (upper left of Figure 5) is the nearest to the light travel horizon and therefore most likely to be contaminated by foreground leakage. This k mode is well above the noise and has a redshift dependence much the same shape as the  $T_{rms}$  curve in Figure 2, with a minimum near redshift 8.5 and a dramatic rise above redshift 10. Longer k modes are all generally positively biased, though none higher than 3  $\sigma$ .

#### 5. CONCLUSIONS

The power of the highly redshifted 21 cm as a cosmic probe lies in its ability to probe a 3D volume by observing at different frequencies. The present analysis extends the work of Parsons et al. (2014), which used a single-baseline delay spectrum analysis to place an upper limit on the 21 cm power at z=7.68 (163 MHz). In this work, we take advantage of the wide bandwidth of the PAPER instrument to place limits on the power spectrum at a range of redshifts 10.7 > z > 7.2.

Most of the power spectrum data points demonstrate a removal of foreground signals to approximately twice the thermal noise limit. The deepest point –at redshift 7.55 and k=0.2– is a  $1\sigma$  upper limit of  $(48\text{mK})^2$ . For comparison, the Lidz et al. (2008) fiducial model of the reionization power spectrum estimates an amplitude of  $(4.4\text{mK})^2$ , for models with 50% reionization at z=7.55. Though most points are apparent detections of residual foregrounds they are still at a much deeper limit then previously found at these redshifts.

The fact that the best upper limit still comes from the z=7.6 band does illustrate that this redshift corresponds to a somewhat special frequency for the PAPER instrument, where the combined contribution of system noise, RFI signals, and residual foregrounds (if any remain) are at a minimum. The success of the filtering style foreground removal heavily leverages the surrounding spectral coverage, and is somewhat less effective at removing foregrounds across the entire band.

Further work will look to expand on this result on several fronts. Alternative algorithms to the wide-band CLEAN could potentially better remove the delay-space covariance introduced by RFI flags in frequency and are under investigation. Another active point of investigation is aimed at alternative windows to the Blackman-Harris used in the wide-band CLEAN (or wholly distinct algorithms) that could better remove foregrounds from the edges of the band while still limiting foreground bleed into the EoR window.

Future work will also include observations from subsequent seasons with more antennas. The observations

reported here demonstrate a nearly noise limited integration over a full season but used only a quarter of the final design antenna count. These 32 antenna observations from 2011 were followed by a full season of 64 antennas in 2012 and 128 in 2013. A second season of observing with the 128 antenna array is now under way. In going from 32 to 128 antenna the mK² sensitivity increases by the expected factor of 4 and by an additional factor of  $\sim$ 2 after accounting for the substantial uv-plane redundancy (Parsons et al. 2012a).

The analysis of the system temperature presented in section 4.1 is one of the most thorough such investigations applied to the PAPER instrument. It confirms the effectiveness of the wideband CLEAN in removing foregrounds over a wide range of the band, while illustrating weaknesses in this approach in pushing towards the edges of PAPER's frequency range. Furthermore, after the application of the fringe rate-based time averaging, we see hints that baseline-to-baseline variations between redundantly spaced baselines are becoming the dominant systematic in the analysis. Techniques like the covariance removal introduced in P14 and also used in this work can differentiate some of this variation from the sky signal; however some combination of improved calibration techniques (Zheng et al. 2014), model building (Sullivan et al. 2012), and a better understanding of error covariance (Liu & Tegmark 2011) may also prove valuable in reducing this systematic.

#### REFERENCES

Bowman, J. et al. 2013, Publications of the Astronomical Society of Australia, 30, 31

Bowman, J., Morales, M., & Hewitt, J. 2006, The Astrophysical Journal, 638, 20

Datta, A., Bowman, J. D., & Carilli, C. L. 2010, The Astrophysical Journal, 724, 526

de Oliveira-Costa, A., Tegmark, M., Gaensler, B. M., Jonas, J., Landecker, T. L., & Reich, P. 2008, Monthly Notices of the Royal Astronomical Society, 388, 247, (c) Journal compilation © 2008 RAS

Dillon, J., Liu, A., & Tegmark, M. 2013, Physical Review D, 87, 43005

Dillon, J. et al. 2014, Physical Review D, 89, 23002 Furlanetto, S. R., Oh, S. P., & Briggs, F. H. 2006, Physics

Reports, 433, 181, elsevier B.V.

Jacobs, D. et al. 2011, The Astrophysical Journal, 734, L34

Jacobs, D. C. et al. 2013, The Astrophysical Journal, 776, 108
Lidz, A., Zahn, O., McQuinn, M., Zaldarriaga, M., & Hernquist,
L. 2008, The Astrophysical Journal, 680, 962

Liu, A., Parsons, A. R., & Trott, C. M. 2014a, Phys. Rev. D, 90, 023018

—. 2014b, Phys. Rev. D, 90, 023019

Liu, A. & Tegmark, M. 2011, Physical Review D, 83, 103006
Liu, A., Tegmark, M., Morrison, S., Lutomirski, A., & Zaldarriaga, M. 2010, Monthly Notices of the Royal Astronomical Society, 408, 1029, (c) Journal compilation © 2010 RAS

Morales, M. F., Hazelton, B., Sullivan, I., & Beardsley, A. 2012, ApJ, 752, 137

Morales, M. F. & Hewitt, J. 2004, The Astrophysical Journal, 615, 7, (c) 2004: The American Astronomical Society

Morales, M. F. & Wyithe, J. S. B. 2010, Annual review of astronomy and astrophysics, 48, 127, oise

Paciga, G. et al. 2013, Monthly Notices of the Royal Astronomical Society, 433, 1427

Parsons, A. et al. 2010, The Astronomical Journal, 139, 1468 Parsons, A., Pober, J., McQuinn, M., Jacobs, D., & Aguirre, J. 2012a, ApJ, 753, 81

Parsons, A. R. & Backer, D. C. 2009, The Astronomical Journal, 138, 219

 $<sup>^{20}</sup>$  Note that this notch is not present in Figure 4 because the foregrounds have been left in for comparison with P14.

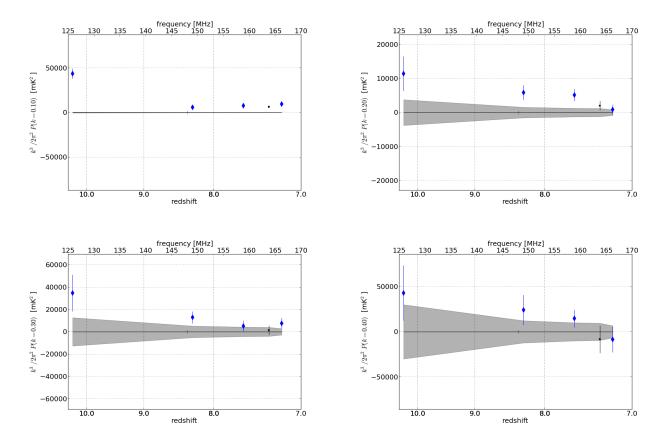


Figure 5. Power spectrum amplitude vs redshift at a selection of k modes. Left to right from top, k=0.1,0.2,0.3,0.4 hMpc<sup>-1</sup>. Parsons 2014 PAPER limit marked with thin black, this work marked with thick blue diamonds. The k=0.1 hMpc<sup>-1</sup> bin (top-left), which samples the delay spectrum at only 2x the maximum horizon delay is foreground dominated with a redshift dependence similar to the  $T_{rms}$  residual in Figure 2. Most have amplitudes at or slightly above the noise level (grey region), particularly at high redshifts suggesting that most are systematic limited. Calculating the theoretical noise requires making several approximations (see Figure 3 and Section 4.2).

Table 1 Power spectrum values. Most are inconsistent with noise at  $\sim\!2\sigma$ , indicating that this full season of integration sees systematics at about twice the level of the noise. In sum the points above k=0.2 are inconsistent with zero at 6.6  $\sigma$ . On average the theoretical and bootstrap estimates of the noise agree to within 15%.

k [hMpc <sup>-1</sup> ]	redshift	$\Delta^2$ [mK <sup>2</sup> ]	bootstrap error $[mK^2]$	noise error [mK <sup>2</sup> ]	model	bootstrap significance	noise model significance
0.1	10.29	43575.6	$\pm 5729.3$		$\pm 676.4$	Det	Det
0.1	8.54	5701.6	$\pm 1168.7$		$\pm 279.7$	Det	Det
0.1	7.94	7695.4	$\pm 2044.6$		$\pm 229.3$	Det	Det
0.1	7.55	9461.8	$\pm 2761.6$		$\pm 153.3$	Det	Det
0.2	10.29	11411.0	$\pm 5039.8$		$\pm 5326.1$	Det	Det
0.2	8.54	5855.9	$\pm 2195.8$		$\pm 2184.4$	Det	Det
0.2	7.94	5143.7	$\pm 1799.7$		$\pm 1785.6$	Det	Det
0.2	7.55	856.3	$\pm 1490.2$		$\pm 1190.0$	ULim	ULim
0.3	10.29	34756.5	$\pm 16384.8$	±	17922.7	Det	Det
0.3	8.54	12923.3	$\pm 5373.5$		$\pm 7339.6$	Det	Det
0.3	7.94	5204.4	$\pm 4834.5$		$\pm 5996.1$	Det	ULim
0.3	7.55	7628.5	$\pm 4810.3$		$\pm 3993.6$	Det	Det
0.4	10.29	42910.0	$\pm 30695.0$	$\pm$	42439.6	Det	Det
0.4	8.54	24115.2	$\pm 16585.6$	±	17370.2	Det	Det
0.4	7.94	14730.7	$\pm 9930.2$	$\pm$	14187.9	Det	Det
0.4	7.55	-8601.8	$\pm 14191.5$	:	$\pm 9447.4$	ULim	ULim

 $<sup>\</sup>overline{}^{a}$  All error bars are 1  $\sigma$ .

<sup>&</sup>lt;sup>b</sup> Det indicates a measurement and error-bar inconsistent with zero, ULim indicates consistency with zero at 1  $\sigma$ .

Parsons, A. R. et al. 2014, ApJ, 788, 106

Parsons, A. R., Pober, J. C., Aguirre, J. E., Carilli, C. L., Jacobs, D. C., & Moore, D. F. 2012b, The Astrophysical Journal, 756, 165

Pober, J. C. et al. 2014, The Astrophysical Journal, 782, 66

—. 2013a, The Astrophysical Journal, 768, L36

—. 2013b, The Astronomical Journal, 145, 65

—. 2012, The Astronomical Journal, 143, 53

Pritchard, J. & Loeb, A. 2008, Physical Review D, 78, 103511 Pritchard, J. R. & Loeb, A. 2012, Reports on Progress in Physics, 75, 6901 Stefan, I. I. et al. 2013, Monthly Notices of the Royal Astronomical Society, 432, 1285

Sullivan, I. S. et al. 2012, The Astrophysical Journal, 759, 17 Thyagarajan, N. et al. 2013, The Astrophysical Journal, 776, 6 Tingay, S. et al. 2013, Publications of the Astronomical Society of

Australia, 30, 7

Trott, C., Wayth, R., & Tingay, S. 2012, The Astrophysical Journal, 757, 101

Vedantham, H., Shankar, N. U., & Subrahmanyan, R. 2012, The Astrophysical Journal, 745, 176

Yatawatta, S. et al. 2013, Astronomy & Astrophysics, 550, 136 Zheng, H. et al. 2014, eprint arXiv, 1405.5527