

Lecture 8: Radio Astronomy - Deconvolution



Images of ALMA from
<https://www.almaobservatory.org/en/images/>

Rutgers Physics 346: Observational Astrophysics
March 23, 2021

Homework for Thursday, Mar. 25

Due: Quiz #8 will appear on Canvas Assignments at 4:40pm, due at noon Wednesday.

Due: Worksheet 2 will appear on Canvas tonight, due at 11:59pm Thursday.

Do: Look over options for Project 2 when they appear on Canvas (will broadcast an Announcement). We'll do signups for Project #2 options at the beginning of Thursday's Project Meeting and then start working. Reminder: if you worked on radio imaging (Projects 1-3) you'll now pursue optical spectroscopy (Projects 7-9). If you worked on optical imaging (Projects 4-6), you'll now pursue radio spectroscopy (Projects 10-12).

Quiz #7: Suppose that an array of telescopes has 8 antennas. What is the maximum number of distinct baselines a configuration of this array could deliver?

A: An antenna cannot create a baseline with itself, and each pair should count only once, so the answer is $8 \times 7 / 2 = 28$ baselines. This can also be calculated as combinations of 2 out of 8 = $8! / (6!)(2!)$.

Aperture synthesis: summary

- any pair of antennas in an array corresponds to a **baseline B** , defined by a length and an orientation
- when viewed “from the point of the source” at a particular time, this becomes a **projected baseline B_p**
- the x and y components of a projected baseline and the observing wavelength λ determine the dimensionless coordinates **$u = B_{p,x}/\lambda$** and **$v = B_{p,y}/\lambda$**
- a given (u,v) pair defines a two-dimensional **spatial frequency**; the power measured at that (u,v) corresponds to how “ripply” the sky brightness distribution is at a particular direction and periodicity

Aperture synthesis: summary

- when we have many pairs of antennas (i.e., baselines) and make measurements at many different times, we can measure power at many spatial frequencies
- **angular resolution $\sim \lambda/B_{p,max}$** – so longer baselines (and shorter wavelengths) give higher resolution!
- **largest angular scale $\sim \lambda/B_{p,min} \geq \lambda/D$** (for D = diameter of an antenna) – so smaller antennas do a better job of recovering smoothly extended emission!
- if the sky brightness distribution includes emission that is larger than λ/D , the interferometer can't see it

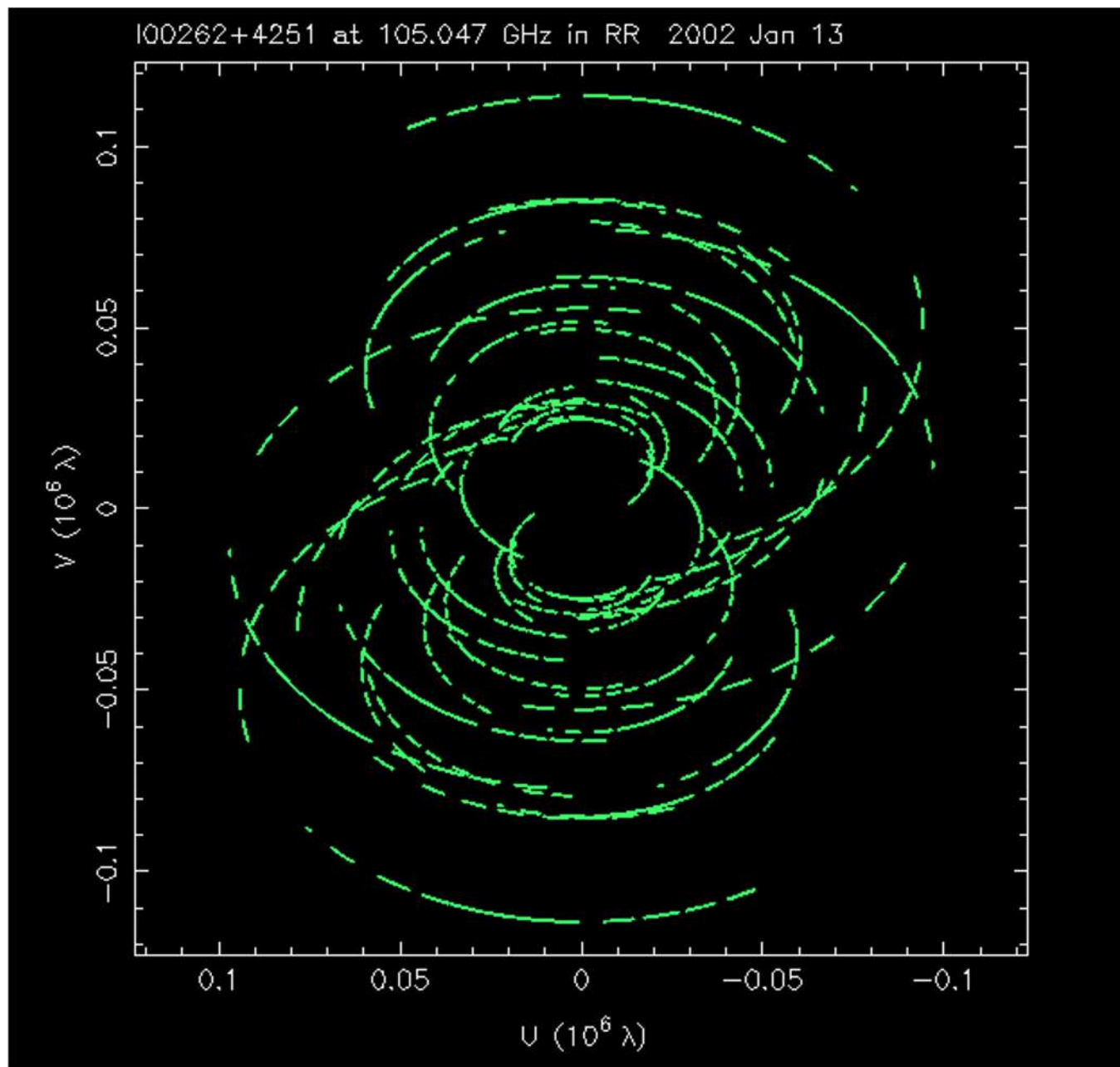
Example of uv sampling

Each arc is produced by a different baseline as the Earth rotates under the source.

Longer baselines reach larger (u,v) .

Central gap due to nonzero antenna diameter.

More sampling is always better!



A crucially important Fourier pair

The complex number that is the response of a (projected) baseline with a given (u,v) to a given sky brightness distribution* $J(l,m)$ (where l and m are small angles on the sky in x and y directions) is the **visibility** $V(u,v)$.

In an ideal world, we'd know $V(u,v)$ at all possible $\{(u,v)\}$, and we could calculate $J(l,m)$ with a simple 2D Fourier transform: (remember $u = B_{p,x}/\lambda$ and $v = B_{p,y}/\lambda$)

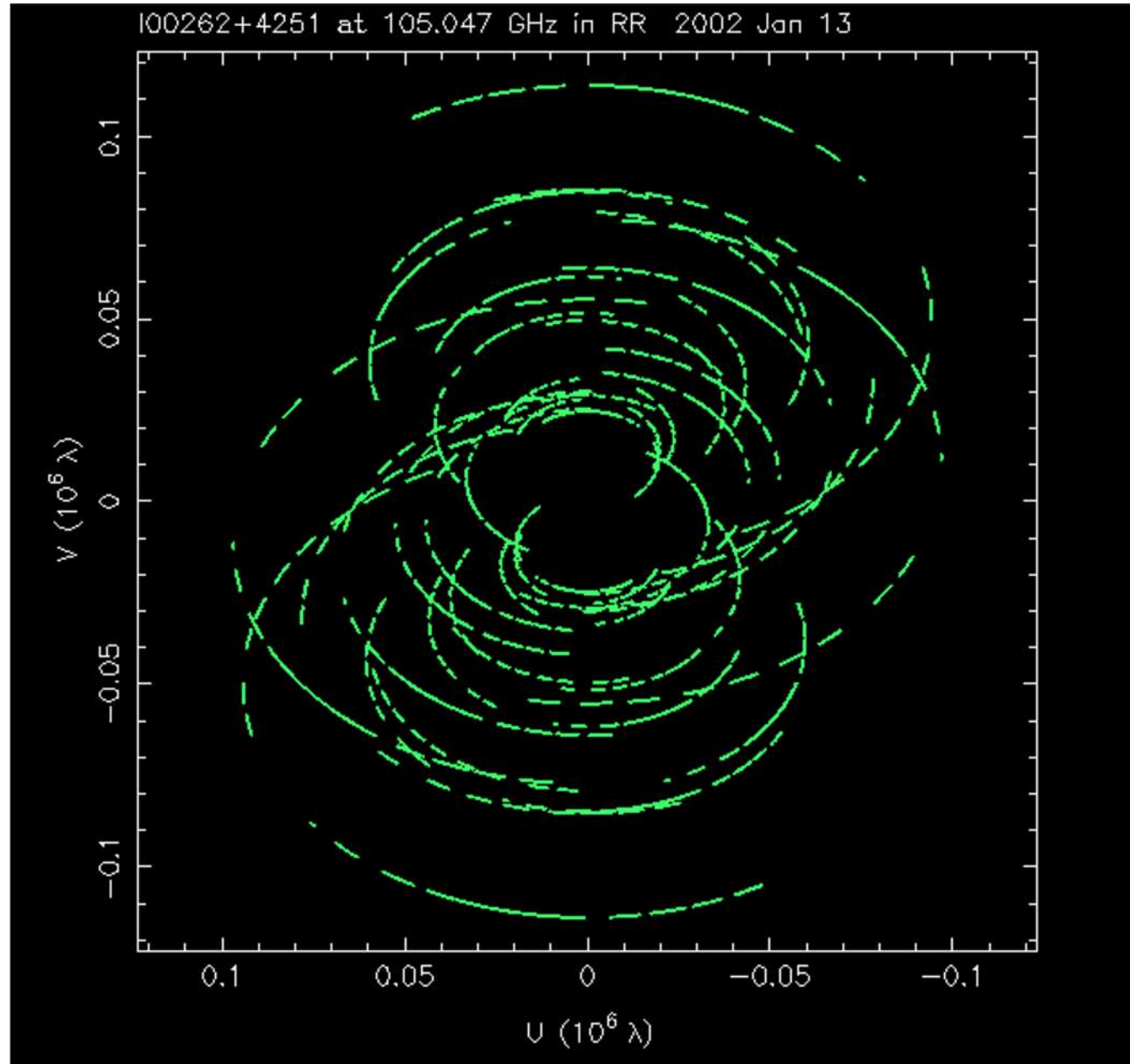
$$J(l,m) = \iint V(u,v) e^{ul+vm} du dv$$

In the real world, however, we don't have measurements of V at all possible $\{(u,v)\}$ – and this is a problem!

*Sky brightness distribution usually written as capital “ I ”...

Incomplete uv sampling

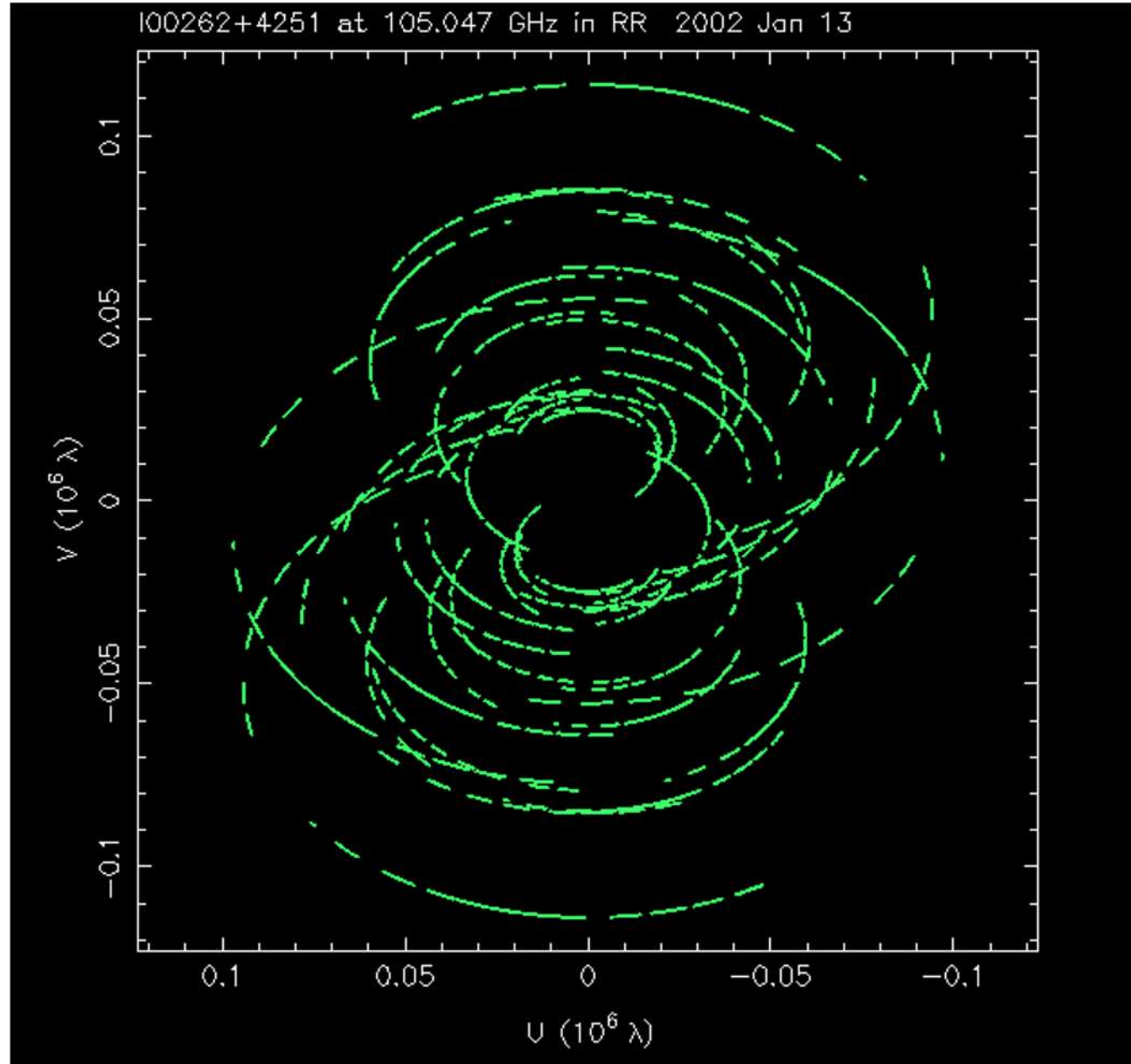
Question: how many possible sky brightness distributions are consistent with a given set of visibilities measured at the $\{(u,v)\}$ represented by the green points?



Incomplete uv sampling

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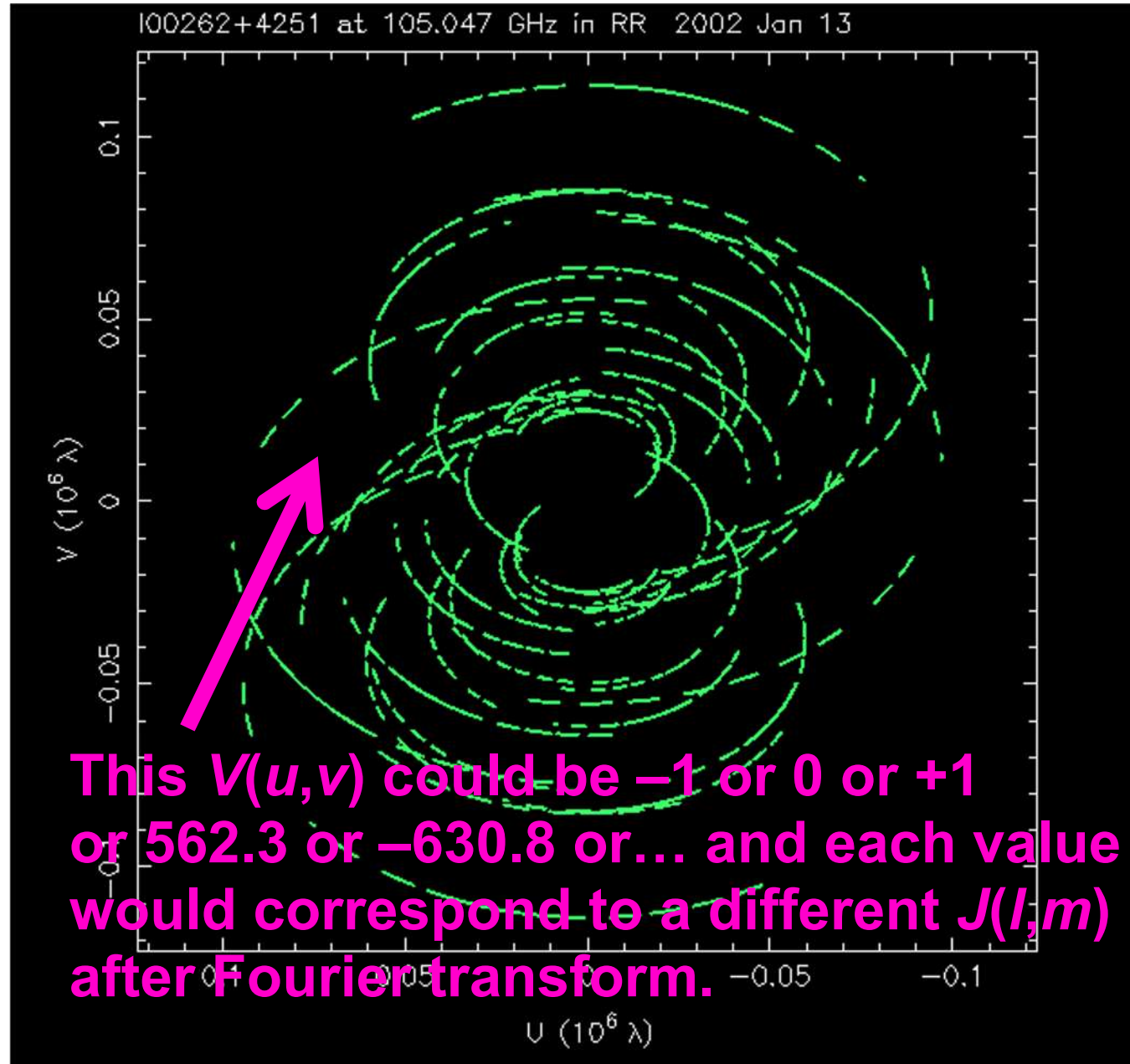
Answer:
infinitely many!



Incomplete uv sampling

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Answer:
infinitely many!



Incomplete uv sampling

Instead of being in the ideal world where we know $V(u,v)$ for all $\{(u,v)\}$ and obtain the sky brightness distribution as

$$J(l,m) = \iint V(u,v) e^{ul+vm} du dv$$

what we actually know in the real world is

$$J_D(l,m) = \iint V(u,v) S(u,v) e^{ul+vm} du dv$$

where $S(u,v)$ is the “sampling function” and has a value of
1 at all $\{(u,v)\}$ where we have data
0 at all $\{(u,v)\}$ where we do not have data

$J_D(l,m)$ is referred to as the “dirty image” or “dirty map.”

Incomplete uv sampling

Let's look a bit more closely at this equation:

$$J_D(l,m) = \iint V(u,v) S(u,v) e^{ul+vm} du dv \quad \text{the “dirty map”}$$

This is effectively a convolution:

$$J_D(l,m) = J(l,m) * B(l,m)$$

in terms of

$$J(l,m) = \iint V(u,v) e^{ul+vm} du dv \quad \text{sky brightness distribution}$$

$$B(l,m) = \iint S(u,v) e^{ul+vm} du dv \quad \text{the “dirty beam”}$$

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$J_D(l,m) = J(l,m) * B(l,m)$ we can calculate this

in terms of $J(l,m)$ we want this

$B(l,m)$ we have measured this

$$J(l,m) = \iint V(u,v) e^{ul+vm} du dv \quad \text{sky brightness distribution}$$

$$B(l,m) = \iint S(u,v) e^{ul+vm} du dv \quad \text{the “dirty beam”}$$

Convolution and deconvolution

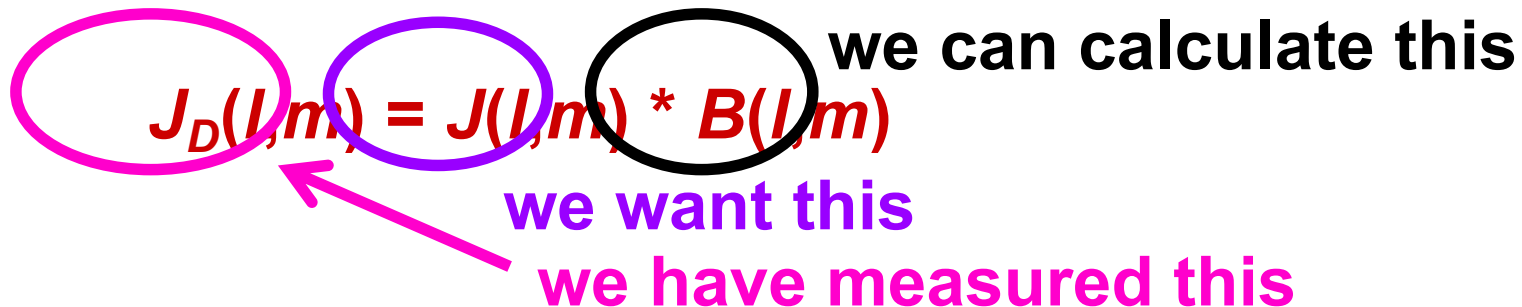
Recall from before: convolution is “smearing” two functions together.

Deconvolution is the process of “unsmearing” the **dirty beam** from the **dirty map** in order to estimate the **sky brightness distribution**.

$J_D(l,m) = J(l,m) * B(l,m)$ we can calculate this

we want this

we have measured this

The diagram shows the equation $J_D(l,m) = J(l,m) * B(l,m)$. The term $J_D(l,m)$ is circled in pink, with a pink arrow pointing to it from the text "we have measured this" below. The term $J(l,m)$ is circled in purple, with the text "we want this" below it. The term $B(l,m)$ is circled in black, with the text "we can calculate this" to its right. The asterisk $*$ is not circled.

The challenge: since there are infinitely many possible sky brightness distributions consistent with measured visibility data and the dirty map $J_D(l,m)$, we have to assume a (Bayesian) prior to guide deconvolution.

image credit: D. Gary

Illustration of dirty map and beam

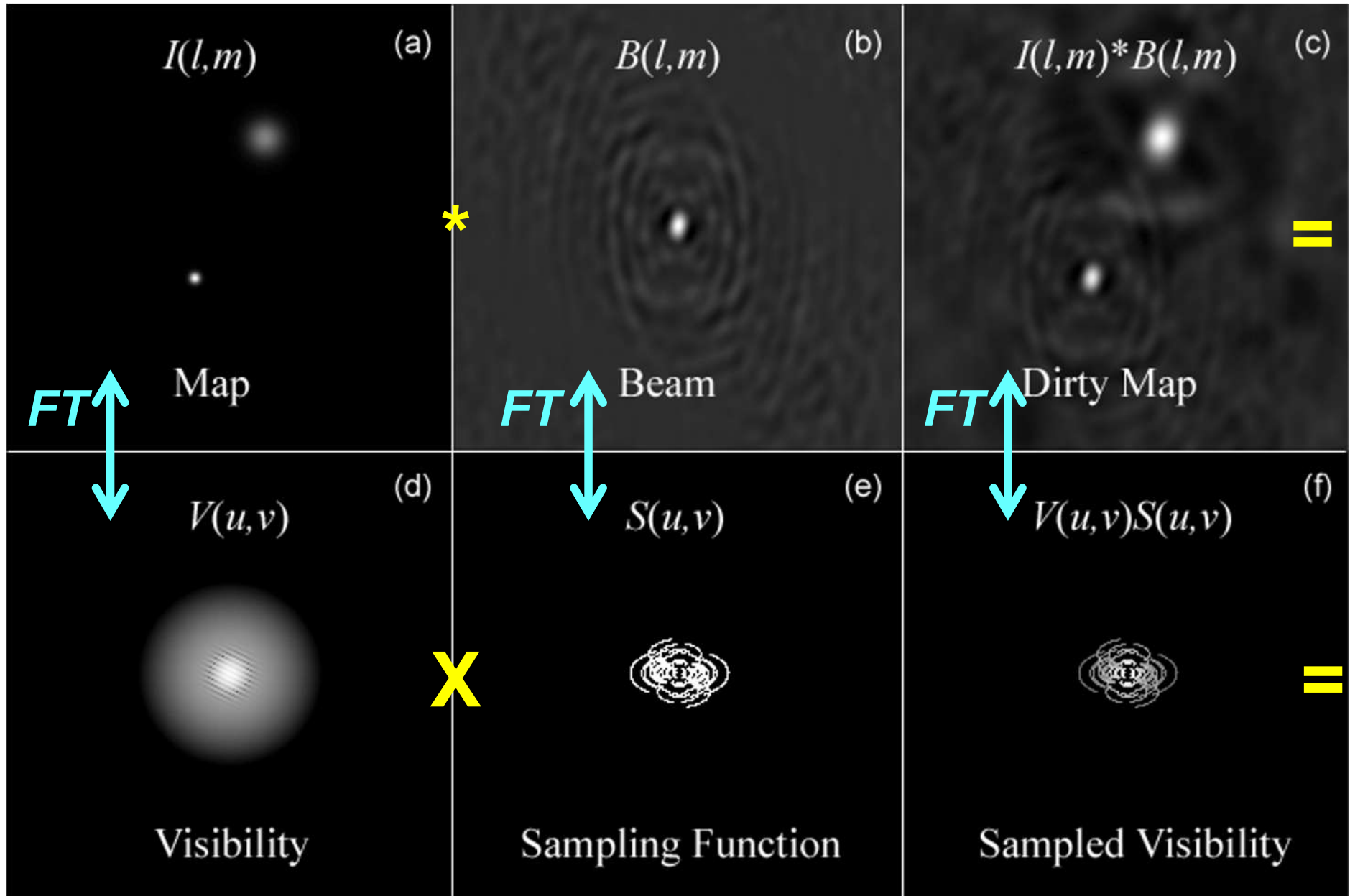


image credit: D. Gary

Illustration of dirty map and beam

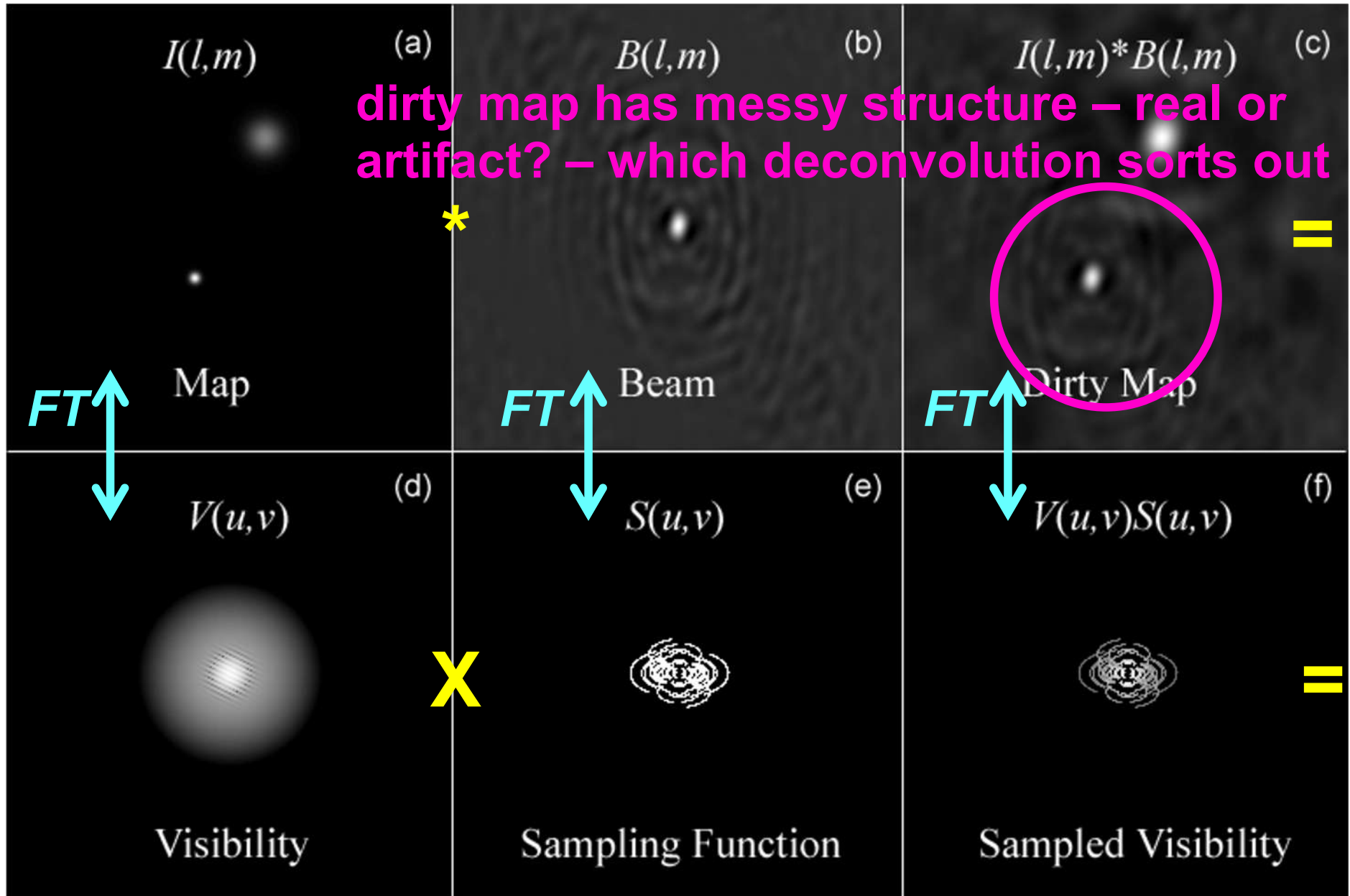
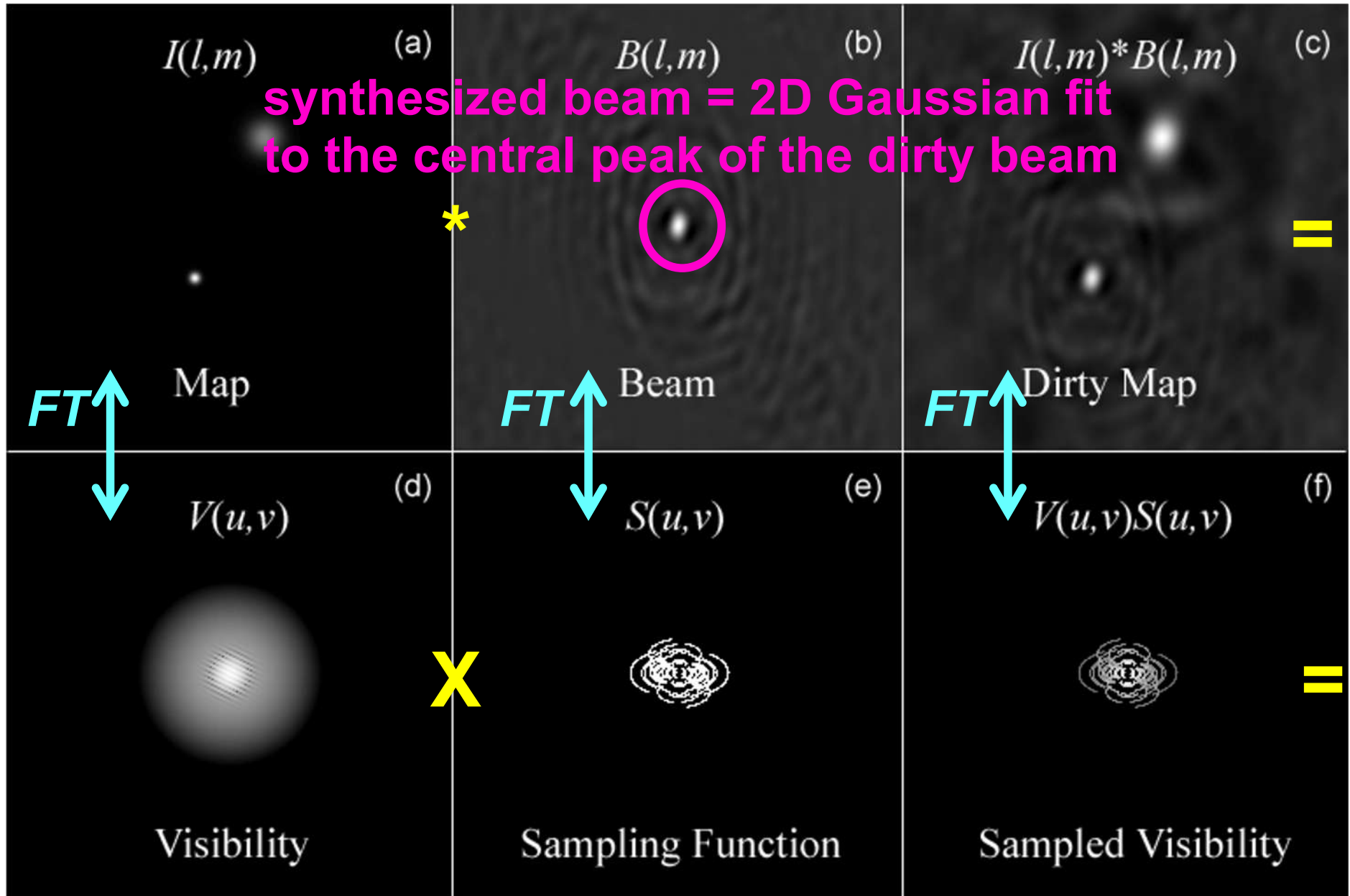


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The challenge: since there are infinitely many possible sky brightness distributions consistent with measured visibility data and the dirty map $J_D(l,m)$, **we have to assume a (Bayesian) prior to guide deconvolution.**

Possible prior # 1: smoothness

One possible prior is that the sky brightness distribution is **maximally smooth** (no sharp edges, etc.).

This prior guides the **maximum entropy method (MEM)** algorithm for deconvolution:

- + leads to reconstructed images that are smooth...
good for imaging structures that extend over wide areas on the sky, but not so good for compact structures
- + algorithm also tends to run away and keep adding more and more emission to reconstructed image, unless a constraint on total flux is available

Possible prior # 2: point sources

Another possible prior is that the sky brightness distribution is well described as a **sum of point sources**.

This prior guides the **CLEAN** algorithm for deconvolution:

- (1) identify the peak pixel remaining in the dirty map
- (2) take the 2D dirty beam, shift it to be centered on the peak pixel, and rescale it to 5–10% of the peak pixel height
- (3) subtract the shifted/scaled dirty beam from the dirty map, and record the corresponding delta function
- (4) are the residuals only noise? **if NO**: go to step (1) and repeat; **if YES**: go to step (5)
- (5) define “clean image” as sum of all delta functions (convolved with synthesized beam) and final residuals

More about CLEAN

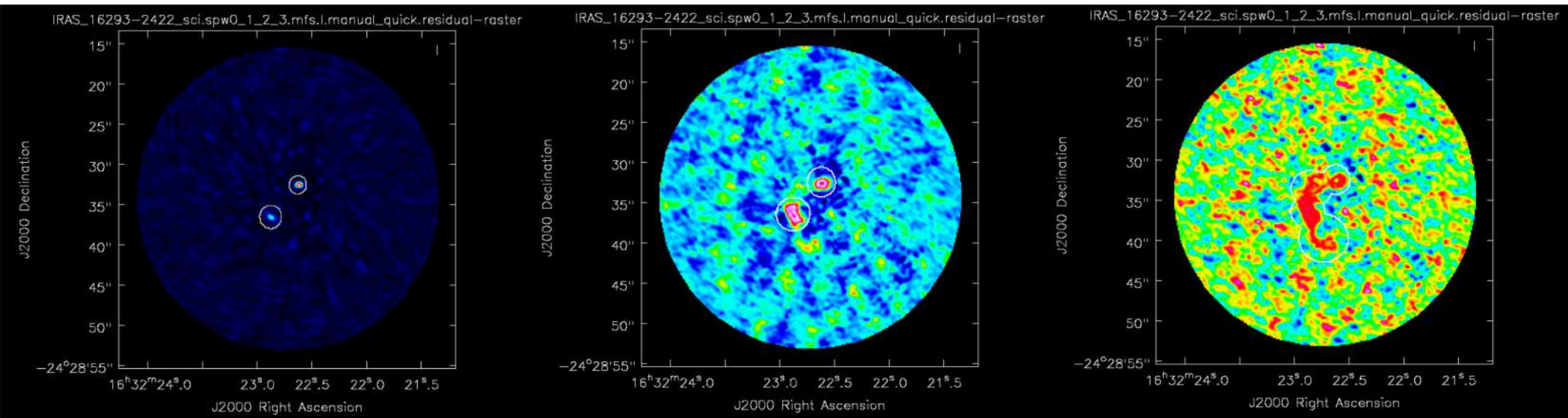
The CLEAN algorithm works quite well for sky brightness distributions that **are** well described in terms of point sources –but not so well for smooth, extended emission. (Fancier algorithms like “multi-scale CLEAN” can help if both point source and extended emission are present.)

One big question: when do we stop repeating steps (1) through (4)?

- + It is possible to **underclean** – then there will be artifacts of the dirty beam left in your “clean” map.
- + It is possible to **overclean** – then some of the “clean components” included in your clean map will actually be noise peaks.
- + This question and other details of CLEANing lead to experimentation and arguments among astronomers...

image credit: ALMA

CLEAN in progress



Left: initial dirty map

Middle: residuals after two clean cycles

+ emission peaks now lower relative to noise

Right: residuals after many clean cycles

**+ low-level extended emission is now visible,
motivating use of larger “clean boxes” in which
to subtract clean components**

Visibility weighting

The 2D Fourier transform of the visibility data is most easily calculated on a regular grid of points – so one of the key steps in processing aperture synthesis data is regridding.

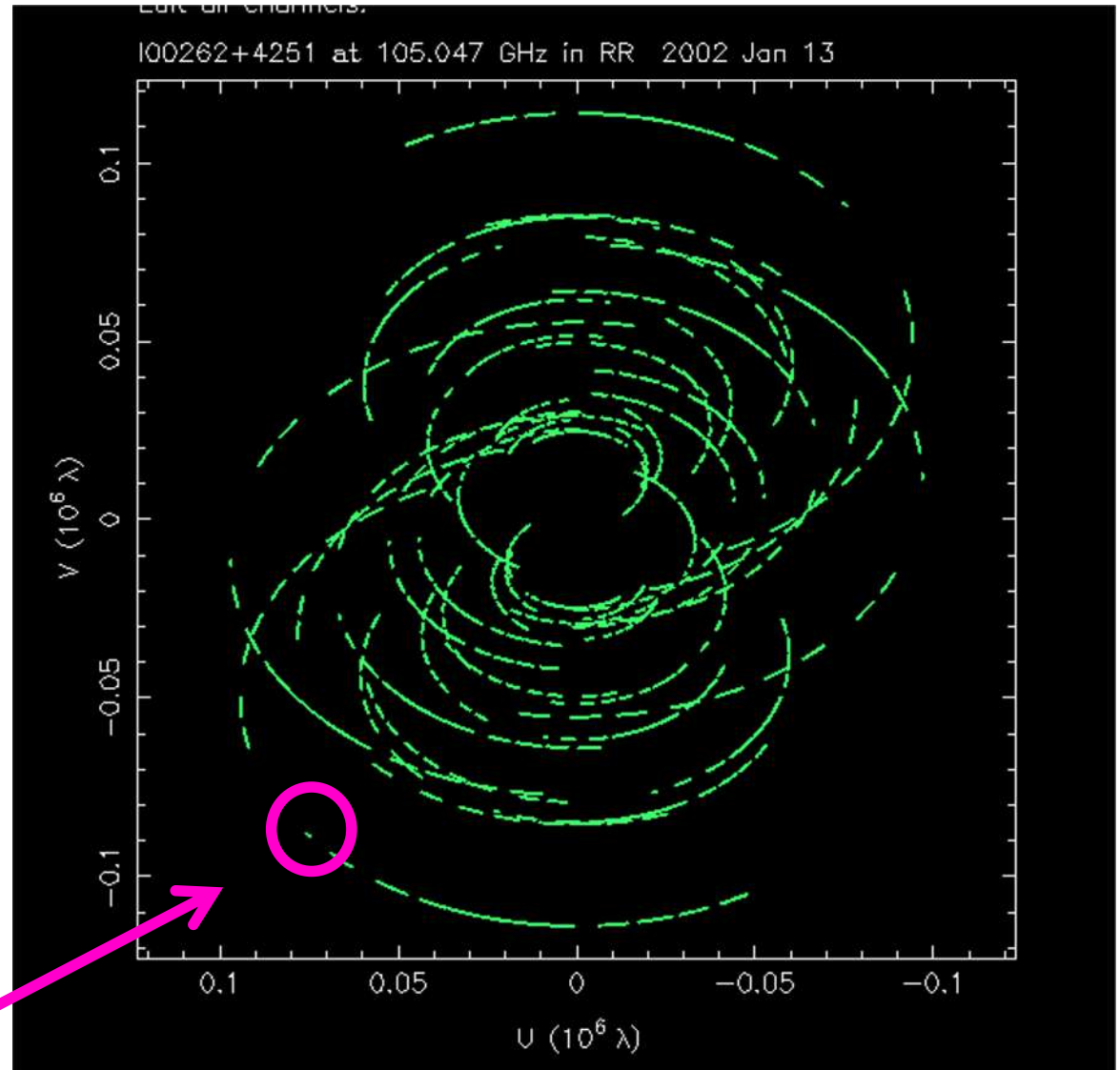
A question now arises: how do we **weight** the individual visibility measurements when regridding the data?

- (1) better quality data should get more weight – so we almost always weight visibilities as $(T_{\text{sys}})^{-2}$ where system temperature is a measure of noise
- (2) we can also think about whether we want to weight visibilities based on their **density**, i.e., how many other visibilities are measured in the same part of the uv plane

Visibility weighting

“Natural” weighting:
every visibility
gets same
density weight
+ optimizes S/N,
limits resolution

“Uniform” weighting:
each visibility is
weighted inversely
proportional to
local density
+ optimizes resolution,
limits S/N

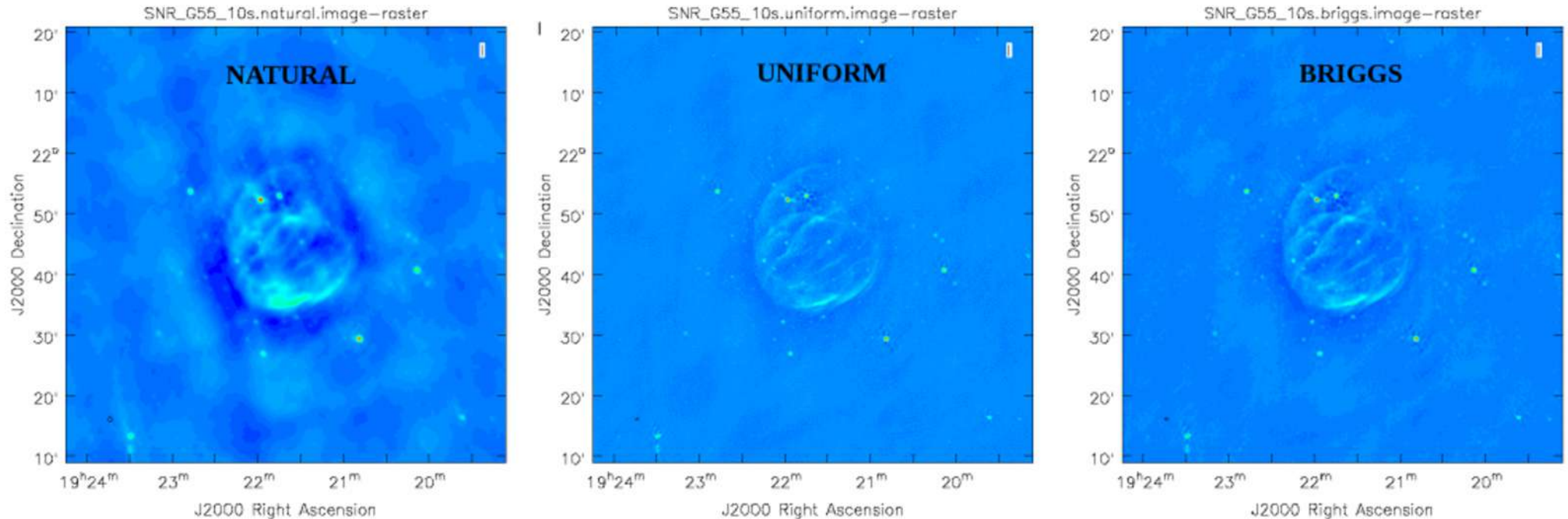


would get high weight for uniform weighting – and more
emphasis on long baselines means higher resolution

image credit: NRAO

Visibility weighting

The possibility of different weighting schemes means that we can use the **same** visibility data to produce (and then deconvolve) **multiple** dirty maps!



“Briggs” weighting is intermediate between natural and uniform (named after the late Dan Briggs).