Wide-band Imaging



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Introduction

This talk is about combining data over a range of frequencies in order to create a single **continuum** image.



N DATA CHANNELS

This process is called **Multi-Frequency Synthesis** (MFS)

(Figure adapted from casadocs .readthedocs.io)



Introduction

For data over a range of frequencies that is used to create a spectral image cube, see the 'Spectral Lines' talks on Tuesday by Ylva Pihlstrom



(Figure adapted from casadocs .readthedocs.io)





Some fundamental terminology:

		VLA L-band	VLA X-band
Frequency range	$oldsymbol{ u}_{min},oldsymbol{ u}_{max}$	1 - 2 GHz	8 - 12 GHz
Bandwidth	$oldsymbol{ u}_{max}$ - $oldsymbol{ u}_{min}$	1 GHz	4 GHz
Bandwidth Ratio	$oldsymbol{ u}_{max}$: $oldsymbol{ u}_{min}$	2:1	1.5 : 1
Fractional Bandwidth	$\frac{\boldsymbol{\nu}_{\text{max}} - \boldsymbol{\nu}_{\text{min}}}{\boldsymbol{\nu}_{\text{o}}}$	66%	40%





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New technologies are enabling even higher (instantaneous) fractional bandwidths, e.g.

- ATCA's 4cm receiver and and ngVLA Band 2 are both 4-12 GHz
- the BIGCAT upgrade for ATCA will provide 8 GHz of correlated bandwidth, and the ngVLA reference design has 20 GHz
- → 100% fractional bandwidth!

However, data over your full frequency range doesn't need to be observed simultaneously- what fundamentally matters to wide-band MFS imaging is only what you are **combining** to make your single continuum image.

This means you can build up larger bandwidths by:

- \rightarrow observing with multiple frequency tunings
- \rightarrow combining data from multiple receiver bands





Why is wide-bandwidth important for continuum imaging?

- Sensitivity
$$\sigma_S = rac{2kT_{
m s}}{A_{
m e}[N\left(N-1
ight)\Delta
u\, au]^{1/2}}.$$

Increasing the bandwidth $\varDelta\nu$ will decrease an image's thermal noise $\sigma_{\scriptscriptstyle S}$

Example: EVLA expansion project, I 28 MHz \rightarrow 8 GHz

$$\sqrt{\frac{8 \text{ GHz}}{128 \text{ MHz}}} \approx 8 \quad \text{x better sensitivity}$$



Why is wide-bandwidth important for continuum imaging?

- UV coverage: the baseline length in wavelengths

$$\begin{bmatrix} u \\ v \\ w \end{bmatrix} = \frac{1}{\lambda} \begin{bmatrix} R(h, \theta) \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \end{bmatrix}$$



(Figure from casadocs.readthedocs.io)





Why is wide-bandwidth important for continuum imaging?

- Spectral information:

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Wide bandwidths provide information about a source's spectral variation, from which we can deduce properties of the emission mechanism(s)



- **Spectral variation** is the primary challenge of wide-band MFS
- Two types of spectral variation:
 - intrinsic: a property of the source itself
 - extrinsic: an instrumental effect of the telescope
- Together, these give the source an *apparent* spectral variation, and it is this net effect that is present in our visibilities

400% flux variation over 160% fractional bandwidth if we observed from 1-10 GHz → (Perley & Butler 2013)

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(Figure by J. Stevens)



• extrinsic: spectral index depends on distance from the beam centre



(Figure by J. Stevens)





Why is spectral variation a problem?

The standard CLEAN model has no frequency dependence!

During the major cycle, the *clean model* is subtracted from the measured visibilities; we call the result of this the *residual data*.

The model will match the data near the center frequency but deviate at the edges.

Flux remaining in the residual data will be convolved by the dirty beam, leading to artifacts in the image

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Why is spectral variation a problem? Convolutional artifacts in the dirty image are dependent on each source's spectral variation \rightarrow each source has a unique PSF

The PSF used by CLEAN will not perfectly subtract from these artifacts, leaving some convolutional pattern behind





We can average channels in either the UV domain or in the image domain The additive property of the FT tells us these two scenarios are equivalent







Why is spectral variation a problem? Convolutional artifacts in the dirty image are dependent on each source's spectral variation \rightarrow each source has a unique PSF

The wide-band PSF can be thought of as the average of each narrow-band PSF

Each source's **dirty beam** is the average of each narrow-band PSF **weighted by the source's spectrum**





Adopt a simple definition using the peak flux S_{peak} and the off-source image RMS σ :

$$\mathsf{DR} = \frac{S_{peak}}{\sigma}$$

 $S_{peak} = 0.6 Jy$ $\sigma = 2.5 uJy$ DR = 240,000

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Image artifacts can prevent you from reaching the theoretical noise limit:

$$DR = \frac{S_{peak}}{\sigma}$$
$$S_{peak} = 0.6 Jy$$
$$\sigma = 81 uJy$$

DR = 7,400

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S_{peak}

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Image artifact due to unmodeled spectral structure \rightarrow





When the dynamic range is lower, the noise is less likely to be dominated by artifacts

$$DR = \frac{S_{peak}}{\sigma}$$
$$S_{peak} = 0.6 Jy$$
$$\sigma = 1.4 mJy$$
$$DR = 430$$

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The same wide-band image artifact is still in this image \rightarrow



How do we prevent convolutional artifacts from limiting dynamic range?

- Understand their nature \rightarrow Tuesday's Error Recognition talk
- If it is a wide-band problem:
 - create a cube of narrow-band images
 - shallow deconvolution
 - loss of resolution
 - use a wide-band deconvolution algorithm

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Example residual image showing a wide-band artifact \rightarrow





Some different wide-band deconvolution algorithms include:

CASA:

- Multi-term Multi-frequency Synthesis (MT-MFS) Rau and Cornwell (2011)
- MTMFS via Cube (MVC) Bhatnagar, Rau and Golap (2013)

WSClean:

- Joined-channel Deconvolution Offringa and Smirnov (2017)

These algorithms are also compatible with *multi-scale* deconvolution





Multi-term Multi-frequency Synthesis (MT-MFS)

Models the spectrum of each clean component by a Taylor series

$$ec{I}^m_
u = \sum_{t=0}^{N_t-1} w^t_
u ec{I}^{sky}_t ext{ where } w^t_
u = \left(rac{
u-
u_0}{
u_0}
ight)^t$$

where I_t^{sky} is the Taylor coefficient image and N_t is the order of the Taylor series expansion, i.e. a polynomial in *linear flux* vs. *linear frequency*



Multi-term Multi-frequency Synthesis (MT-MFS)

The *major cycle* produces a set of N_t Taylor coefficient images and PSFs by gridding with the applied weighting factor w_v^t

The *minor cycle* proceeds iteratively, constructing a frequency-dependent model image and subtracting PSFs from the residual images.

The *restored* Taylor coefficient images can be related to a power law:

$$I_{\nu}^{sky} = I_{\nu_0}^{sky} \left(\frac{\nu}{\nu_0}\right)^{\alpha + \beta \log(\nu/\nu_0)} \qquad \qquad I_{\nu_0}^{m} = I_0^m \\ I_{\alpha}^m = I_1^m/I_0^m \\ I_{\beta}^m = \left[I_2^m/I_0^m\right] - \left[I_{\alpha}^m(I_{\alpha}^m - 1)/2\right]$$

The primary beam's effect on α and β will need to be estimated and removed





MT-MFS via Cube (MVC)

The wide bandwidth is partitioned into a small number (\sim 5-10) of frequency intervals

The *major cycle* produces one image per interval, resulting in a *coarse cube* of PSFs and images, using standard narrow-band techniques

Each channel of the image cube is corrected for the primary beam



MT-MFS via Cube (MVC)

The frequency weighting factor w_v^t is applied to each channel of the coarse cubes, then channels are averaged over to produce N_t Taylor coefficient images and PSFs.

The *minor cycle* and *restore* steps proceed in the same way as MT-MFS

No further correction is needed for the primary beam's effect on α and β



Joined-channel Deconvolution

The *major cycle* creates a coarse cube of PSFs and images, and also collapses the cube to form a single image at the reference frequency.

The *minor cycle* finds the brightest peak in the single image and does a model fit to the cube at that peak position. Polynomial and power-law models are supported.

Each channel is deconvolved with its own PSF which is scaled to the fitted model at that channel's frequency and multiplied by a gain factor γ . The new residual cube is then collapsed to form a new single image at the reference frequency.



Examples

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3C 286: 15 minutes of VLA L-Band data (1-2 GHz)



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Model Intensity Image



Model Spectral Index



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New Mexico Tech

Model spectral curvature



(Figure by U. Rao)

MT-MFS results for a 1-2 GHzVLA simulation (no noise)

multi-scale





point-source











20th Synthesis Imaging Workshop – Wide-band



Abell 2256 (Owen et al 2014) VLA L, S & C bands (1-2, 2-4, 4-8 GHz)



Intensity

Intensity weighted Spectral Index





Summary

- Wide-band imaging is a fundamental concern for high dynamic range continuum imaging
- Unmodeled spectral structure causes convolutional artifacts to persist in the final (cleaned) image
- Wide-band imaging and deconvolution algorithms account for each source's spectral variation, both intrinsic and apparent, in the model used for deconvolution





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