# EVN Primary Beam Model

### Jack Radcliffe

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## 1 Introduction

This is the documentation for the primary beam model for the EVN developed by Jack Radcliffe as part of the Hubble Deep Field-North (HDF-N) wide-field VLBI project. Due to the lack of accurate primary beam models for many EVN telescopes, the primary beam can only be approximated as a Gaussian which is then applied directly to the upmost CL table in AIPS. To run the primary beam correction, you must have the packages described in Section 2.

## 2 Requirements

- 1. Python version 2.7 (3+ is not supported) with packages:
  - numpy
  - astropy (version 1.3.3 is recommended, 2+ does not work at the moment, it is a known bug)
  - scipy
  - matplotlib
- 2. AIPS (31DEC16 or newer)
- 3. ParselTongue
- 4. AIPS

# 3 Theory

Each phase centre is primary beam corrected by following similar steps to Cao+2014. Due to the lack of accurate primary beam models for many EVN telescopes, the primary beam power response of each telescope were approximated by using a normalized, symmetric, 2D Gaussian of the form,

$$P(\theta, \phi) \approx \exp\left(\frac{(\theta - \theta_0)^2 + (\phi - \phi_0)^2}{2\sigma^2}\right),$$
 (1)

where  $P(\theta, \phi)$  is the relative power response.  $\theta$  and  $\phi$  are the respective azimuthal and polar angular distances from the antennas' pointing centers. The azimuthal and polar coordinates of the telescope's pointing centers are defined by  $\theta_0$  and  $\phi_0$  respectively. The standard deviation,  $\sigma$ , can be related to the FWHM of the primary beam,  $\theta_{1/2}$ , through the expression,

$$\sigma^2 = \frac{\theta_{1/2}^2}{8 \ln 2},\tag{2}$$

where the FWHM of the primary beam is defined as,

$$\theta_{1/2} = K \frac{\lambda_{\rm c}}{D_{\rm eff}}.\tag{3}$$

Here,  $\lambda_c$ , is the central wavelength of the observation and  $D_{\text{eff}}$  is the effective aperture diameter that takes into account any tapering of the telescope (Keimpema 2015, priv. comm.). The effective aperture diameters

Telescopes	Country	Diameter (Effective)
Ef	Germany	100 (78)
Wb	Netherlands	25
On	Sweden	25
Nt	Italy	32
$\operatorname{Tr}$	Poland	32
Sv	Russia	32
$\operatorname{Bd}$	Russia	32
Zc	Russia	32
$\operatorname{Sh}$	China	25 (22.5)
Jb1	United Kingdom	76 (67)

Table 1: Sub-sample of EVN telescopes. Effective diameters are derived from fitting to beam models (A. Keimpema priv. comm.)

in use are shown in Table ??. A small correction factor, K = 1.05, is used to take into account any aperture blockages (Wrigley et al., in prep.).  $P(\theta, \phi)$  was derived for each telescope at the centre of every phase centre. These correction factors are then applied using AIPS task CLCOR, where the gain amplitudes for each antenna were multiplied by the correction factors,  $P(\theta,\phi)^{-1/2}$ , of the two telescopes that form the baseline.

#### Algorithm 4

1. pbcor\_model.py will take the inputs specified in this file and and uses astropy.modeling 2D Gaussians to model the primary beam shape. The standard deviation is entered as:

$$\sigma_{\text{Dec}} = (1/\cos(\text{Dec})) \frac{\theta_{1/2}}{(2\sqrt{2\log(2)})}$$

$$\sigma_{\text{RA}} = \frac{\theta_{1/2}}{(2\sqrt{2\log(2)})}$$
(5)

$$\sigma_{\rm RA} = \frac{\theta_{1/2}}{(2\sqrt{2\log(2)})}\tag{5}$$

and the full block of Python code implementing the model is:

```
1
           def station_HPBW(station, frequency):
2
                   HPBW = ((constants.c/frequency)/station)*(180/np.pi)
                   return HPBW
3
4
           xstd= station_HPBW(stations[telescope], frequency)/(2*np.sqrt(2*np.
5
              log(2)))*u.degree*(1/np.cos((phase_centers[i].dec.radian)))
           ystd =station_HPBW(stations[telescope], frequency)/(2*np.sqrt(2*np.
6
              \log(2))*u.degree
           models.Gaussian2D(amplitude=1, \
7
           x_mean=phase_centers[i].ra.degree,\
8
           v_mean=phase_centers[i].dec.degree,\
9
10
           x_stddev=xstd,
11
           v_stddev=ystd, theta=0)
```

Examples of the Effelsberg (100m, effective 78m) and Medicina (32m) beams are shown in Figure ??

- 2. This code will then grid the model to the observational FoV (for either single or multiple pointings) and then use a user-defined number of FITS images to extract filenames and RA and Declinations.
- 3. These are evaluated per pointing and per telescope, where the square root is taken so that the value represents the \*\*VOLTAGE\*\* beam. This creates a pickle file which contains the corrections for each phase center and each telescope.

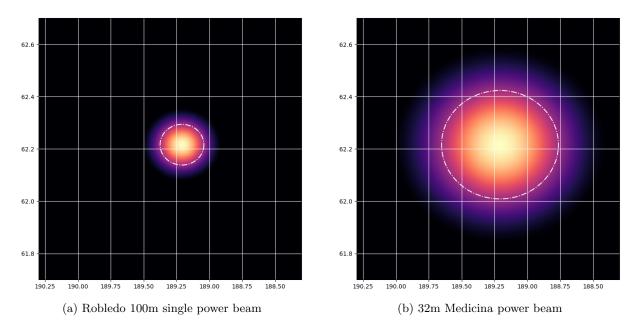


Figure 1

4. apply\_clcor\_Parseltongue.py is then used to apply this to all the calibrated data sets. The data set (with only one phase centre) is moved to the current directory and the filename is checked against the filename of the fitsfiles (not pretty I know). The correction values for each telescope and pointing centre are parsed. The correction is applied to each telescope using AIPS Task CLCOR with OPCODE='GAIN'. This is applied to the upmost CL table.

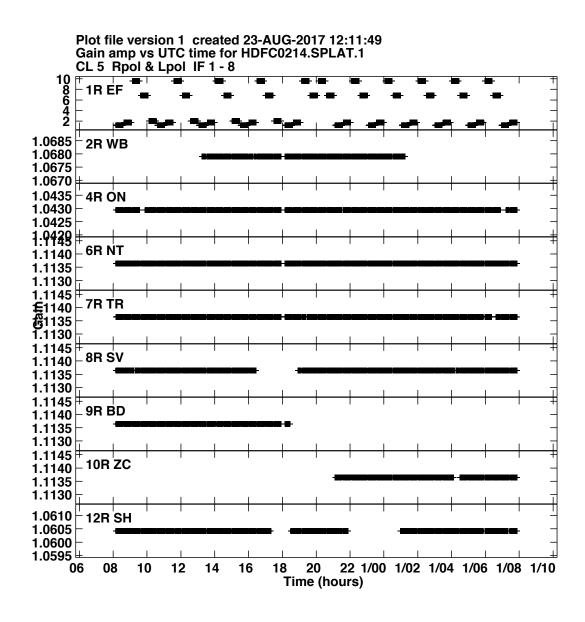


Figure 2: CL table for HDF-N source HDFC0214. The EF telescope nods between 5 different pointings where three are close to the position of the source and 2 are distant ~20arcmin. These furthest pointings are flagged in the observations subsequently as the primary beam model will diverge rapidly past the HPBW.