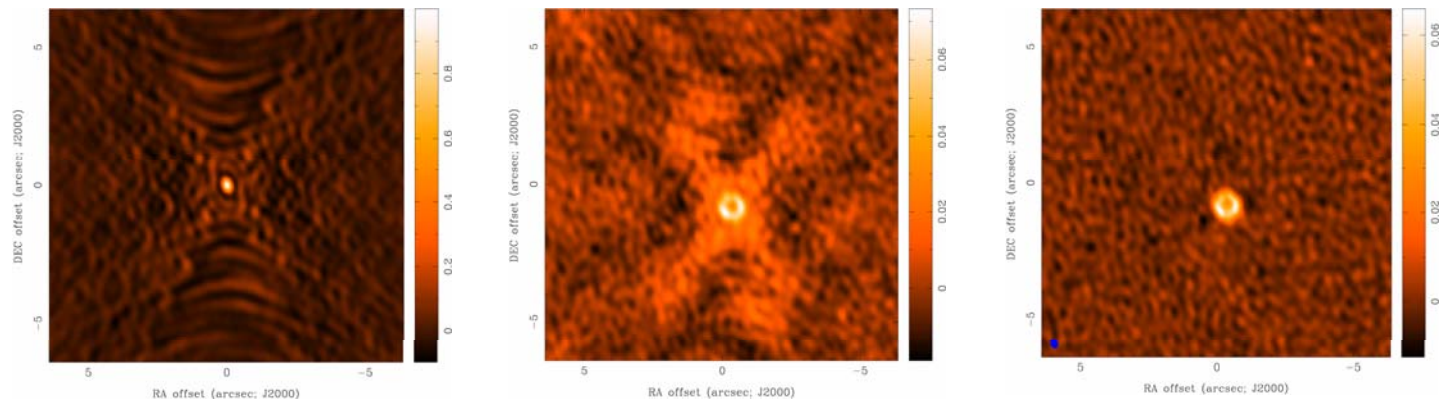


# Deconvolution

Amy Mioduszewski  
National Radio Astronomy Observatory

## Synthesis Imaging in Radio Astronomy (based on a talk given by David Wilner (CfA) at the NRAO's 2010 Synthesis Imaging Workshop)



# The Fourier Domain

- acquire comfort with the Fourier domain...
  - in older texts, functions and their Fourier transforms occupy *upper* and *lower* domains, as if “functions circulated at ground level and their transforms in the underworld” (Bracewell 1965)



# Visibility and Sky Brightness

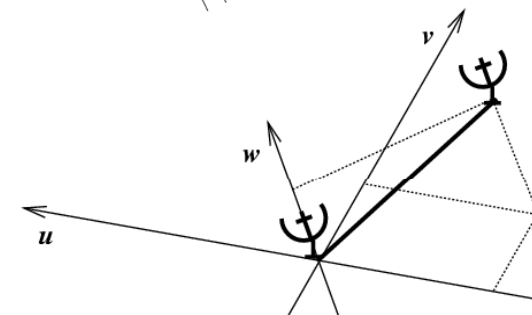
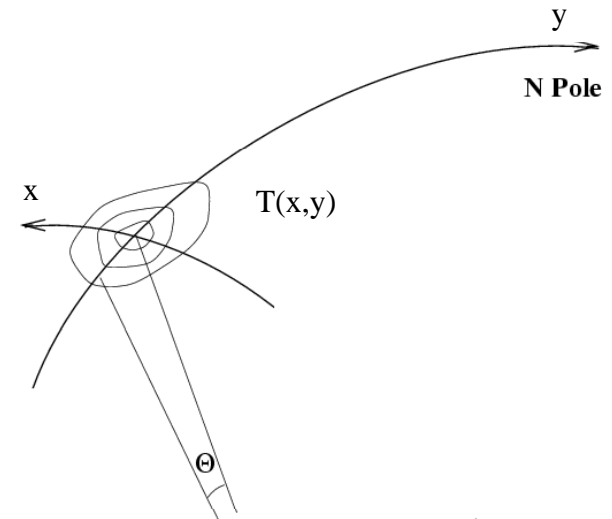
- from the van Cittert-Zernike theorem (TMS Chapter 14)

- for small fields of view:  
the complex visibility,  $V(u,v)$ ,  
is the 2D Fourier transform of  
the brightness on the sky,  $T(x,y)$

$$V(u, v) = \iint T(x, y) e^{2\pi i(ux+vy)} dx dy$$

$$T(x, y) = \iint V(u, v) e^{-2\pi i(ux+vy)} du dv$$

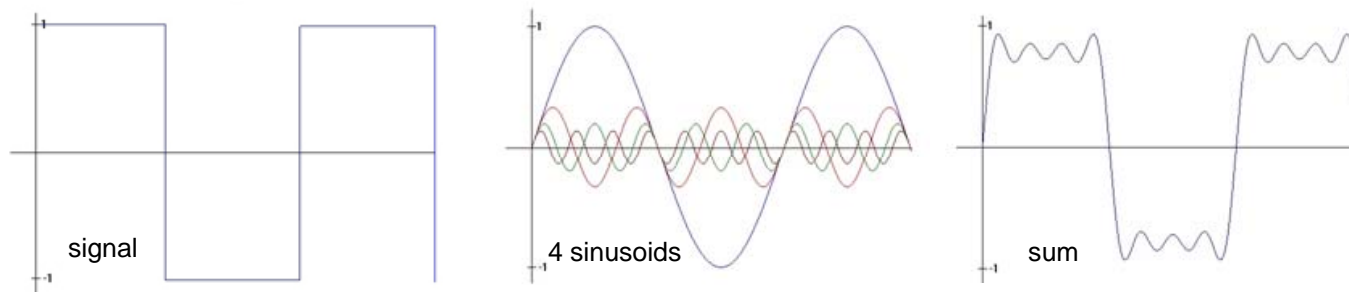
- $u, v$  (wavelengths) are spatial frequencies in  
E-W and N-S directions, i.e. the baseline lengths
- $x, y$  (rad) are angles in tangent plane relative to  
a reference position in the E-W and N-S directions



$$V(u, v) \rightleftharpoons T(x, y)$$

# The Fourier Transform

- Fourier theory states that any signal (including images) can be expressed as a sum of sinusoids

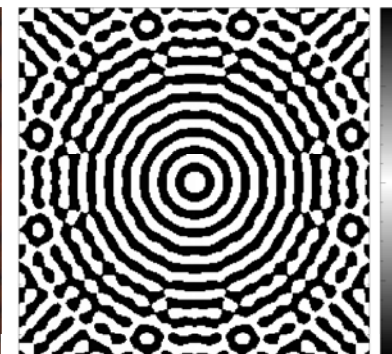
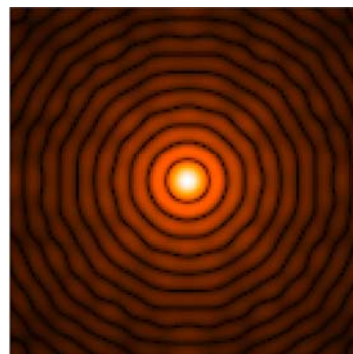


**Jean Baptiste  
Joseph Fourier**  
1768-1830

- (x,y) plane and (u,v) plane are conjugate coordinates

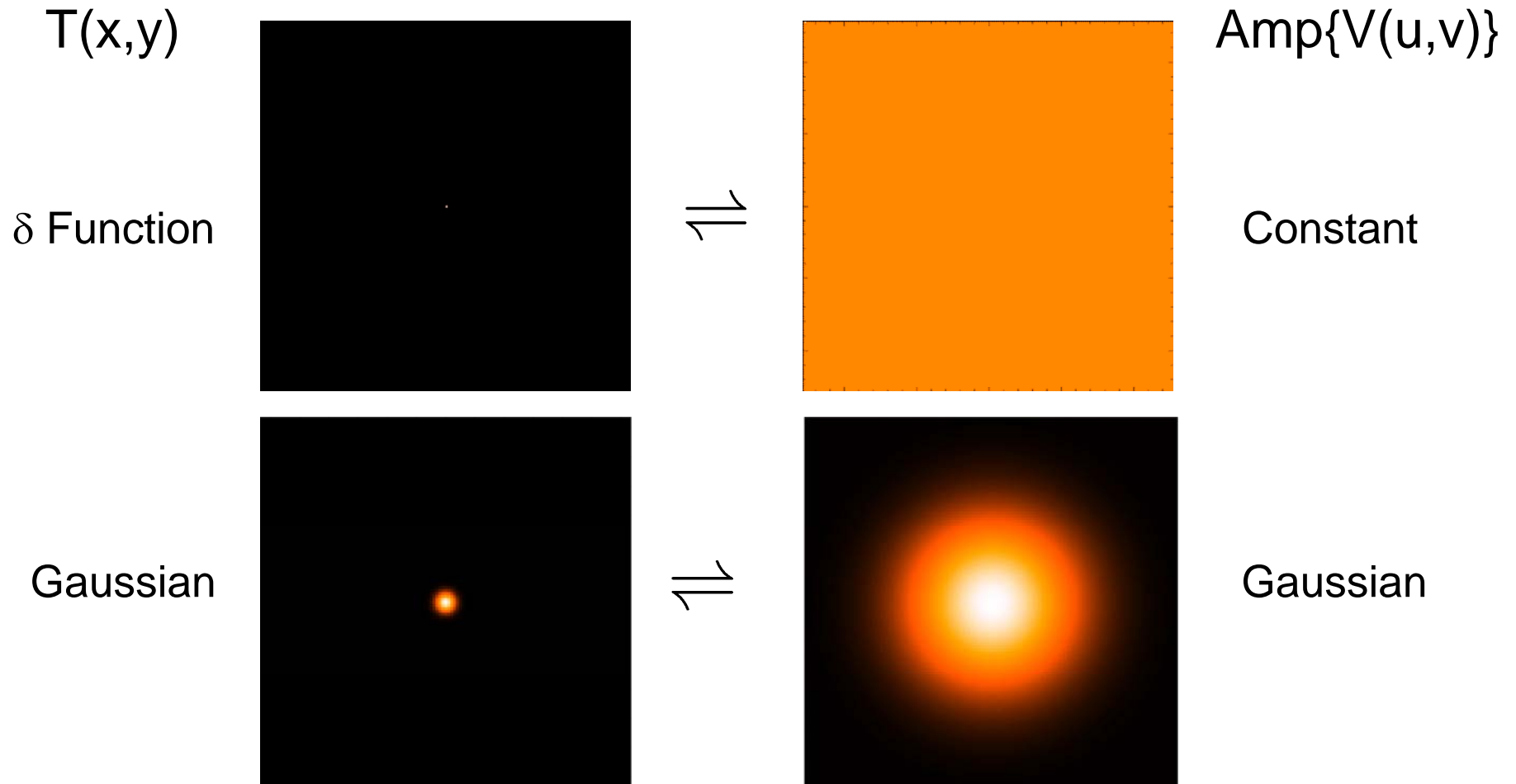
$T(x,y)$

$V(u,v) = \text{FT}\{T(x,y)\}$



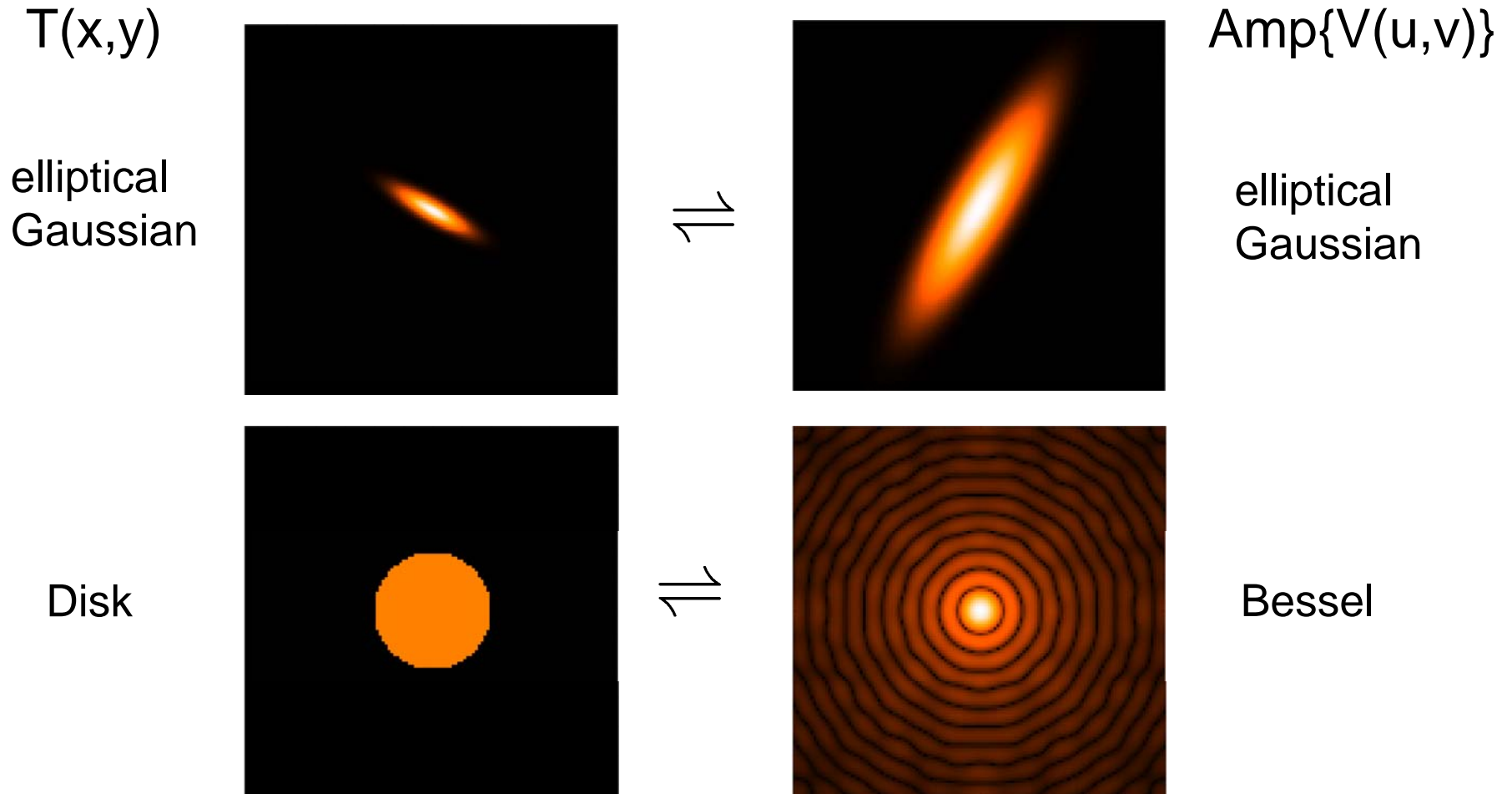
- the Fourier Transform contains all information of the original

# Some 2D Fourier Transform Pairs



narrow features transform to wide features (and vice-versa)

# More 2D Fourier Transform Pairs



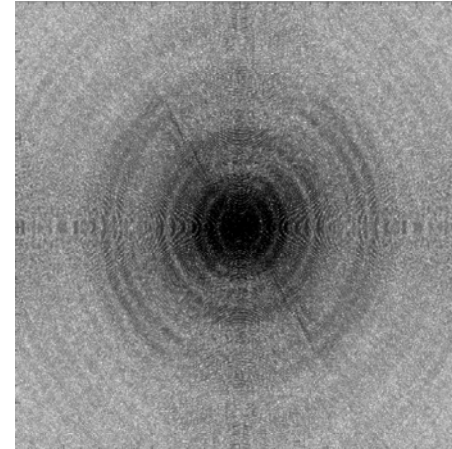
sharp edges result in many high spatial frequencies

# More 2D Fourier Transform Pairs

$T(x,y)$



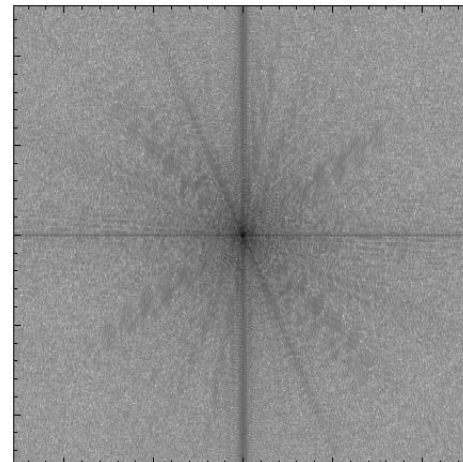
$\hat{=}$



$\text{Amp}\{V(u,v)\}$



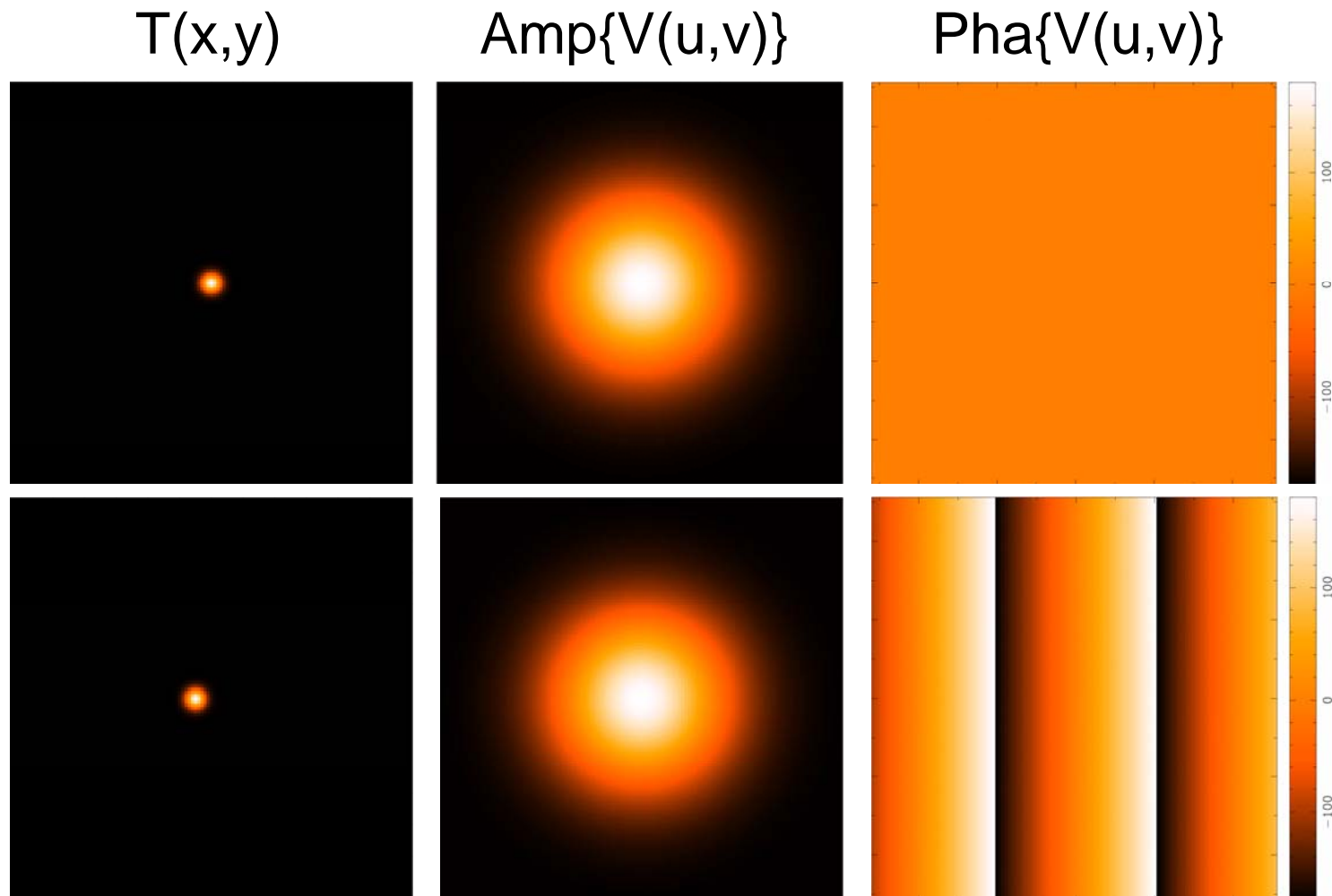
$\hat{=}$



complicated structure on many scales

# Amplitude and Phase

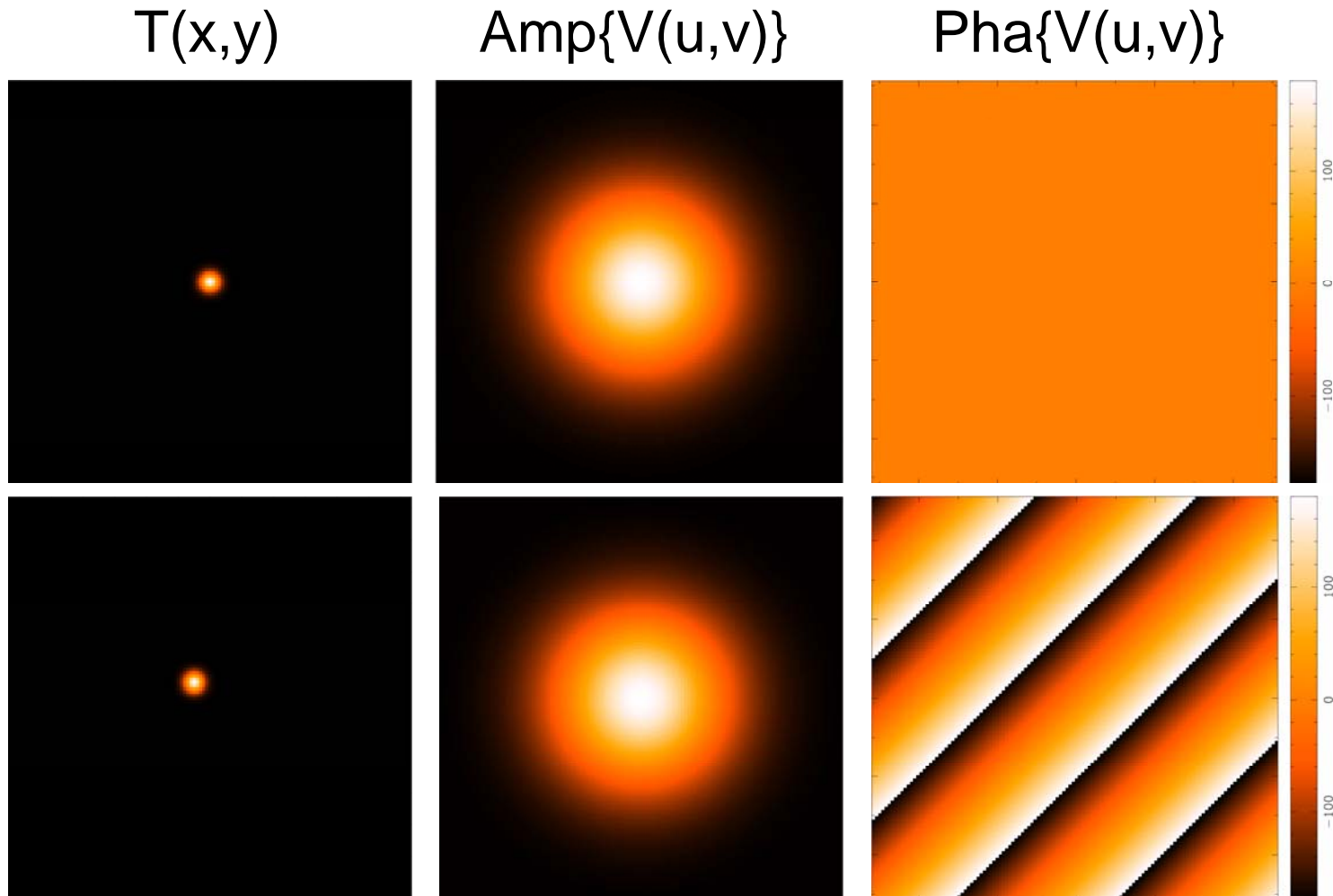
- complex numbers: (real, imaginary) or (amplitude, phase)
  - amplitude tells “how much” of a certain frequency component
  - phase tells “where” this component is located





# Amplitude and Phase

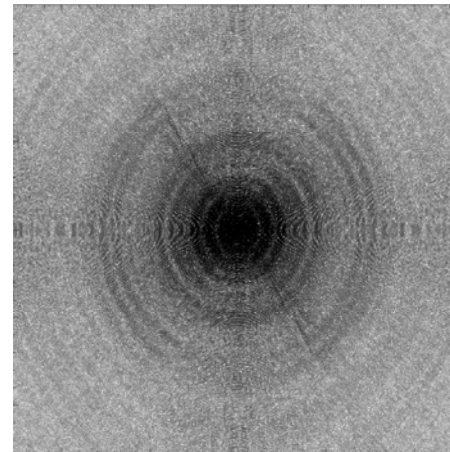
- complex numbers: (real, imaginary) or (amplitude, phase)
  - amplitude tells “how much” of a certain frequency component
  - phase tells “where” this component is located



# Two Visibilities for One Measurement

- $T(x,y)$  is real, but  $V(u,v)$  is complex (in general)
- $V(-u,-v) = V^*(u,v)$  where  $*$  is complex conjugation

$T(x,y)$

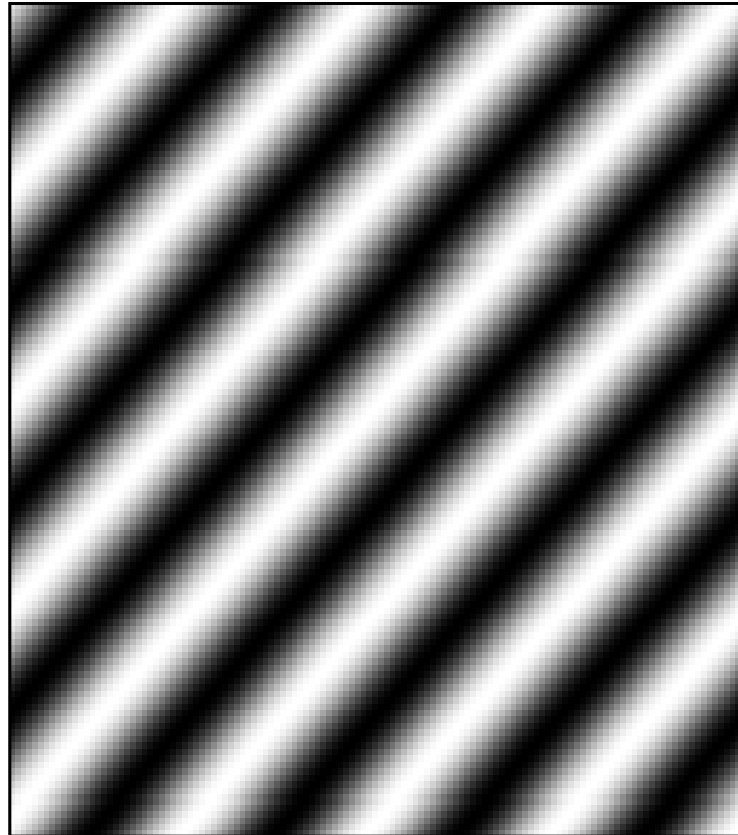


$\text{Amp}\{V(u,v)\}$

- $V(u=0,v=0) \rightarrow \text{integral of } T(x,y)dx dy = \text{total flux}$

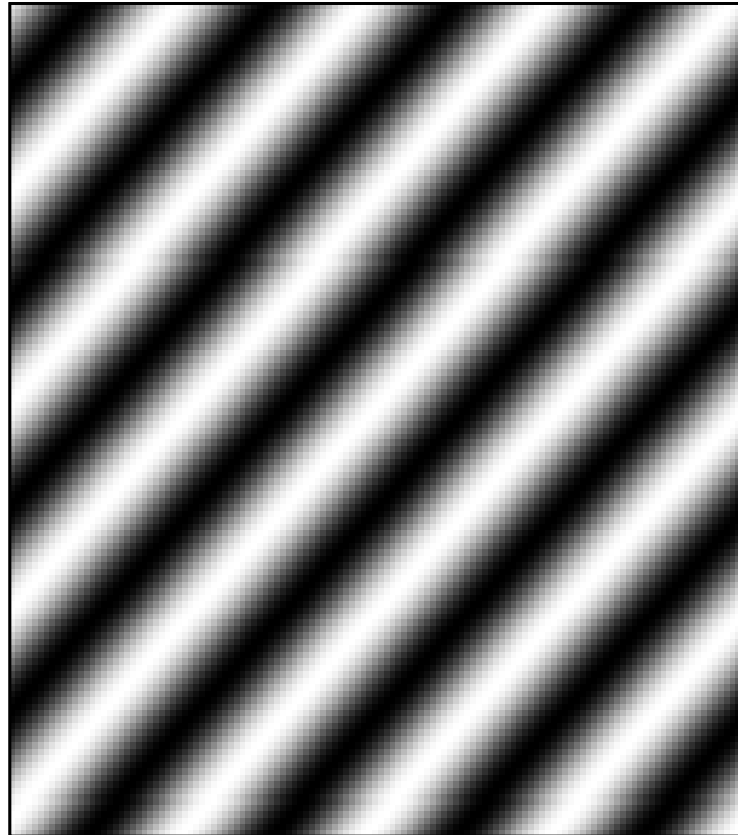
# Visibility and Sky Brightness

$$V(u, v) = \int \int T(x, y) e^{2\pi i(ux+vy)} dx dy$$



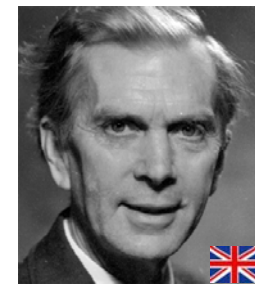
# Visibility and Sky Brightness

$$V(u, v) = \int \int T(x, y) e^{2\pi i(ux+vy)} dx dy$$

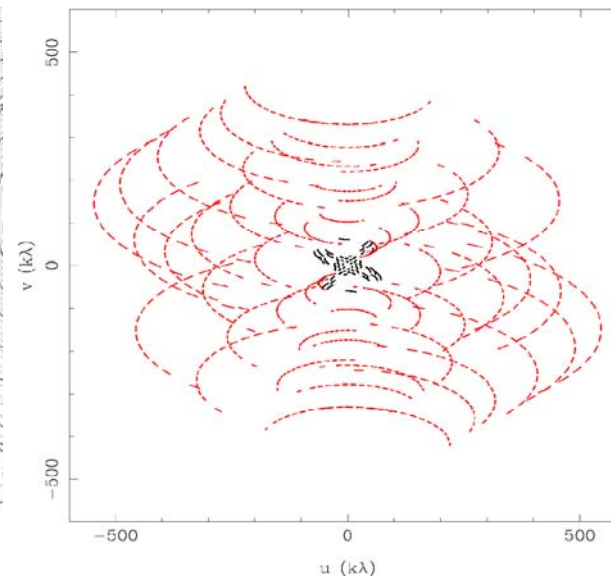
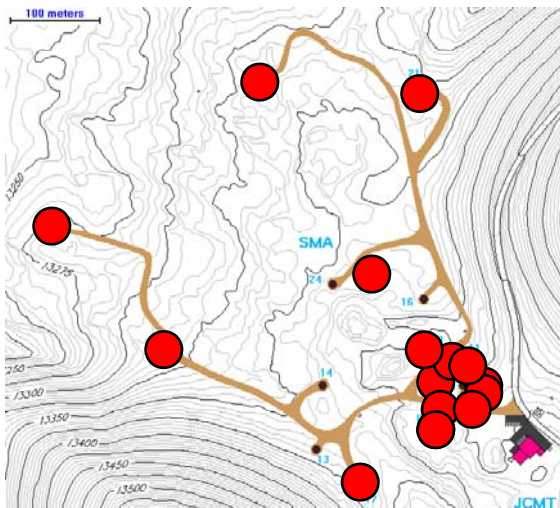


# Aperture Synthesis

- sample  $V(u,v)$  at enough points to synthesis the equivalent large aperture of size  $(u_{\max}, v_{\max})$ 
  - 1 pair of telescopes  $\rightarrow$  1  $(u,v)$  sample at a time
  - $N$  telescopes  $\rightarrow$  number of samples =  $N(N-1)/2$
  - fill in  $(u,v)$  plane by making use of Earth rotation:  
Sir Martin Ryle, 1974 Nobel Prize in Physics
  - reconfigure physical layout of  $N$  telescopes for more

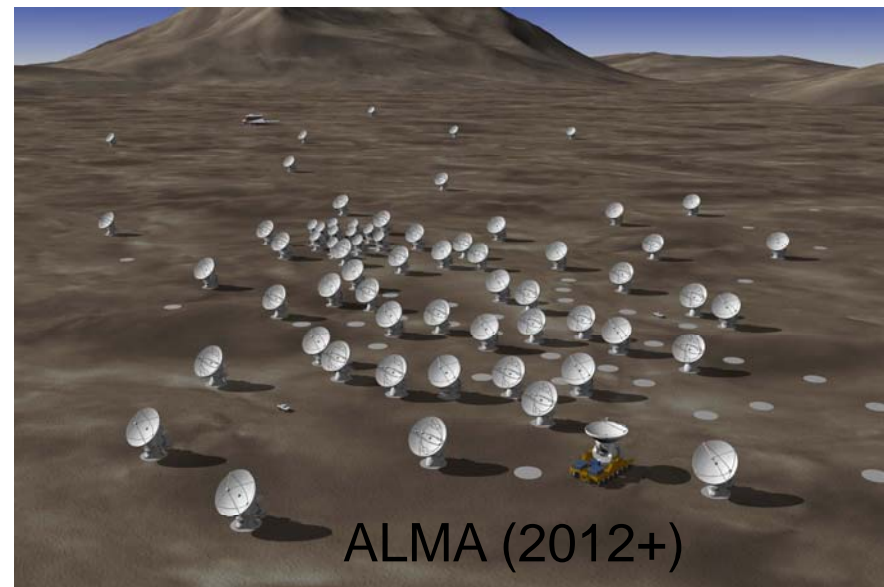


**Sir Martin Ryle**  
1918-1984



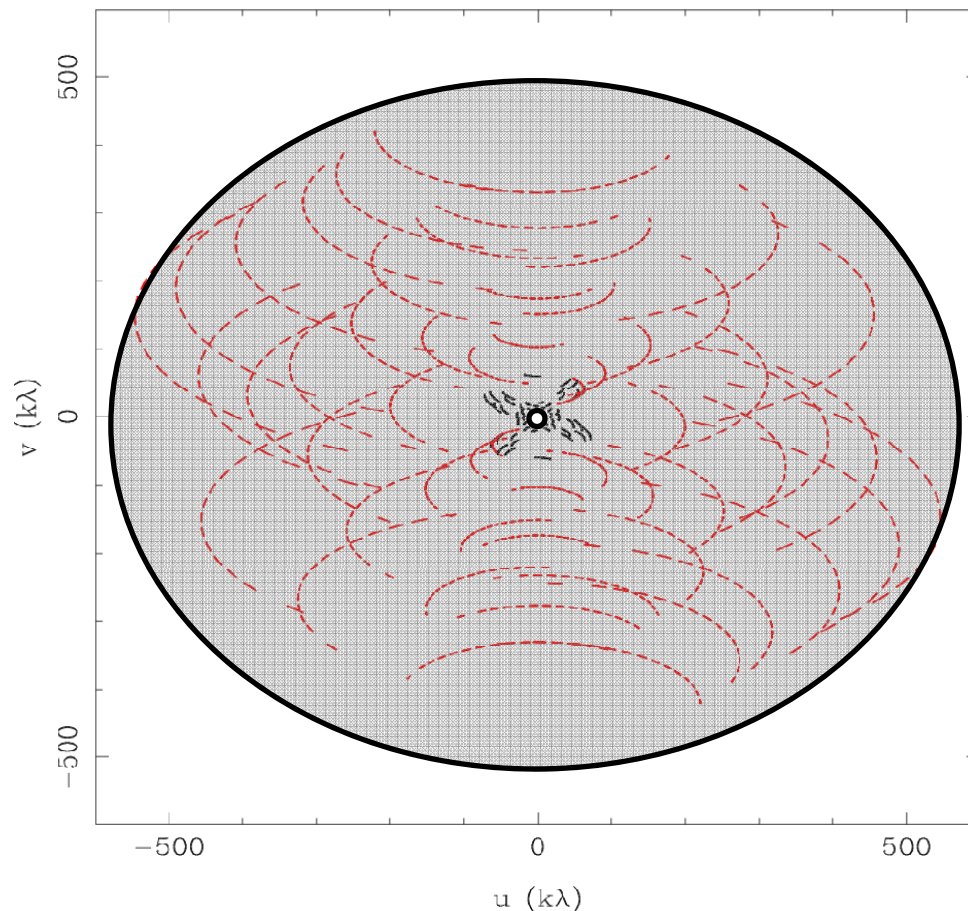
2 configurations  
of 8 SMA antennas  
345 GHz  
Dec = -24 deg

# Examples of Millimeter Aperture Synthesis Telescopes



# Imaging: (u,v) plane Sampling

- in aperture synthesis,  $V(u,v)$  samples are limited by number of telescopes, and Earth-sky geometry



- high spatial frequencies
  - maximum angular resolution
- low spatial frequencies
  - extended structures invisible
- irregular within high/low limits
  - sampling theorem violated
  - information missing



# Formal Description

- sample Fourier domain at discrete points

$$B(u, v) = \sum_k (u_k, v_k)$$

- the inverse Fourier transform is

$$T^D(x, y) = FT^{-1}\{B(u, v) \times V(u, v)\}$$

- the convolution theorem tells us

$$T^D(x, y) = b(x, y) \otimes T(x, y)$$

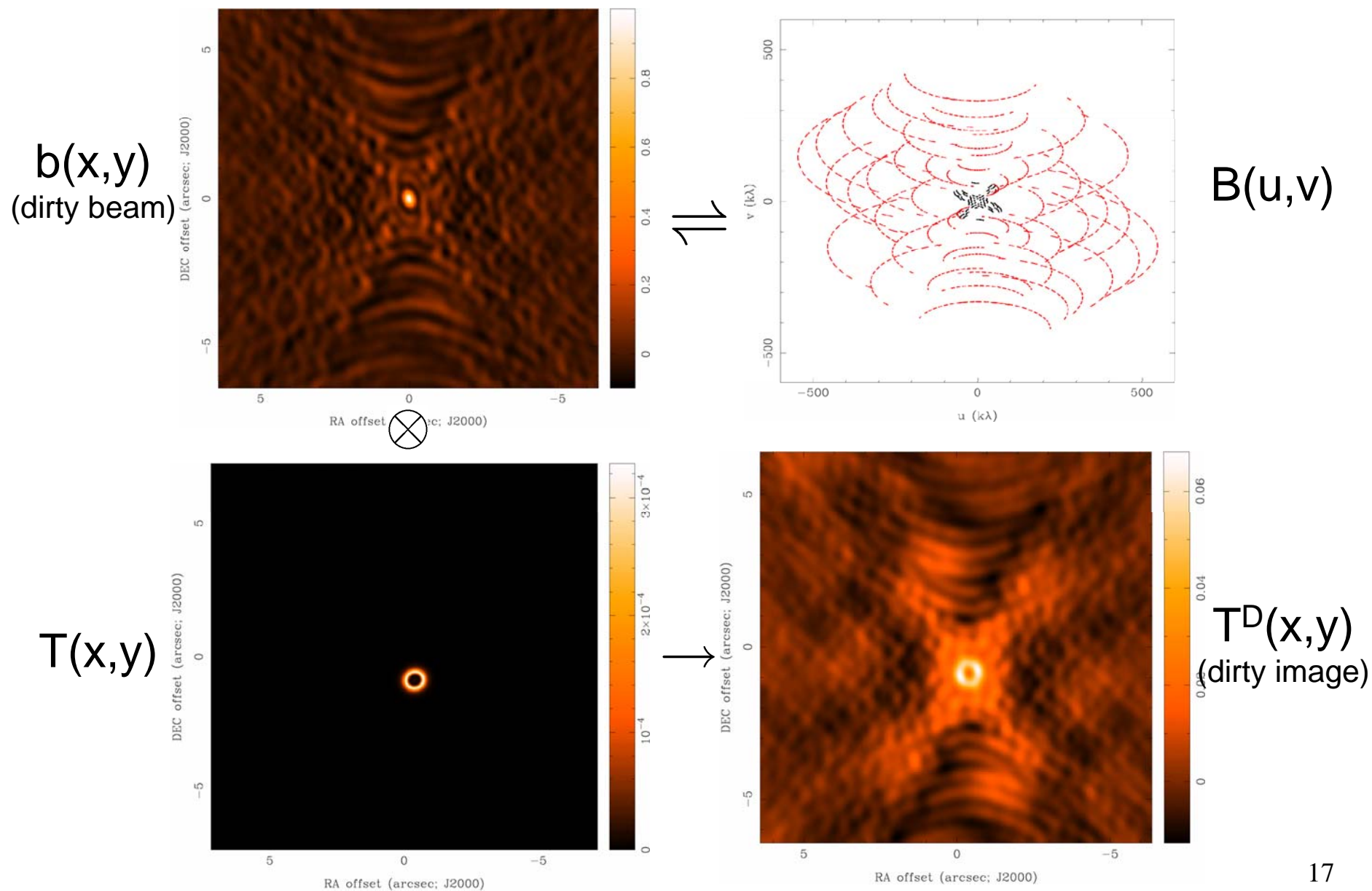
where  $b(x, y) = FT^{-1}\{B(u, v)\}$  (the point spread function)

Fourier transform of sampled visibilities yields the true sky  
brightness convolved with the point spread function

(the “dirty image” is the true image convolved with the “dirty beam”)

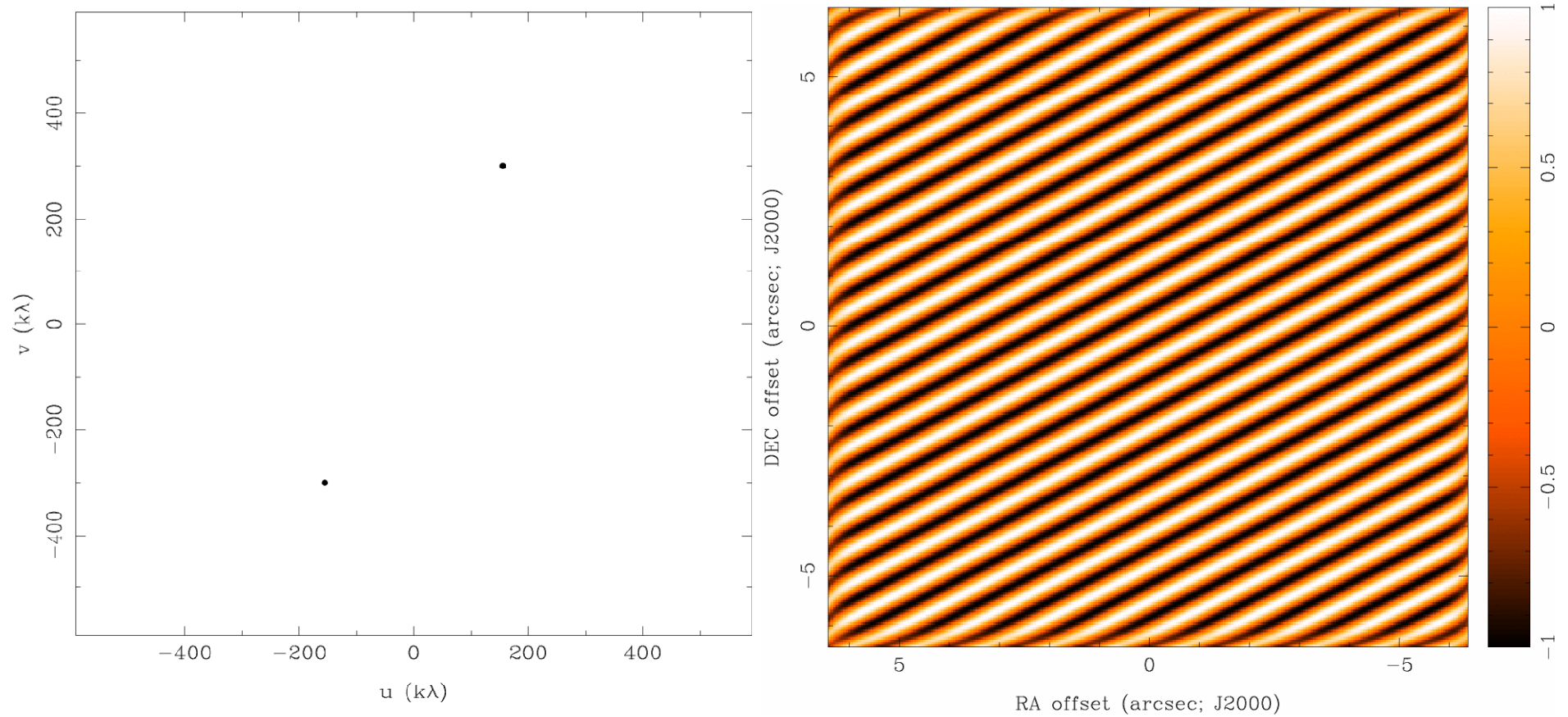


# Dirty Beam and Dirty Image



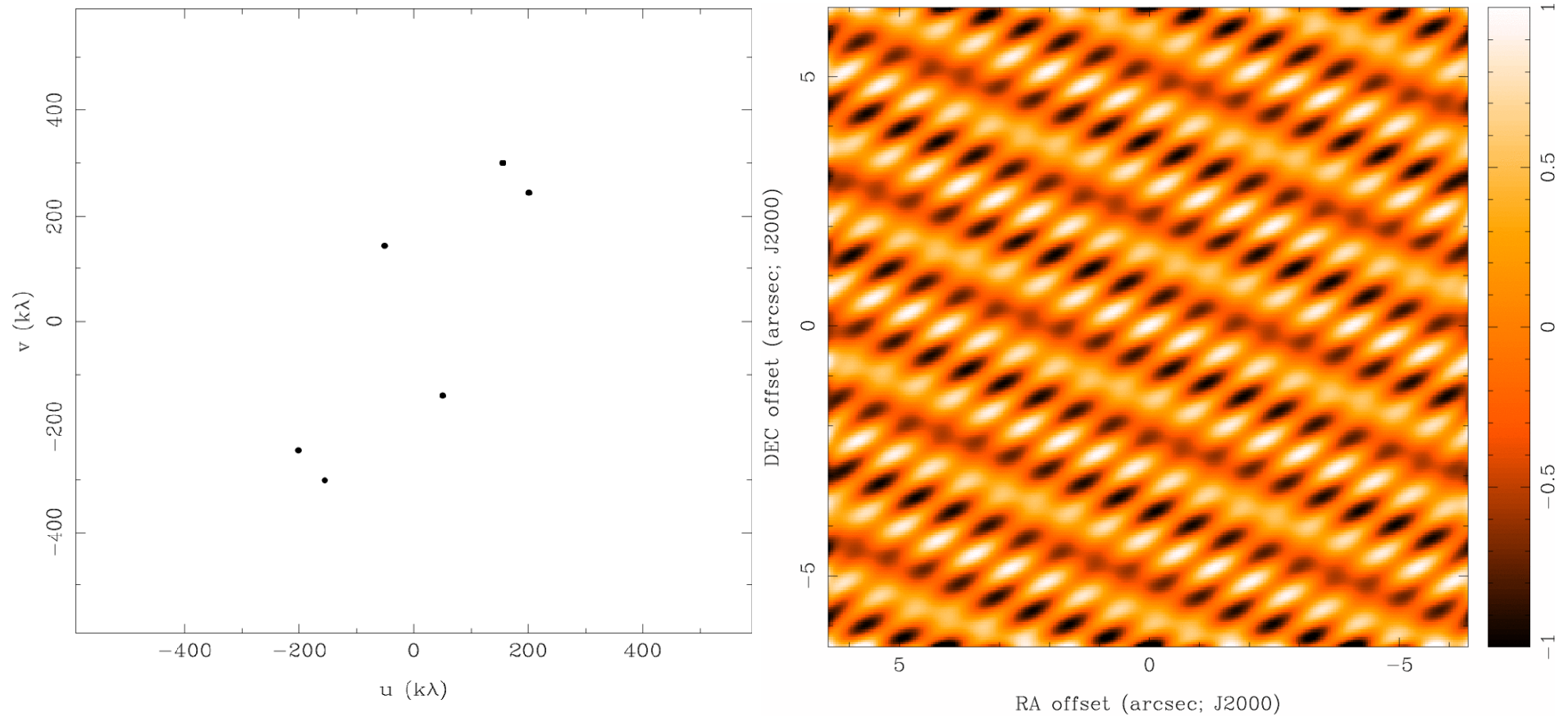
# Dirty Beam Shape and N Antennas

## 2 Antennas



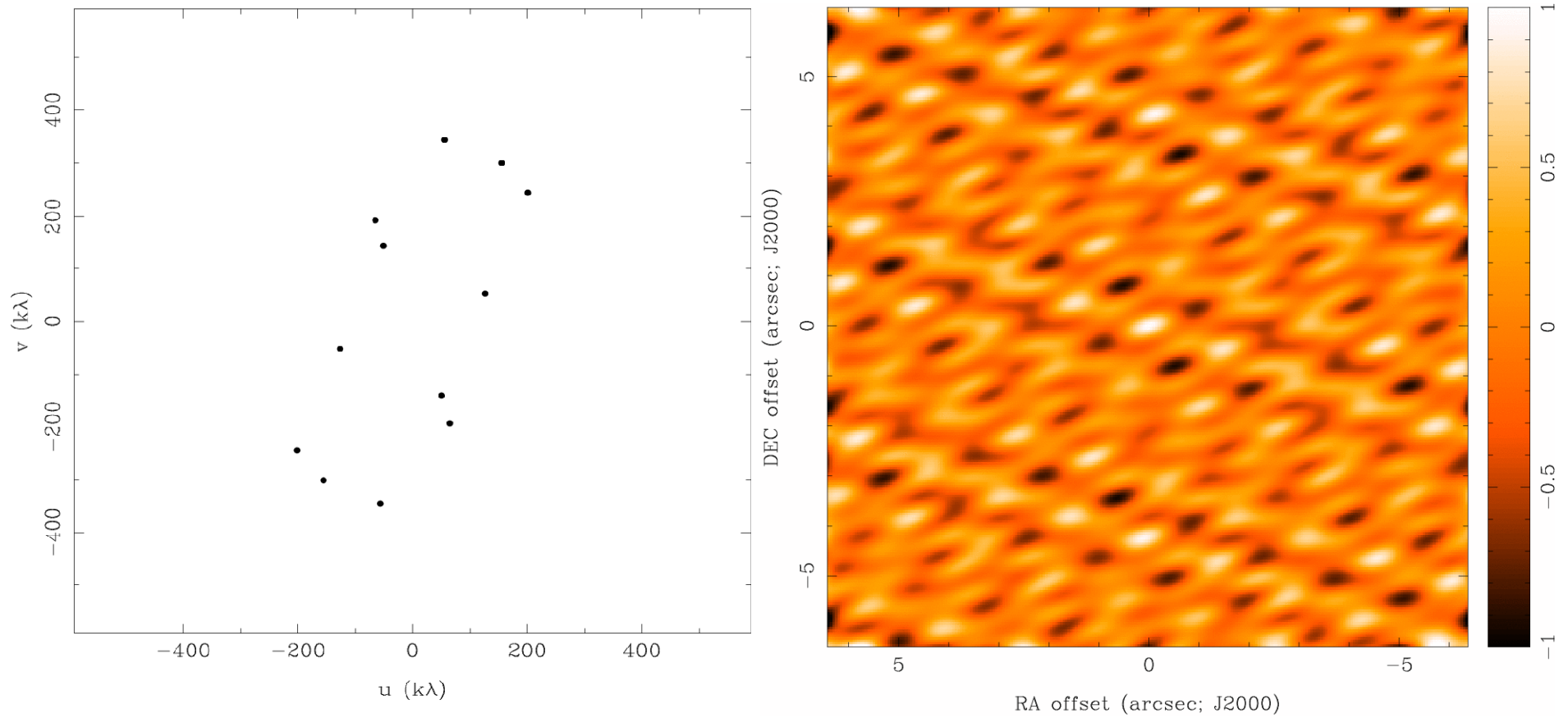
# Dirty Beam Shape and N Antennas

## 3 Antennas



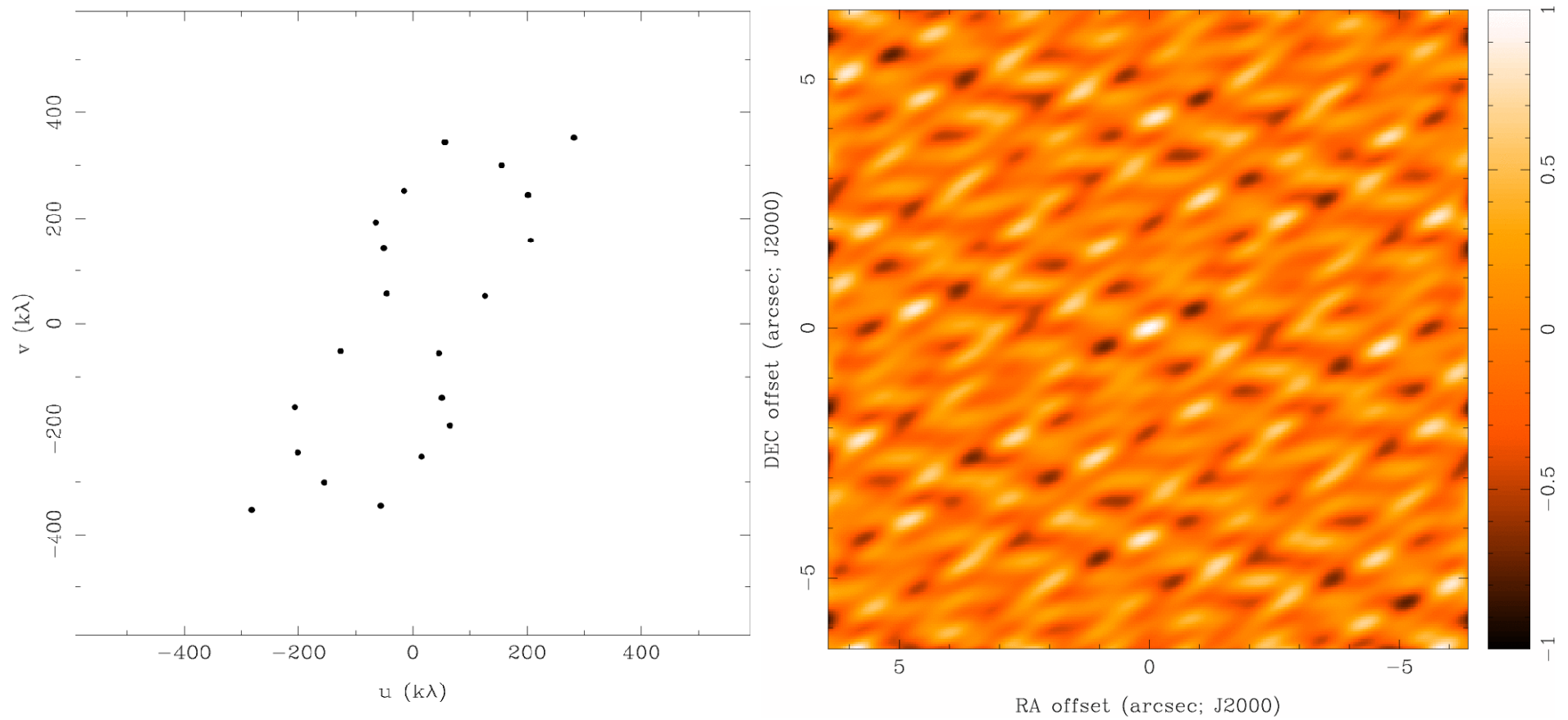
# Dirty Beam Shape and N Antennas

## 4 Antennas



# Dirty Beam Shape and N Antennas

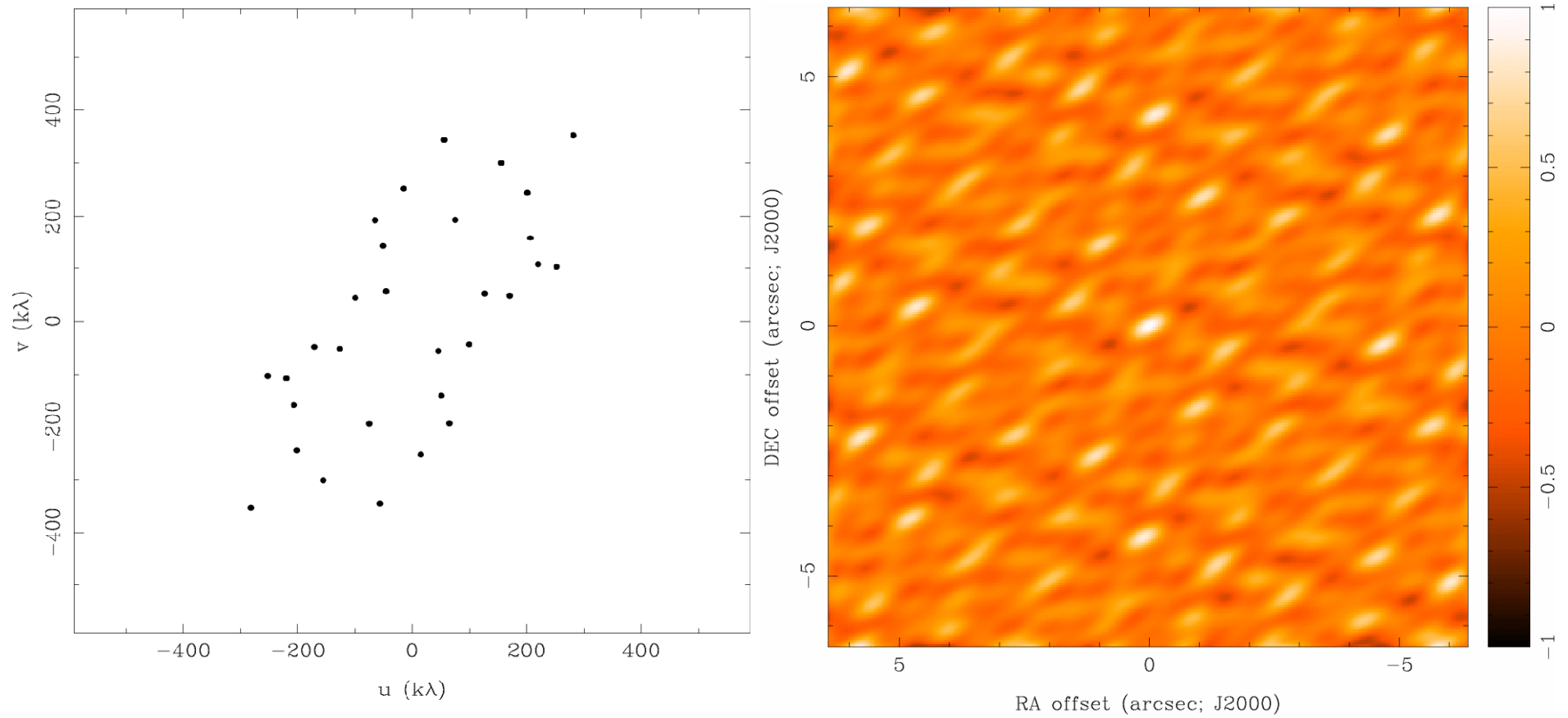
## 5 Antennas





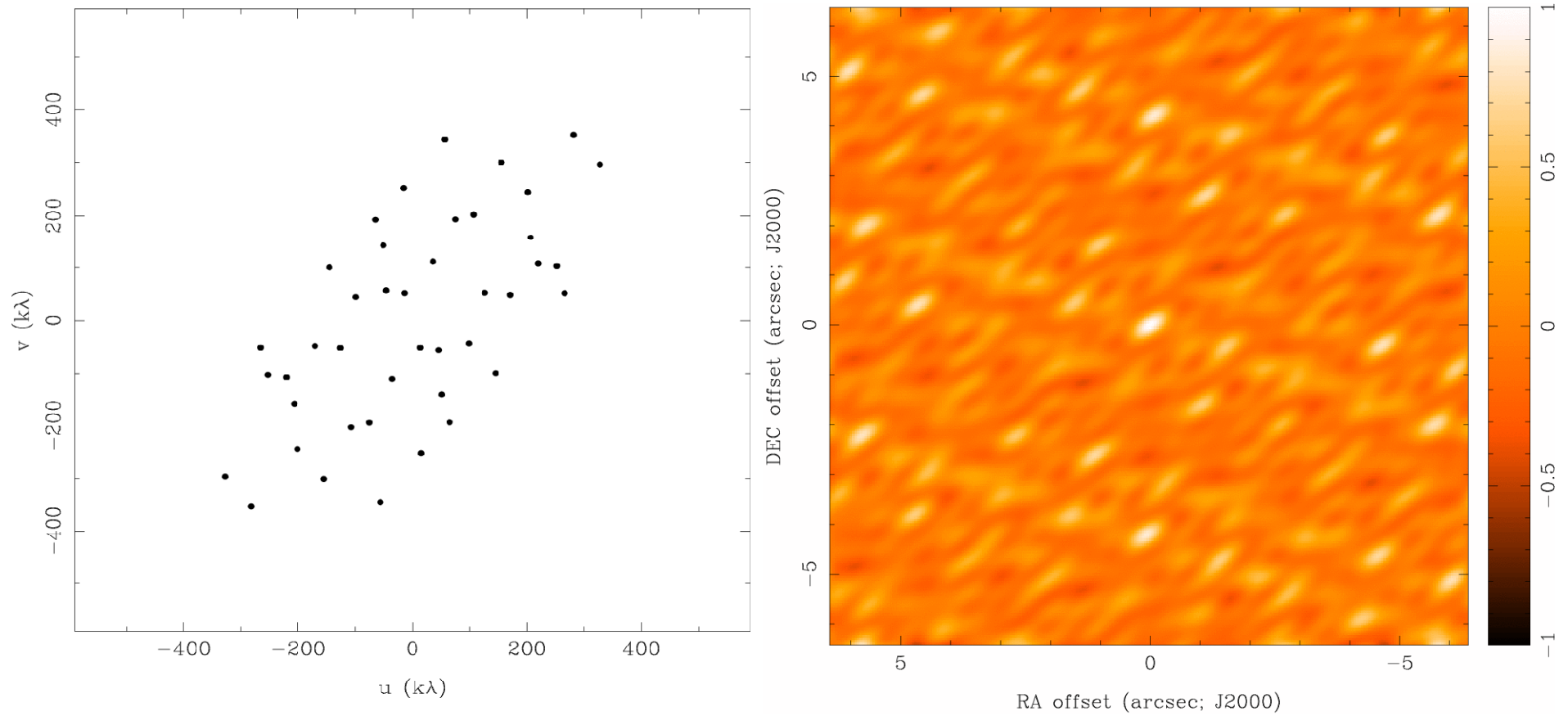
# Dirty Beam Shape and N Antennas

## 6 Antennas



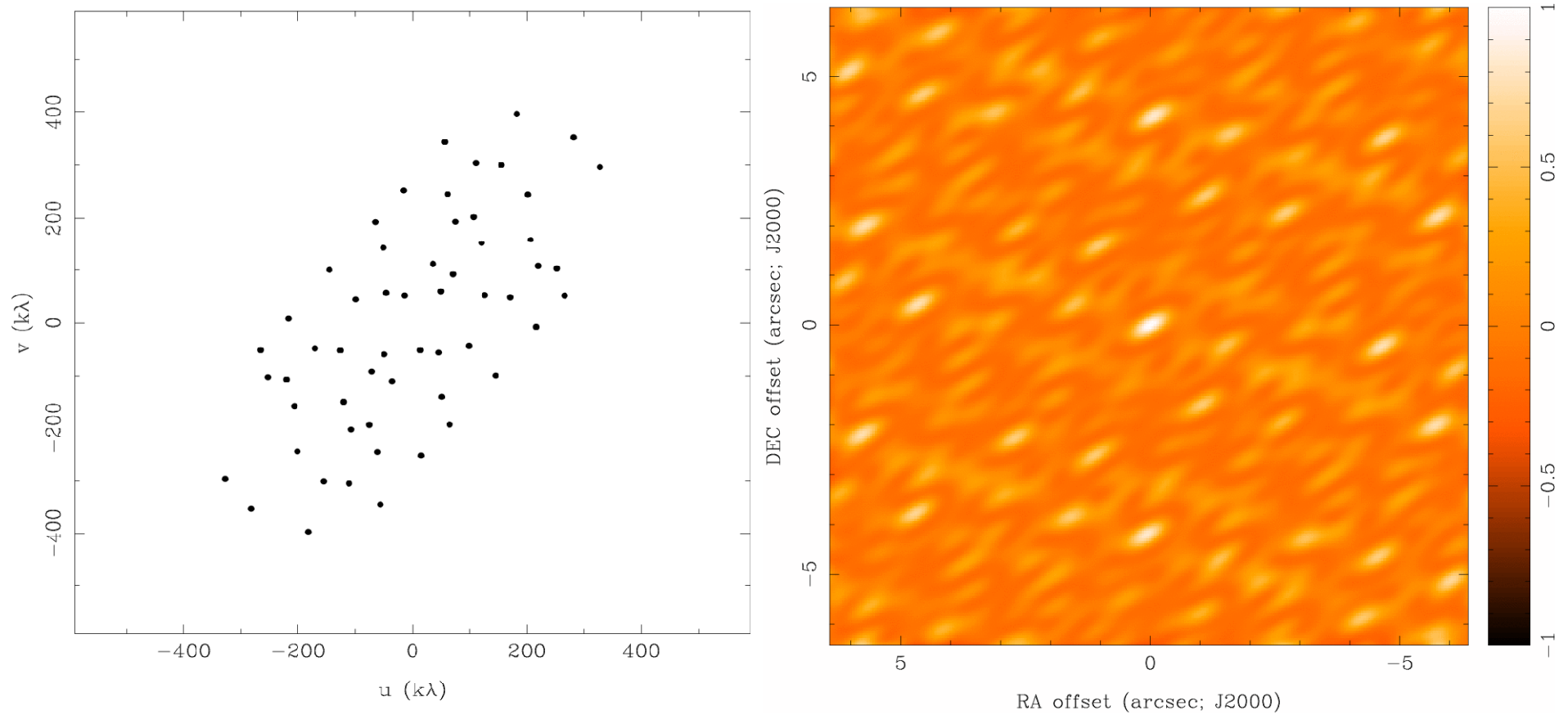
# Dirty Beam Shape and N Antennas

## 7 Antennas



# Dirty Beam Shape and N Antennas

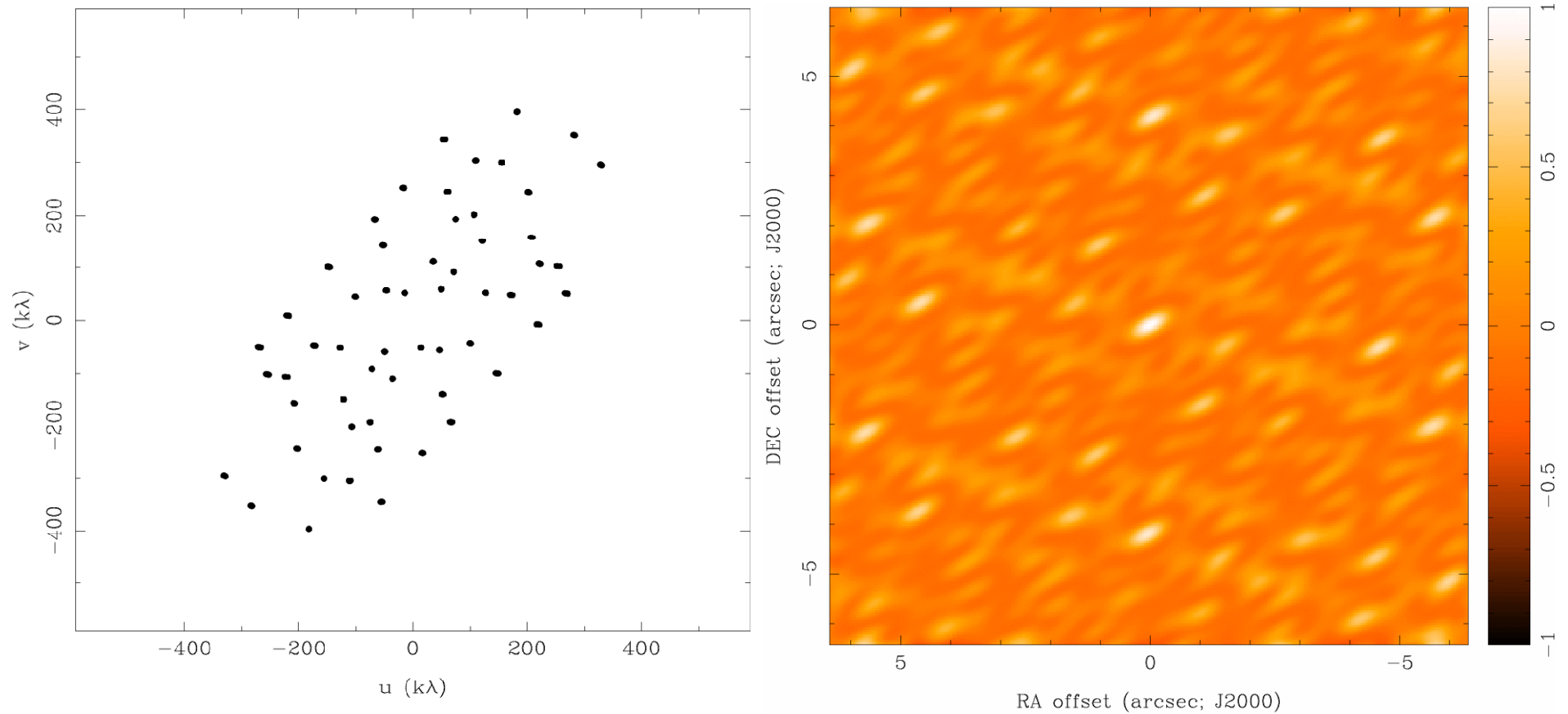
## 8 Antennas





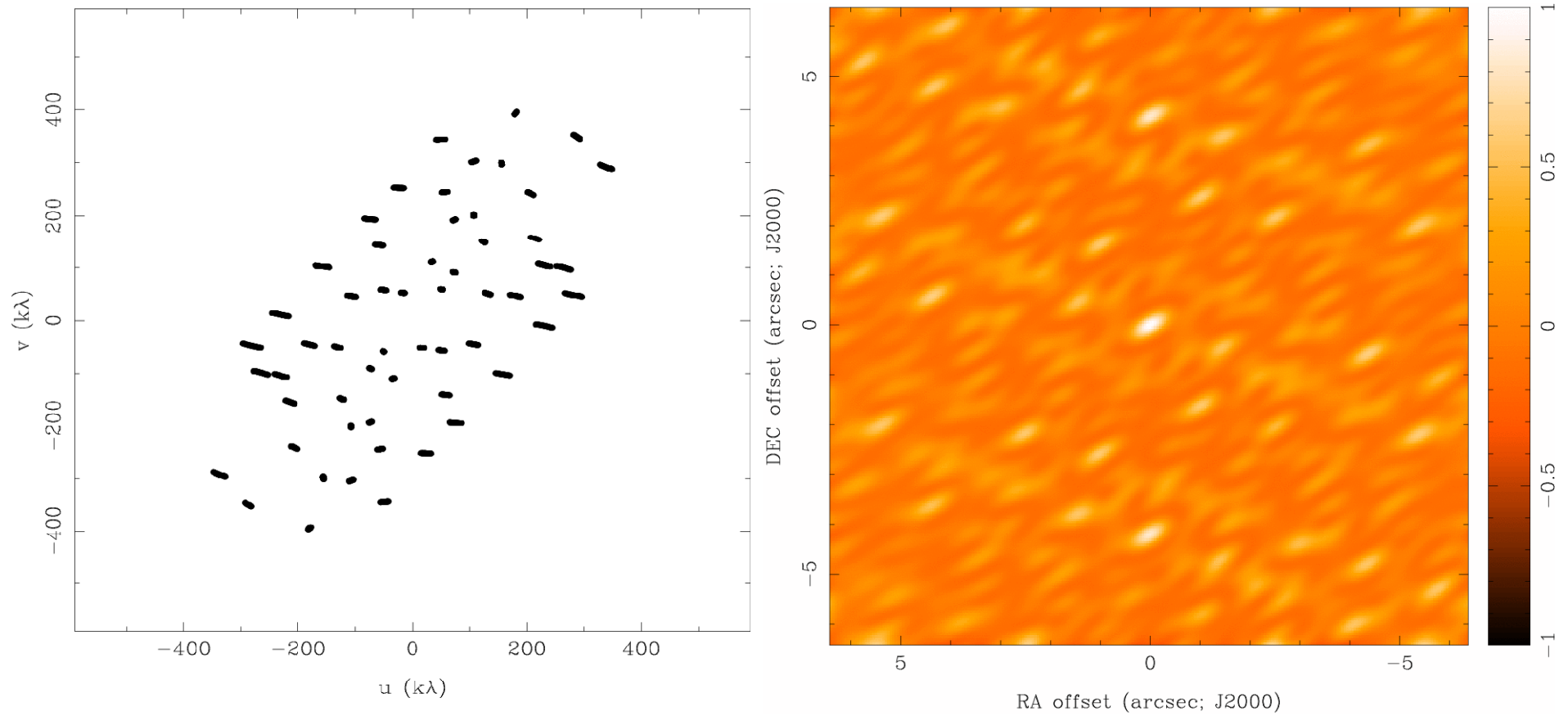
# Dirty Beam Shape and N Antennas

8 Antennas x 6 Samples



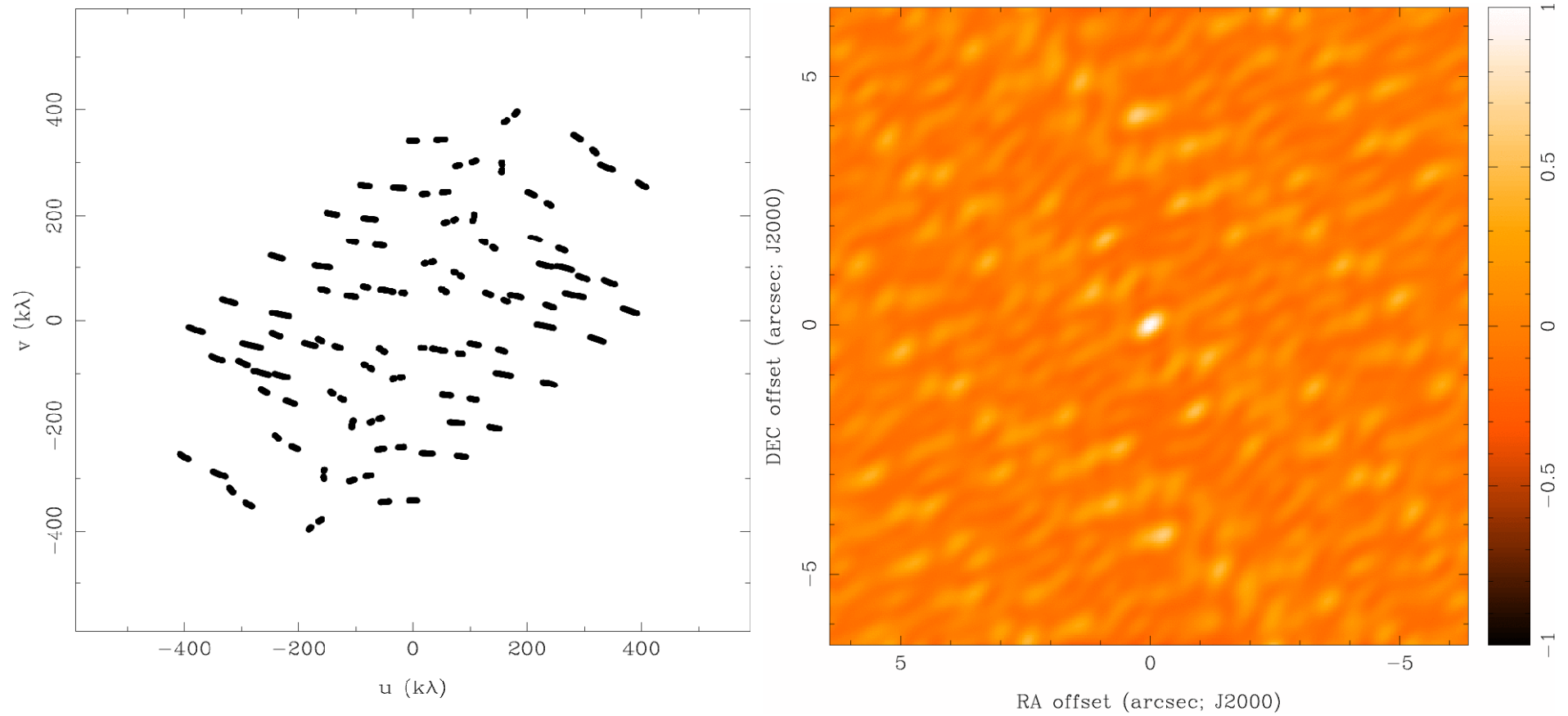
# Dirty Beam Shape and N Antennas

8 Antennas x 30 Samples



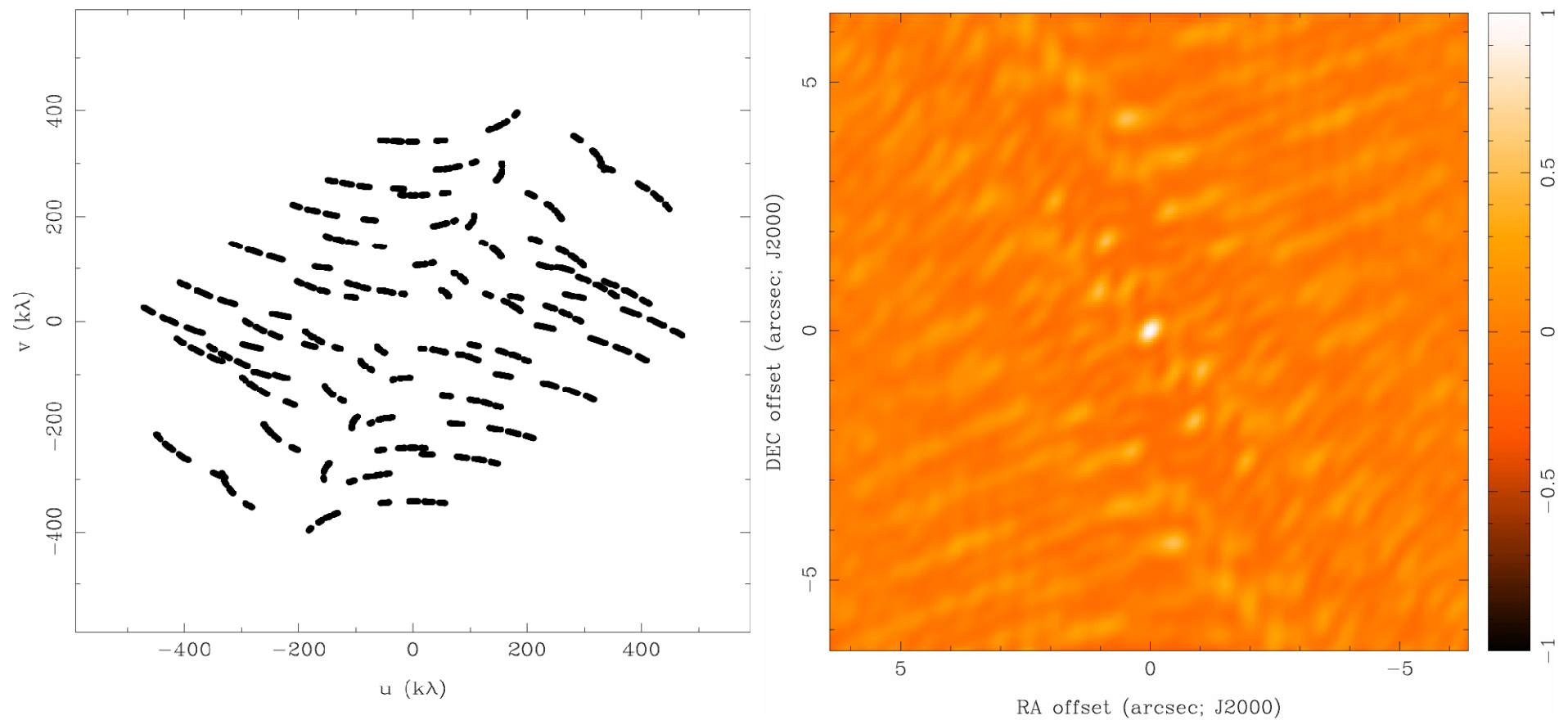
# Dirty Beam Shape and N Antennas

8 Antennas x 60 Samples



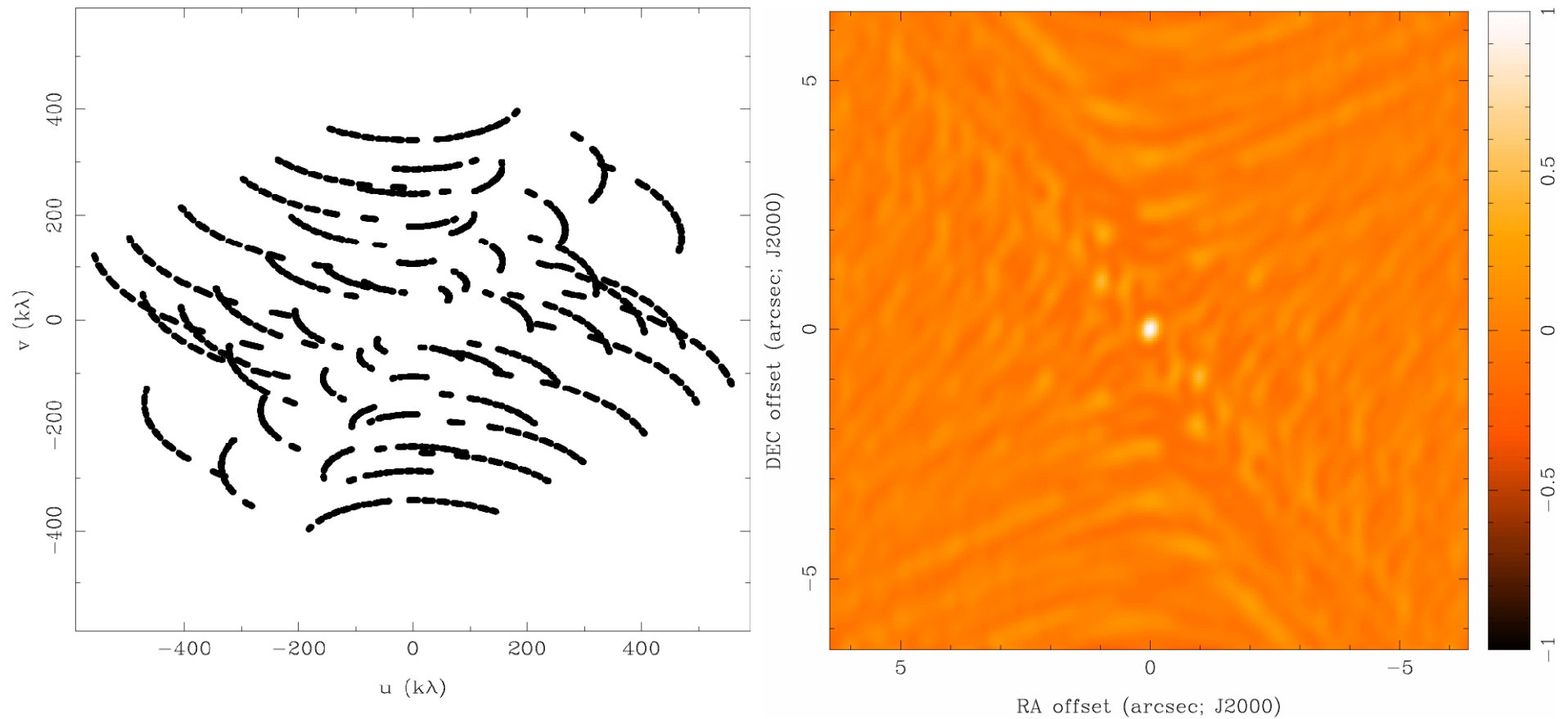
# Dirty Beam Shape and N Antennas

8 Antennas x 120 Samples



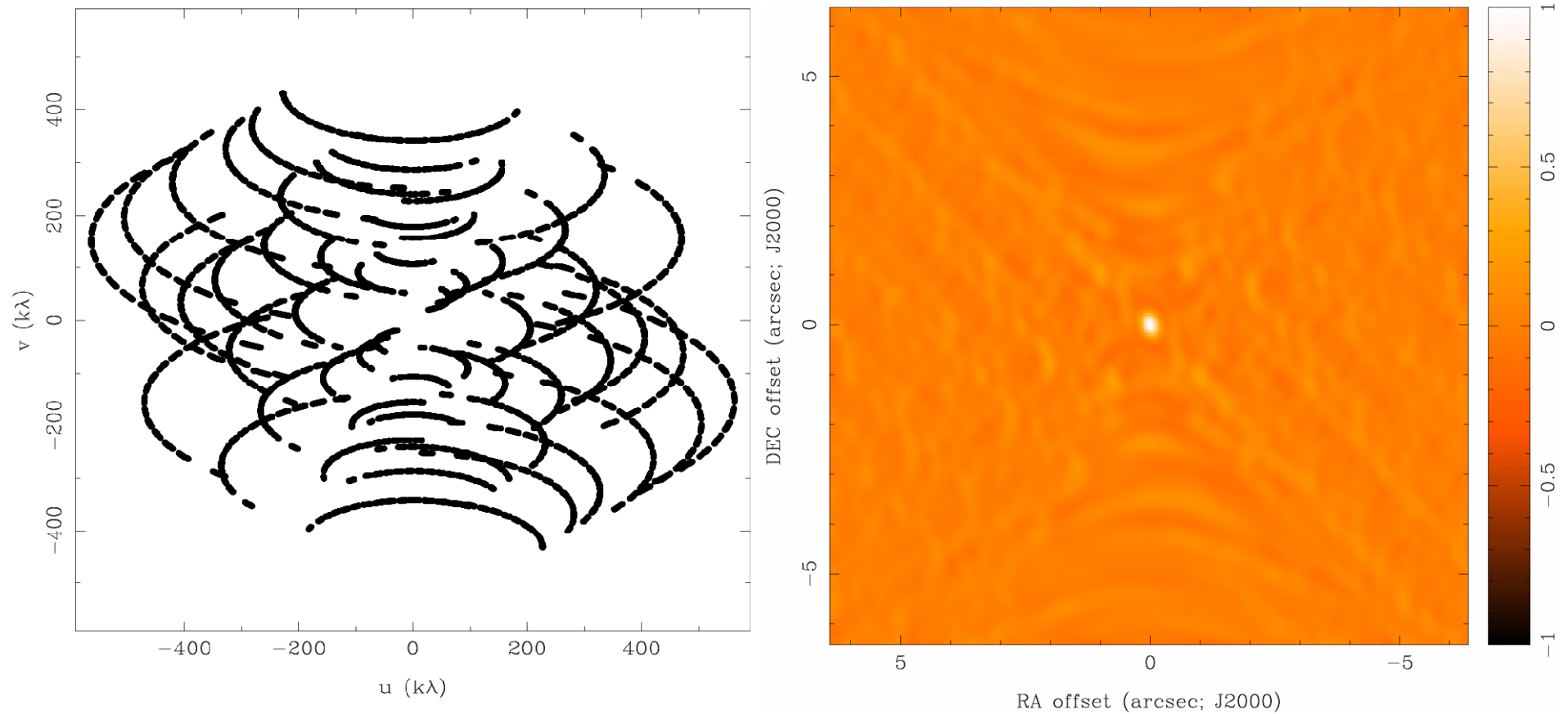
# Dirty Beam Shape and N Antennas

8 Antennas x 240 Samples



# Dirty Beam Shape and N Antennas

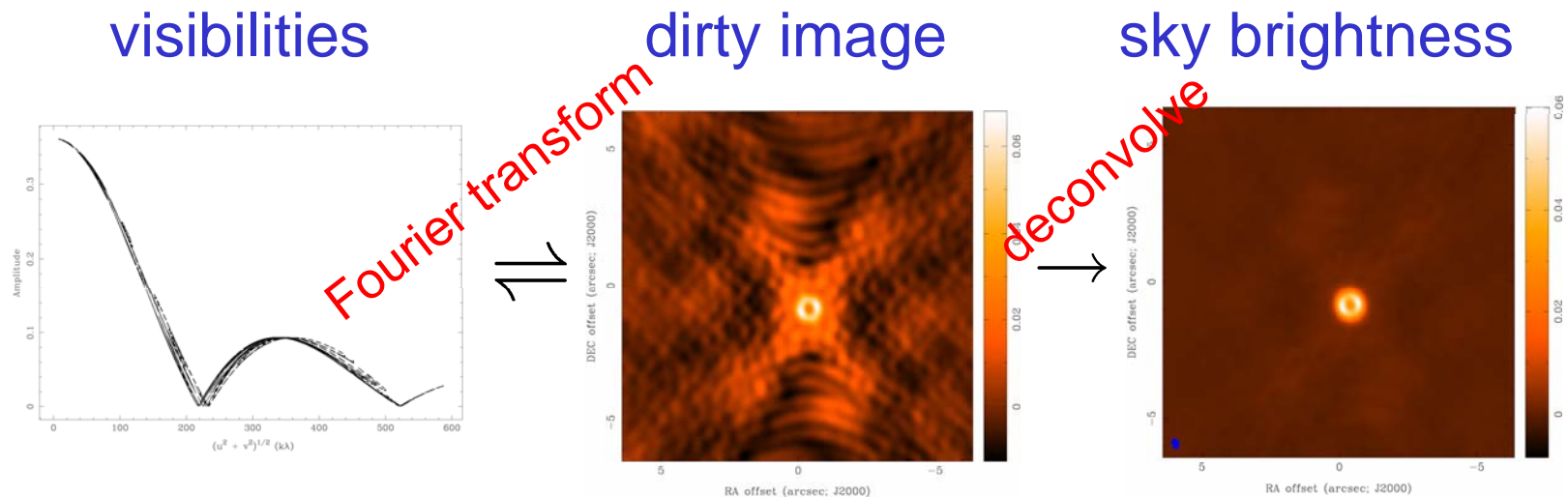
8 Antennas x 480 Samples





# How to analyze interferometer data?

- uv plane analysis
  - best for “simple” sources, e.g. point sources, disks
- image plane analysis
  - Fourier transform  $V(u,v)$  samples to image plane, get  $T^D(x,y)$
  - but difficult to do science on dirty image
  - deconvolve  $b(x,y)$  from  $T^D(x,y)$  to determine (model of)  $T(x,y)$



# Details of the Dirty Image

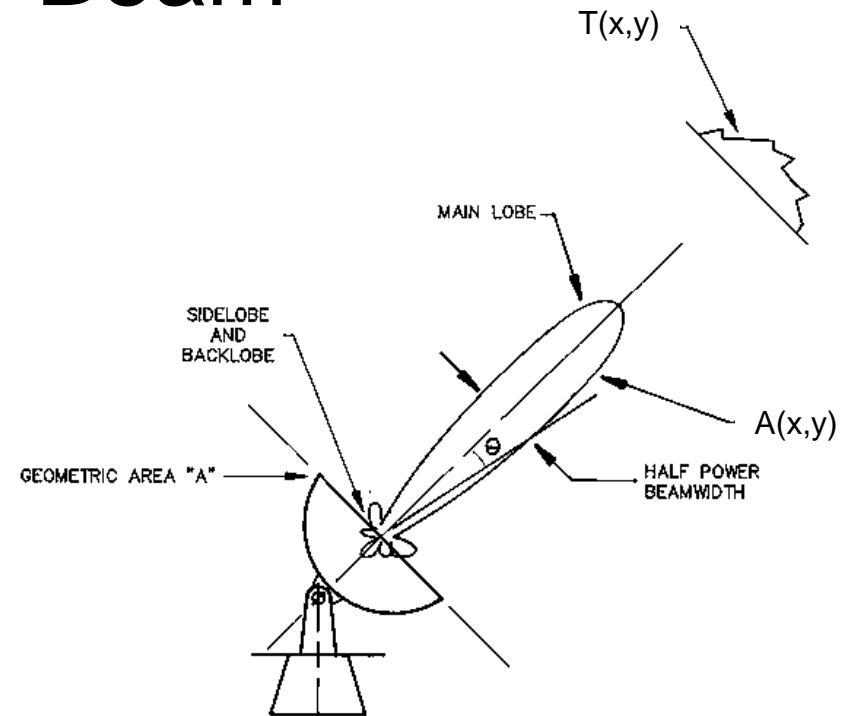
- Fourier Transform
  - Fast Fourier Transform (FFT) much faster than simple Fourier summation,  $O(N \log N)$  for  $2^N \times 2^N$  image
  - FFT requires data on regularly spaced grid
  - aperture synthesis observations not on a regular grid...
- “Gridding” is used to resample  $V(u,v)$  for FFT
  - customary to use a convolution technique
    - visibilities are noisy samples of a smooth function
    - nearby visibilities not independent
  - use special (“Spheroidal”) functions with nice properties
    - fall off quickly in  $(u,v)$  plane (not too much smoothing)
    - fall off quickly in image plane (avoid aliasing)

$$V^G(u, v) = V(u, v)B(u, v) \otimes G(u, v) \rightleftharpoons T^D(x, y)g(x, y)$$

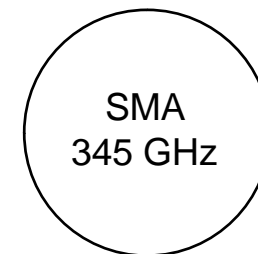


# Primary Beam

- A telescope does not have uniform response across the entire sky
  - main lobe approximately Gaussian,  $\text{fwhm} \sim 1.2\lambda/D$ , where  $D$  is ant diameter = “primary beam”
  - limited field of view
  - sidelobes, error beam (sometimes important)
- primary beam response modifies sky brightness:  $T(x,y) \rightarrow A(x,y)T(x,y)$ 
  - correct with division by  $A(x,y)$  in image plane



$T(x,y)$



large  $A(x,y)$



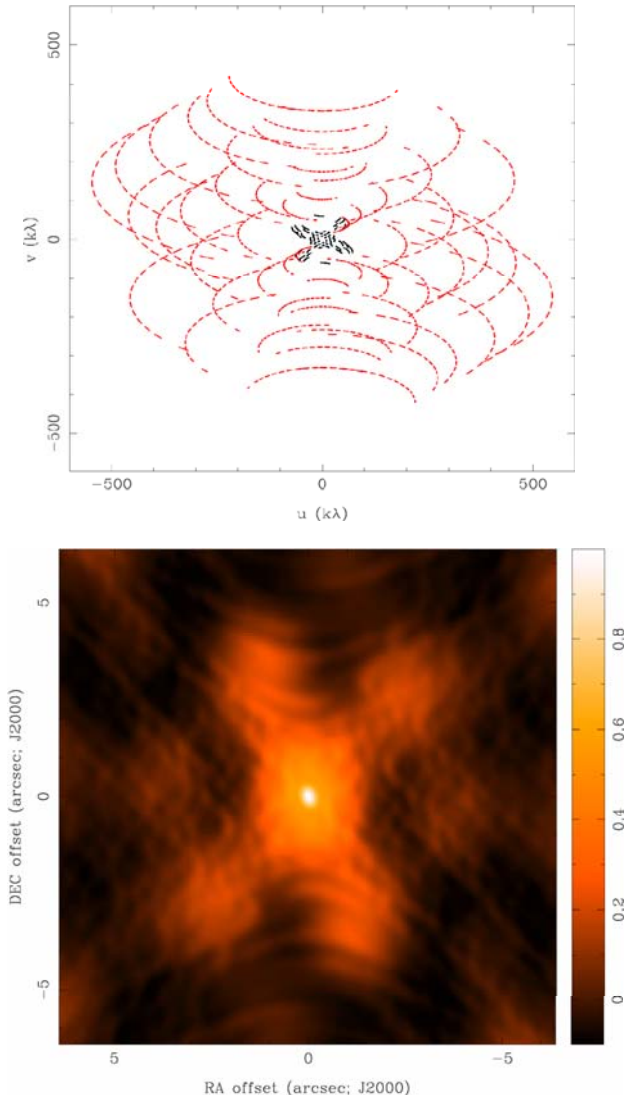
small  $A(x,y)$

# Pixel Size and Image Size

- pixel size
  - should satisfy sampling theorem for the longest baselines,  
 $\Delta x < 1/2 u_{\max}$  ,  $\Delta y < 1/2 v_{\max}$
  - in practice, 3 to 5 pixels across the main lobe of the dirty beam  
(to aid deconvolution)
  - e.g., SMA: 870  $\mu\text{m}$ , 500 m baselines  $\rightarrow 600 k\lambda \rightarrow < 0.1$  arcsec
- image size
  - natural resolution in (u,v) plane samples  $\text{FT}\{A(x,y)\}$ , implies  
image size 2x primary beam
  - e.g., SMA: 870  $\mu\text{m}$ , 6 m telescope  $\rightarrow 2 \times 35$  arcsec
  - if there are bright sources in the sidelobes of  $A(x,y)$ , then they  
will be aliased into the image (need to make a larger image)

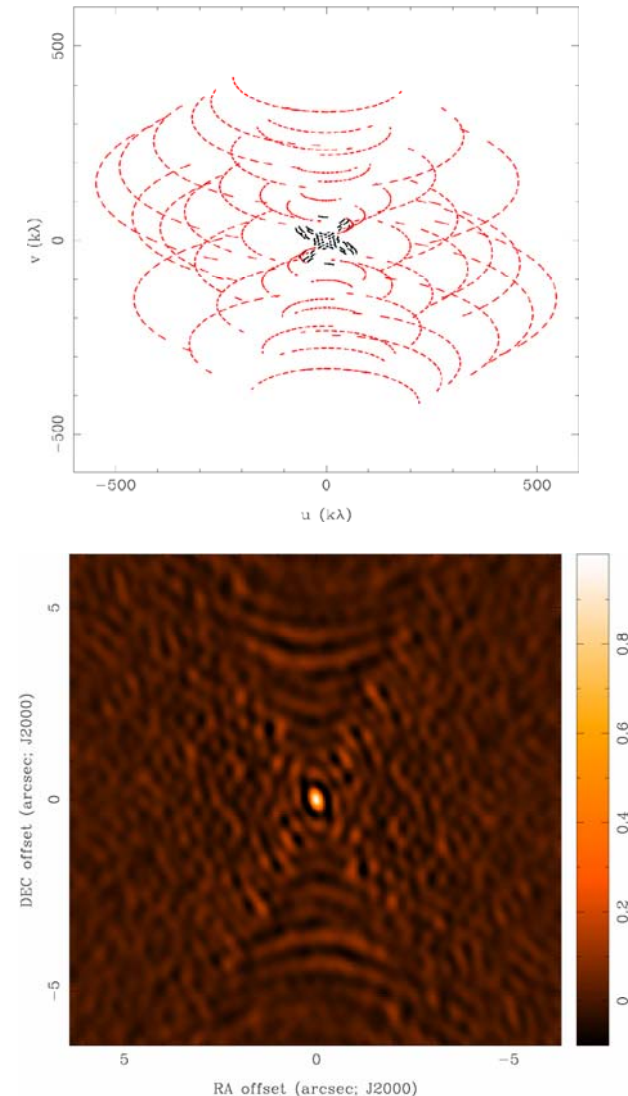
# Dirty Beam Shape and Weighting

- introduce weighting function  $W(u,v)$   
$$b(x,y) = FT^{-1}\{W(u,v)B(u,v)\}$$
  - $W$  modifies sidelobes of dirty beam  
( $W$  is also gridded for FFT)
- “Natural” weighting
  - $W(u,v) = 1/\sigma^2(u,v)$  at points with data and zero elsewhere, where  $\sigma^2(u,v)$  is the noise variance of the  $(u,v)$  sample
  - maximizes point source sensitivity  
(lowest rms in image)
  - generally more weight to short baselines  
(large spatial scales), degrades resolution



# Dirty Beam Shape and Weighting

- “Uniform” weighting
  - $W(u,v)$  is inversely proportional to local density of  $(u,v)$  points, so sum of weights in a  $(u,v)$  cell is a constant (or zero)
  - fills  $(u,v)$  plane more uniformly, so (outer) sidelobes are lower
  - gives more weight to long baselines and therefore higher angular resolution
  - degrades point source sensitivity (higher rms in image)
  - can be trouble with sparse sampling: cells with few data points have same weight as cells with many data points

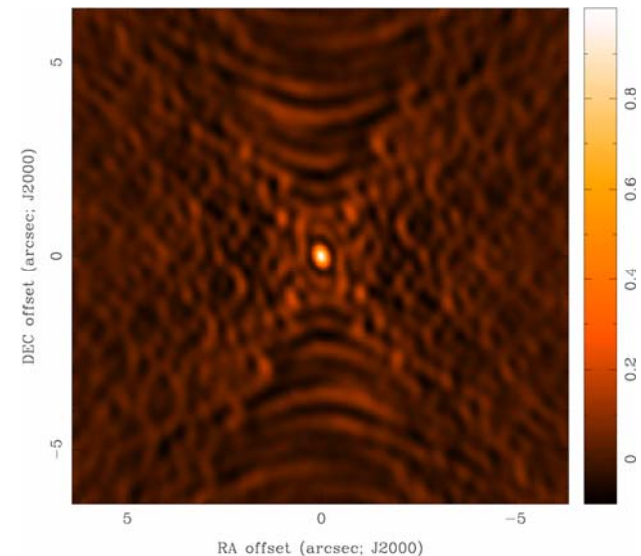
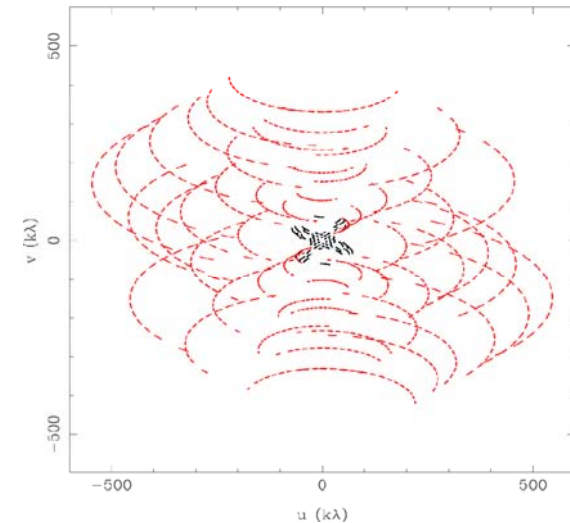


# Dirty Beam Shape and Weighting

- “Robust” (Briggs) weighting
  - variant of “uniform” that avoids giving too much weight to cell with low natural weight
  - implementations differ, e.g.  $S_N$  is natural weight of a cell,  $S_t$  is a threshold

$$W(u, v) = \frac{1}{\sqrt{1 + S_N^2 / S_{thresh}^2}}$$

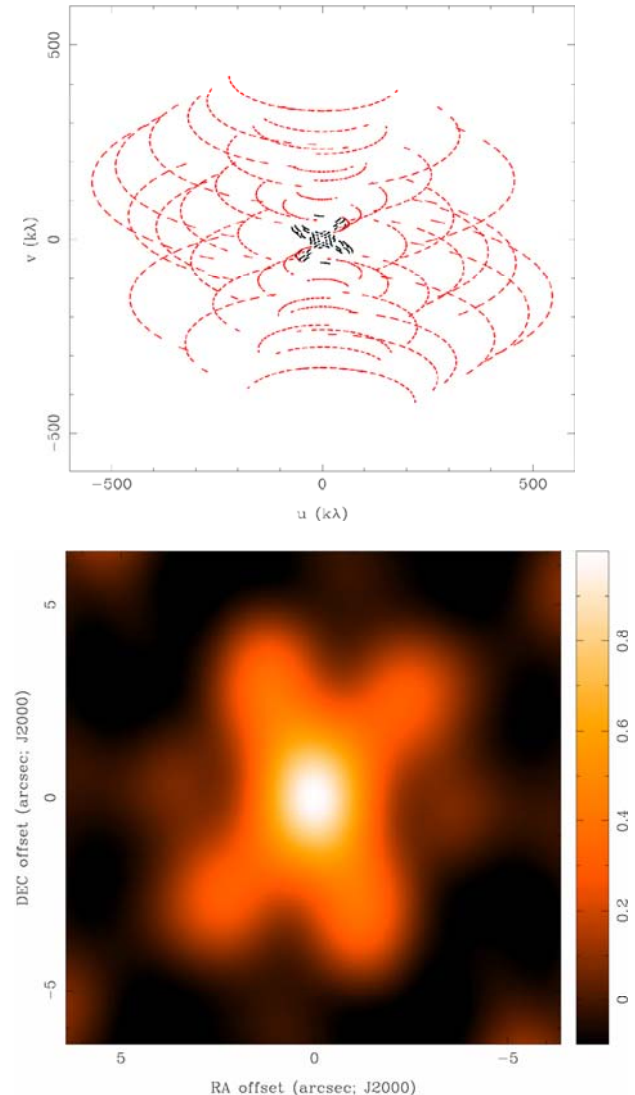
- large threshold  $\rightarrow$  natural weighting
- small threshold  $\rightarrow$  uniform weighting
- an adjustable parameter that allows for continuous variation between highest angular resolution and optimal point source sensitivity



# Dirty Beam Shape and Weighting

- “Tapering”
  - apodize the  $(u,v)$  sampling by a Gaussian
$$W(u, v) = \exp \left\{ -\frac{(u^2 + v^2)}{t^2} \right\}$$

$t$  = tapering parameter (in  $k\lambda$ ; arcsec)
  - like smoothing in the image plane (convolution by a Gaussian)
  - gives more weight to short baselines, degrades angular resolution
  - degrades point source sensitivity but can improve sensitivity to extended structure
  - could use elliptical Gaussian, other function
  - limits to usefulness

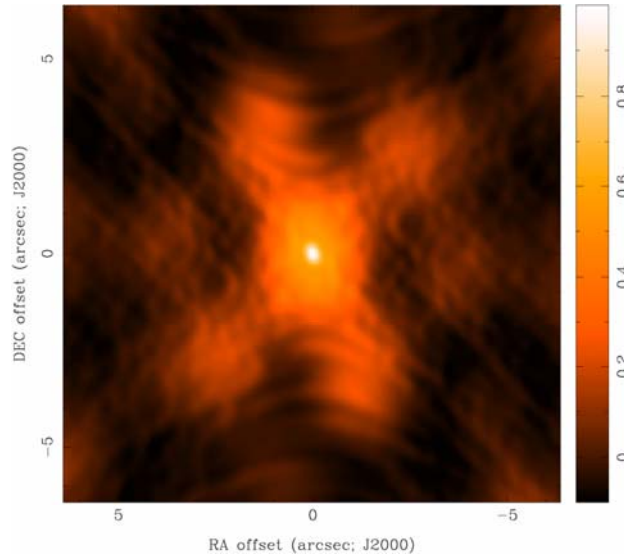




# Weighting and Tapering: Noise

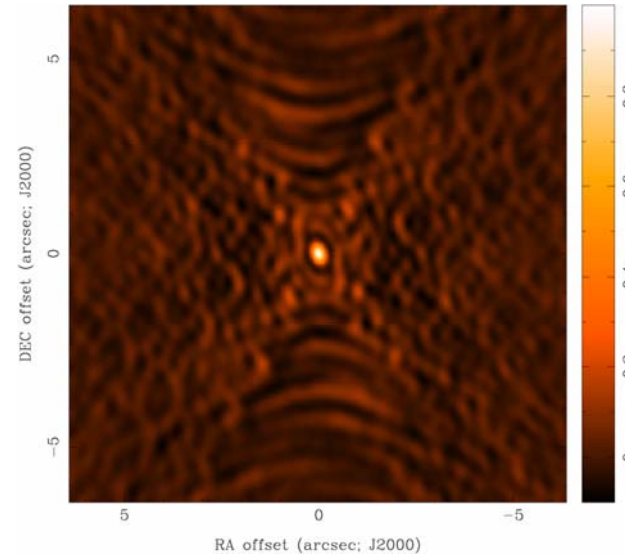
Natural  
0.77x0.62

$\sigma=1.0$



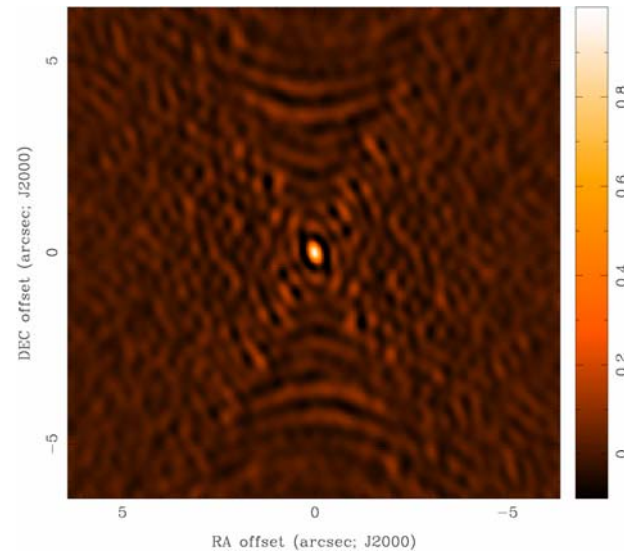
Robust 0  
0.41x0.36

$\sigma=1.6$



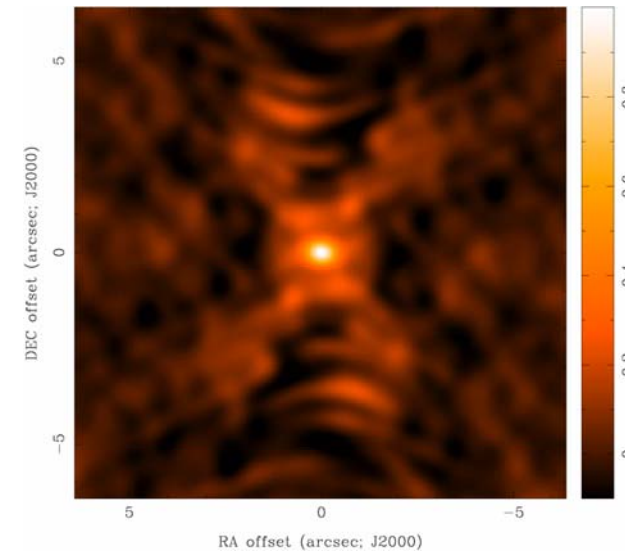
Uniform  
0.39x0.31

$\sigma=3.7$



Robust 0  
+ Taper  
0.77x0.62

$\sigma=1.7$



# Weighting and Tapering: Summary

- imaging parameters provide a lot of freedom
- appropriate choice depends on science goals

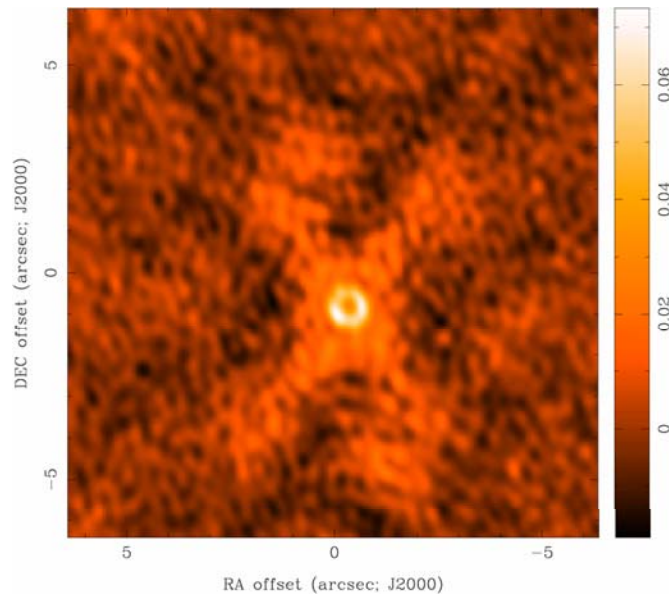
	Robust/Uniform	Natural	Taper
Resolution	higher	medium	lower
Sidelobes	lower	higher	depends
Point Source Sensitivity	lower	maximum	lower
Extended Source Sensitivity	lower	medium	higher



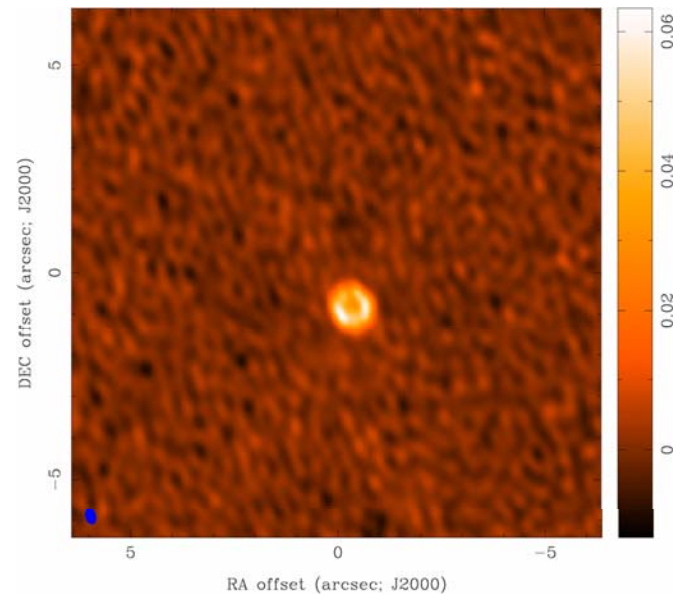
# Deconvolution

- difficult to do science on dirty image
- deconvolve  $b(x,y)$  from  $T^D(x,y)$  to recover  $T(x,y)$
- information is missing, so be careful!  
(there's noise, too)

dirty image



“CLEAN” image



# Deconvolution Philosophy

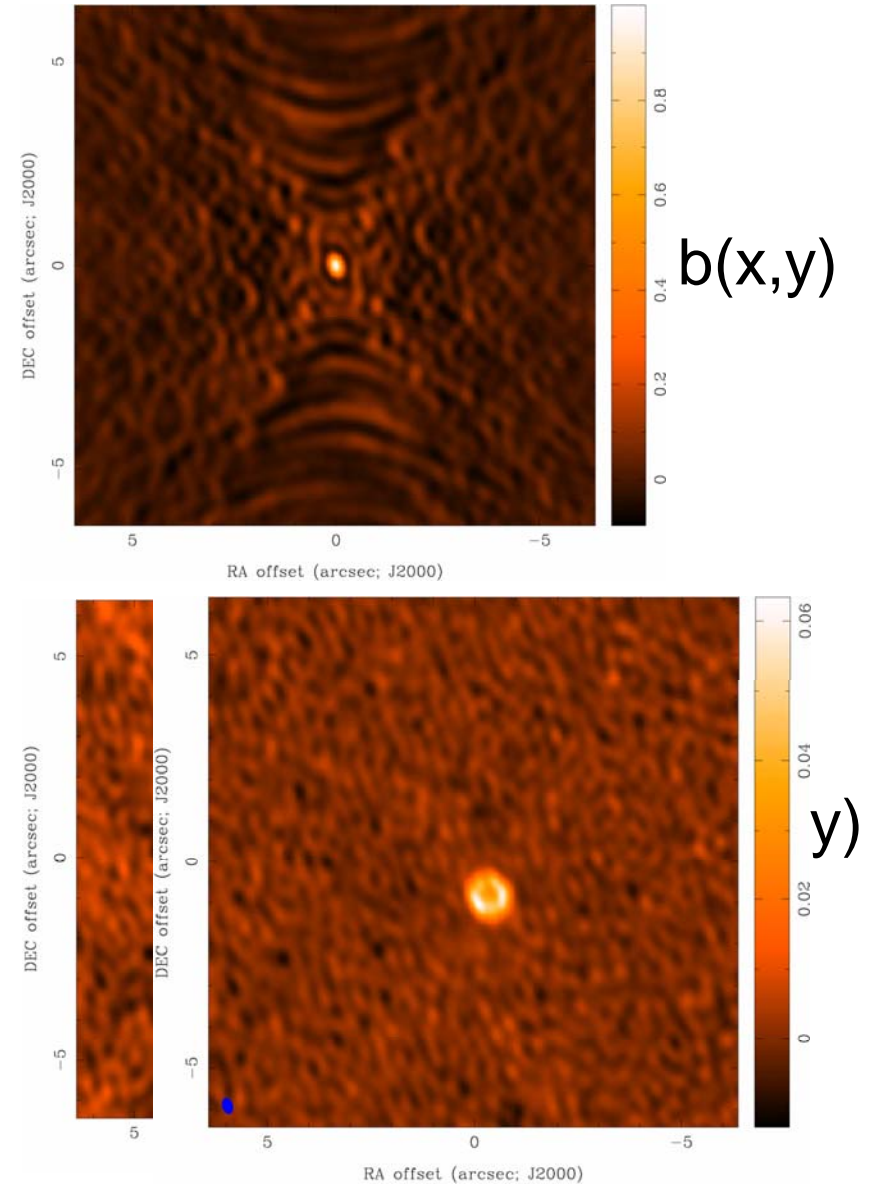
- to keep you awake at night
  - $\exists$  an infinite number of  $T(x,y)$  compatible with sampled  $V(u,v)$ , i.e. “invisible” distributions  $R(x,y)$  where  $b(x,y) \otimes R(x,y) = 0$ 
    - no data beyond  $u_{\max}, v_{\max} \rightarrow$  unresolved structure
    - no data within  $u_{\min}, v_{\min} \rightarrow$  limit on largest size scale
    - holes between  $u_{\min}, v_{\min}$  and  $u_{\max}, v_{\max} \rightarrow$  sidelobes
  - noise  $\rightarrow$  undetected/corrupted structure in  $T(x,y)$
  - no unique prescription for extracting optimum estimate of true sky brightness from visibility data
- deconvolution
  - uses non-linear techniques effectively interpolate/extrapolate samples of  $V(u,v)$  into unsampled regions of the  $(u,v)$  plane
  - aims to find a **sensible** model of  $T(x,y)$  compatible with data
  - requires *a priori* assumptions about  $T(x,y)$

# Deconvolution Algorithms

- most common algorithms in radio astronomy
  - CLEAN (Högbom 1974)
    - *a priori* assumption:  $T(x,y)$  is a collection of point sources
    - variants for computational efficiency, extended structure
  - Maximum Entropy (Gull and Skilling 1983)
    - *a priori* assumption:  $T(x,y)$  is smooth and positive
    - vast literature about the deep meaning of entropy (Bayesian)
  - hybrid approaches of these can be effective
- deconvolution requires knowledge of beam shape and image noise properties (usually OK for aperture synthesis)
  - atmospheric seeing can modify effective beam shape
  - deconvolution process can modify image noise properties

# Basic CLEAN Algorithm

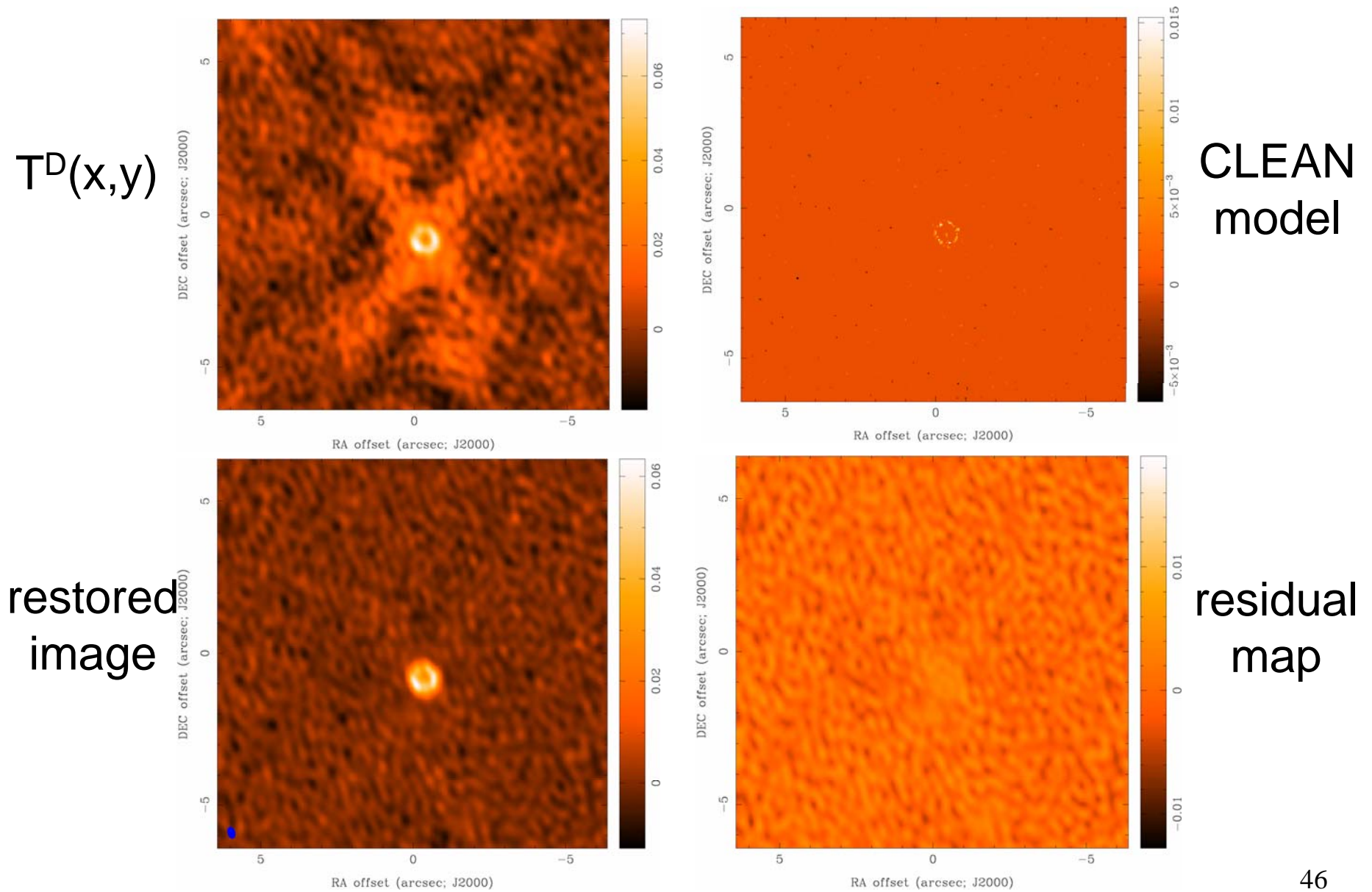
1. Initialize
  - a *residual* map to the dirty map
  - a *Clean component* list to empty
2. Identify strongest feature in *residual* map as a point source
3. Add a fraction  $g$  (the loop gain) of this point source to the clean component list
4. Subtract the fraction  $g$  times  $b(x,y)$  from *residual* map
5. If stopping criteria not reached, goto step 2 (an iteration)
6. Convolve *Clean component* (cc) list by an estimate of the main lobe of the dirty beam (the “Clean beam”) and add *residual* map to make the final “restored” image



# Basic CLEAN Algorithm (cont)

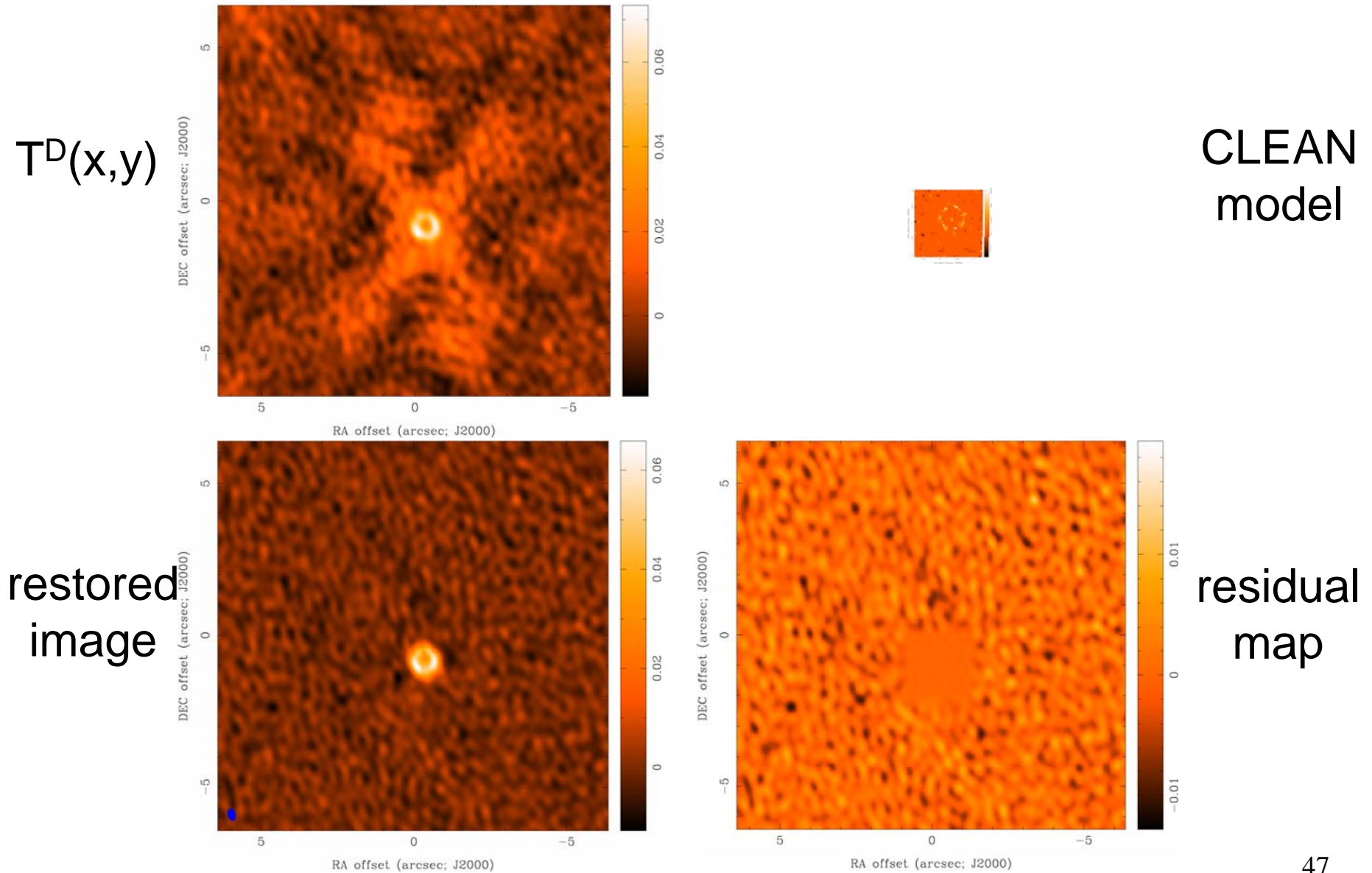
- stopping criteria
  - *residual* map max < multiple of rms (when noise limited)
  - *residual* map max < fraction of dirty map max (dynamic range limited)
  - max number of clean components reached (no justification)
- loop gain
  - good results for  $g \sim 0.1$  to  $0.3$
  - lower values can work better for smoother emission,  $g \sim 0.05$
- easy to include *a priori* information about where to search for clean components (“clean boxes”)
  - very useful but potentially dangerous!
- Schwarz (1978): CLEAN is equivalent to a least squares fit of sinusoids, in the absense of noise

# CLEAN

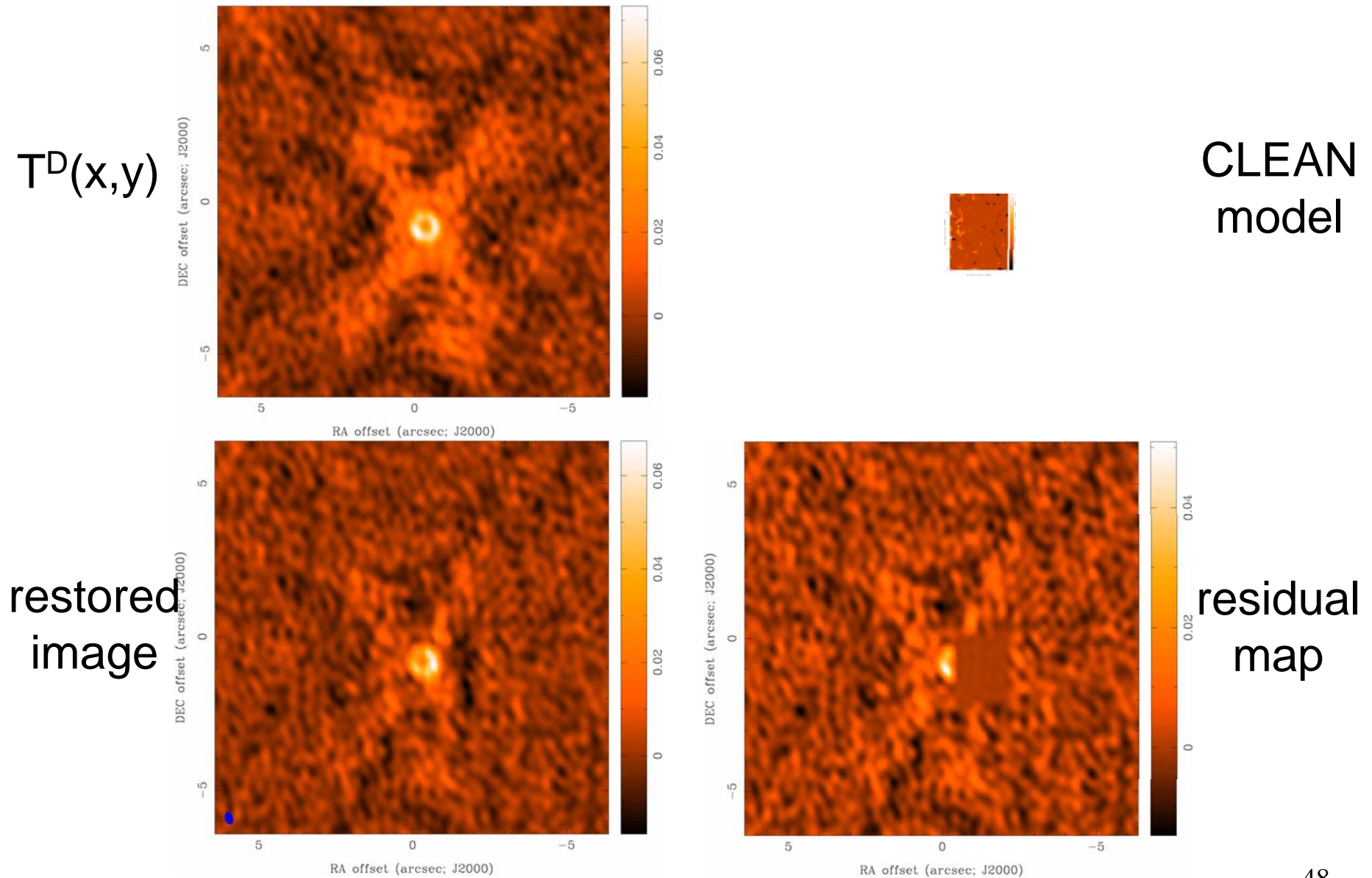




# CLEAN with Box



# CLEAN with Poor Choice of Box





# CLEAN Variants

- Clark CLEAN
  - aims at faster speed for large images
  - Högbom-like “minor” cycle w/ truncated dirty beam, subset of largest residuals
  - in “major” cycle, cc’s are FFT’d and subtracted from the FFT of the residual image from the previous “major” cycle
- Cotton-Schwab CLEAN (MX)
  - in “major” cycle, cc’s are FFT’d and subtracted from ungridded visibilities
  - more accurate but slower (gridding steps repeated)
- Steer, Dewdney, Ito (SDI) CLEAN
  - aims to suppress CLEAN “stripes” in smooth, extended emission
  - in “minor” cycles, any point in the residual map greater than a fraction ( $<1$ ) of the maximum is taken as a cc
- Multi-Resolution CLEAN
  - aims to account for coupling between pixels by extended structure
  - independently CLEAN a smooth map and a difference map, fewer cc’s

# “Restored” Images

- CLEAN beam size:
  - natural choice is to fit the central peak of the dirty beam with elliptical Gaussian
  - unit of deconvolved map is Jy per CLEAN beam area (= intensity, can convert to brightness temperature)
  - minimize unit problems when adding dirty map residuals
  - modest super resolution often OK, but be careful
- photometry should be done with caution
  - CLEAN does not conserve flux (extrapolates)
  - extended structure missed, attenuated, distorted
  - phase errors (e.g. seeing) can spread signal around

# Noise in Images

- point source sensitivity: straightforward
  - telescope area, bandwidth, integration time, weighting
  - in image, modify noise by primary beam response
- extended source sensitivity: problematic
  - not quite right to divide noise by  $\sqrt{n}$  beams covered by source: smoothing = tapering, omitting data  $\rightarrow$  lower limit
  - Interferometers always missing flux at some spatial scale
- be careful with low signal-to-noise images
  - if position known,  $3\sigma$  OK for point source detection
  - if position unknown, then  $5\sigma$  required (flux biased by  $\sim 1\sigma$ )
  - if  $< 6\sigma$ , cannot measure the source size (require  $\sim 3\sigma$  difference between “long” and “short” baselines)
  - spectral lines may have unknown position, velocity, width

# Maximum Entropy Algorithm

- Maximize a measure of smoothness (the entropy)

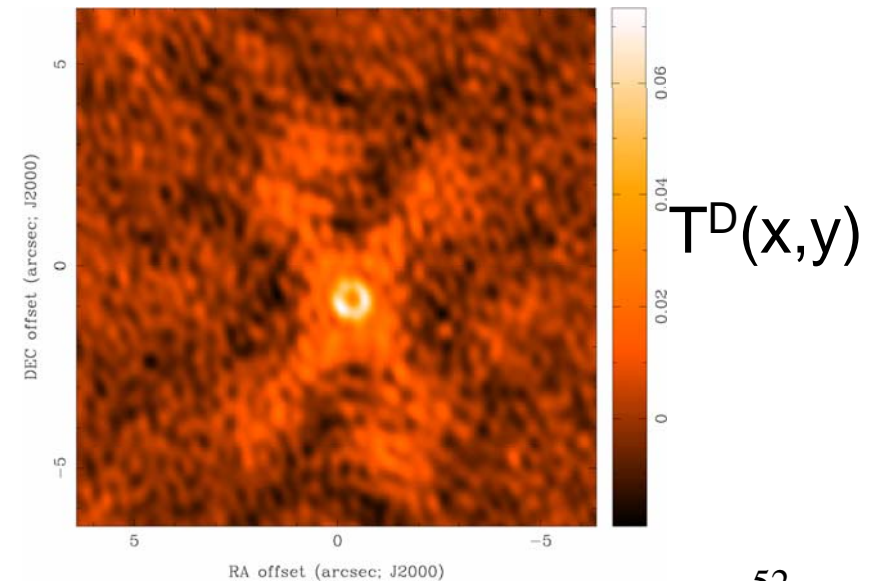
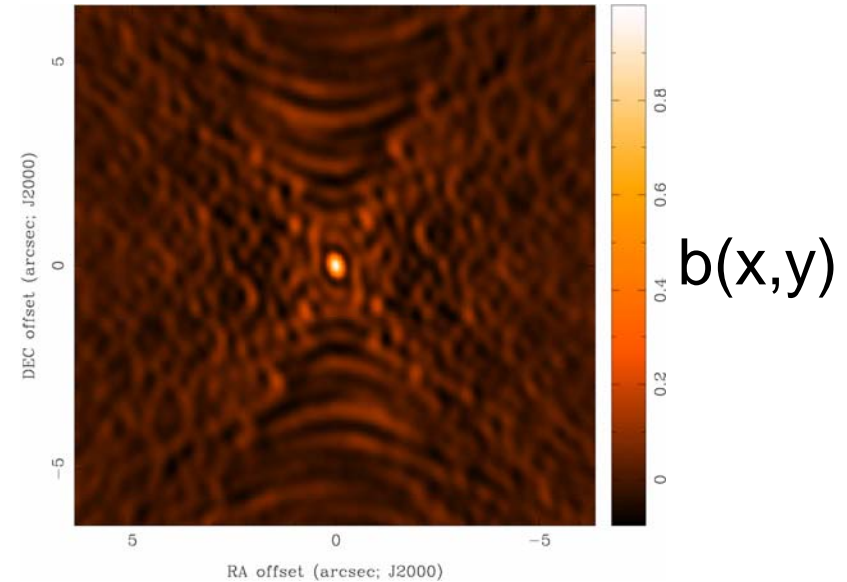
$$H = - \sum_k T_k \log \left( \frac{T_k}{M_k} \right)$$

subject to the constraints

$$\chi^2 = \sum_k \frac{|V(u_k, v_k) - \text{FT}\{T\}|^2}{\sigma_k^2}$$

$$F = \sum_k T_k$$

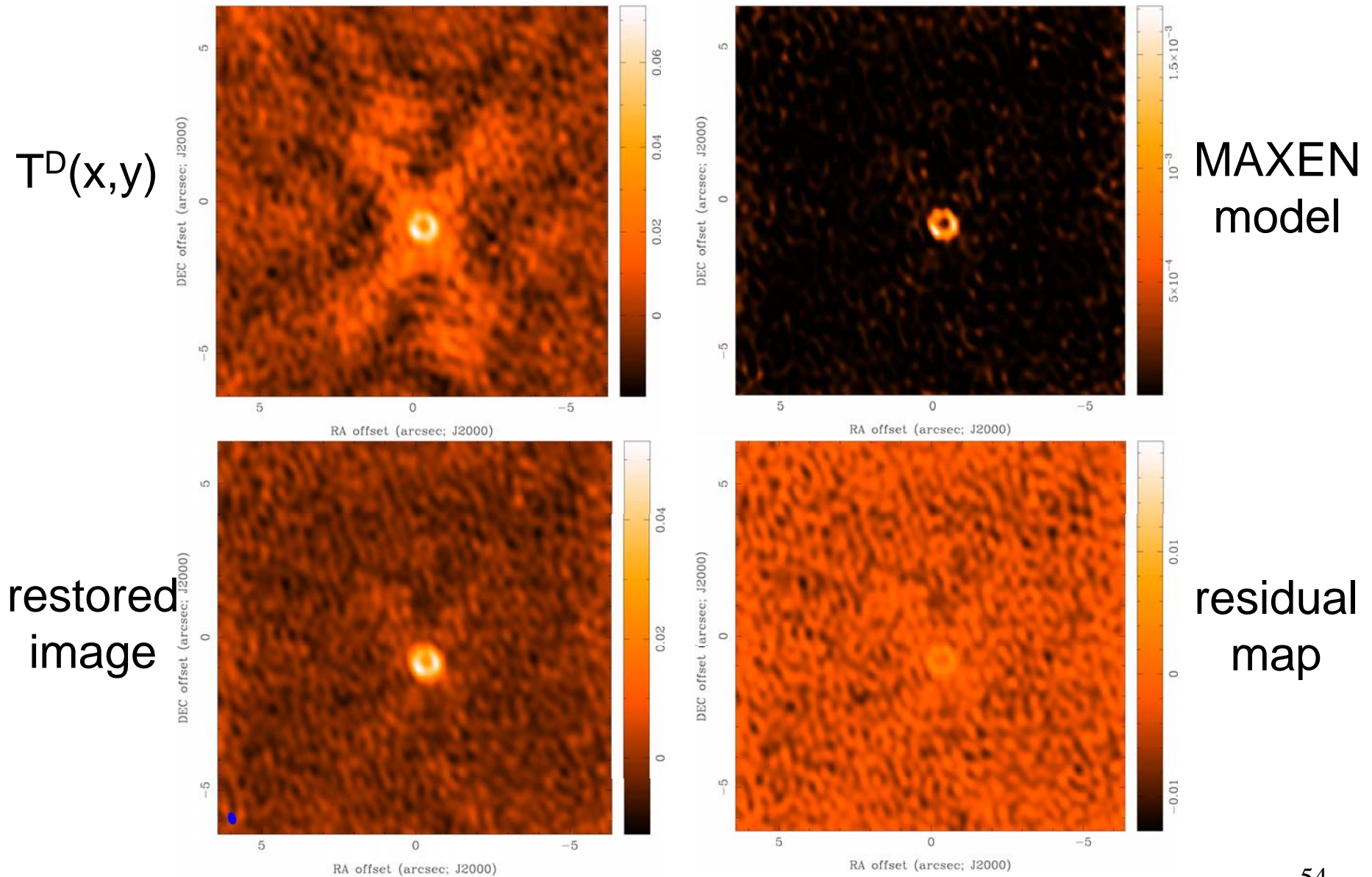
- M is the “default image”
- fast (NlogN) non-linear optimization solver due to Cornwell and Evans (1983)
- optional: convolve with Gaussian beam and add residual map to make image



# Maximum Entropy Algorithm (cont)

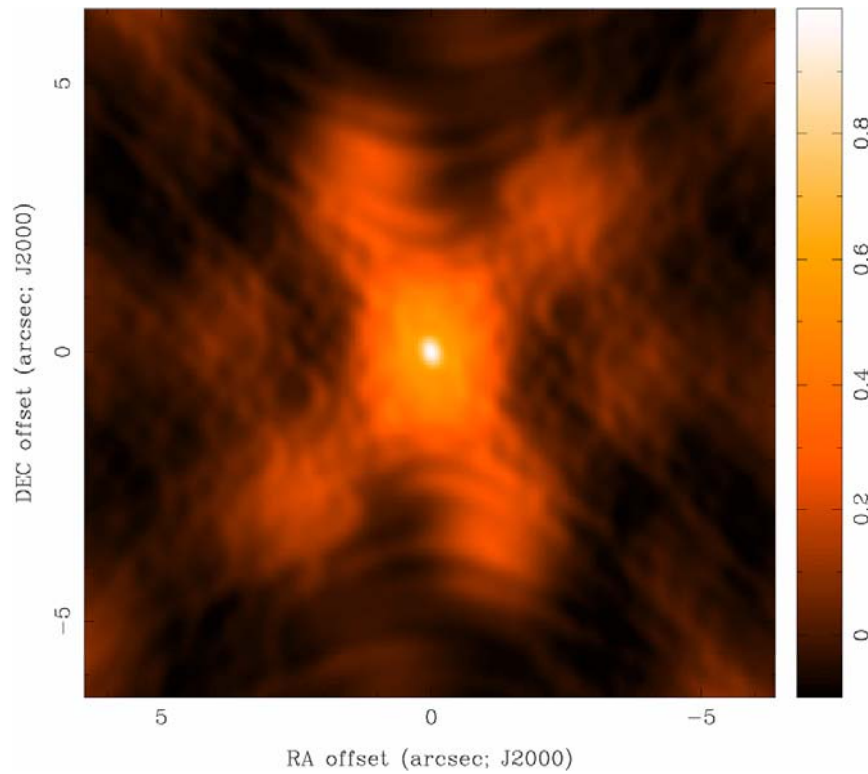
- easy to include *a priori* information with default image
  - flat default best only if nothing known (or nothing observed!)
- straightforward to generalize  $\chi^2$  to combine different observations/telescopes and obtain optimal image
- many measures of “entropy” available
  - replace log with cosh  $\rightarrow$  “emptiness” (does not enforce positivity)
- less robust and harder to drive than CLEAN
- works well on smooth, extended emission
- trouble with point source sidelobes
- no noise estimate possible from image

# Maximum Entropy

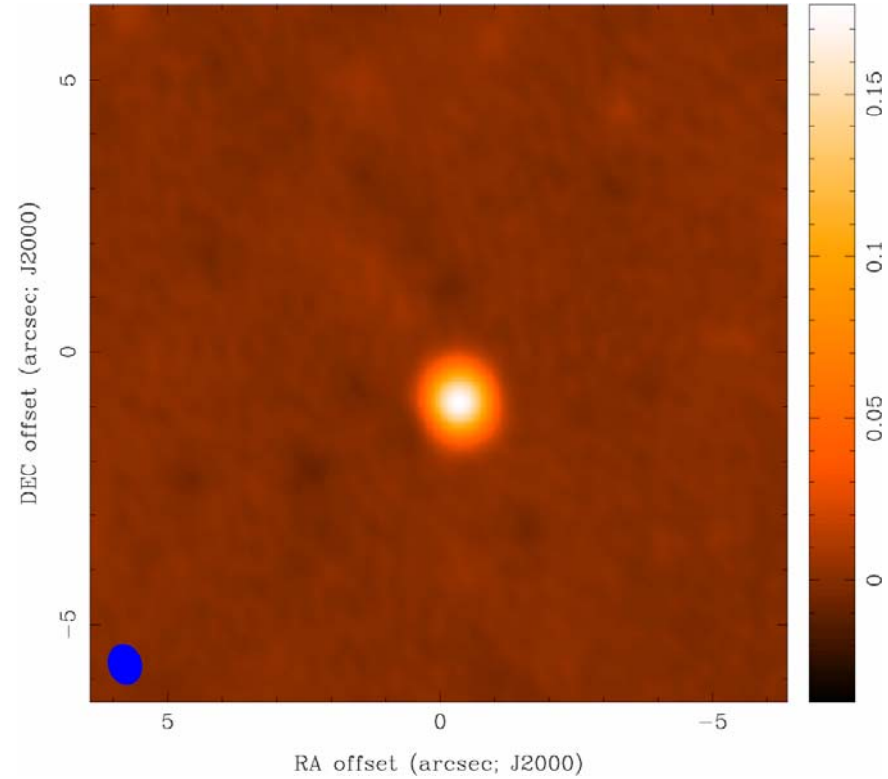


# Imaging Results

## Natural Weight Beam



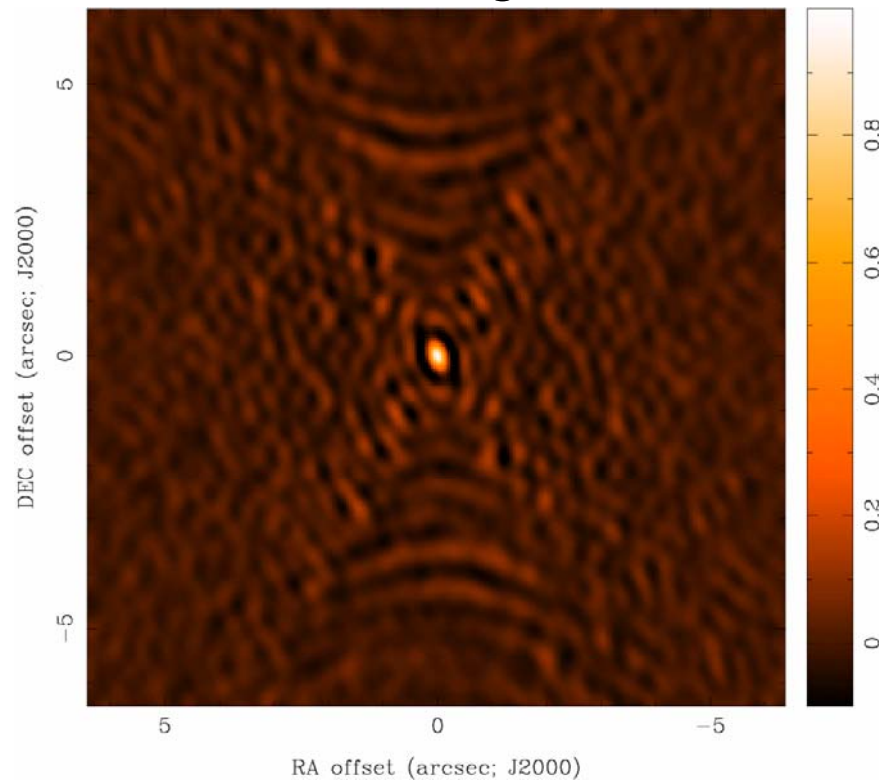
## CLEAN image



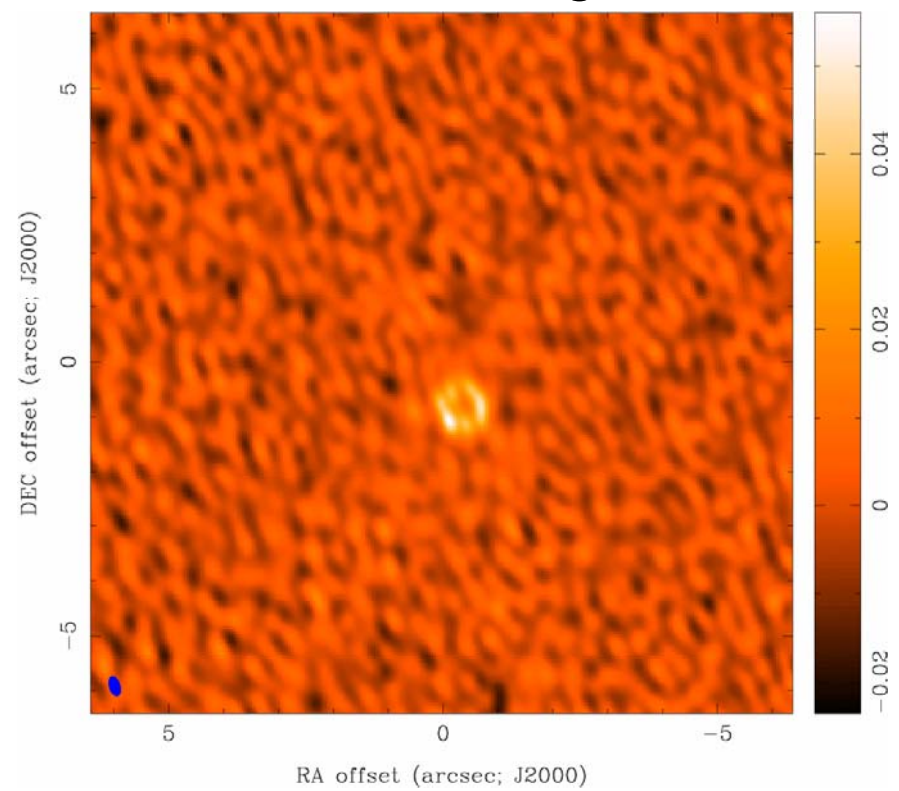


# Imaging Results

## Uniform Weight Beam

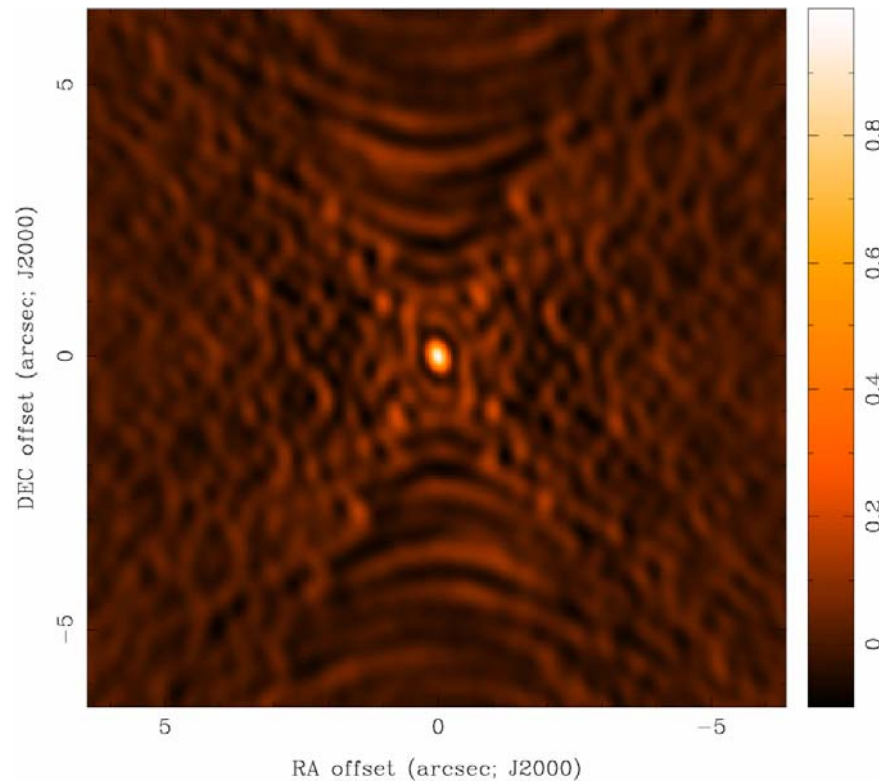


## CLEAN image

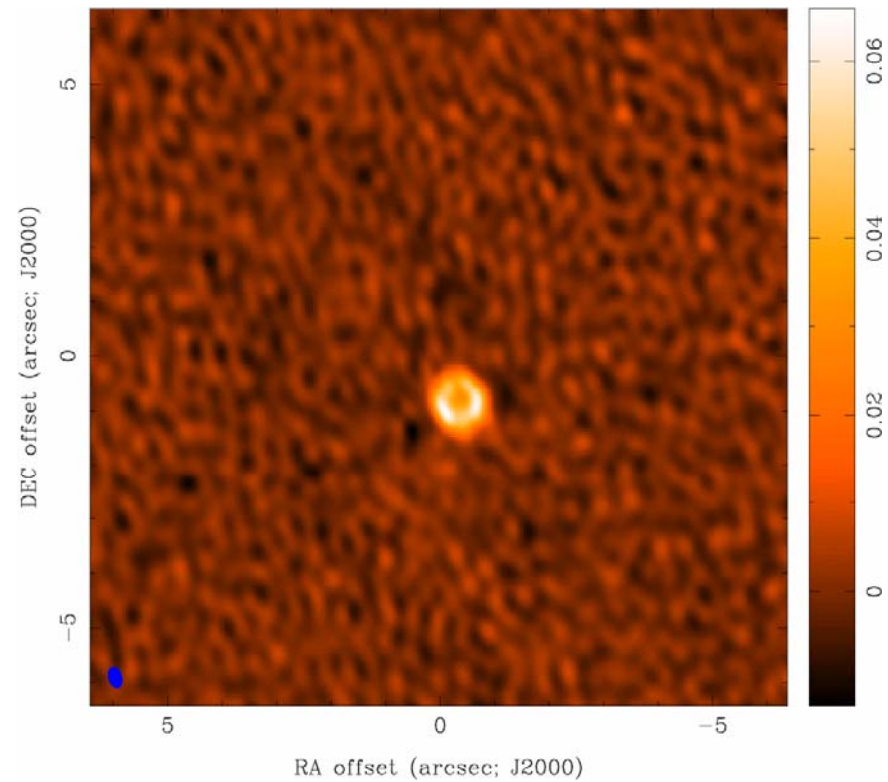


# Imaging Results

Robust=0 Beam

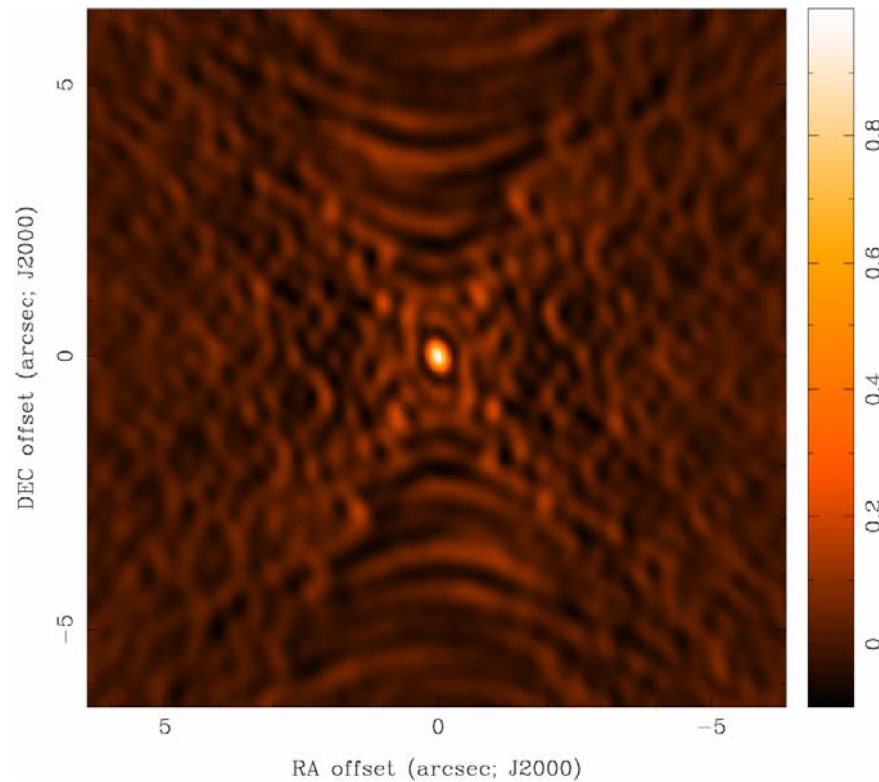


CLEAN image

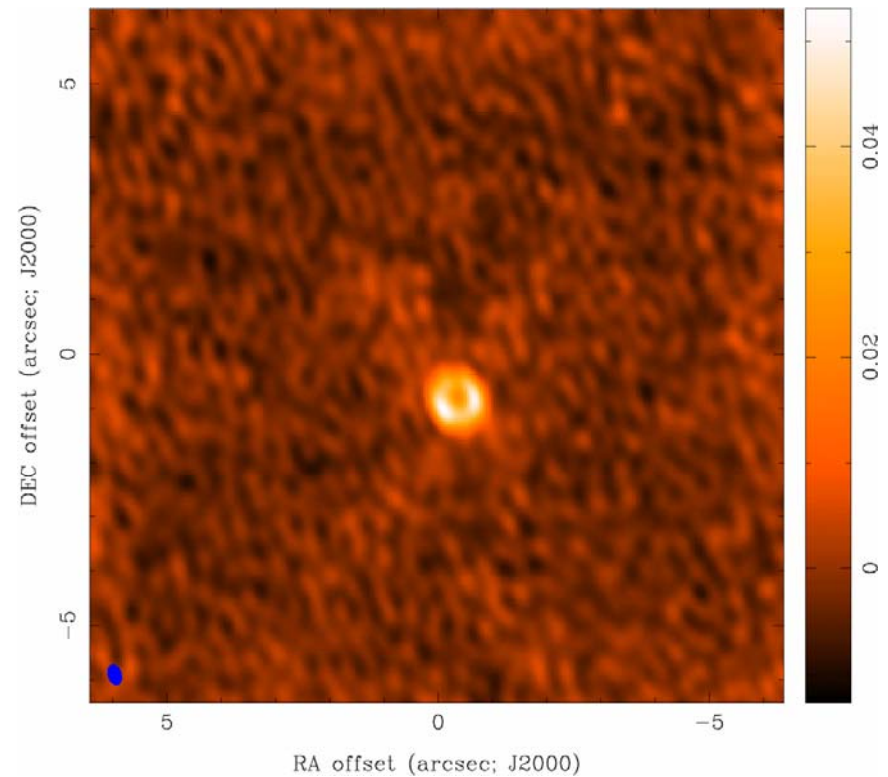


# Imaging Results

Robust=0 Beam

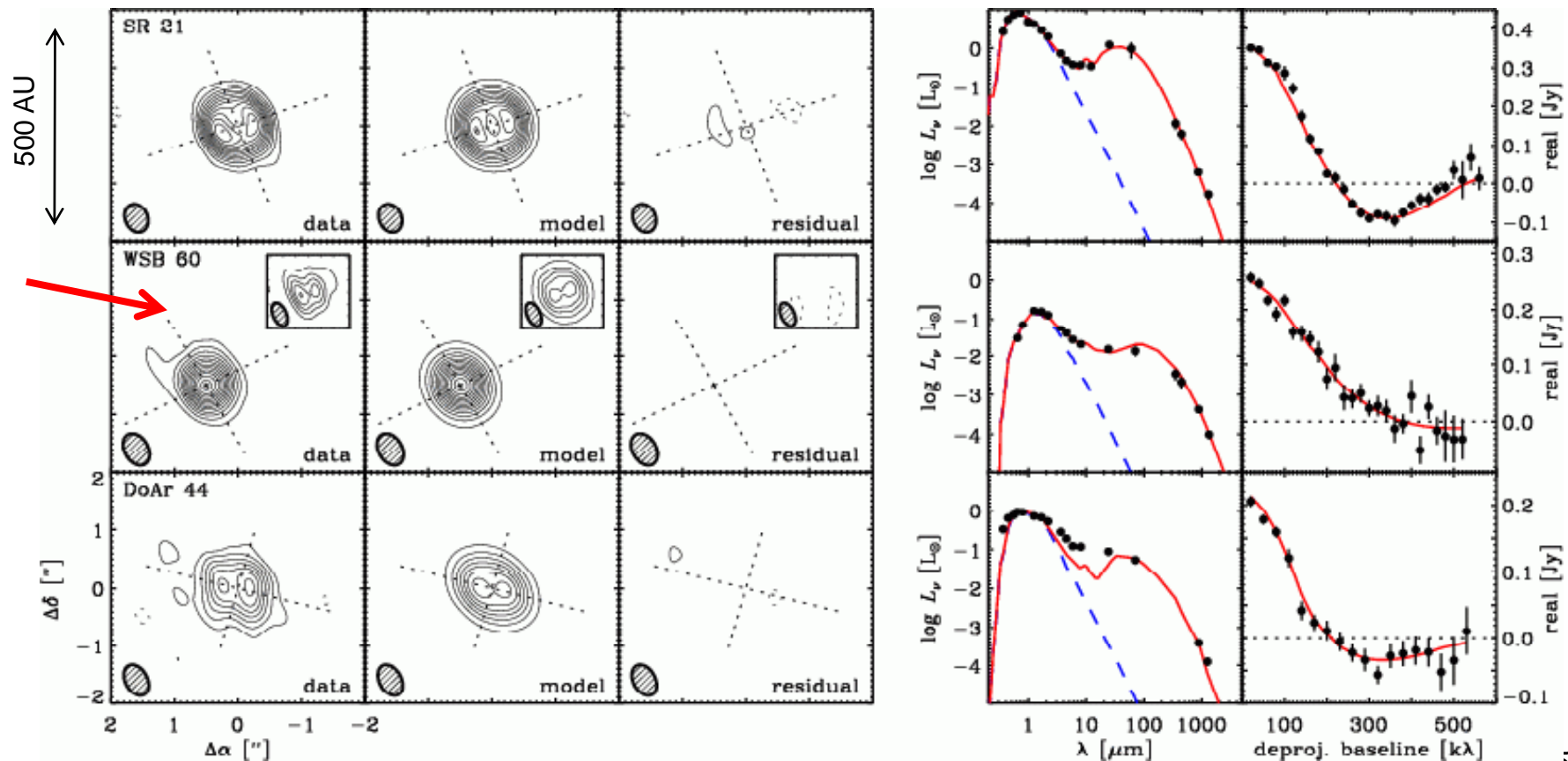


MAXEN image



# Tune Resolution/Sensitivity to suit Science

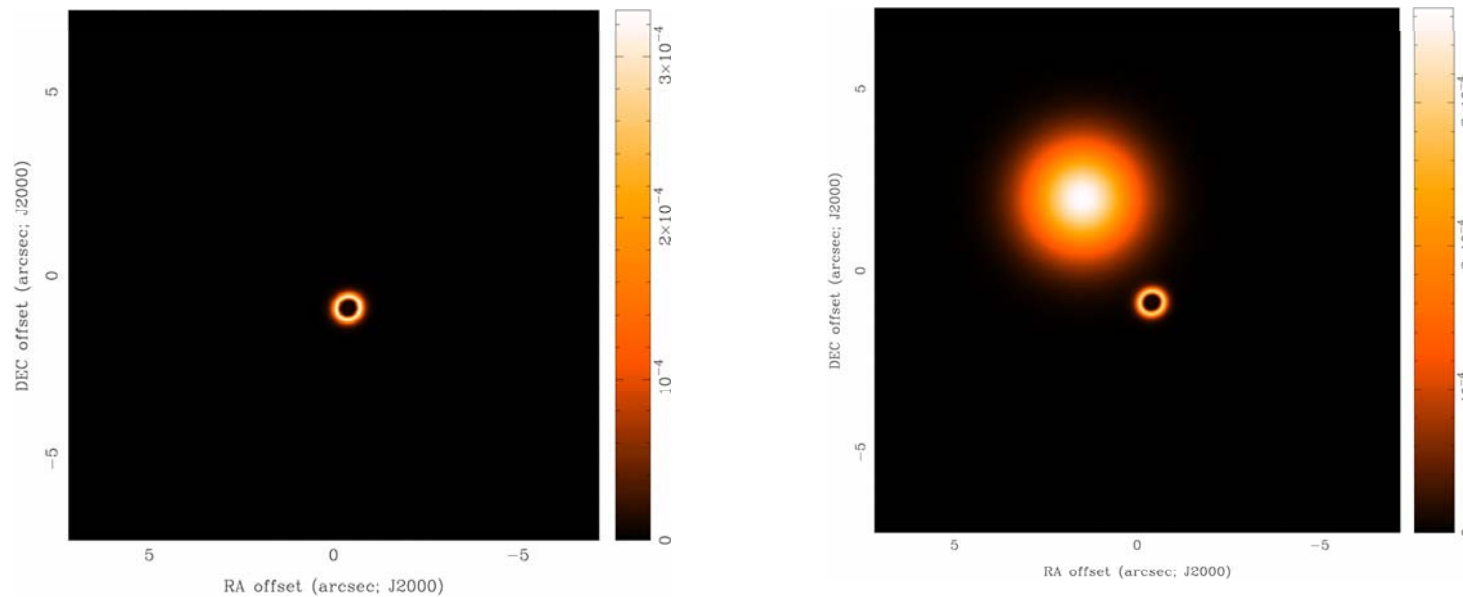
- e.g. Andrews, Wilner et al. 2009, ApJ, 700, 1502
  - SMA 870  $\mu\text{m}$  images of “transitional” protoplanetary disks with resolved inner holes, note images of WSB 60





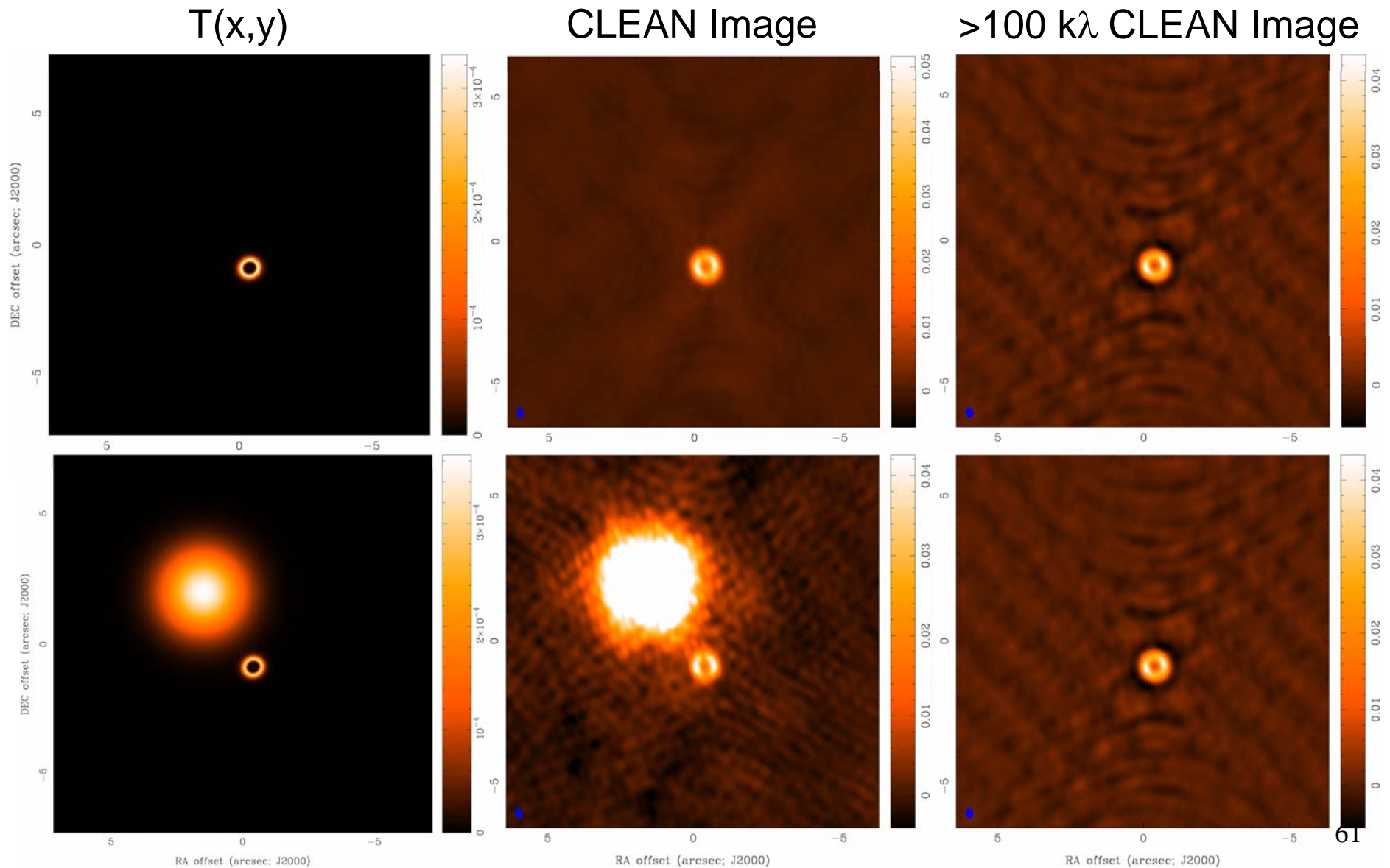
# Missing Short Spacings

Do the visibilities in the example discriminate between these models of the sky brightness distribution,  $T(x,y)$ ?



Yes... but only on baselines shorter than  $\sim 100 \text{ k}\lambda$ .

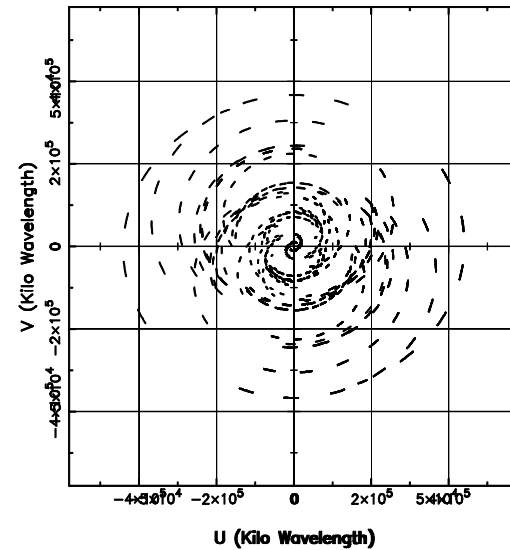
# Missing Short Spacings: Demonstration



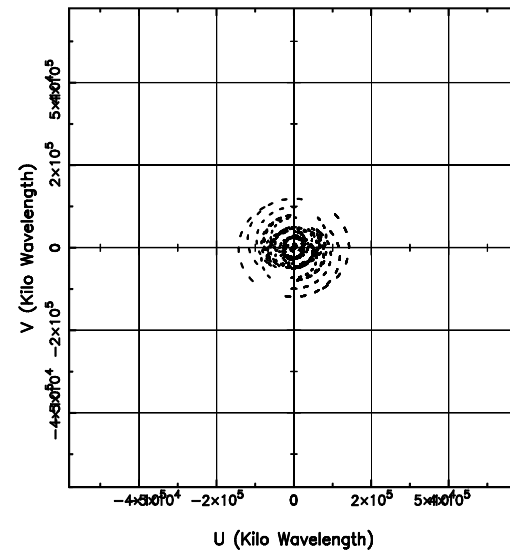
# Missing Short Spacings: Real Data

- Observations of X-ray transient CI Cam at 15 and 5 GHz with the VLBA+VLA1
- **Because of the different frequencies they have different uv-coverage**

u-v coverage  
for 15 GHz

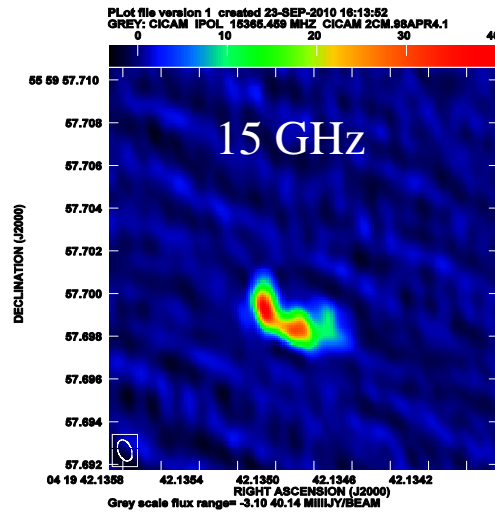


u-v coverage  
for 5 GHz





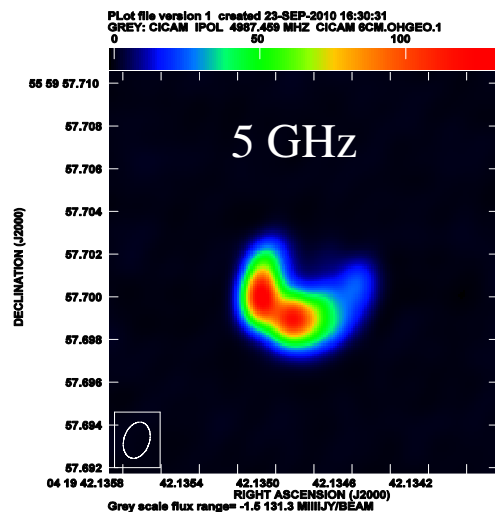
# Missing Short Spacings: Real Data



Two possible explanations of differences between frequencies:

1. Real differences in the spectral index across the source
2. u-v coverage differences

How do we tell the difference?



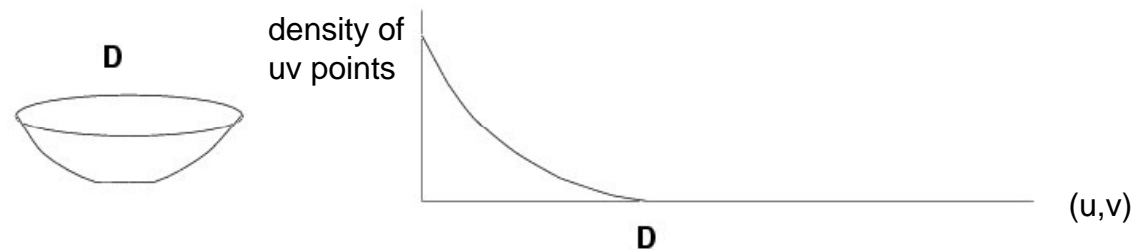
Unfortunately, the “standard” wisdom is that we do is convolve the 15 GHz image with the 5 GHz beam.

This takes account of the differences in resolution and ignores the hole in the u-v coverage.

Another approach is to put a model of the 5 GHz source through the u-v coverage of the 15 GHz image.

# Low Spatial Frequencies (I)

- Large Single Telescope
  - make an image by scanning across the sky
  - all Fourier components from 0 to  $D$  sampled, where  $D$  is the telescope diameter (weighting depends on illumination)



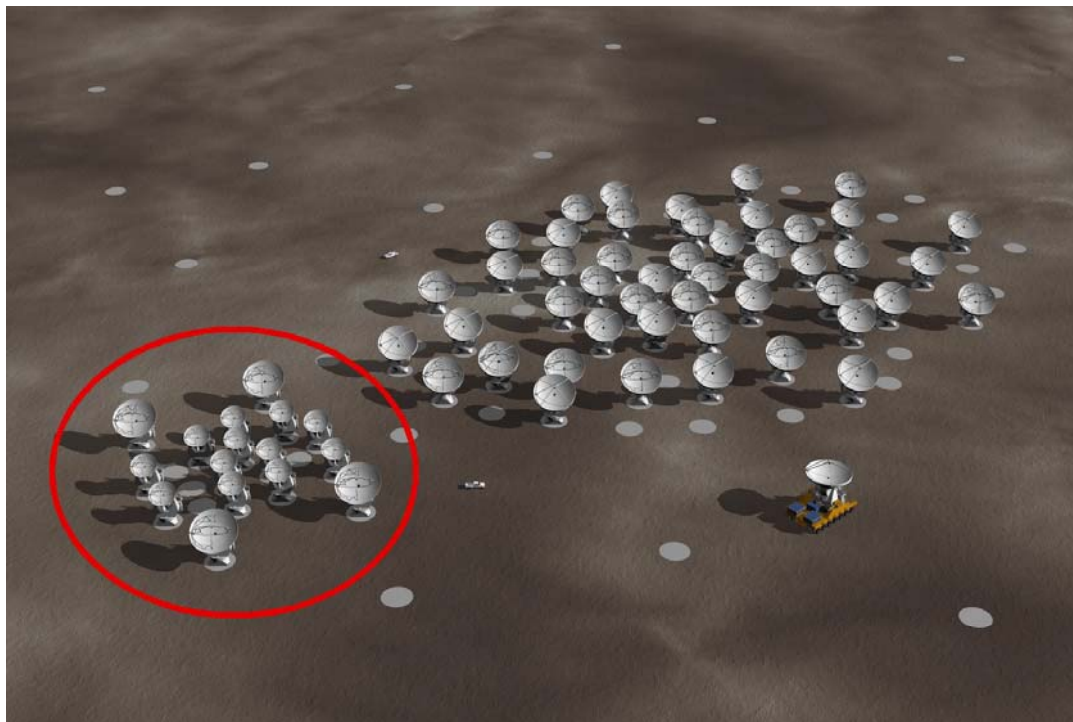
- Fourier transform single dish map =  $T(x,y) \otimes A(x,y)$ , then divide by  $a(x,y) = \text{FT}\{A(x,y)\}$ , to estimate  $V(u,v)$

$$\hat{V}(u, v) = \frac{[V(u, v)a(u, v)]}{\hat{a}(u, v)}$$

- choose  $D$  large enough to overlap interferometer samples of  $V(u,v)$  and avoid using data where  $a(x,y)$  becomes small

# Low Spatial Frequencies (II)

- separate array of smaller telescopes
  - use smaller telescopes observe short baselines not accessible to larger telescopes
  - shortest baselines from larger telescopes total power maps



## ALMA with ACA

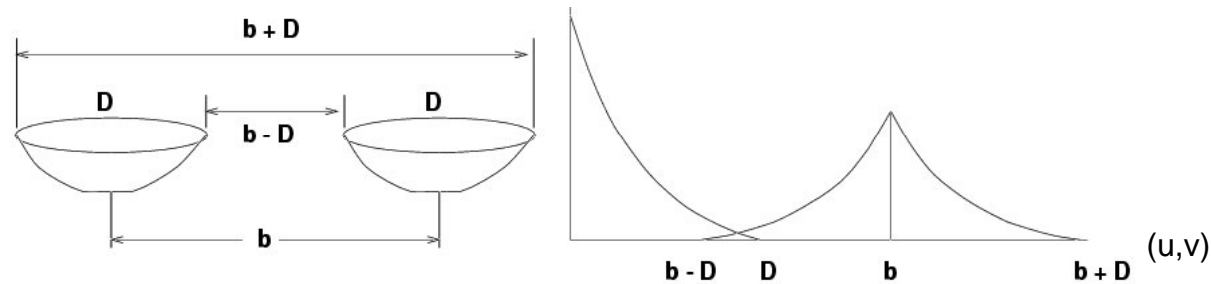
50 x 12 m: 12 m to 14 km

+12 x 7 m: fills 7 to 12 m

+ 4 x 12 m: fills 0 to 7 m

# Low Spatial Frequencies (III)

- mosaic with a homogeneous array
  - recover a range of spatial frequencies around the nominal baseline  $b$  using knowledge of  $A(x,y)$  (Ekers and Rots 1979) (and get shortest baselines from total power maps)



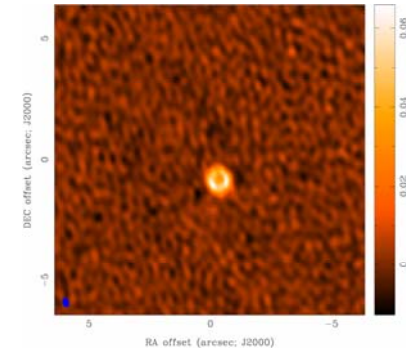
- $V(u,v)$  is linear combination of baselines from  $b-D$  to  $b+D$
- depends on pointing direction  $(x_o, y_o)$  as well as  $(u,v)$
- Fourier transform with respect to pointing direction  $(x_o, y_o)$

$$V(u - u_o, v - v_o) = \frac{\int \int V(u, v; x_o, y_o) e^{2\pi i(u_o x_o + v_o y_o)} dx_o dy_o}{a(u_o, v_o)}$$

# Measures of Image Quality

- “dynamic range”

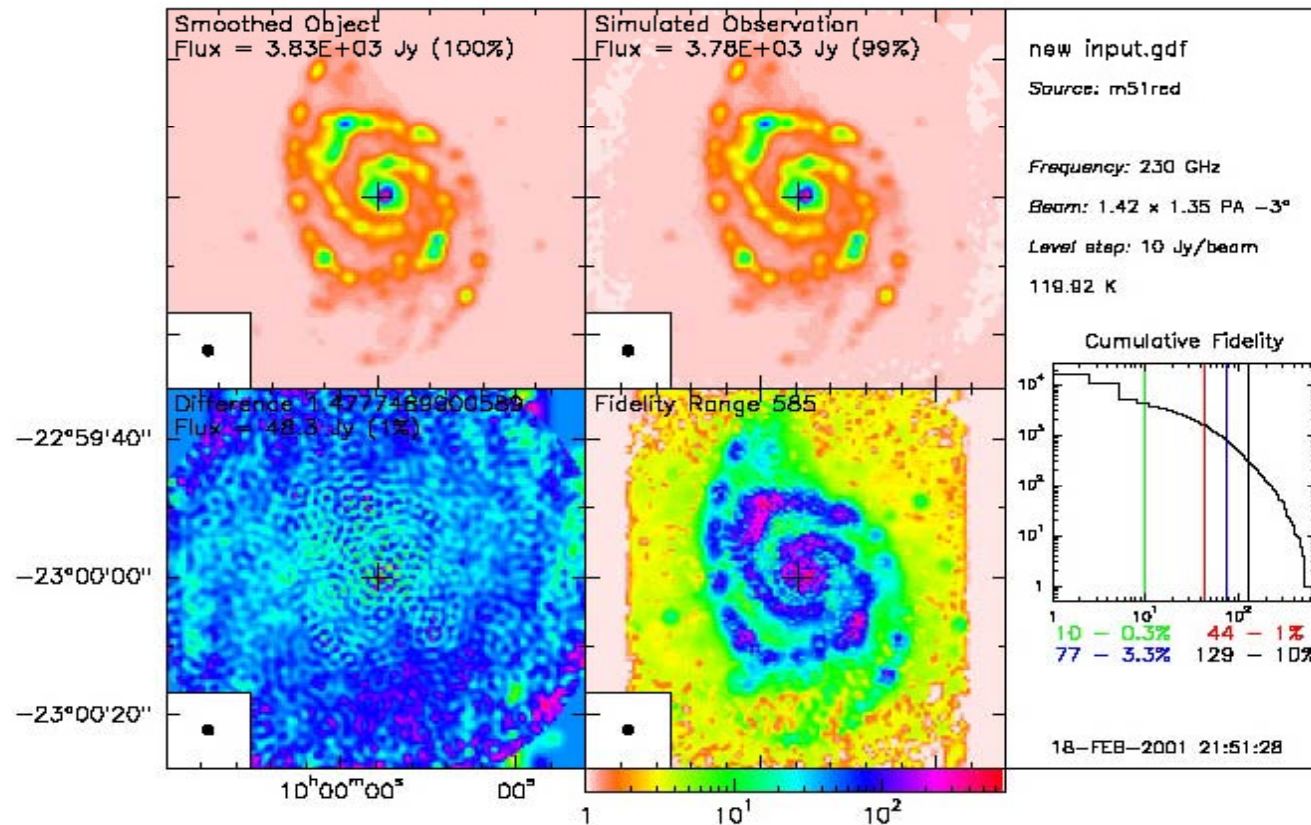
- ratio of peak brightness to rms noise in a region void of emission (common in astronomy)
- an easy to calculate lower limit to the error in brightness in a non-empty region



- “fidelity”

- difference between any produced image and the correct image
- a convenient measure of how accurately it is possible to make an image that reproduces the brightness distribution on the sky
- need a priori knowledge of correct image to calculate
- fidelity image = input model / difference  
=  $\text{model} \otimes \text{beam} / \text{abs}(\text{model} \otimes \text{beam} - \text{reconstruction})$
- fidelity is the inverse of the relative error
- in practice, lowest values of difference need to be truncated

# Measures of Image Quality



ALMA Memo #387  
Pety et al.

- ALMA Level 1 Science Goal #3
  - ALMA will have: The ability to provide precise images at an angular resolution of 0.1". Here the term precise image means accurately representing the sky brightness at all points where the brightness is greater than 0.1% of the peak image brightness.

# Self Calibration

- a priori calibration not perfect
  - interpolated from different time, different sky direction from source
- basic idea of self calibration
  - correct for antenna-based errors *together with imaging*
- works because
  - at each time, measure  $N$  complex gains and  $N(N-1)/2$  visibilities
  - source structure represented by small number of parameters
  - highly overconstrained problem if  $N$  large and source simple
- in practice, an iterative, non-linear relaxation process
  - assume initial model  $\rightarrow$  solve for time dependent gains  $\rightarrow$  form new sky model from corrected data using e.g. CLEAN  $\rightarrow$  solve for new gains...
  - requires sufficient signal-to-noise ratio for each solution interval
- loses absolute phase and therefore position information
- dangerous with small  $N$ , complex source, low signal-to-noise



# Concluding Remarks

- interferometry samples visibilities that are related to a sky brightness image by the Fourier transform
- deconvolution corrects for incomplete sampling
- remember... there are usually an infinite number of images compatible with the sampled visibilities
- astronomer must use judgement in imaging process
- imaging is generally fun (compared to calibration)
- many, many issues not covered today (see References)

# References

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