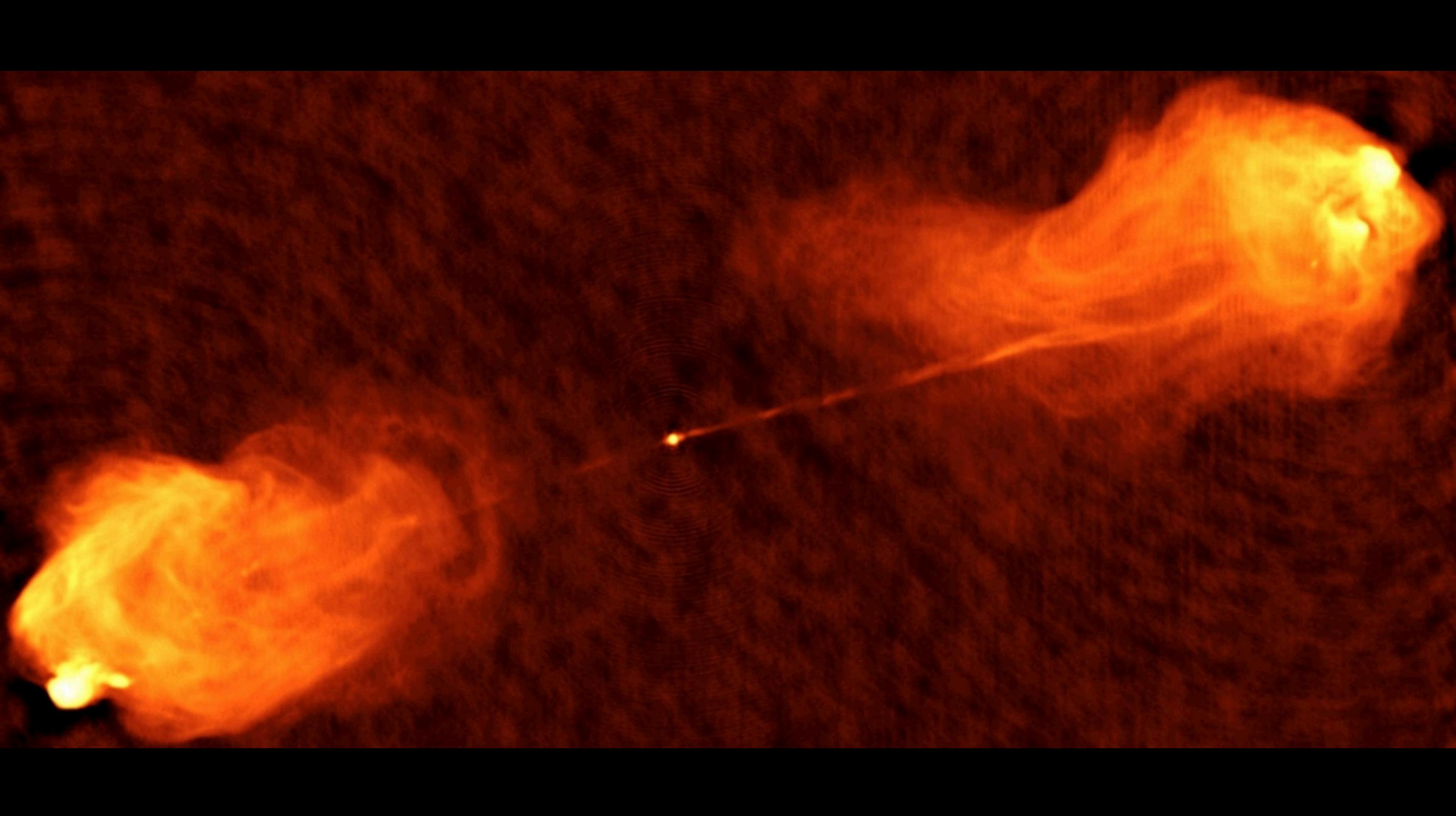


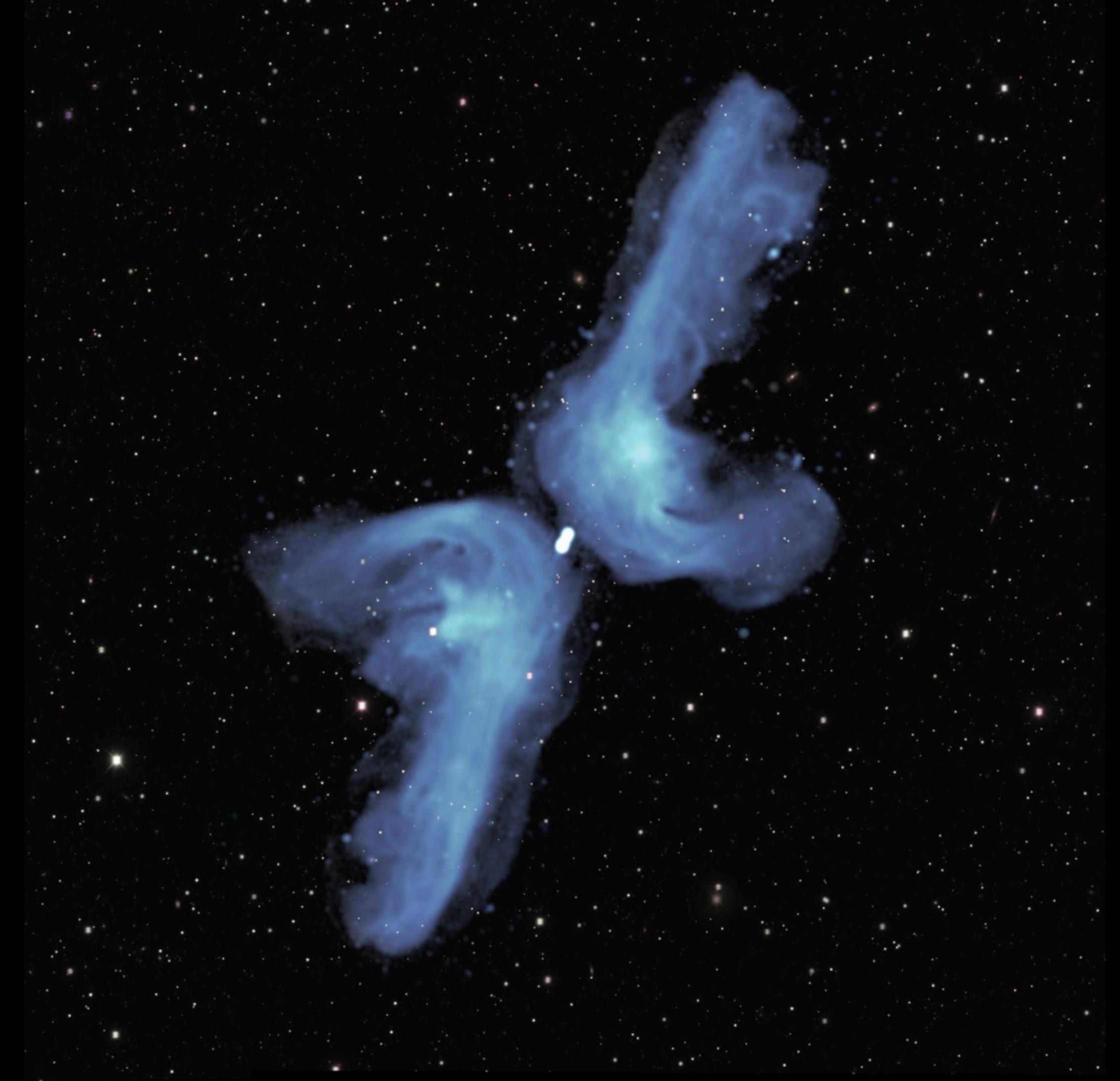


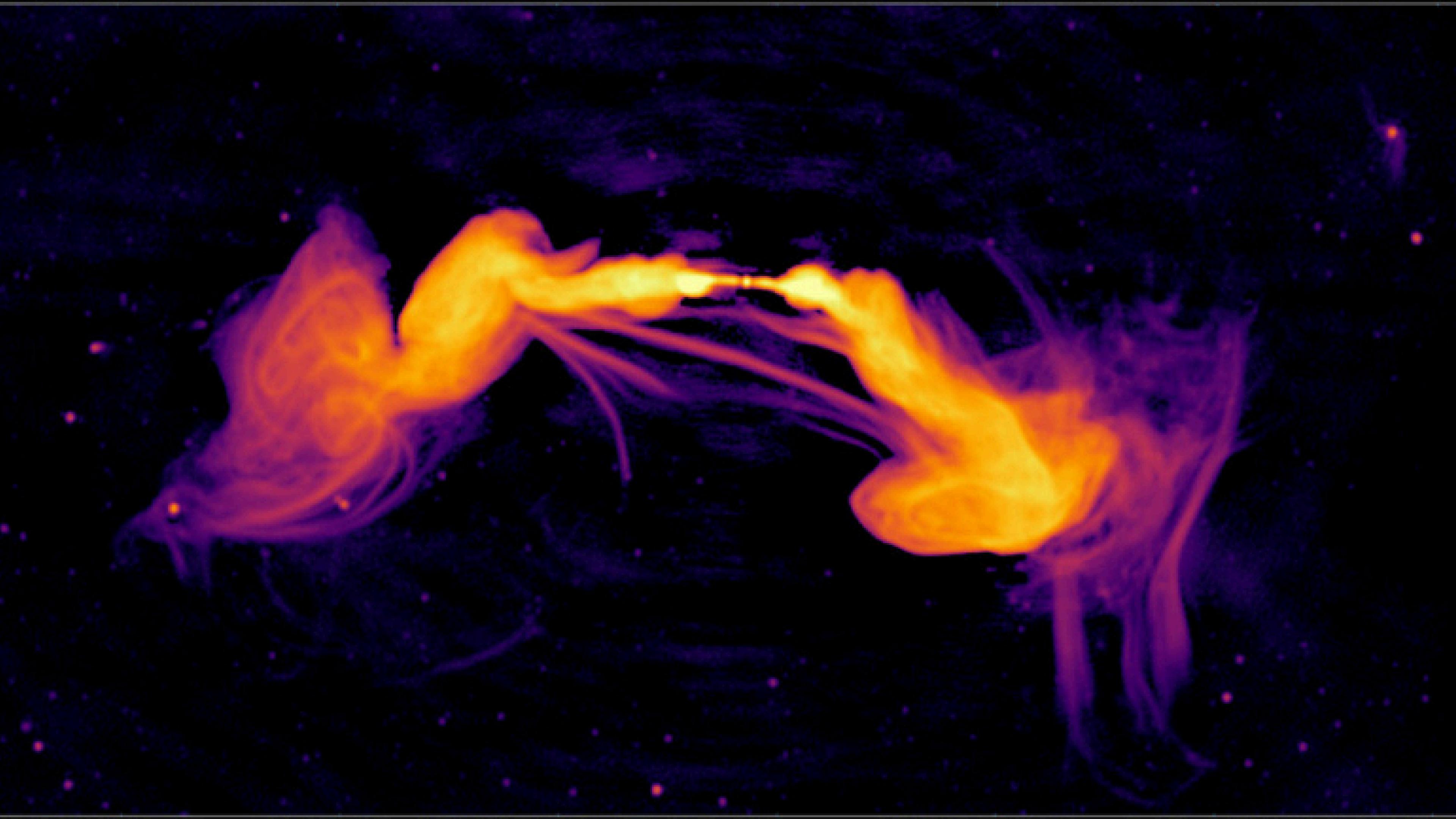
Deloitte.

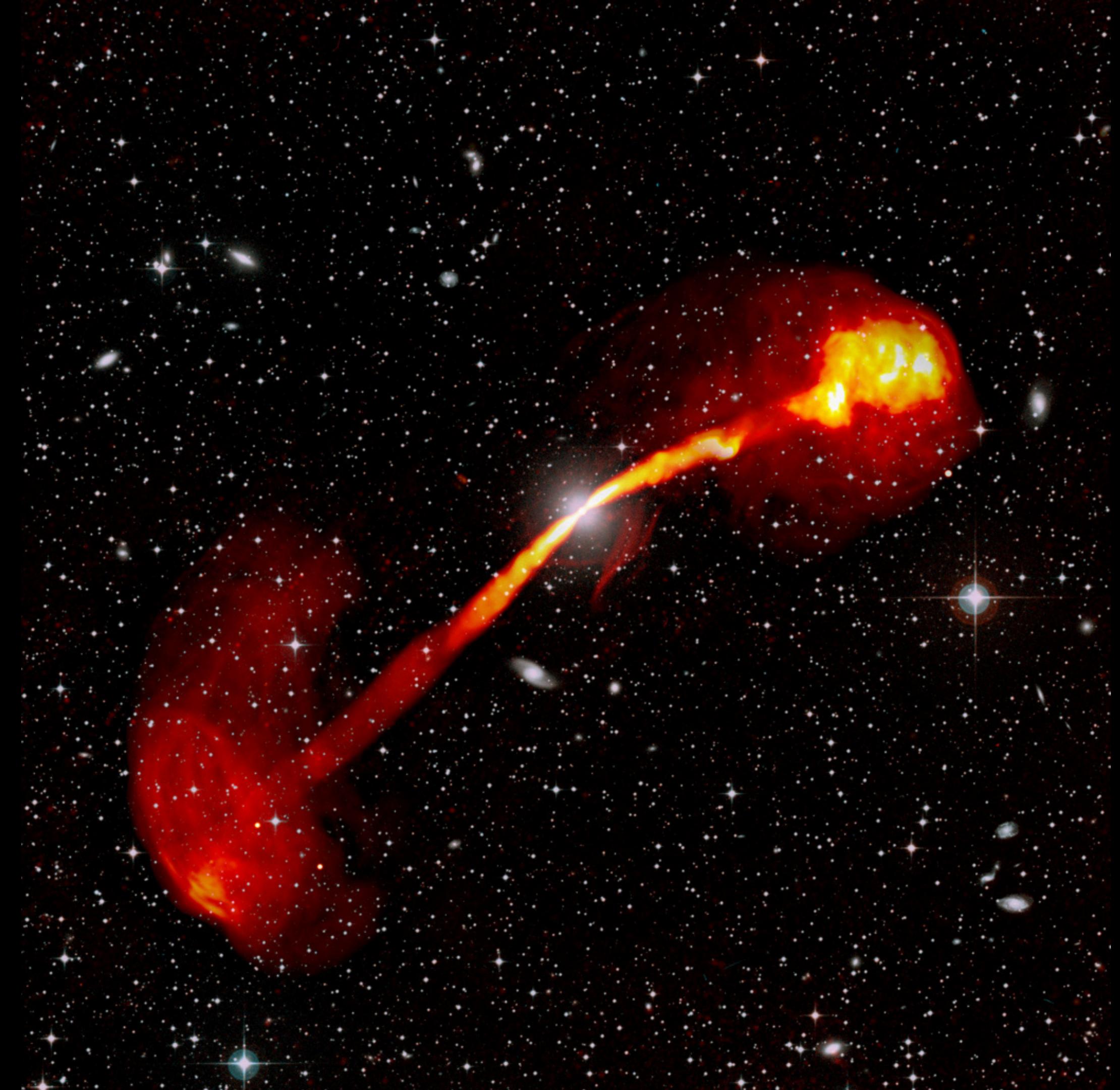


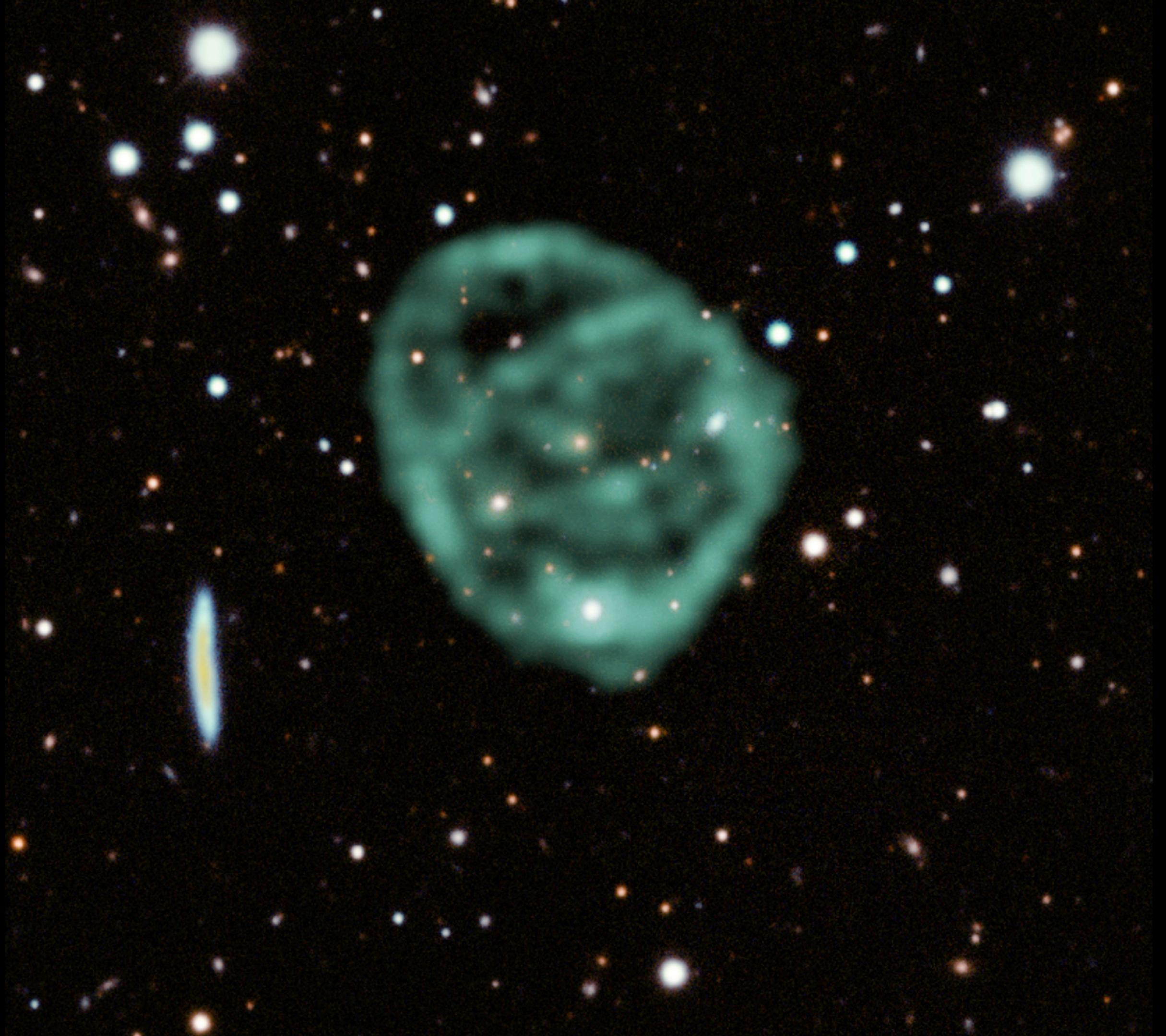
IMAGING



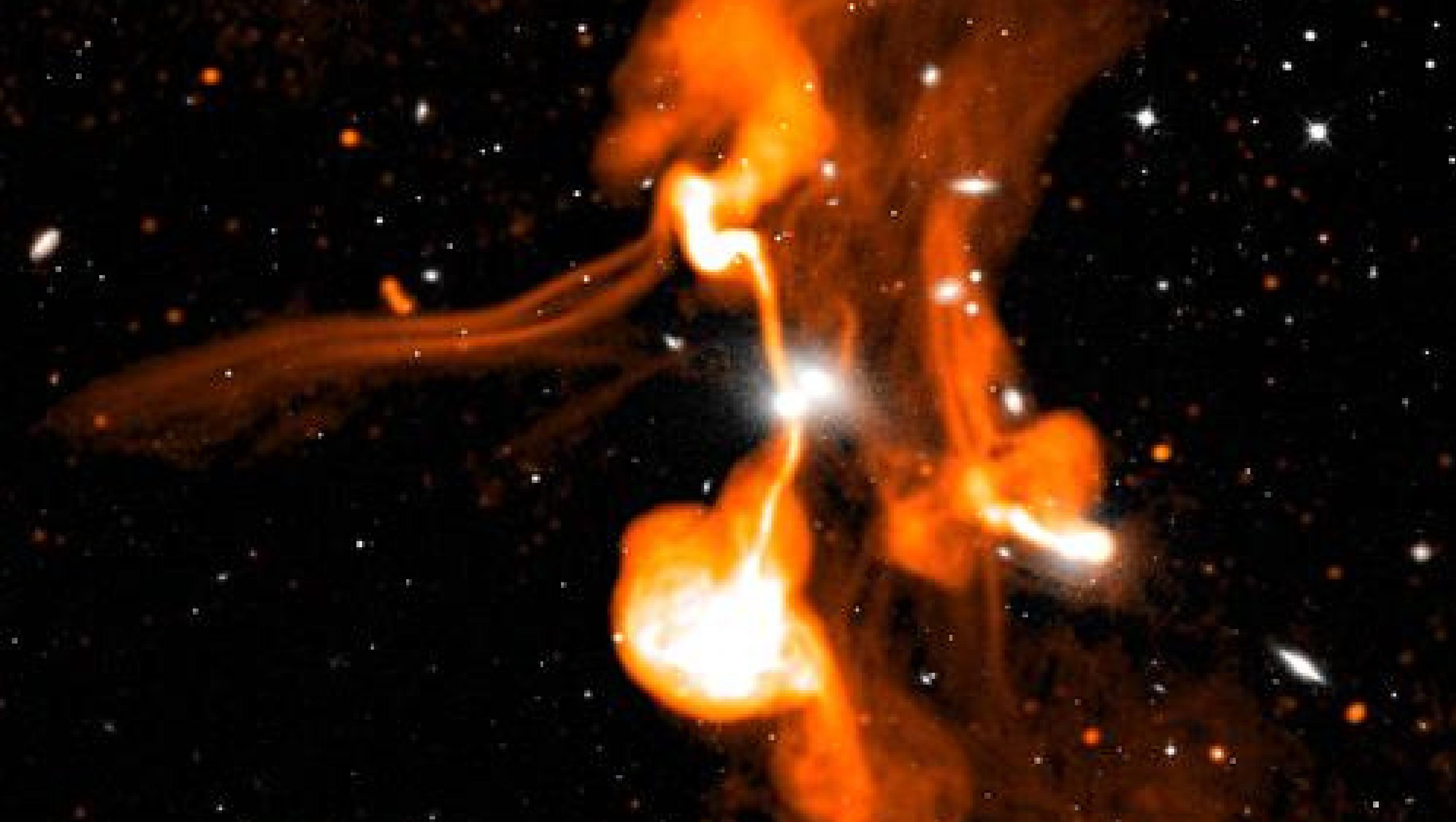






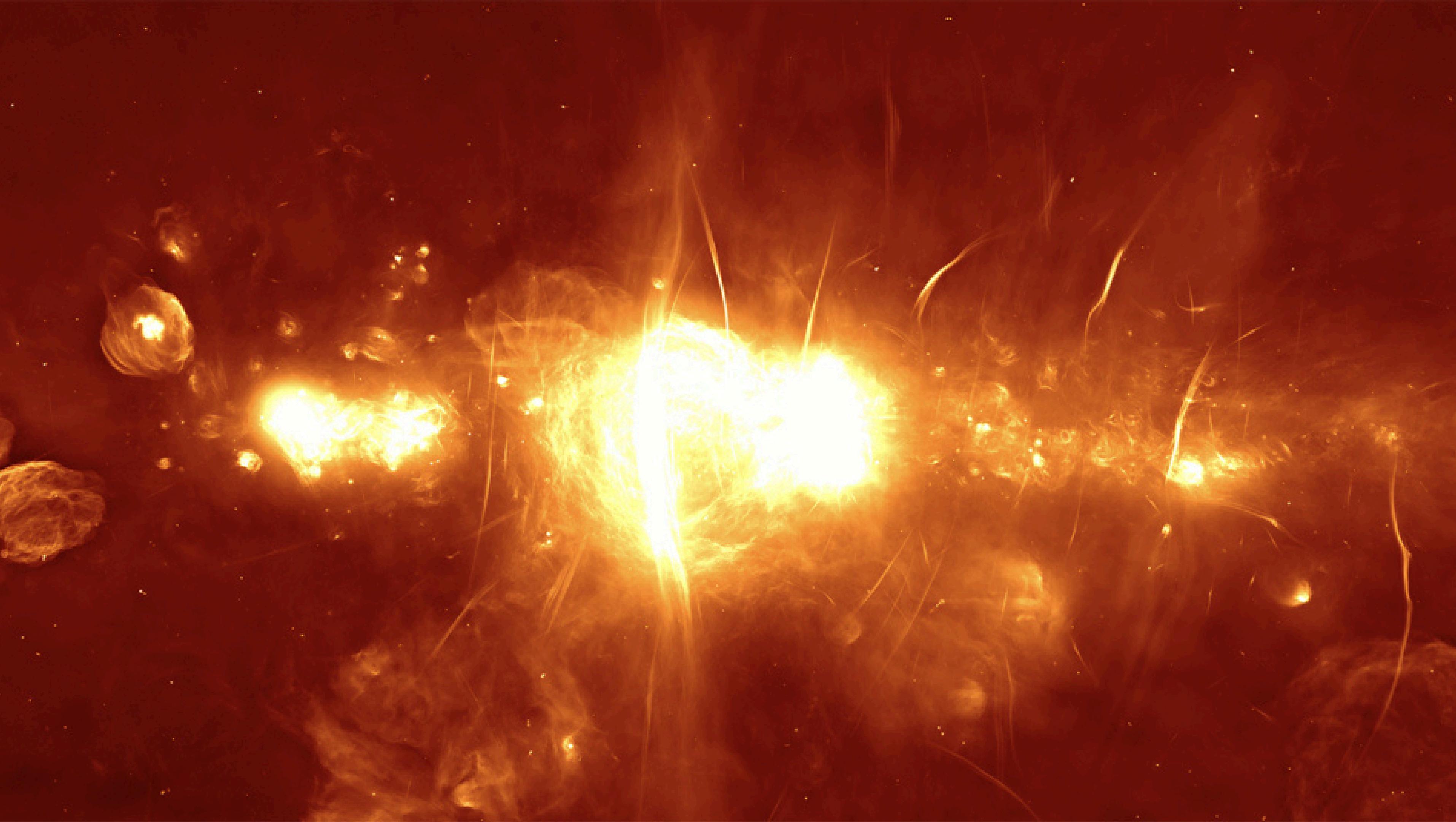








“From The Cradle To The Grave” (Ben Hugo, SARA0 & Rhodes)



MOTIVATION FOR INTERFEROMETRY :

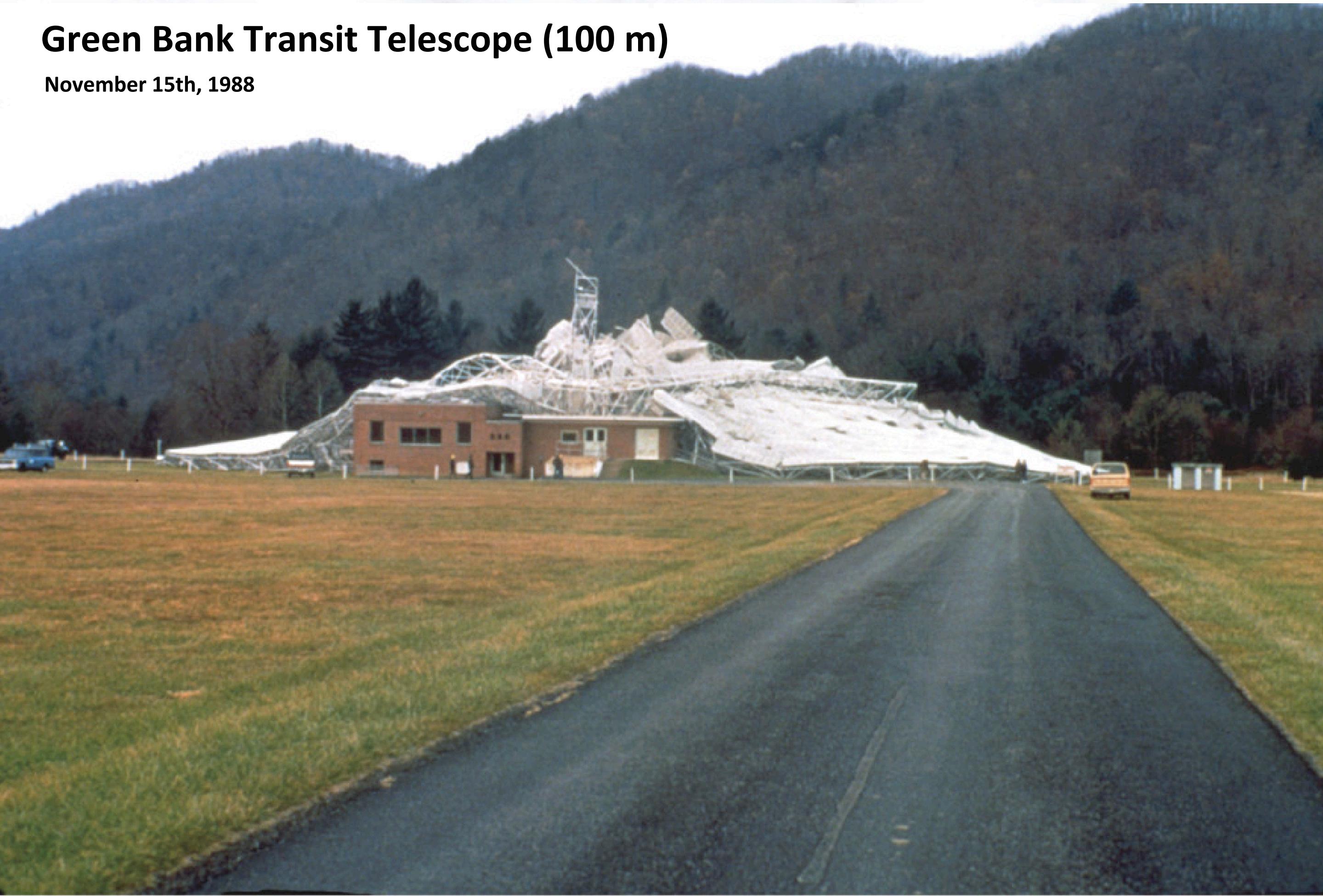
Green Bank Transit Telescope (100 m)



MOTIVATION FOR INTERFEROMETRY :

Green Bank Transit Telescope (100 m)

November 15th, 1988



MOTIVATION FOR INTERFEROMETRY :

Green Bank Transit Telescope (100 m)

November 15th, 1988

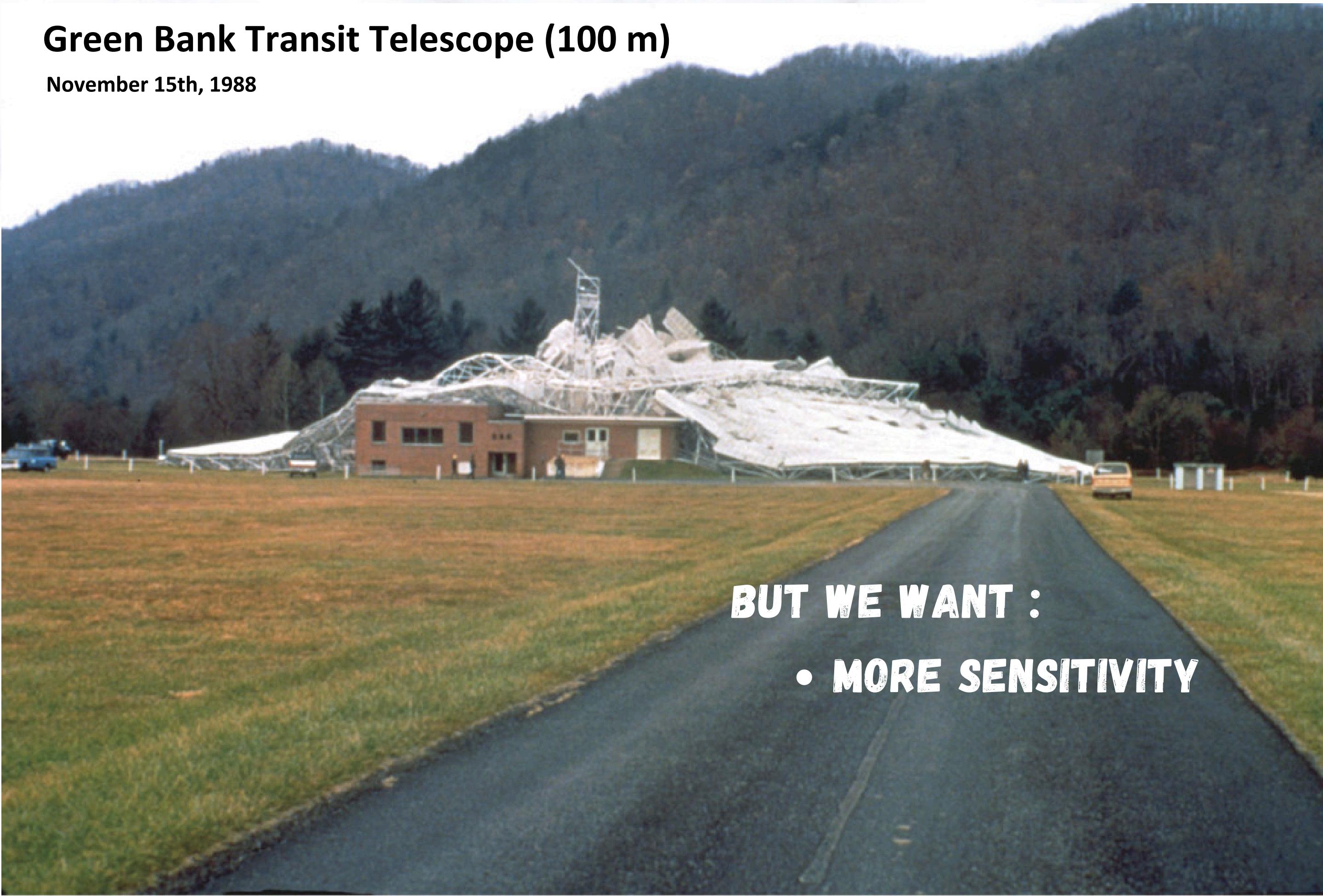


BUT WE WANT :

MOTIVATION FOR INTERFEROMETRY :

Green Bank Transit Telescope (100 m)

November 15th, 1988



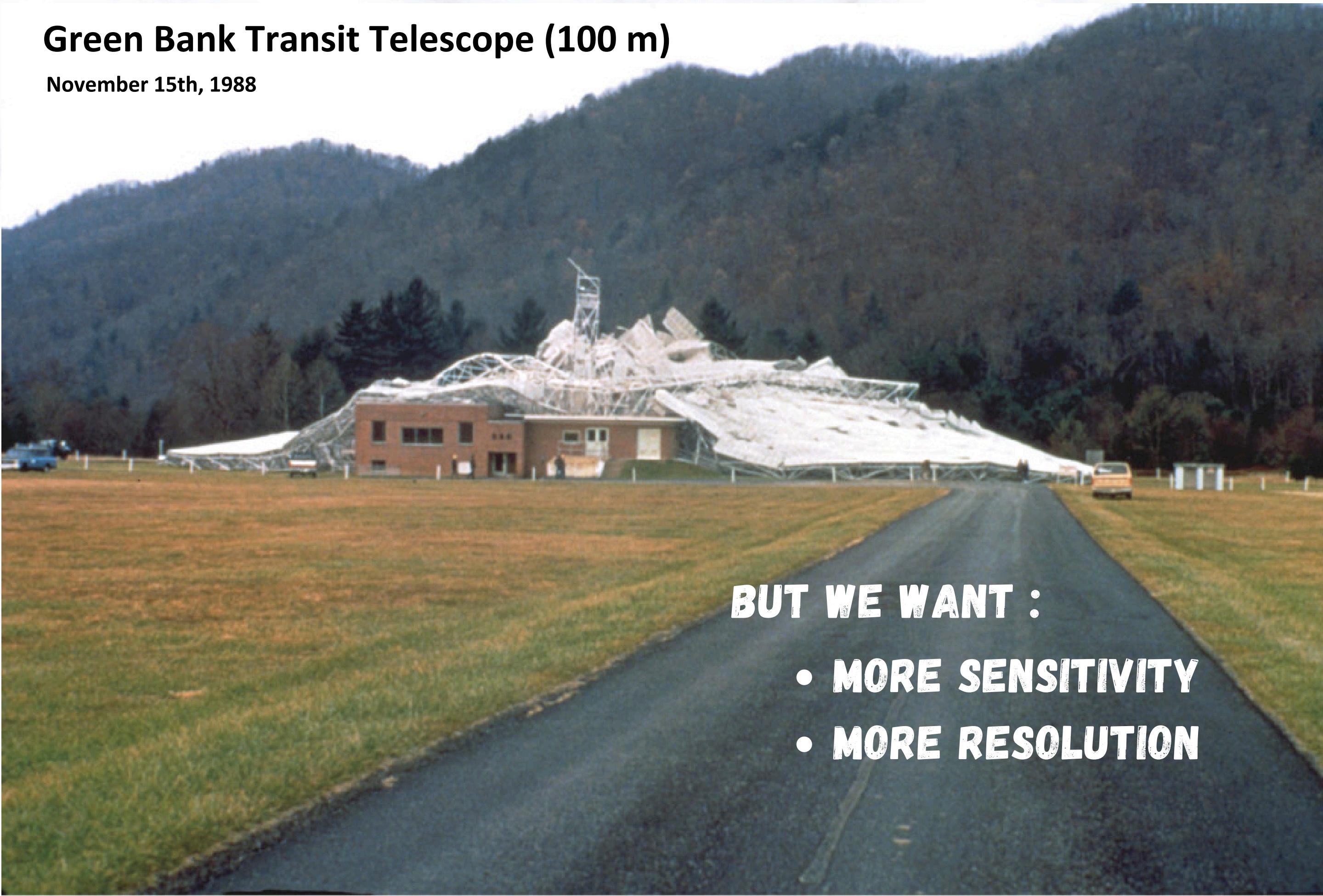
BUT WE WANT :

- **MORE SENSITIVITY**

MOTIVATION FOR INTERFEROMETRY :

Green Bank Transit Telescope (100 m)

November 15th, 1988



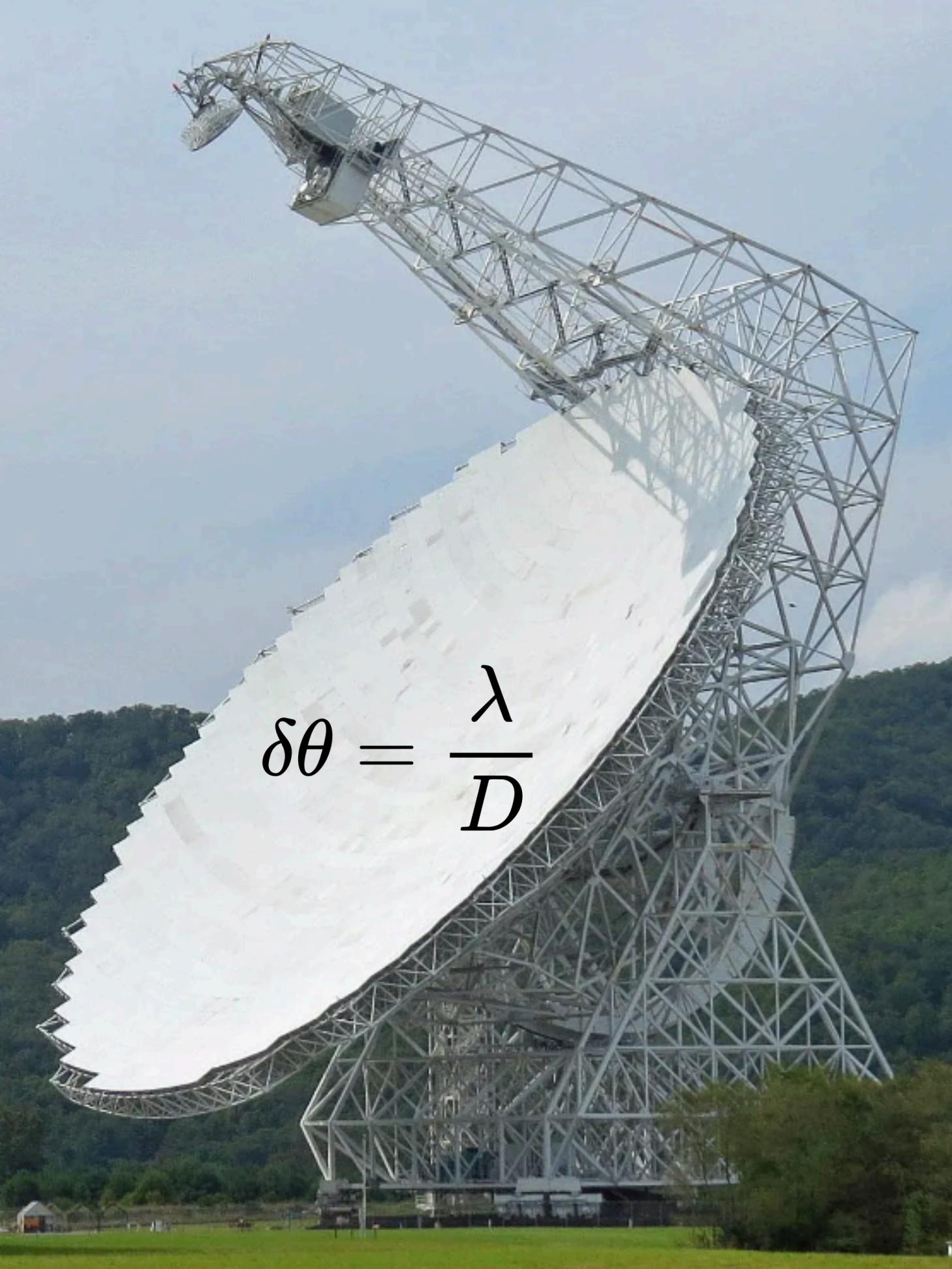
BUT WE WANT :

- MORE SENSITIVITY
- MORE RESOLUTION



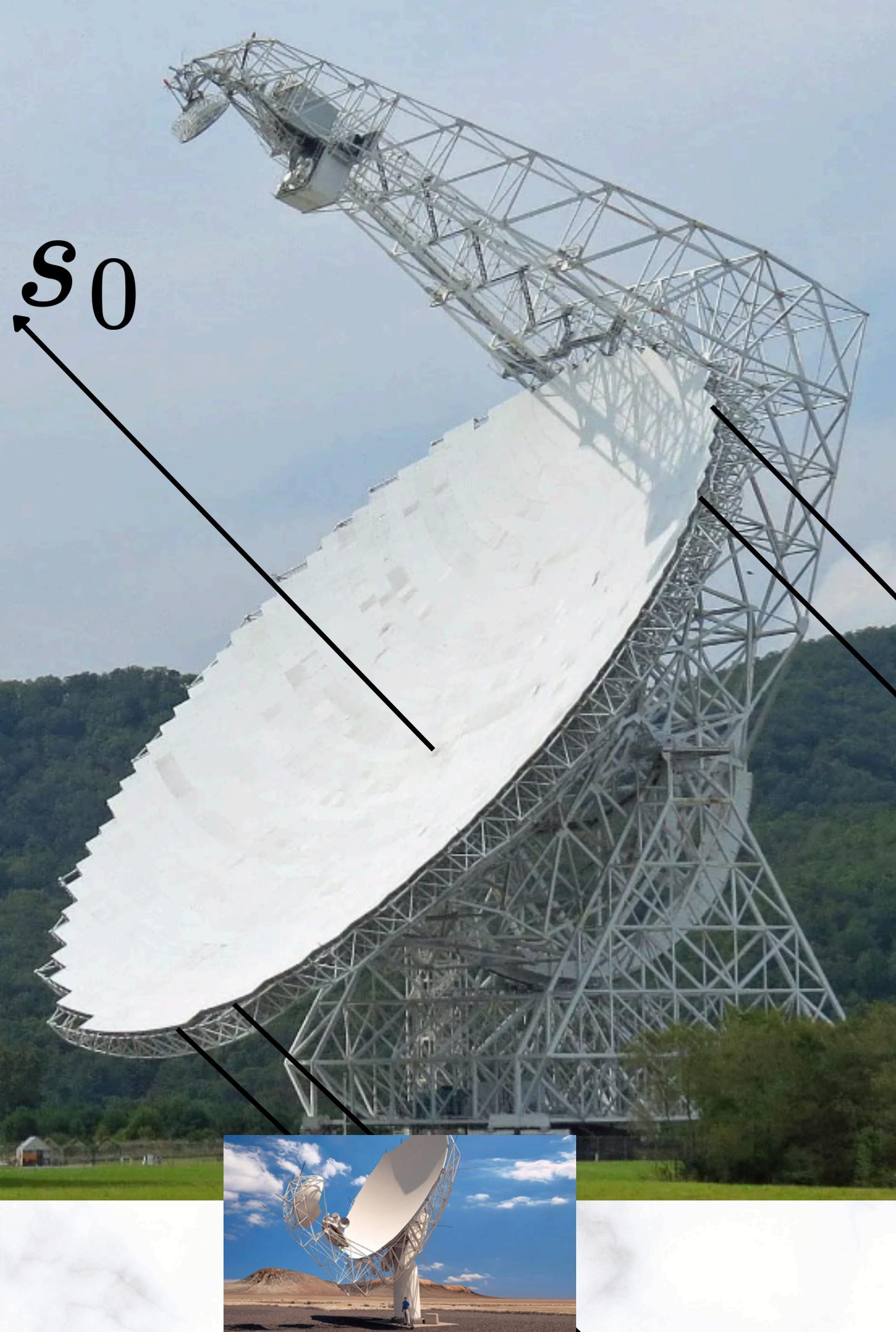


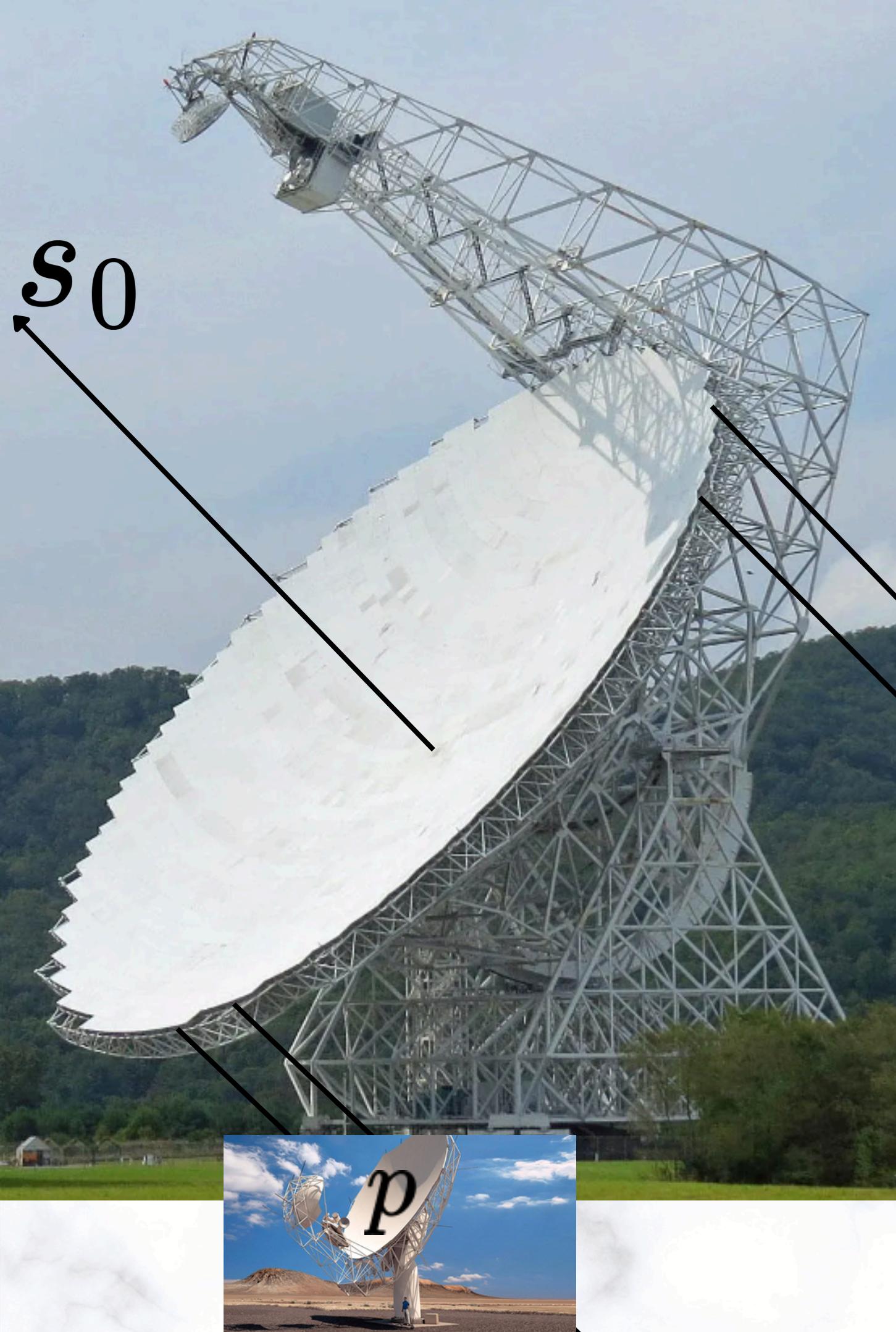


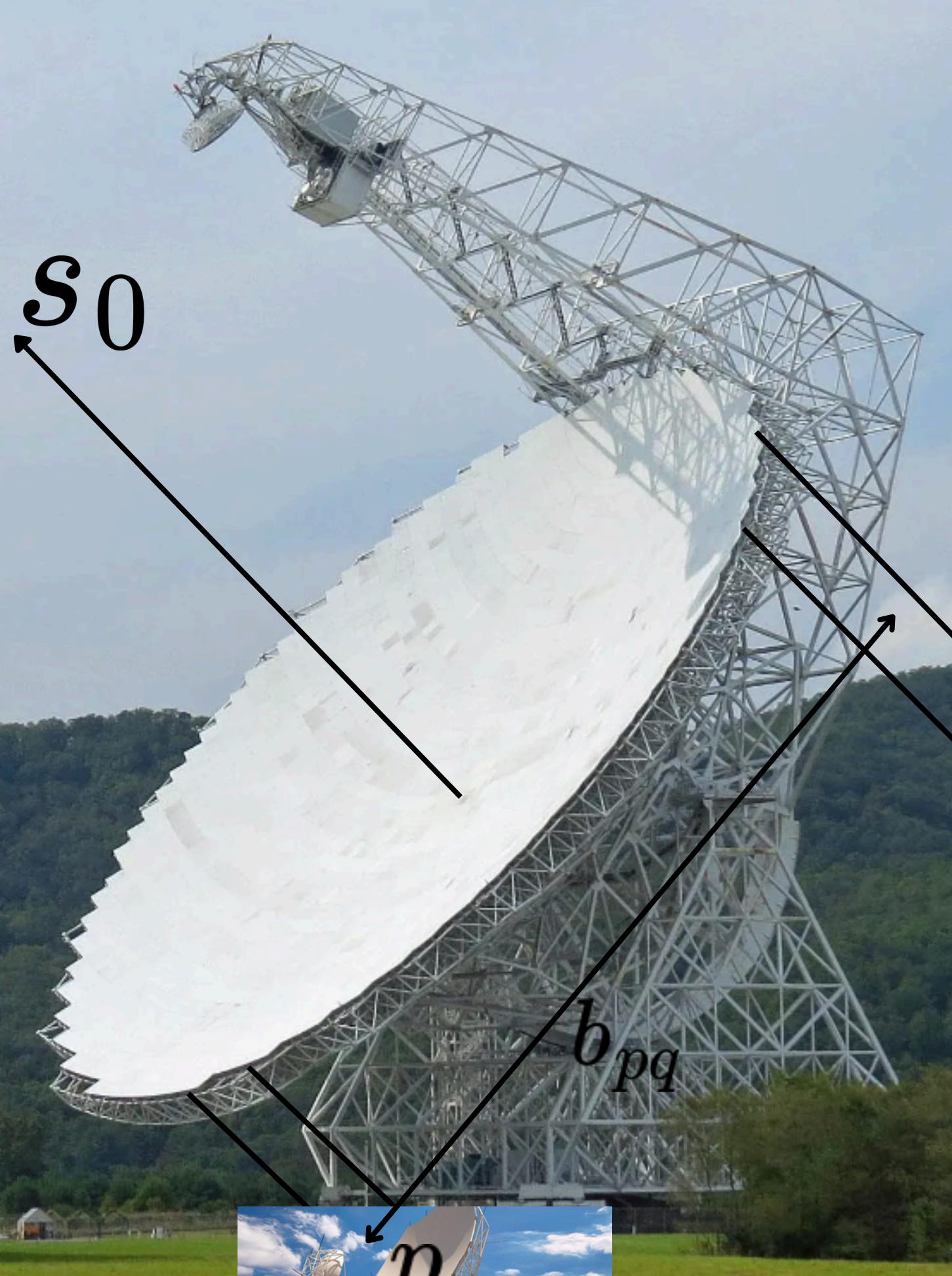


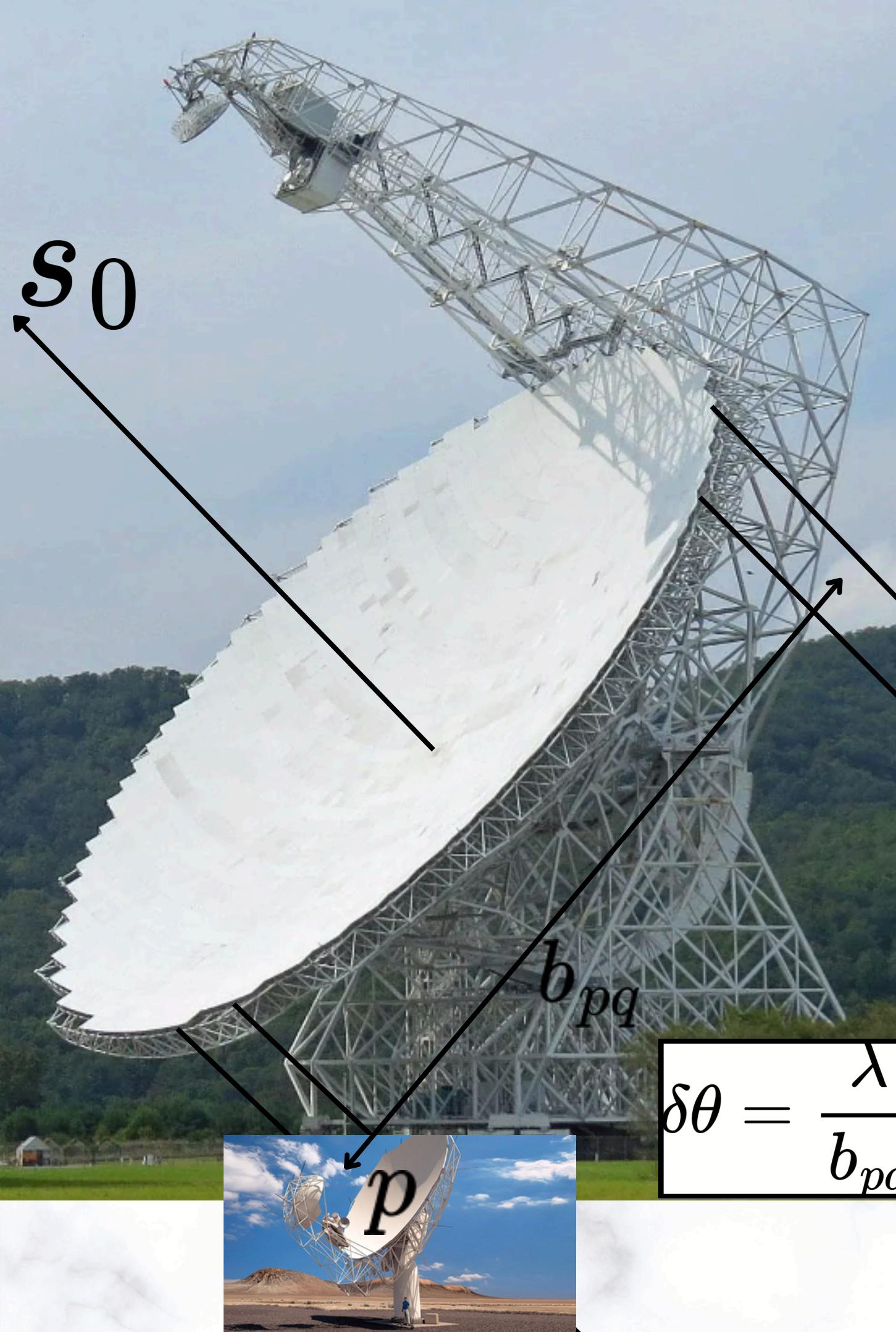
$$\delta\theta = \frac{\lambda}{D}$$





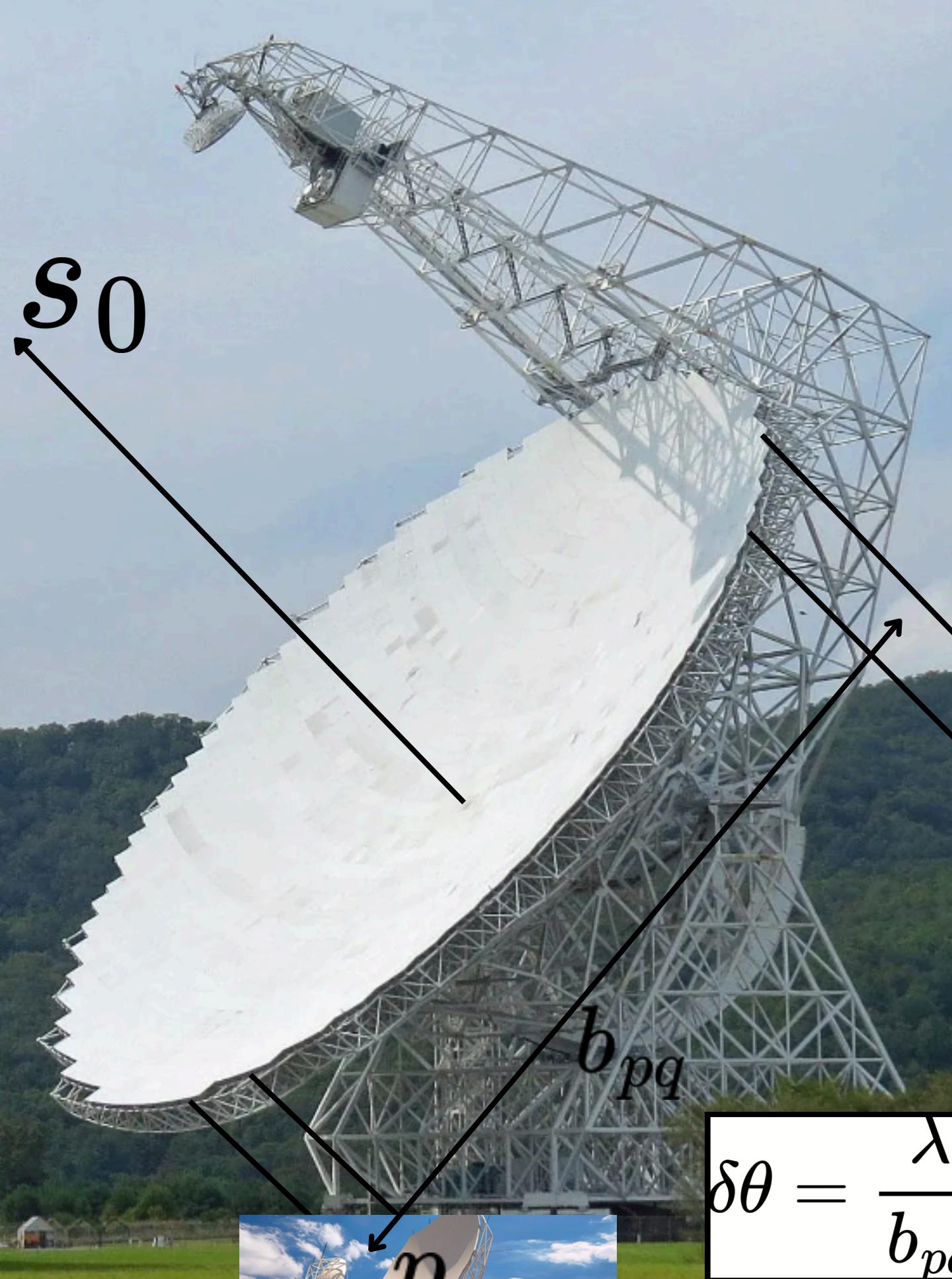




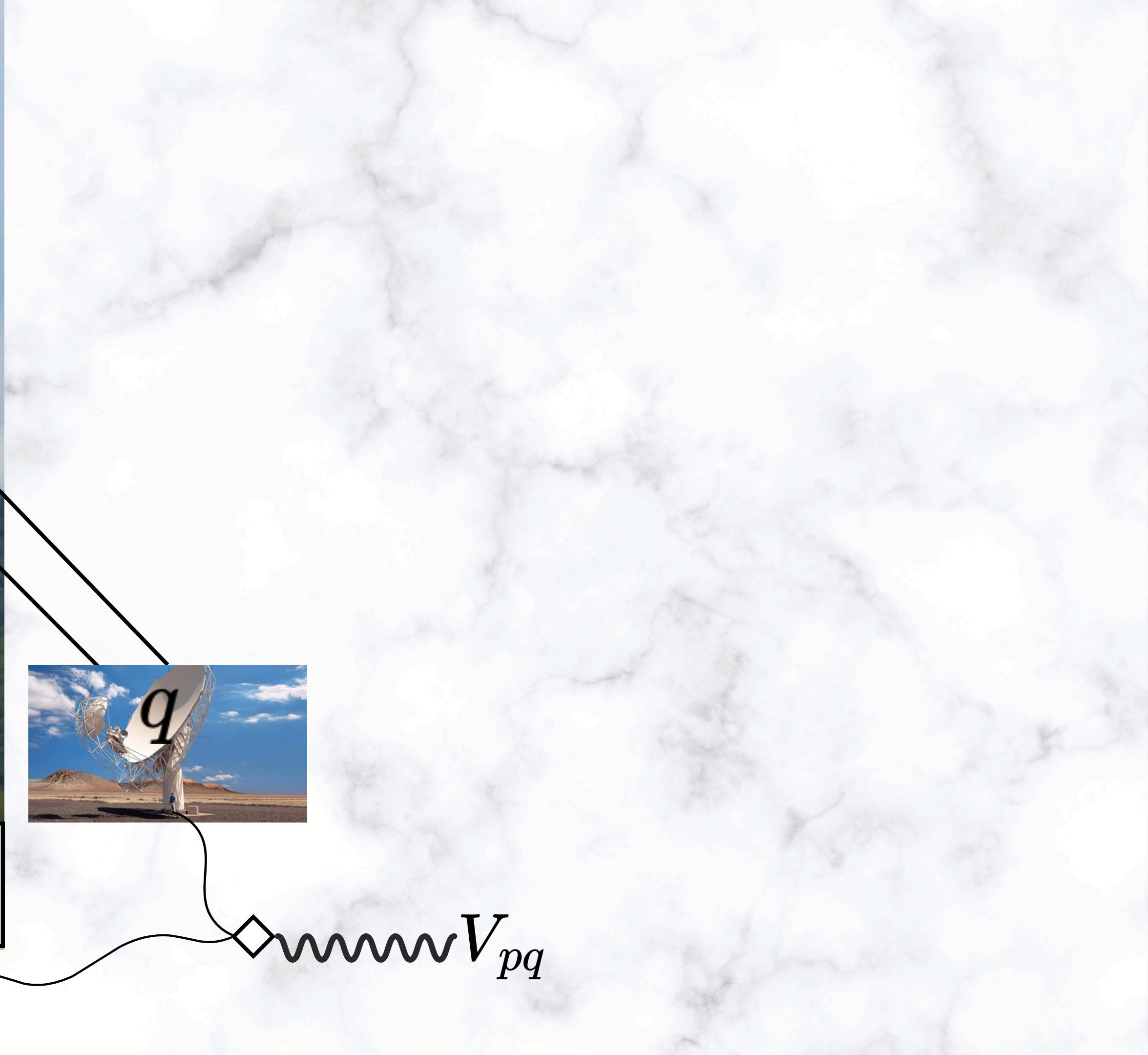


$$\delta\theta = \frac{\lambda}{b_{pq}}$$

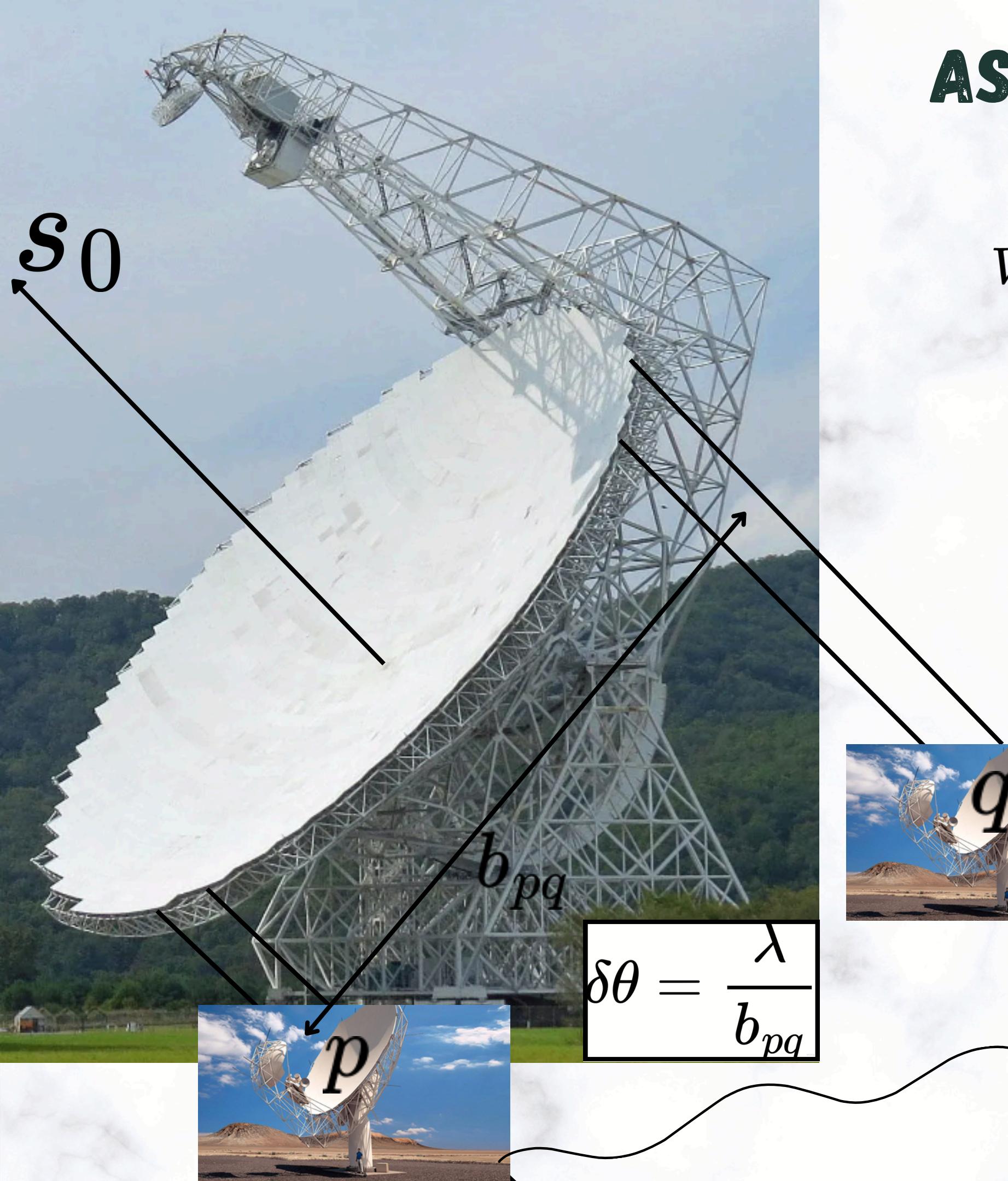




$$\delta\theta = \frac{\lambda}{b_{pq}}$$



ASSUMING PERFECT CALIBRATION



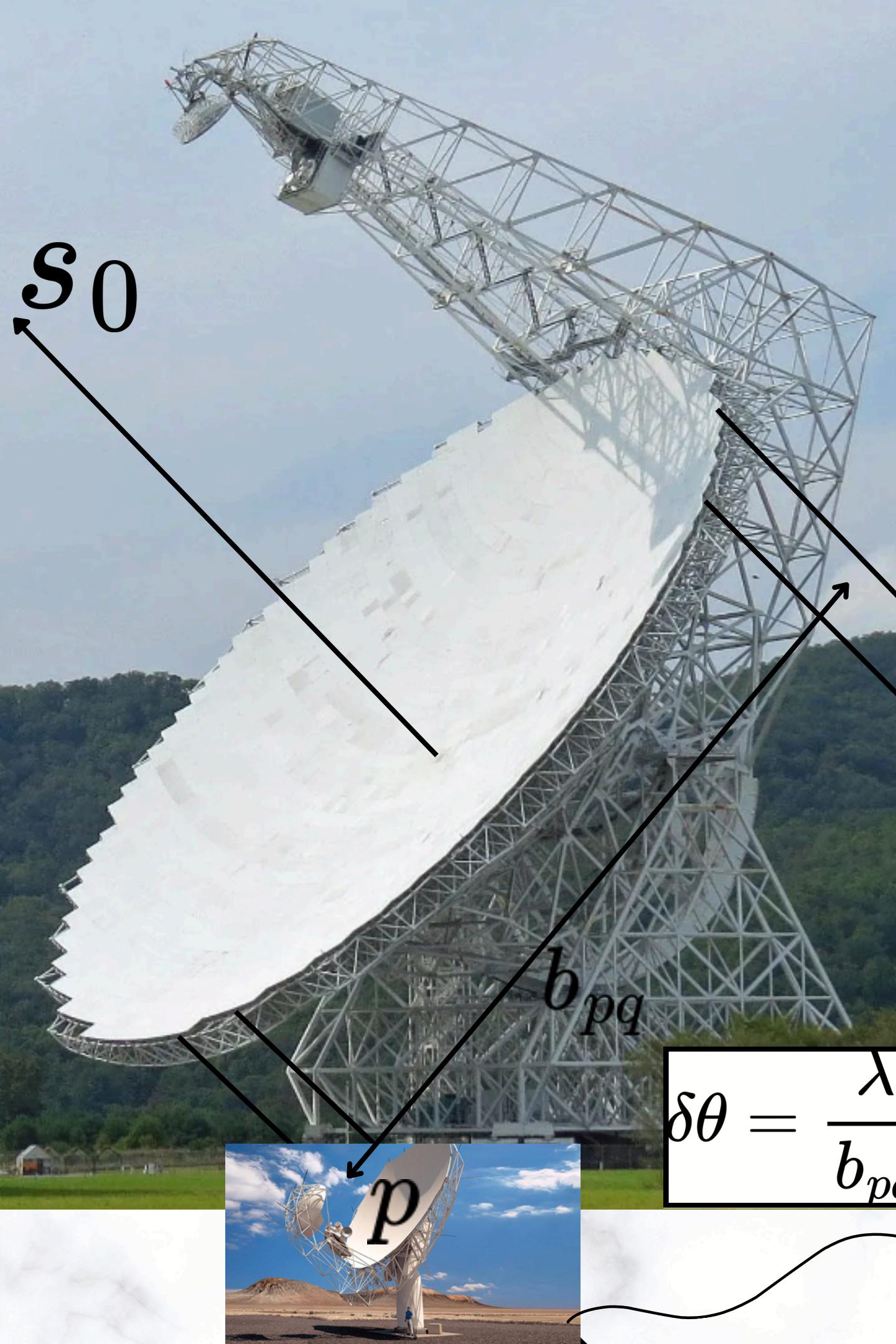
$$\delta\theta = \frac{\lambda}{b_{pq}}$$

$$V_{pqt} = \int I(s) \exp \left(-2\pi i \frac{\nu}{c} (b_{pqt}^T (s - s_0)) \right) ds$$



$$V_{pq}$$

ASSUMING PERFECT CALIBRATION



$$\delta\theta = \frac{\lambda}{b_{pq}}$$

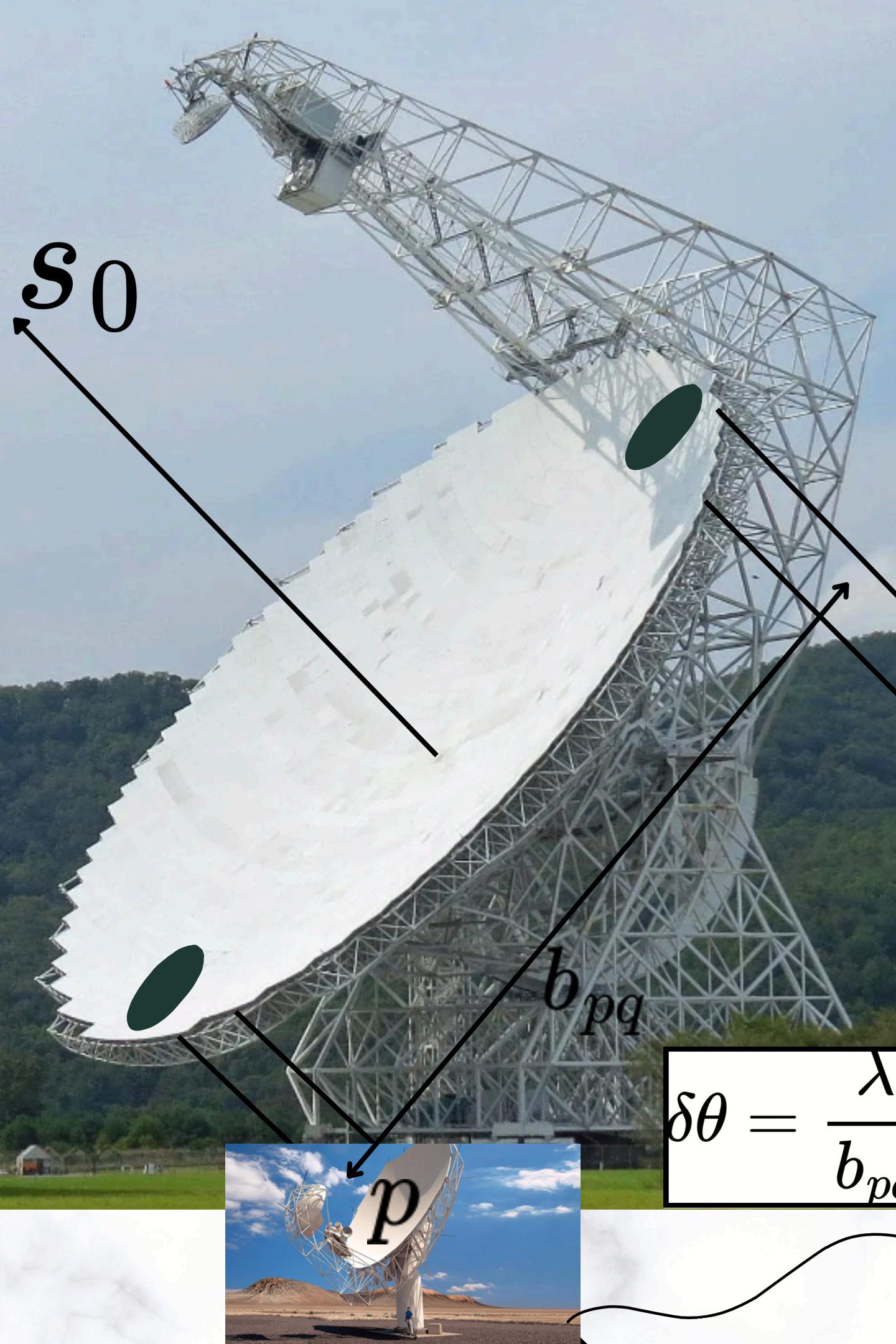
$$V_{pqt} = \int I(s) \exp \left(-2\pi i \frac{\nu}{c} (b_{pqt}^T (s - s_0)) \right) ds$$

**RESOLUTION DOES
NOT COME FOR FREE!**



$$V_{pq}$$

ASSUMING PERFECT CALIBRATION



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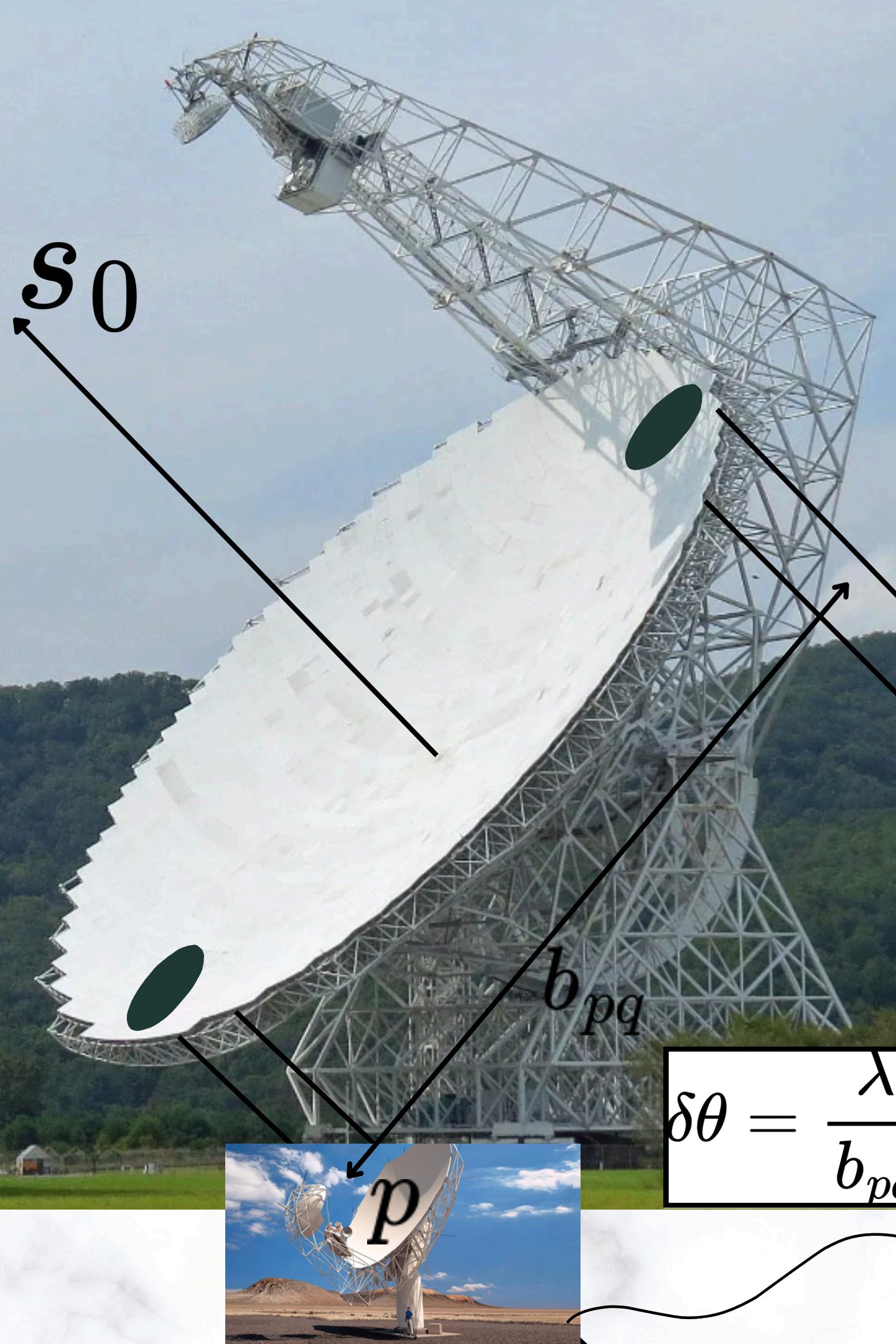
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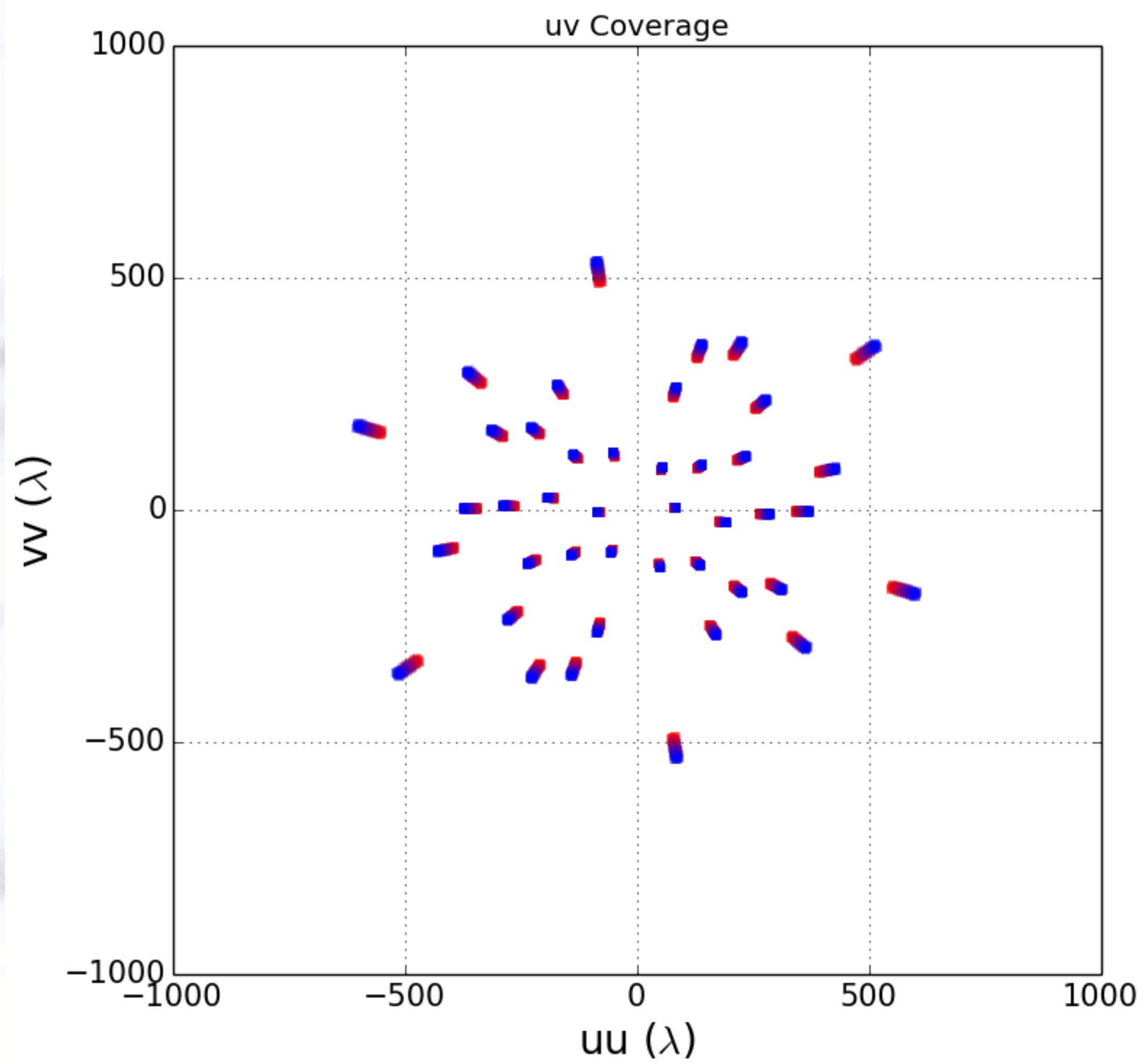
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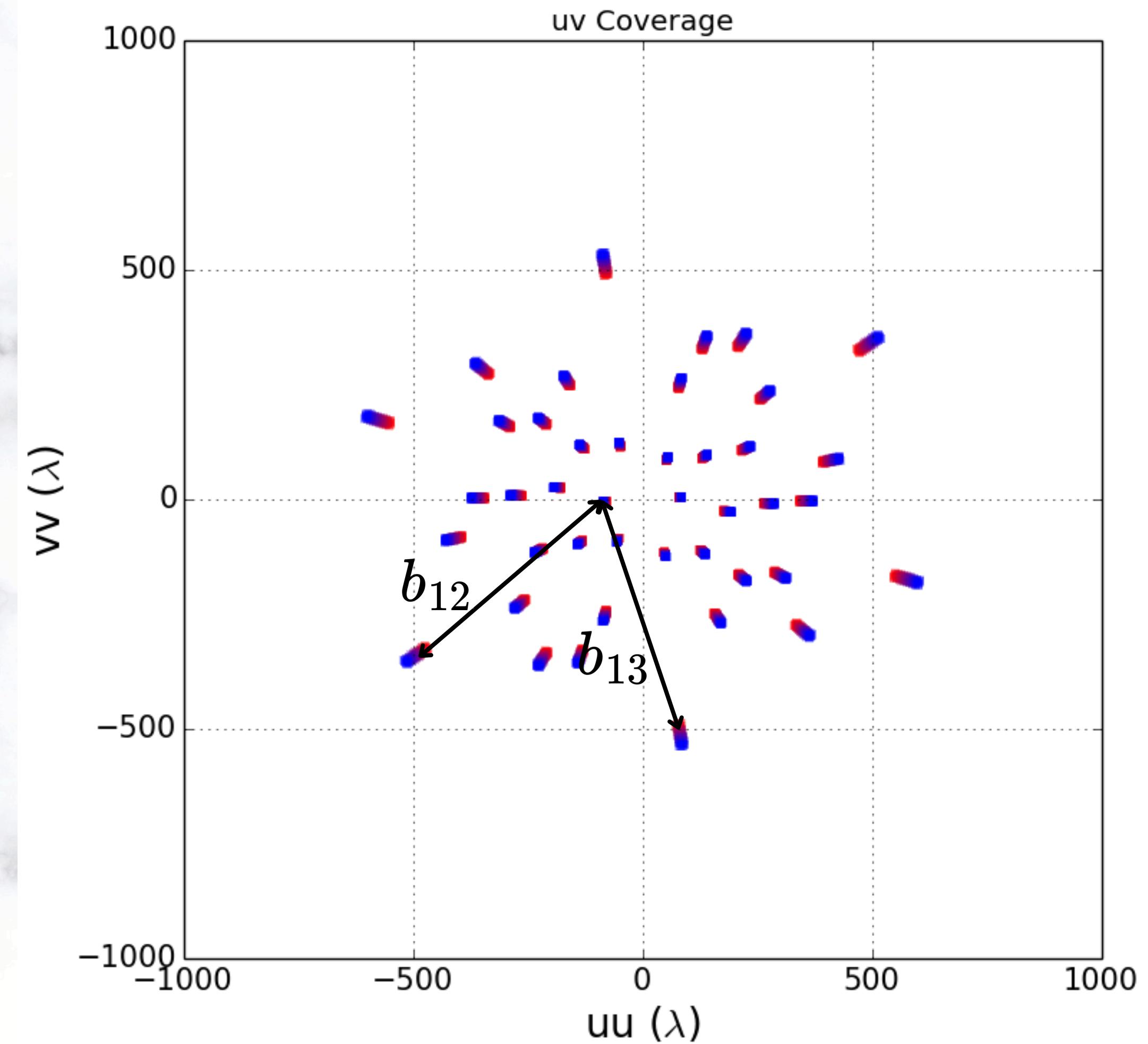
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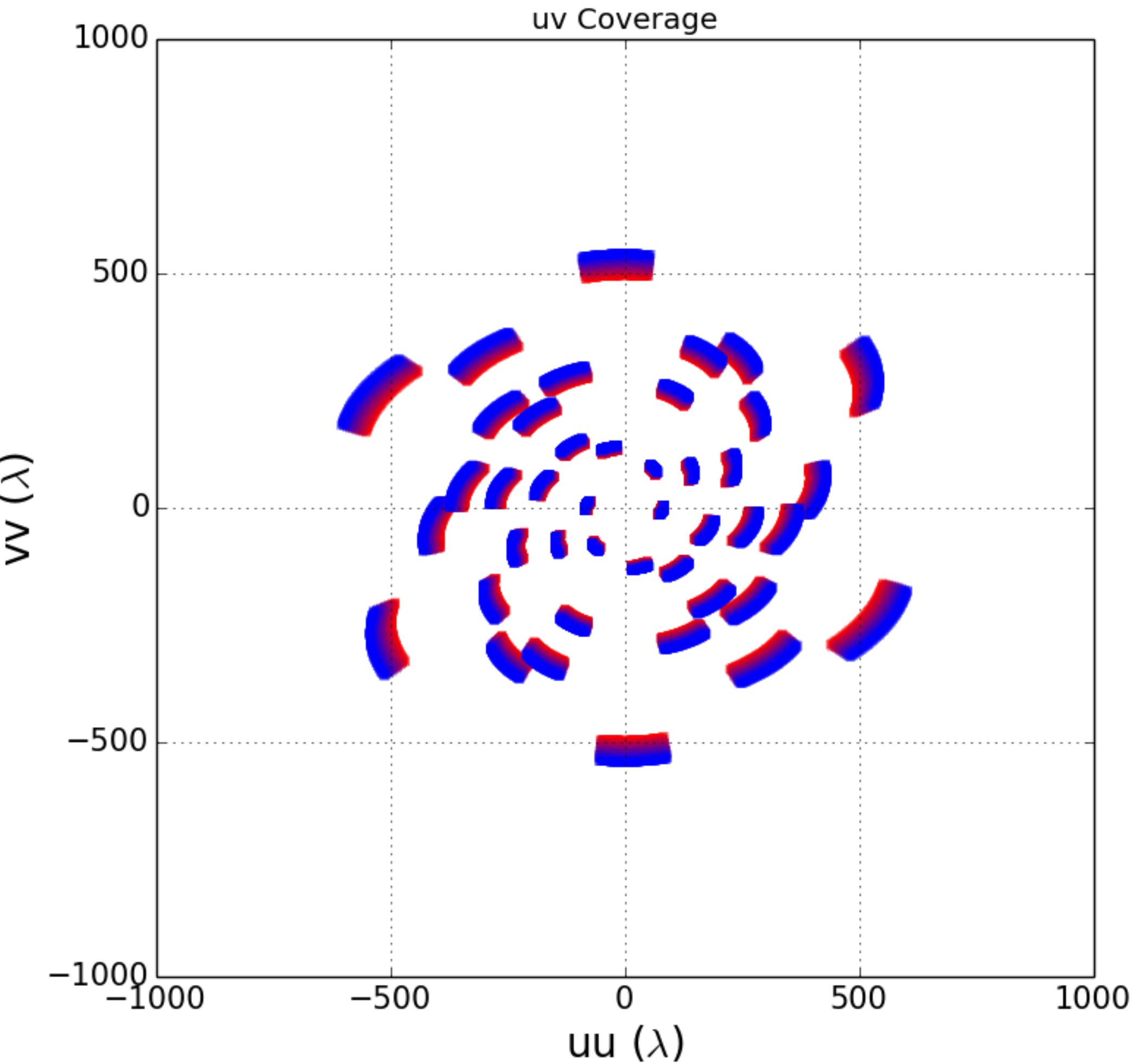
MORE ANTENNAS
+
EARTH ROTATION
+
FREQUENCY COVERAGE
=
FEWER HOLES



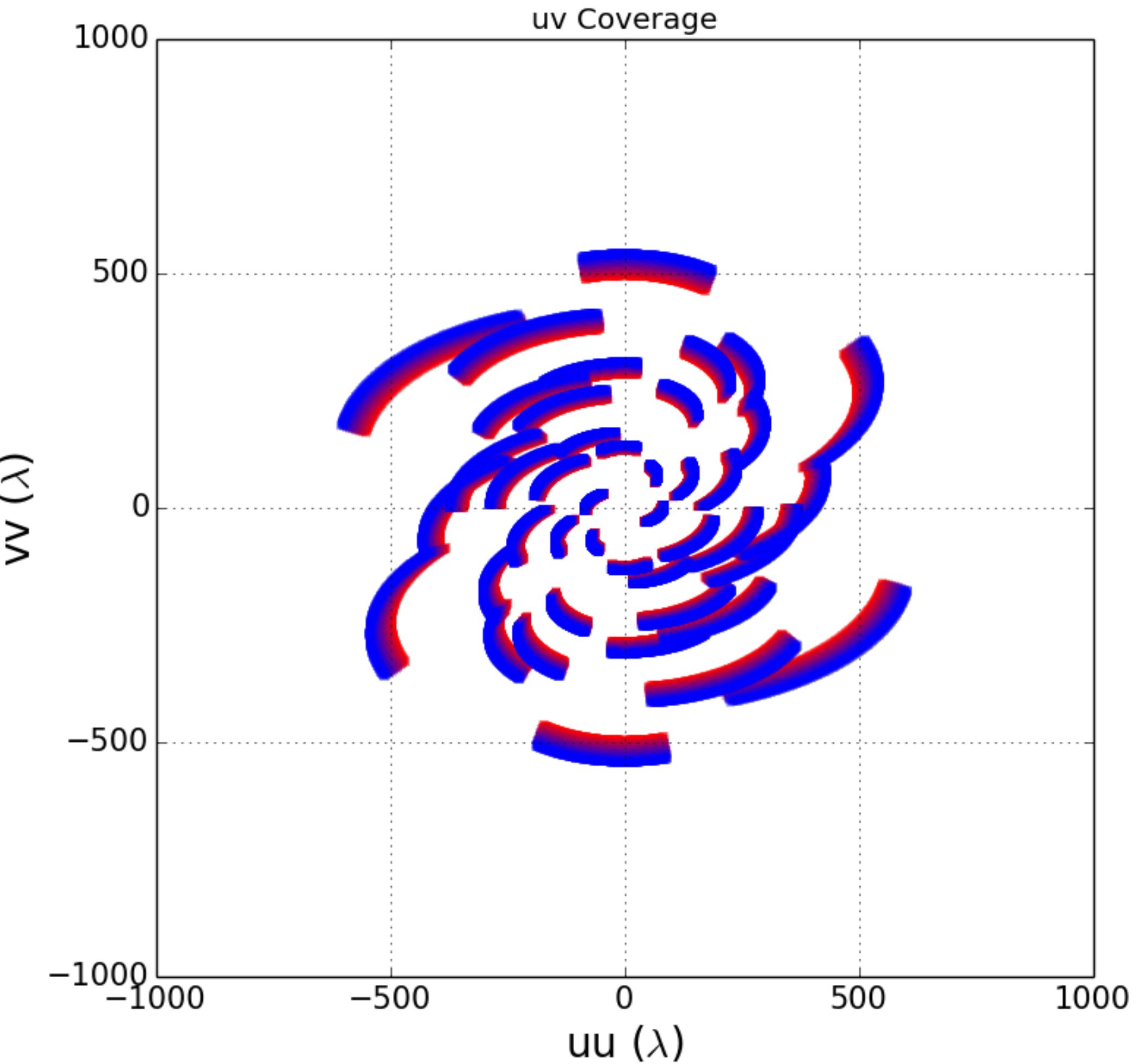
- SAMPLING PATTERN DEFINED BY
RELATIVE POSITION OF ANTENNAS



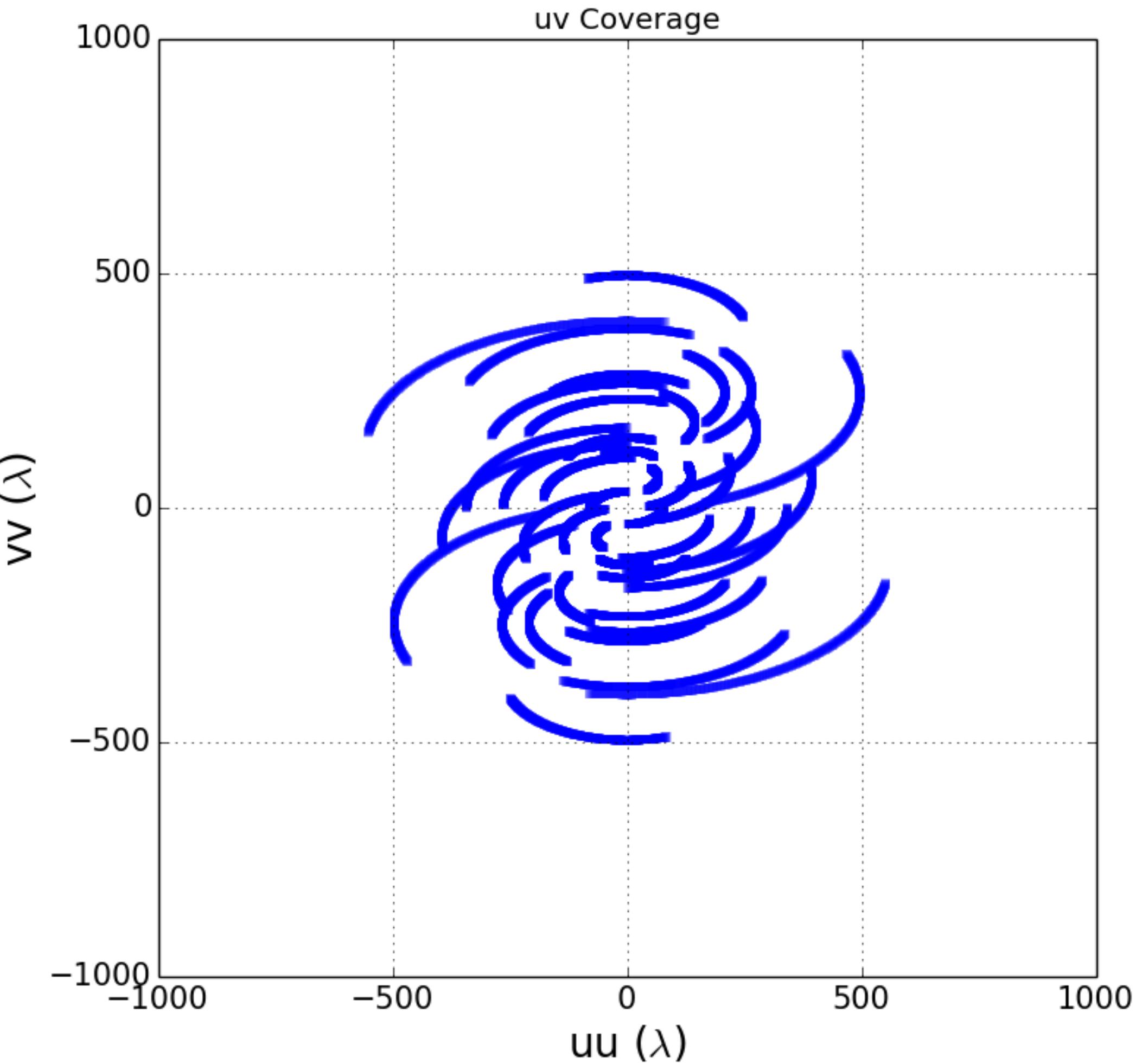
- SAMPLING PATTERN DEFINED BY
RELATIVE POSITION OF ANTENNAS
- EARTH ROTATION CAUSES BASELINES
TO TRACE OUT ELLIPTICAL TRACKS



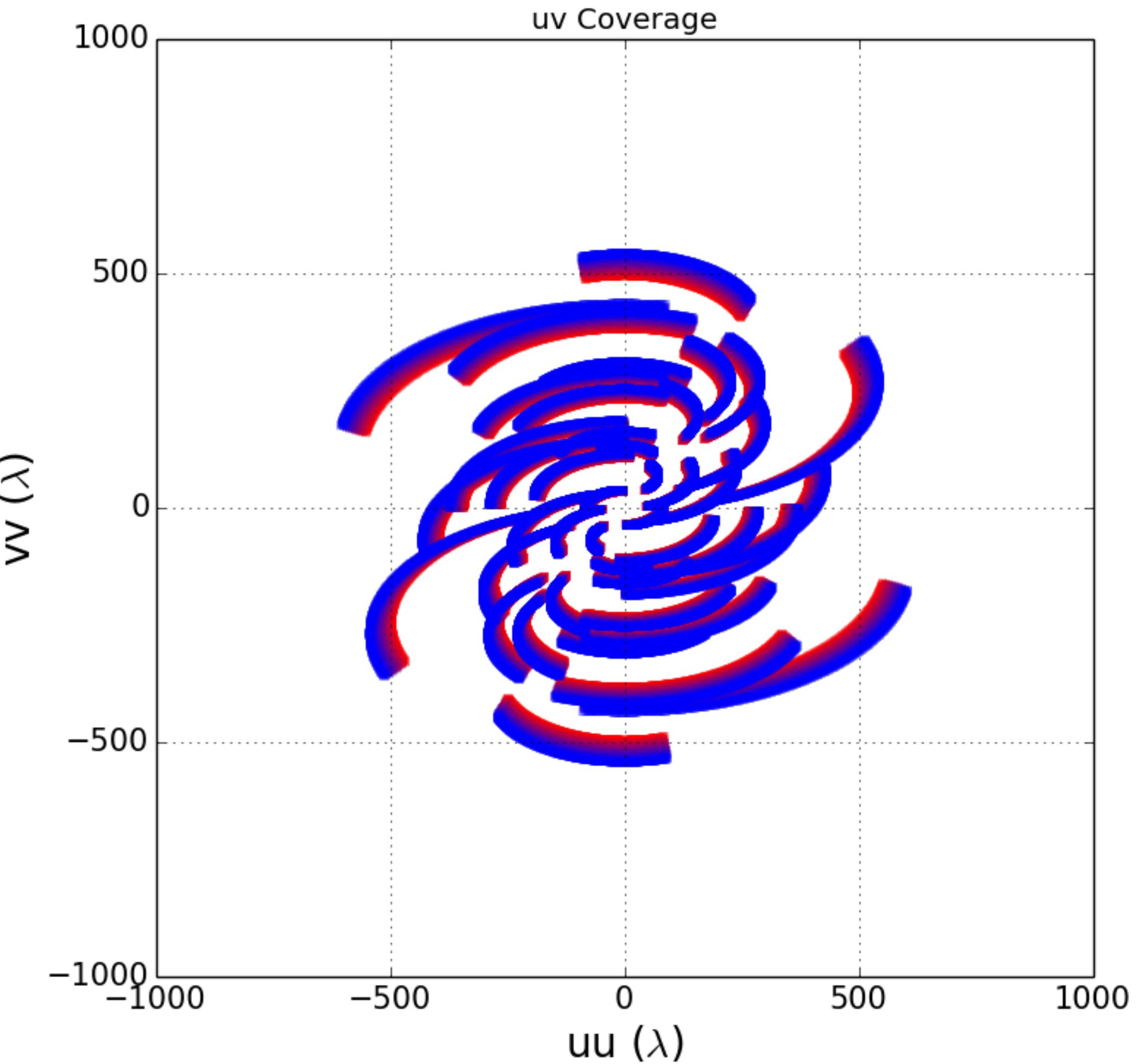
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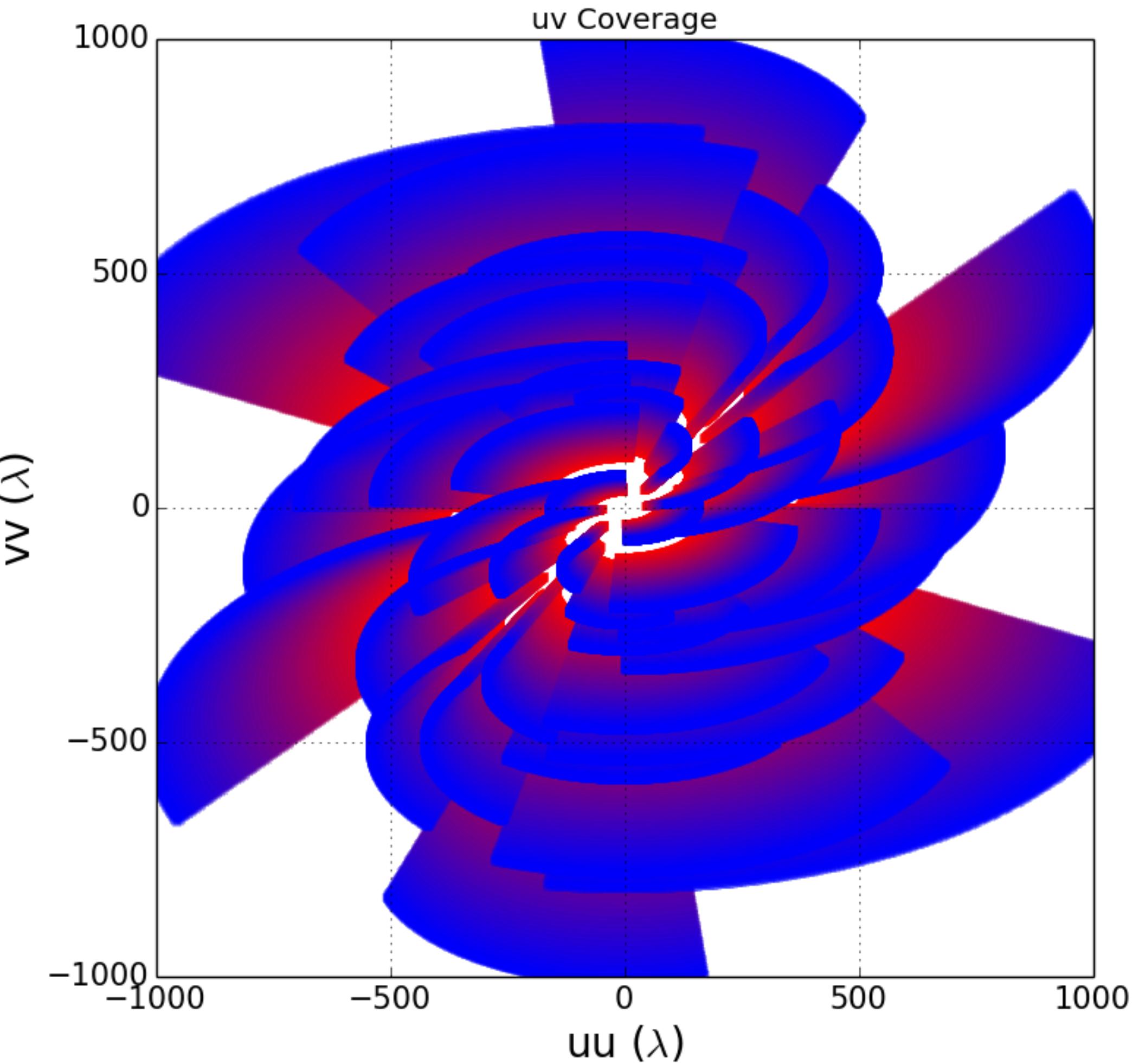
- SAMPLING PATTERN DEFINED BY
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TO TRACE OUT ELLIPTICAL TRACKS
- FREQUENCY COVERAGE FILLS THE
UV-PLANE RADIALLY



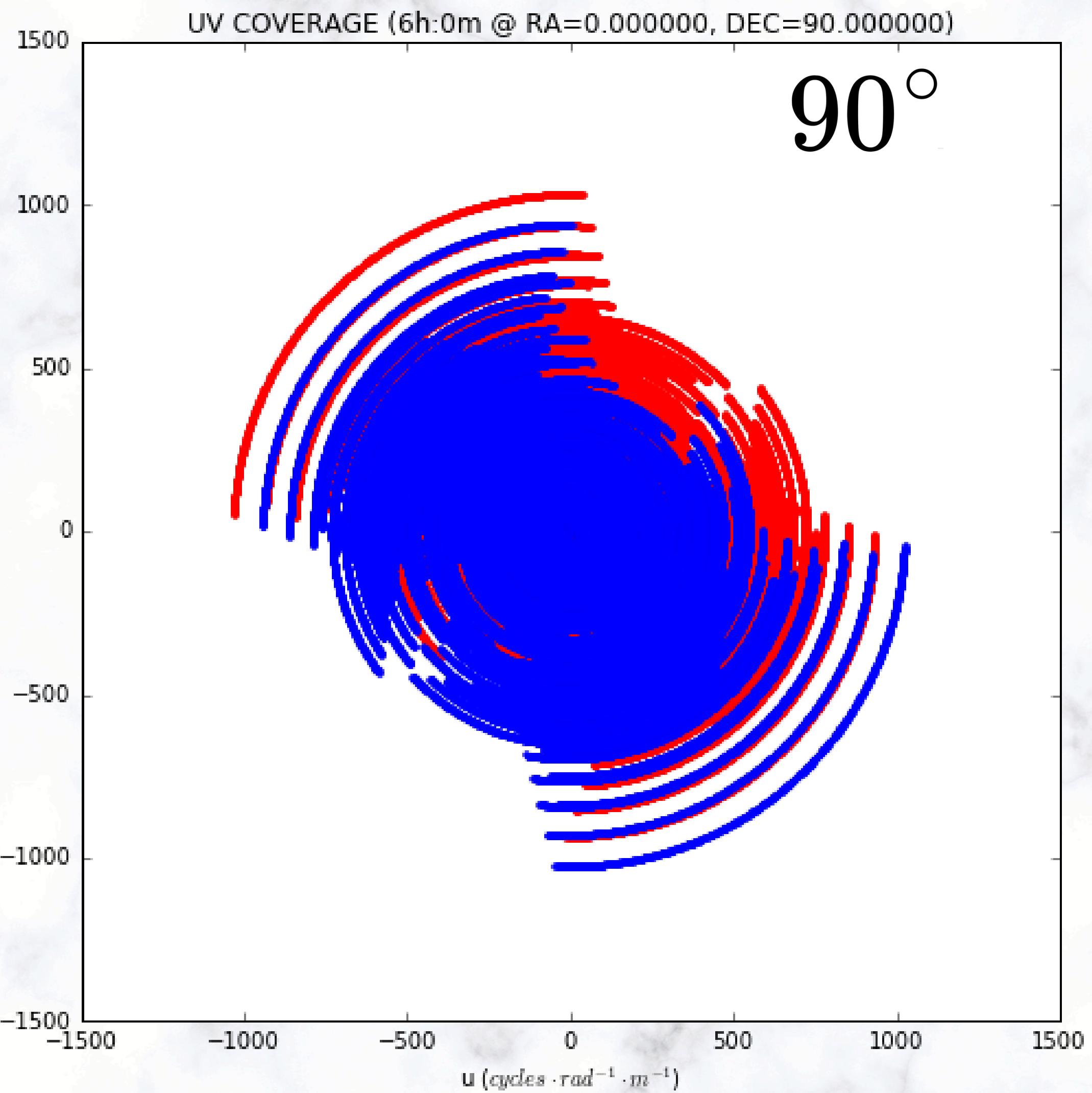
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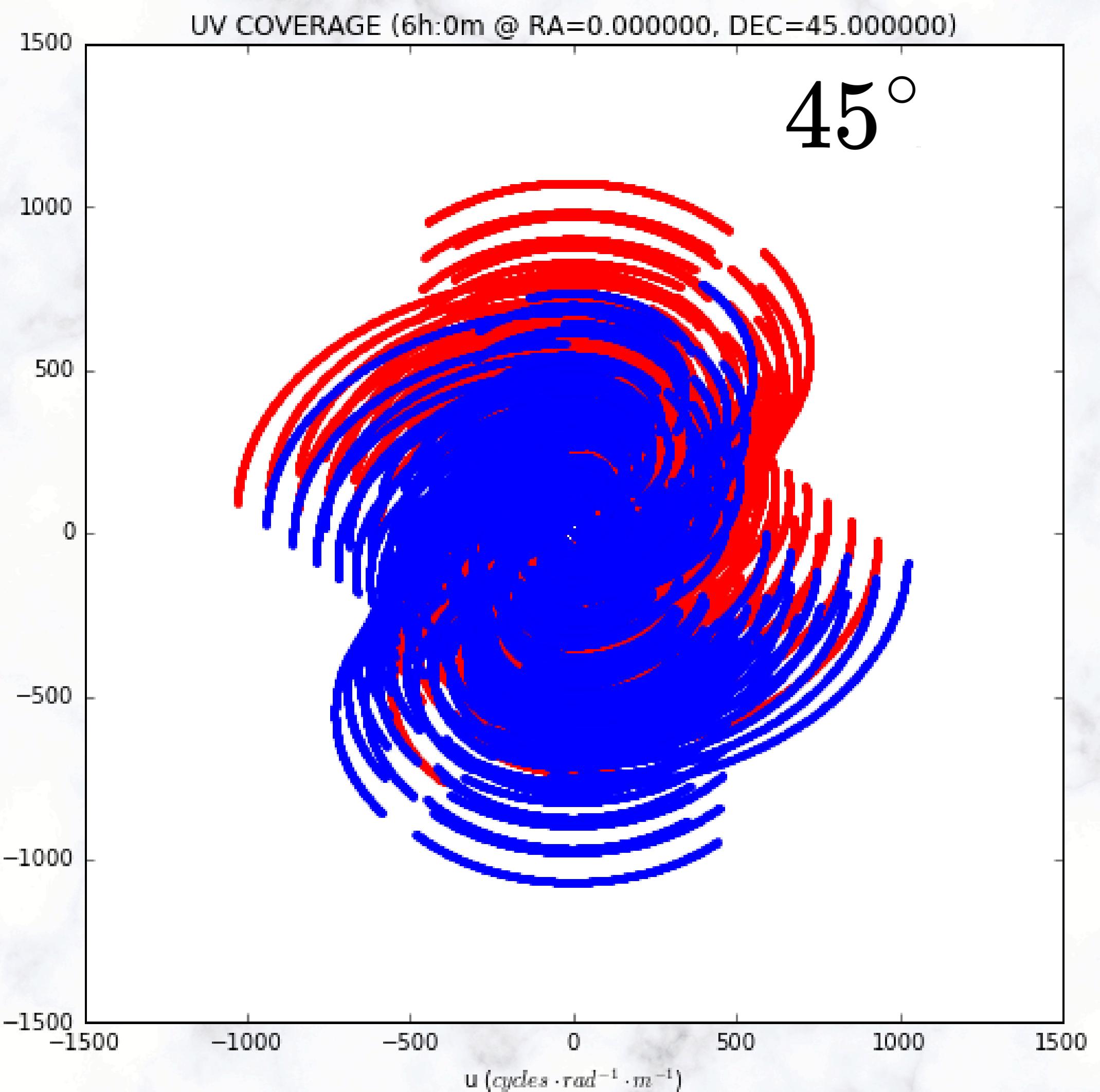
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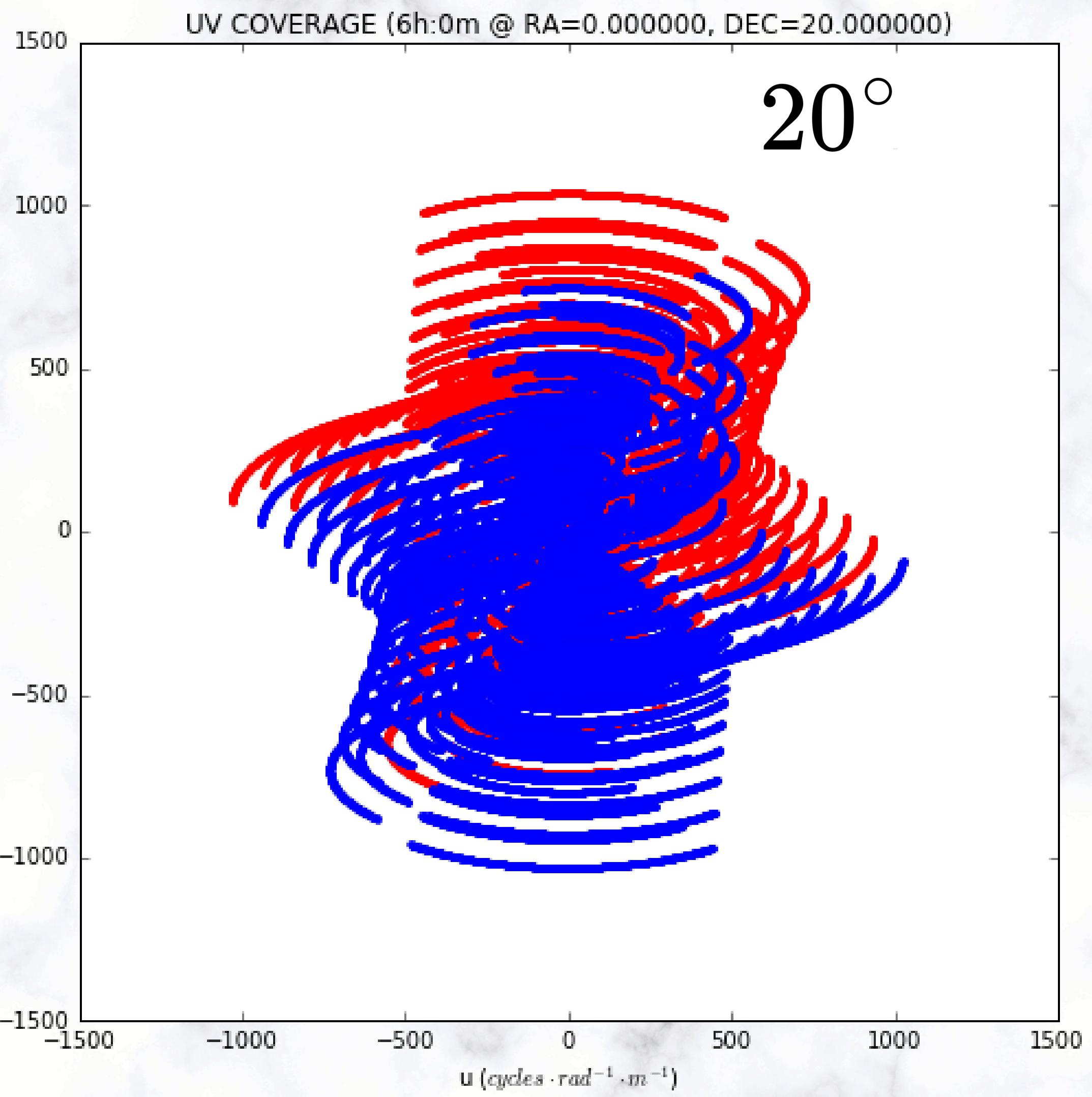
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- SAMPLING PATTERN DEPENDS ON
DECLINATION



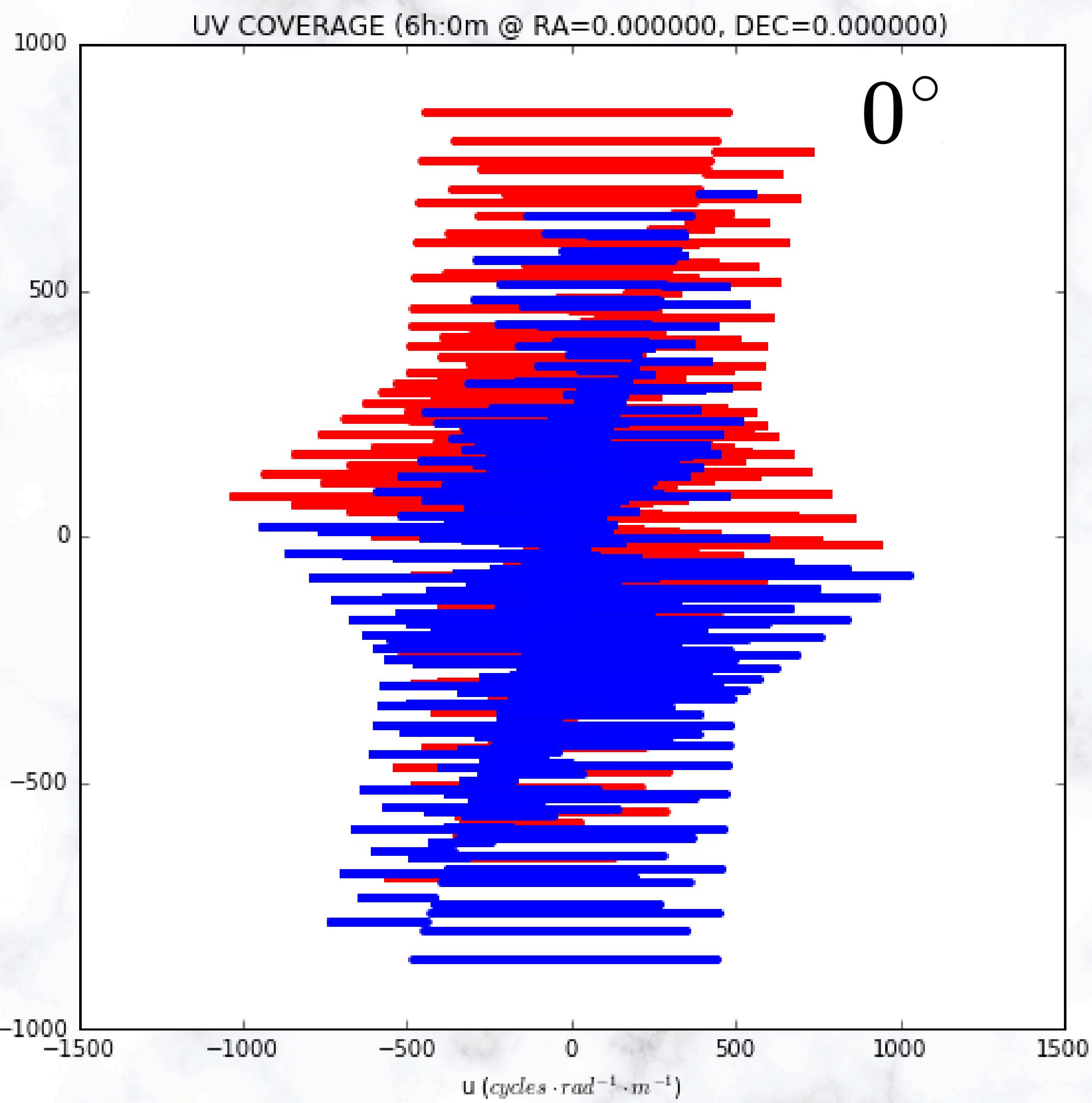
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VAN CITTERT ZERNIKE

$$V_{pq} = \int I(l, m) \exp \left(-2\pi i \frac{\nu}{c} (u_{pq}l + v_{pq}m + w_{pq}(n - 1)) \right) \frac{dl dm}{n^{\text{miro}}}$$

VISIBILITIES
(DATA)

VAN CITTERT ZERNIKE

$$V_{pq} = \int I(l, m) \exp \left(-2\pi i \frac{\nu}{c} (u_{pq}l + v_{pq}m + w_{pq}(n - 1)) \right) \frac{dl dm}{n}$$

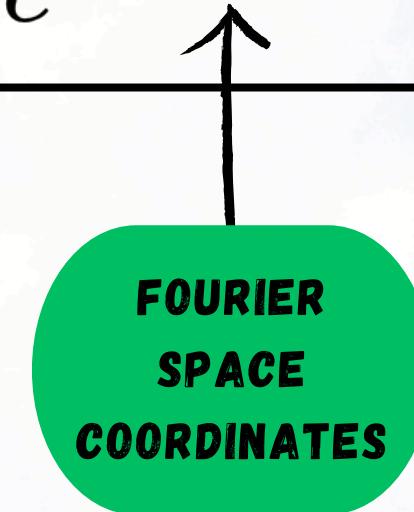
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TOTAL
INTENSITY
IMAGE

VAN CITTERT ZERNIKE

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VAN CITTERT ZERNIKE

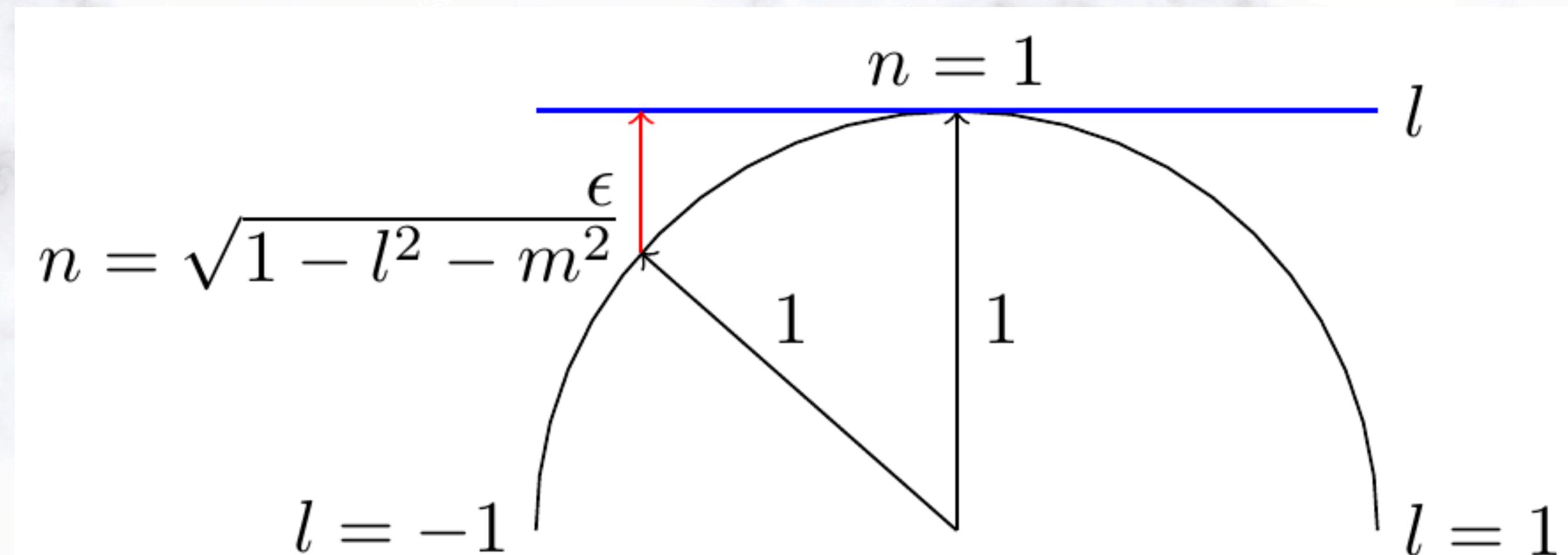
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IMAGE
SPACE
COORDINATES

VAN CITTERT ZERNIKE

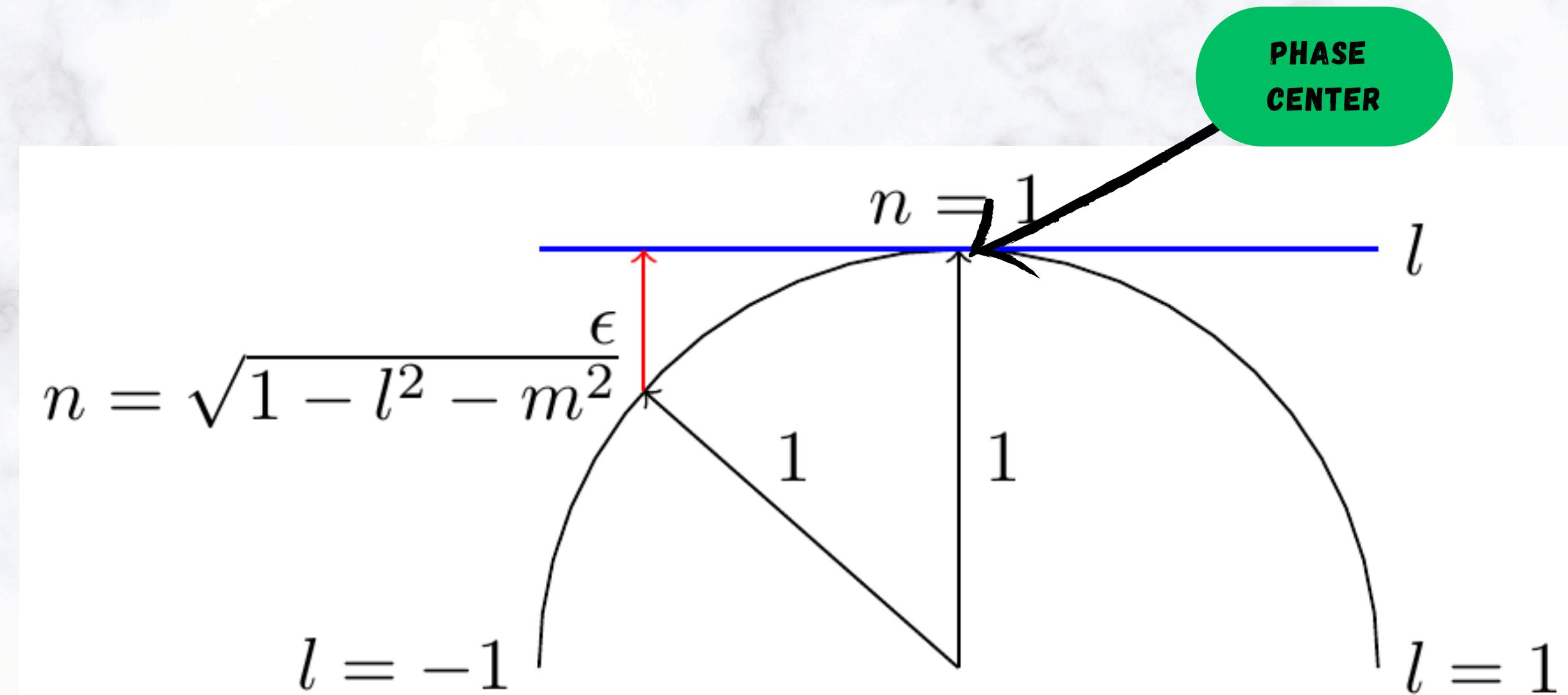
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IMAGE
SPACE
COORDINATES



VAN CITTERT ZERNIKE

$$V_{pq} = \int I(l, m) \exp \left(-2\pi i \frac{\nu}{c} (u_{pq}l + v_{pq}m + w_{pq}(n - 1)) \right) \frac{dl dm}{n}$$



VAN CITTERT ZERNIKE

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**IMAGING -> HOW TO OBTAIN IMAGE
FROM MEASURED VISIBILITIES?**

Some Fourier Transform Pairs:

$x(t)$	$X(\nu)$
$A_0\delta(t - t_0)$	$A_0e^{-2\pi i\nu t_0}$
$\Pi(t)$	$\text{sinc}(\nu)$
$\exp(-at^2)$	$\sqrt{\frac{\pi}{a}} \exp(-\pi^2\nu^2/a)$
\vdots	\vdots

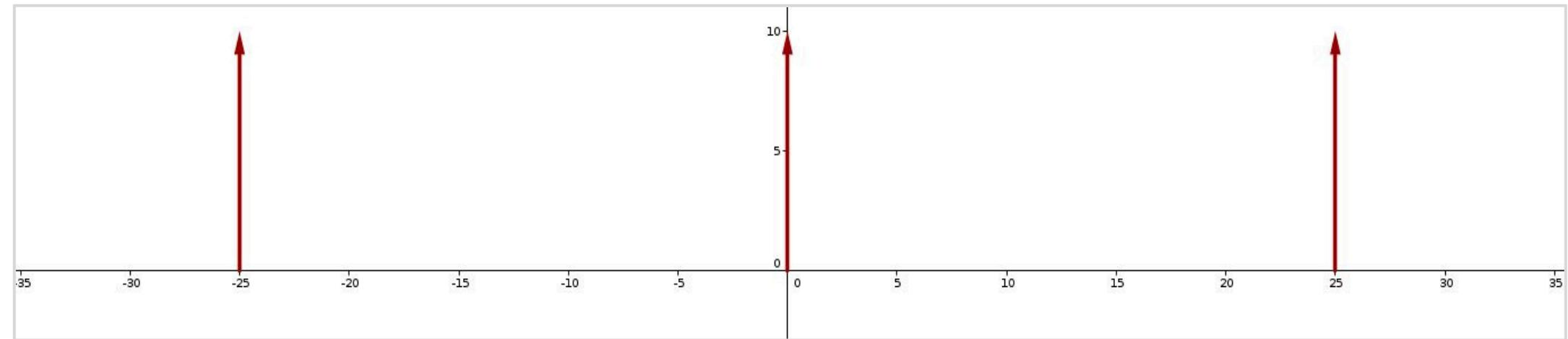
Fourier Theorems

Linearity	$\mathcal{F}[ax(t) + by(t)](s) = a\mathcal{F}[x(t)](s) + b\mathcal{F}[y(t)](s)$
Shift	$\mathcal{F}[x(t - t_0)] = e^{-2\pi i s t_0} \mathcal{F}[x(t)]$
Similarity	$\mathcal{F}[x(at)](s) = \frac{1}{ a } \mathcal{F}[x(t)]\left(\frac{s}{a}\right)$
Convolution	$\mathcal{F}[x(t)y(t)](s) = \mathcal{F}[x(t)](s) * \mathcal{F}[y(t)](s)$
Parseval	$\langle x, y \rangle = x^\dagger y = \langle \mathcal{F}[x], \mathcal{F}[y] \rangle$

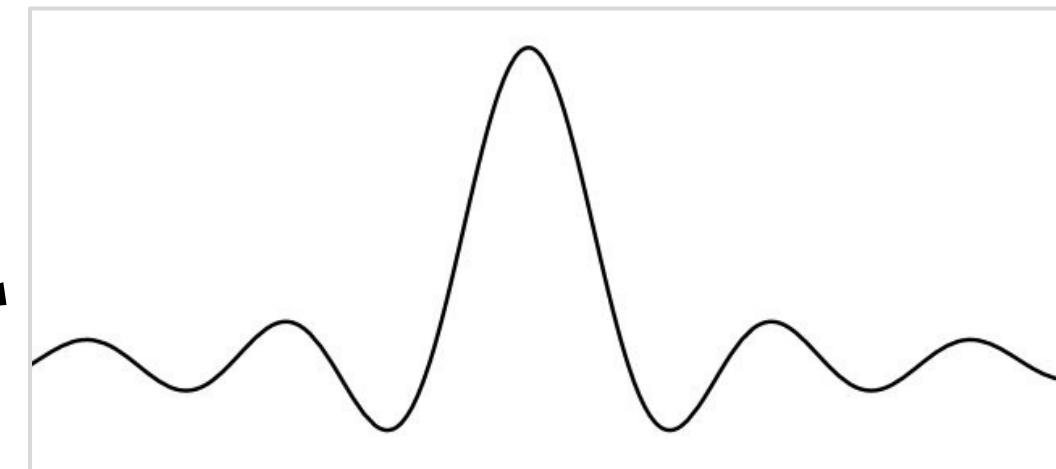
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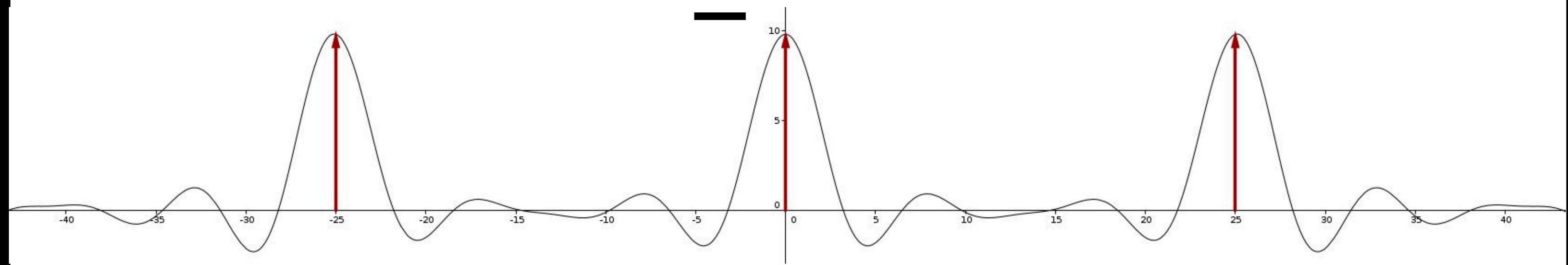
Example - Bed of nails convolved with sinc:



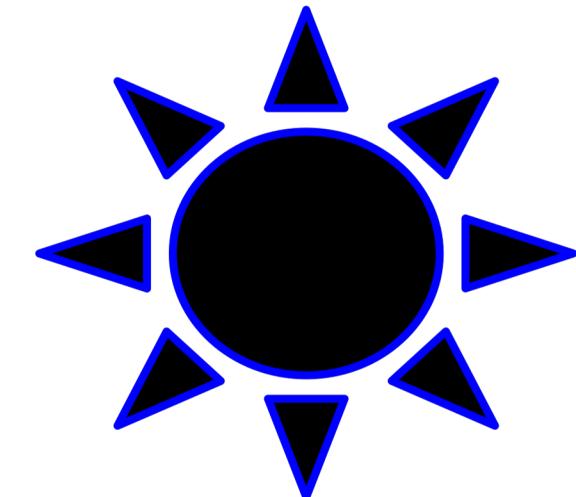
*



=



Discrete Fourier Transform



$$X_k = \sum_{n=0}^{N-1} x_n e^{-2\pi i \frac{kn}{N}}$$

$$x_n = \sum_{k=0}^{N-1} X_k e^{2\pi i \frac{kn}{N}}$$

$$t_n = t_0 + n\Delta t$$

$$\nu_k = \frac{k f_s}{N} \quad f_s = \frac{1}{\Delta t}$$

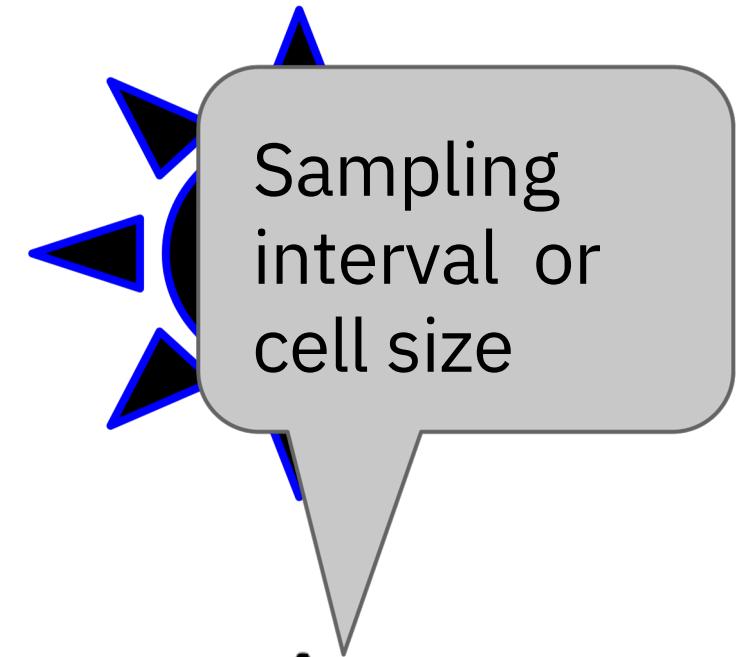
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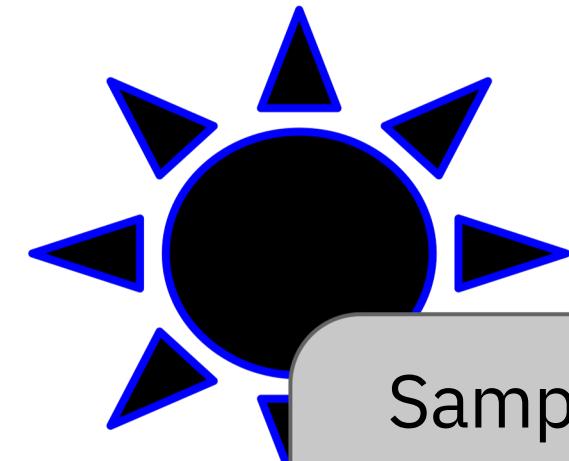
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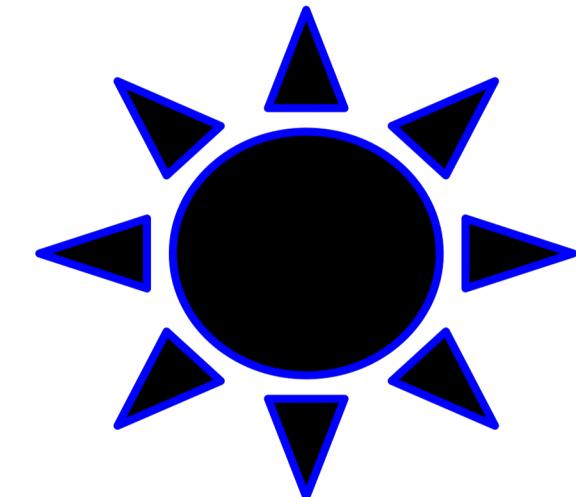
$$t_n = t_0 + n\Delta t$$

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Sampling rate
= maximum
frequency in
spectrum

Discrete Fourier Transform



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$$x_n = \sum_{k=0}^{N-1} X_k e^{2\pi i \frac{kn}{N}}$$

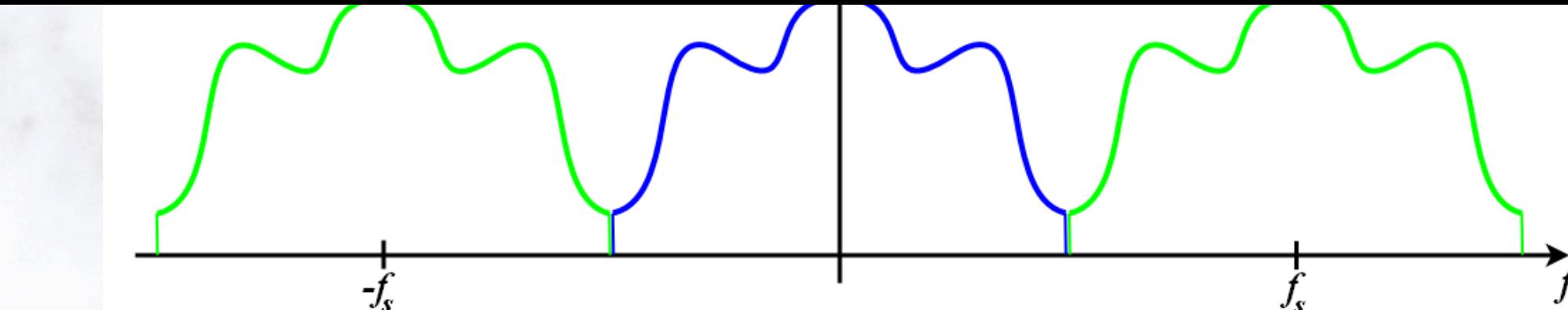
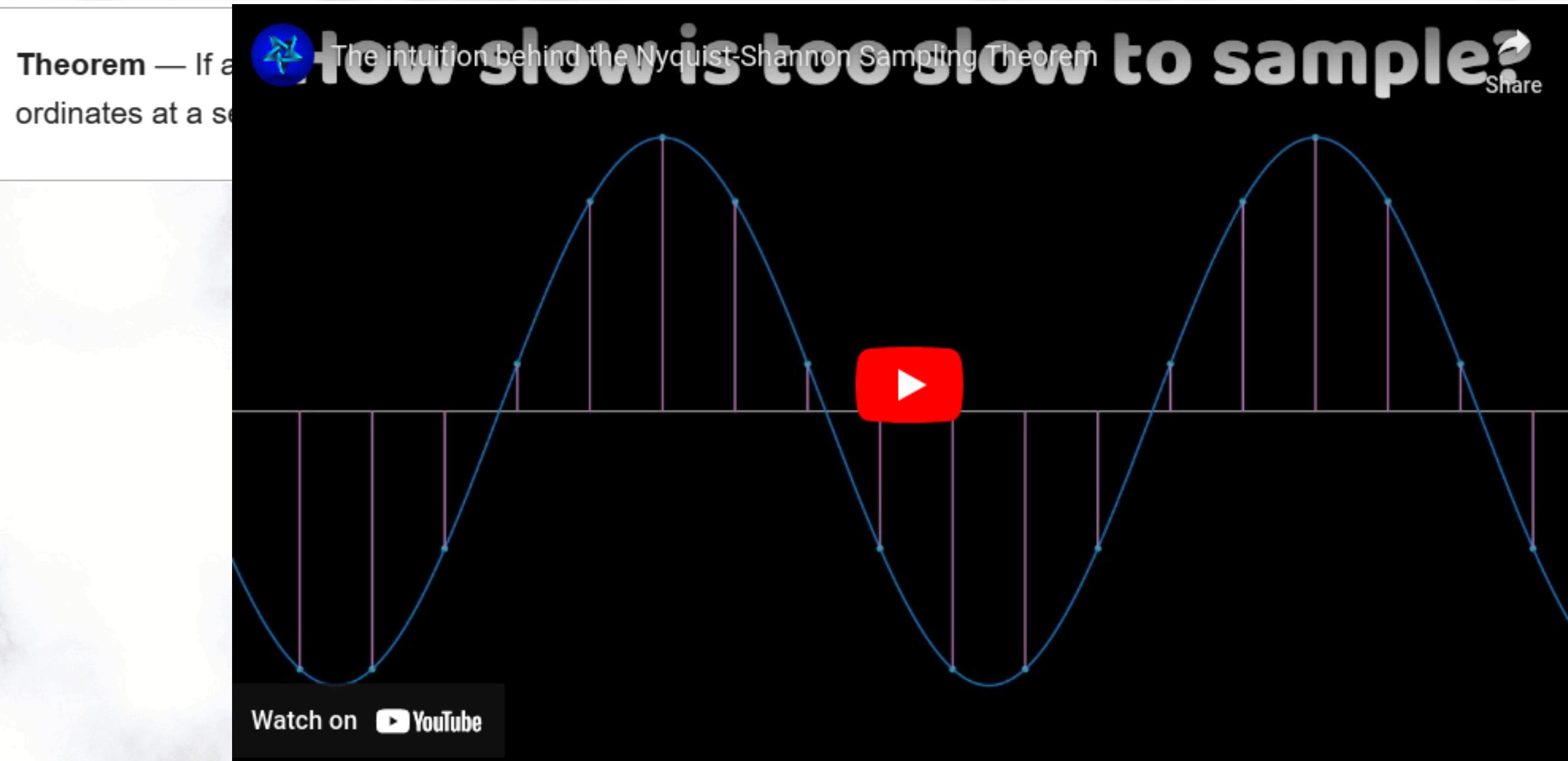
$$t_n = t_0 + n\Delta t$$

$$\nu_k = \frac{k f_s}{N} \quad f_s = \frac{1}{\Delta t}$$

Implemented using FFT which has computational complexity $\mathcal{O}(N \log_2 N)$

SAMPLING THEOREM AND ALIASING

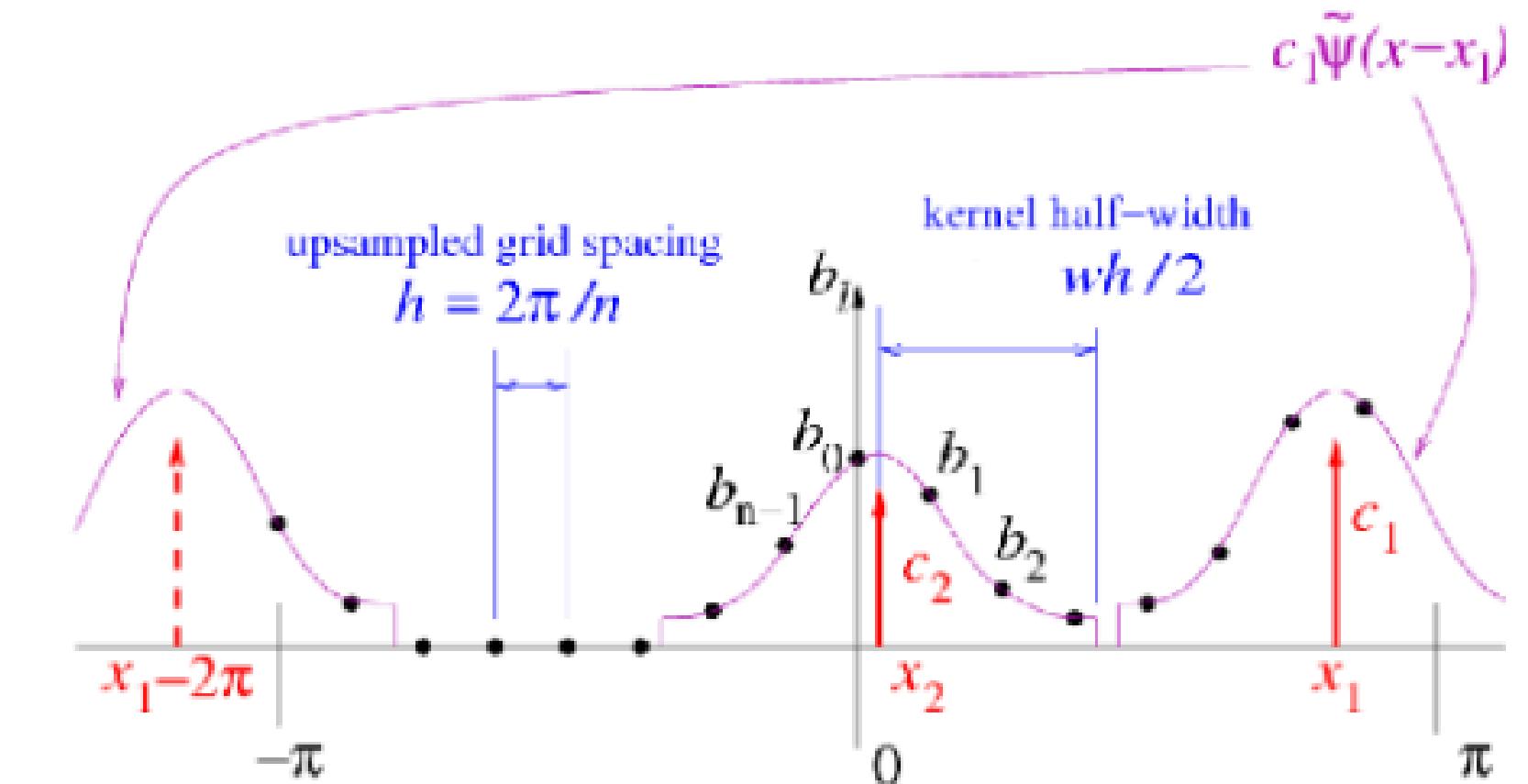
FROM WIKIPEDIA



NON-UNIFORM FFT

SEE EG.

Flatiron Institute Nonuniform Fast Fourier Transform



IMPLEMENTED EFFICIENTLY USING
CONVOLUTIONAL GRIDDING + FFT

VAN CITTERT ZERNIKE

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VAN CITTERT ZERNIKE

$$V_{pq} = \int I(l, m) \exp \left(-2\pi i \frac{\nu}{c} (u_{pq}l + v_{pq}m + w_{pq}(-1)) \right) \frac{dl dm}{n}$$

NUFFT IN ABSENCE OF WIDE-FIELD EFFECTS

VAN CITTERT ZERNIKE

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NUFFT IN ABSENCE OF WIDE-FIELD EFFECTS

**UV-COVERAGE DETERMINED BY INSTRUMENT -> SET CELL
SIZE BASED ON NYQUIST CRITERION**

VAN CITTERT ZERNIKE

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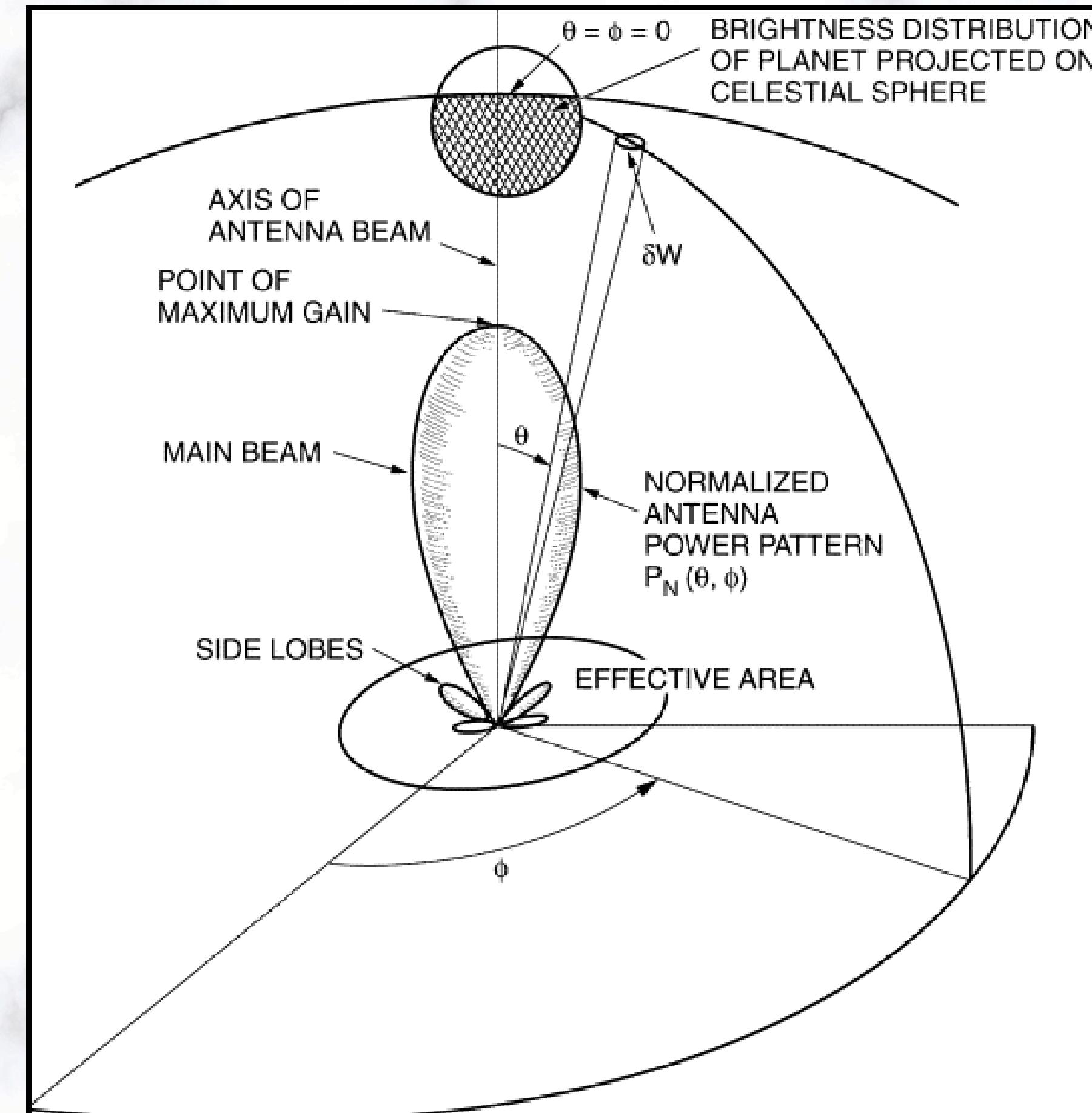
NUFFT IN ABSENCE OF WIDE-FIELD EFFECTS

**UV-COVERAGE DETERMINED BY INSTRUMENT -> SET CELL
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WHAT ABOUT THE FIELD OF VIEW (I.E. NPIX)?

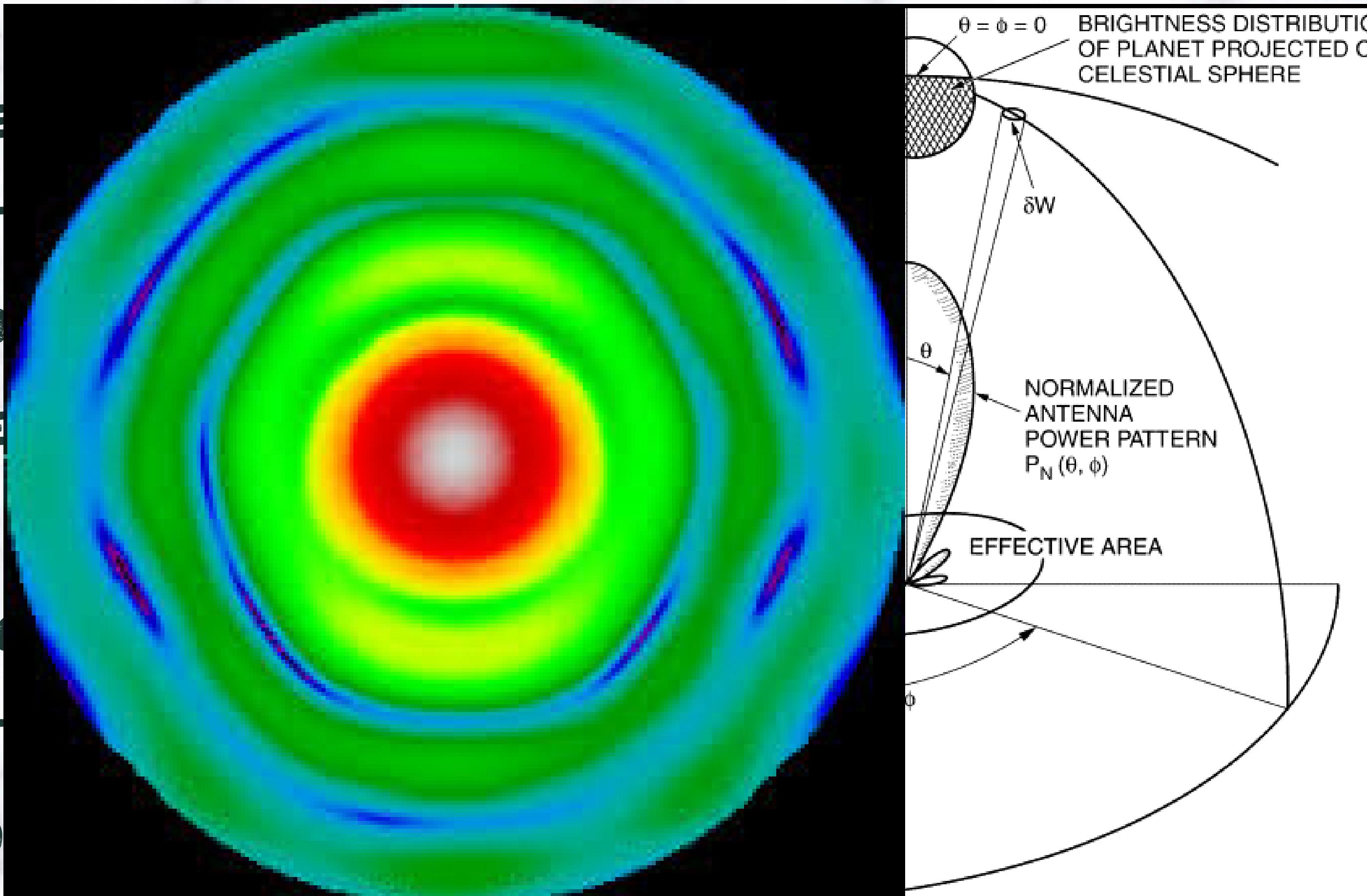
PRIMARY BEAM PATTERN

- SHAPE OF DISH DETERMINES ANTENNA SENSITIVITY PATTERN (YET MORE FOURIER TRANSFORMS)
- TIME AND FREQUENCY DEPENDENT!
- PRIMARY SOURCE OF DIRECTION DEPENDENT EFFECTS FOR MEERKAT
- THIS LARGELY DETERMINES THE REQUIRED FIELD OF VIEW



PRIMARY BEAM PATTERN

- SHAPE OF DISH DETERMINES SENSITIVITY PATTERN
- FOURIER TRANSFORM
- TIME AND FREQUENCY DEPENDENT EFFECTS
- PRIMARY SOURCE DEPENDENT EFFECTS
- THIS IS WHAT DETECTORS REQUIRE FIELD OF



IMAGING IN A NUTSHELL

IMAGING IN A NUTSHELL

DISCRETIZED MEASUREMENT MODEL FOR IMAGING

$$y = Rx + \epsilon, \quad \epsilon \sim \mathcal{N}(0, \Sigma)$$

IMAGING IN A NUTSHELL

DISCRETIZED MEASUREMENT MODEL FOR IMAGING

$$y = Rx + \epsilon, \quad \epsilon \sim \mathcal{N}(0, \Sigma)$$

DISTRIBUTION OF THE NOISE GIVES

$$\chi^2(x) \propto (y - Rx)^\dagger \Sigma^{-1} (y - Rx) \rightarrow \nabla_x \chi^2 = R^\dagger \Sigma^{-1} (y - Rx)$$

IMAGING IN A NUTSHELL

DISCRETIZED MEASUREMENT MODEL FOR IMAGING

$$y = Rx + \epsilon, \quad \epsilon \sim \mathcal{N}(0, \Sigma)$$

DISTRIBUTION OF THE NOISE GIVES

$$\chi^2(x) \propto (y - Rx)^\dagger \Sigma^{-1} (y - Rx) \rightarrow \nabla_x \chi^2 = R^\dagger \Sigma^{-1} (y - Rx)$$

SETTING THE GRADIENT TO ZERO

$$\nabla_x \chi^2 = 0 \rightarrow R^\dagger \Sigma^{-1} y = R^\dagger \Sigma^{-1} Rx \rightarrow I^D \approx I^{PSF} \star x$$

RESOLUTION

ROBUSTNESS

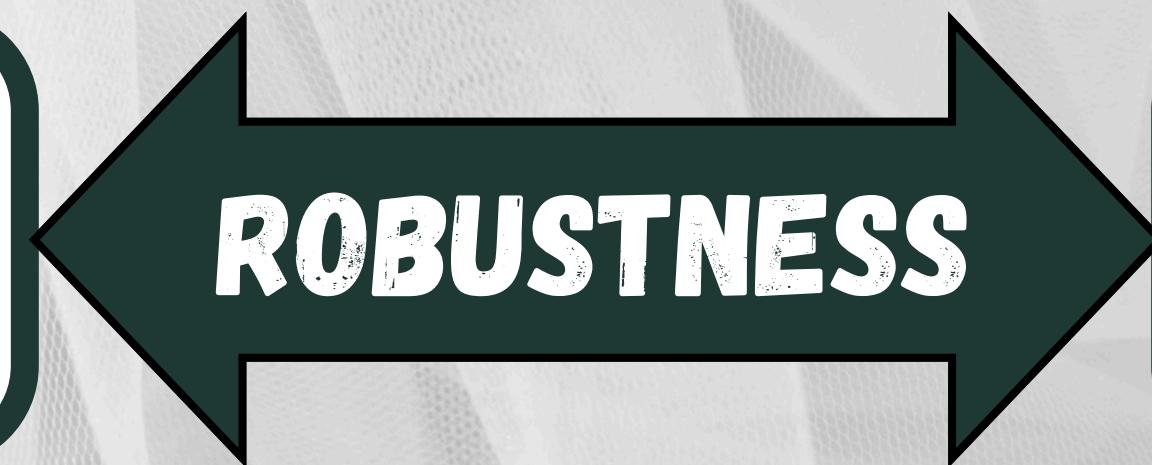
SENSITIVITY

UNIFORM

RESOLUTION

ROBUSTNESS

SENSITIVITY



UNIFORM

RESOLUTION

ROBUSTNESS

SENSITIVITY

NATURAL

UNIFORM

RESOLUTION

NATURAL

ROBUSTNESS

SENSITIVITY

WHY NOT BOTH?

MeerKAT Array Layout

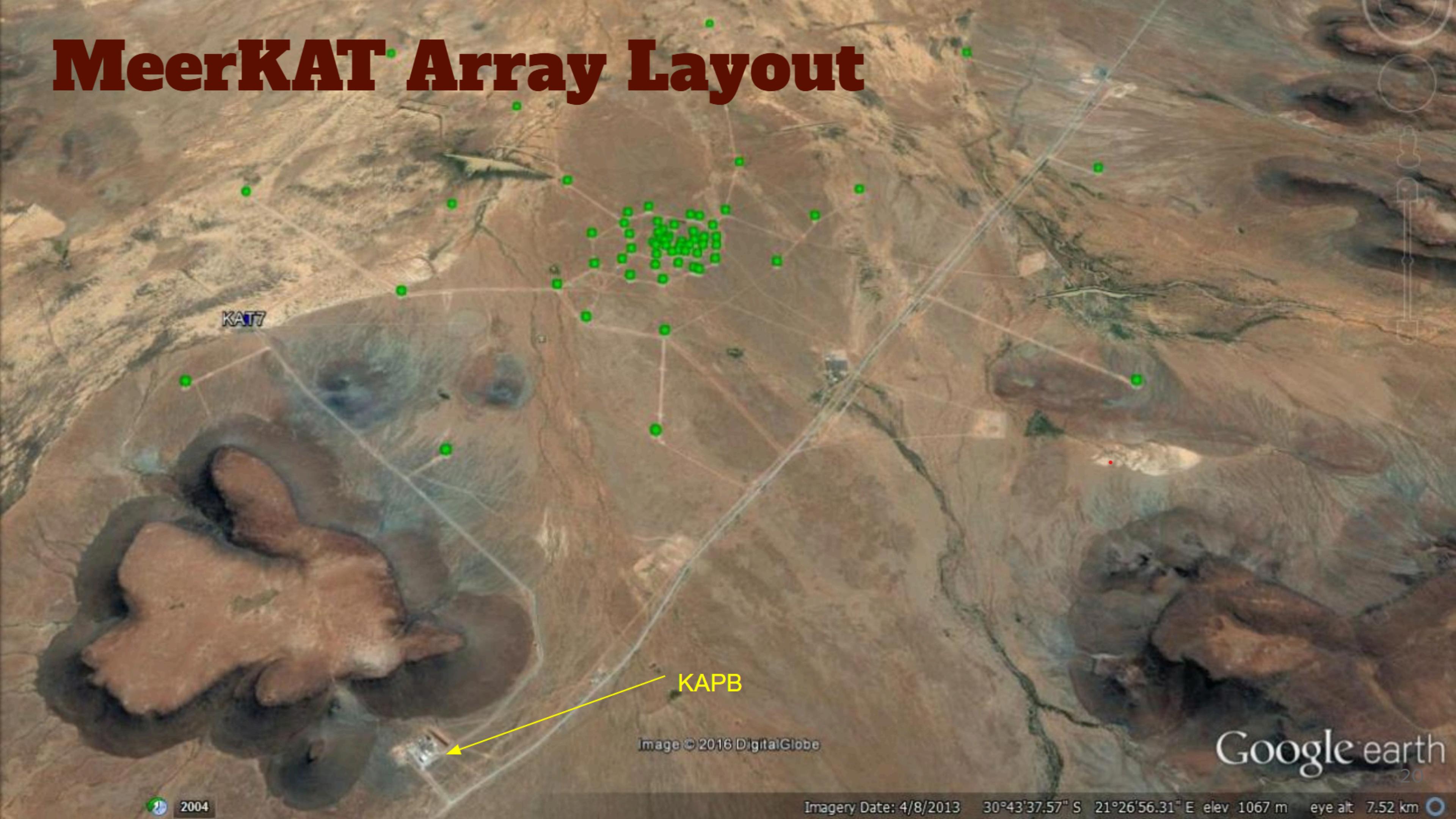
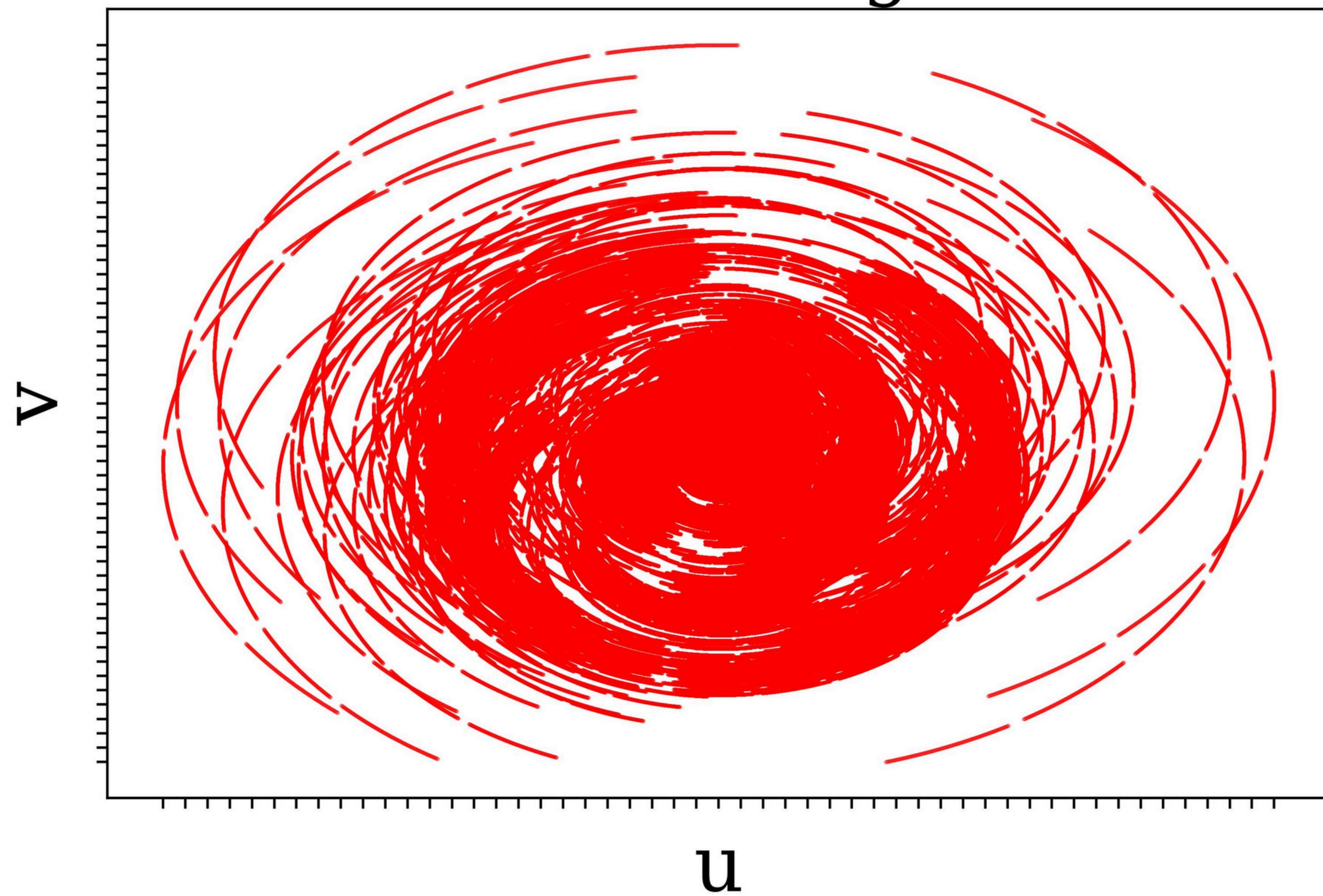


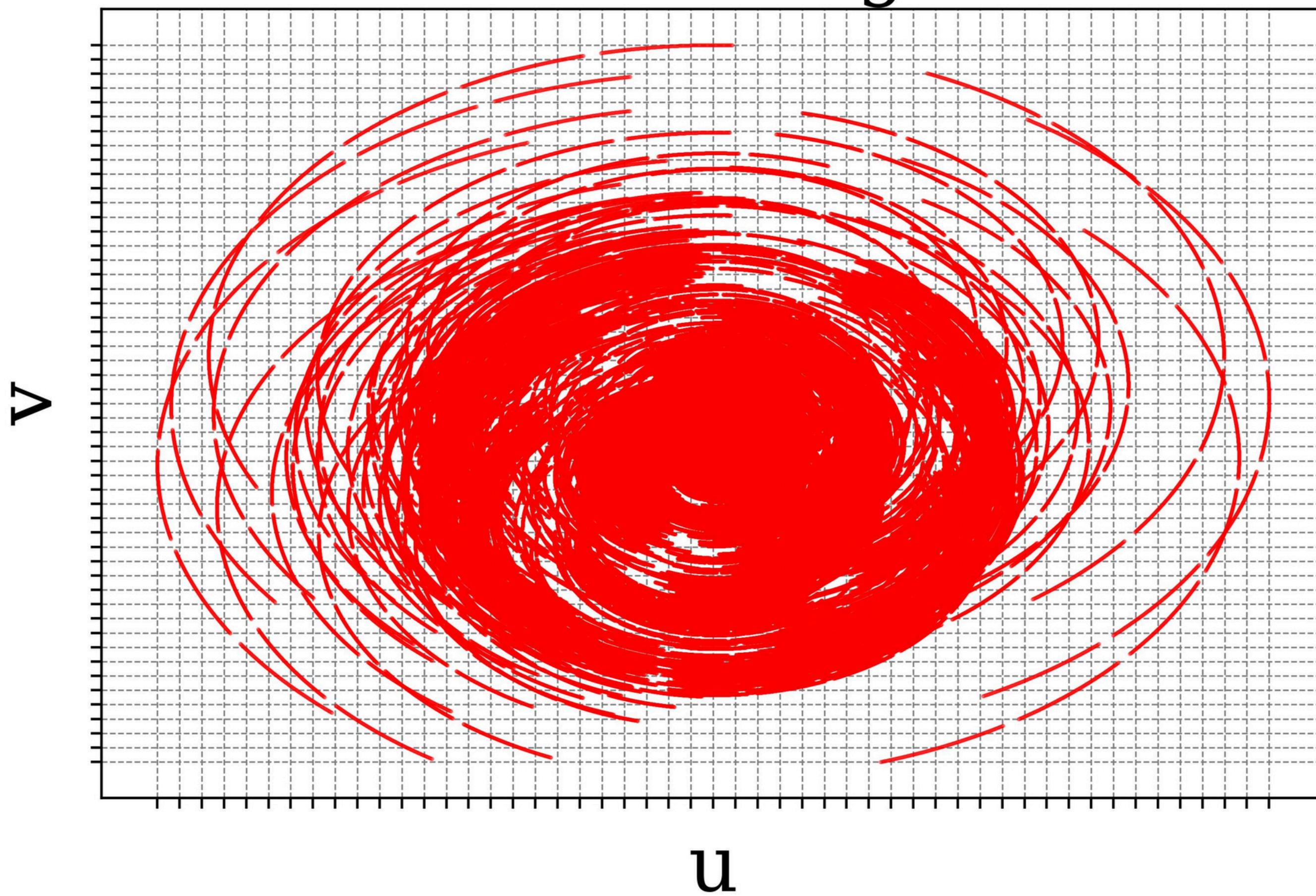
Image © 2016 DigitalGlobe

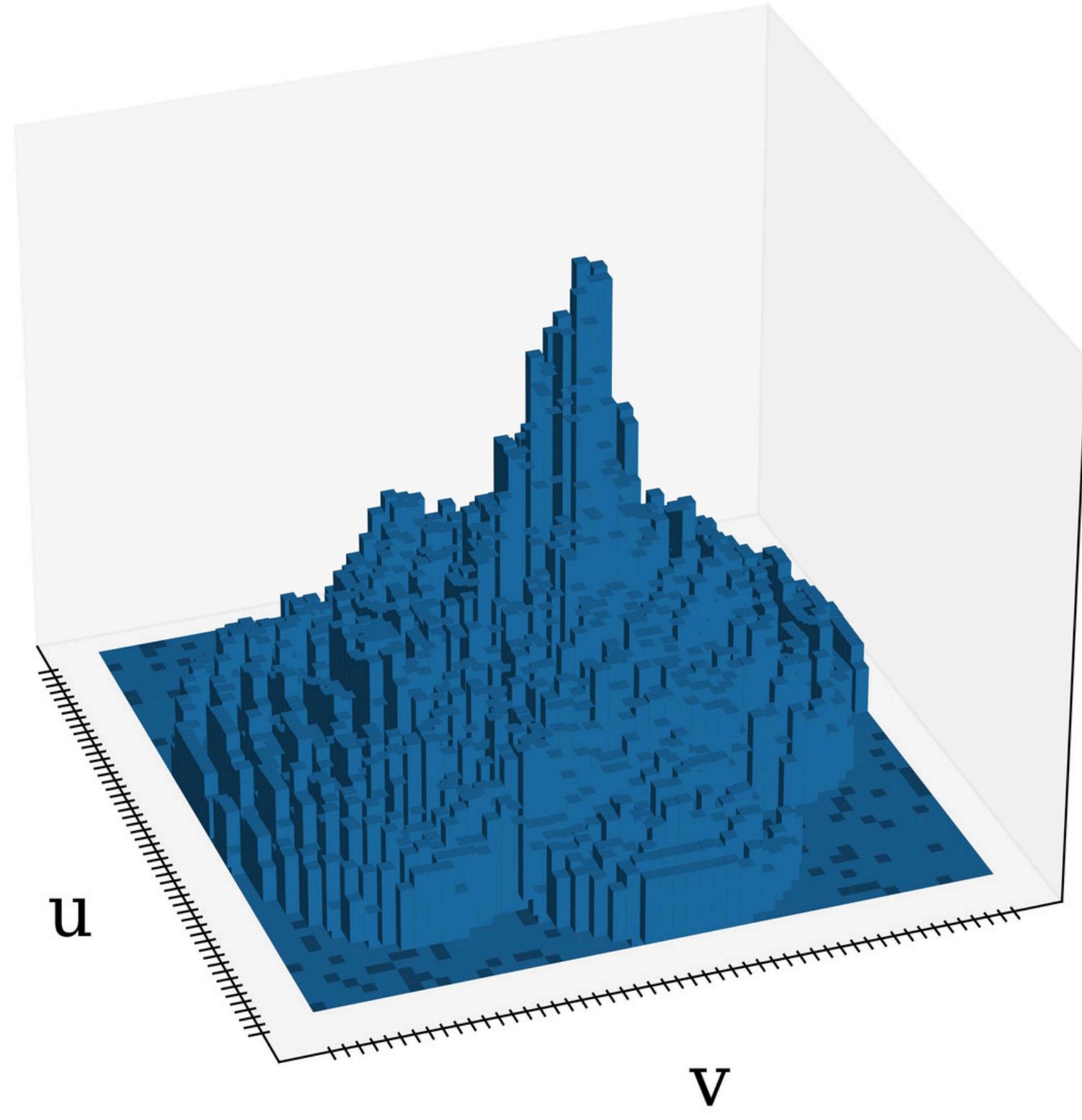
Google earth
20

uv-coverage



uv-coverage



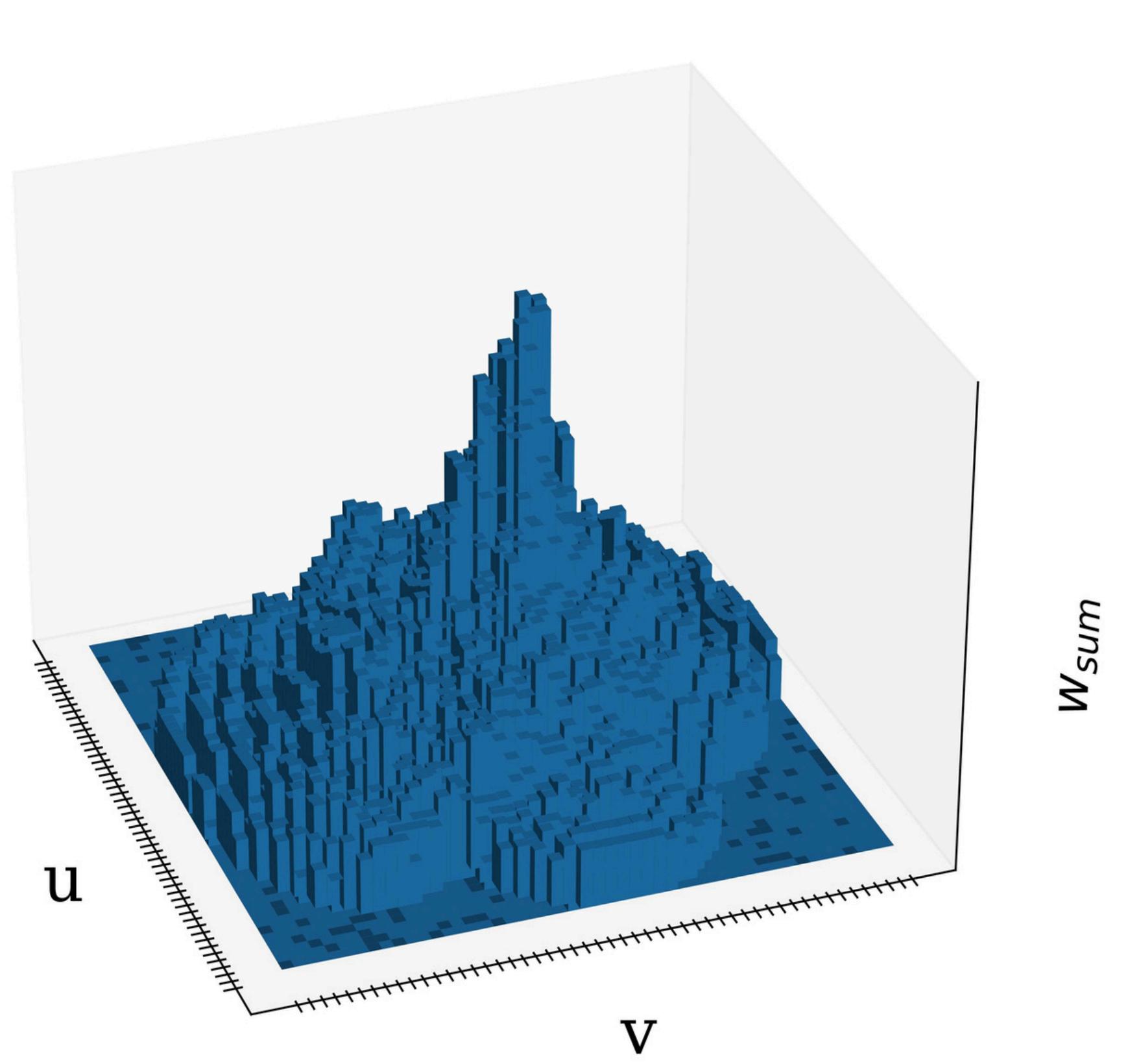


W_{sum}

u

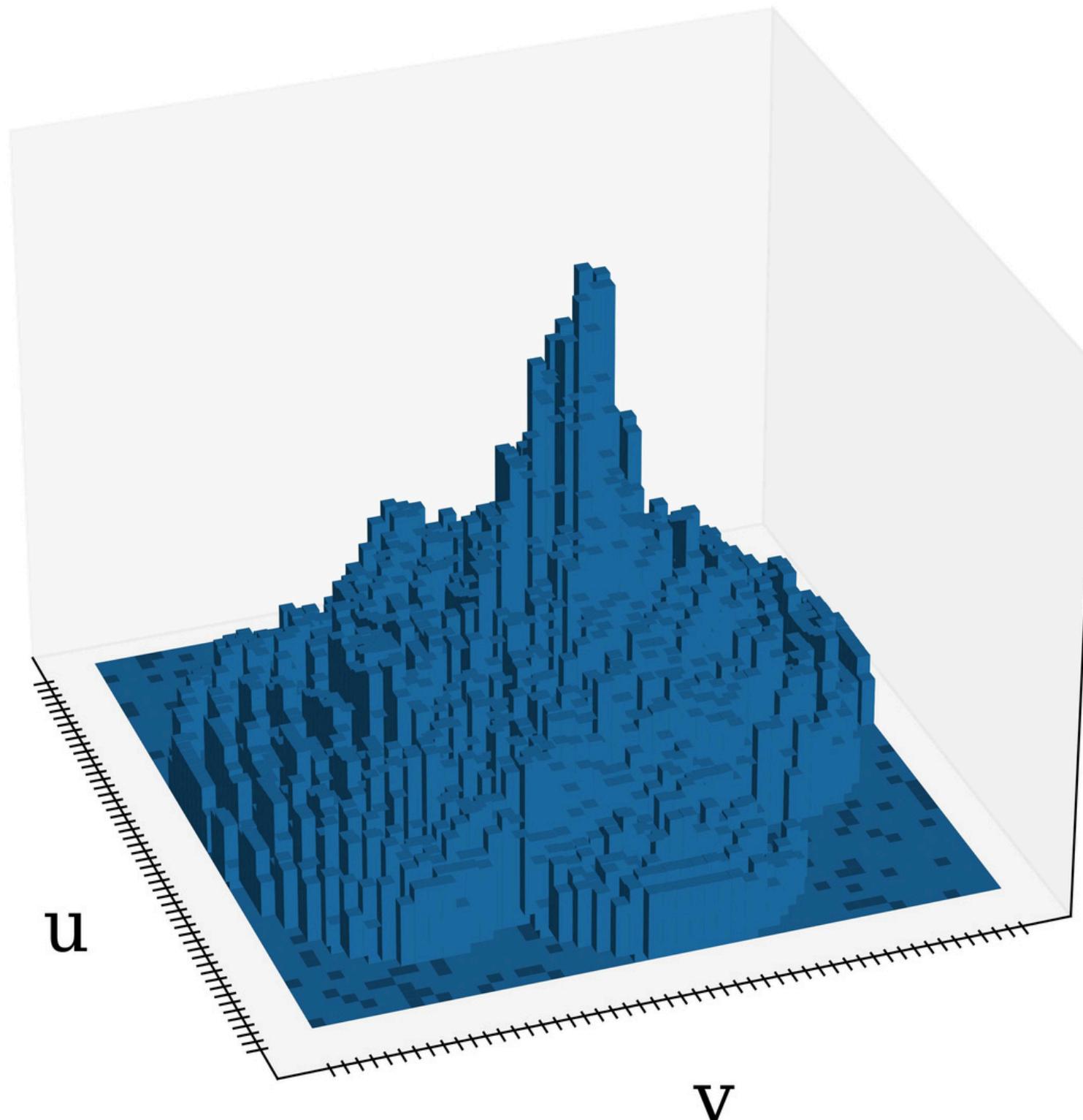
v

NATURAL WEIGHTING

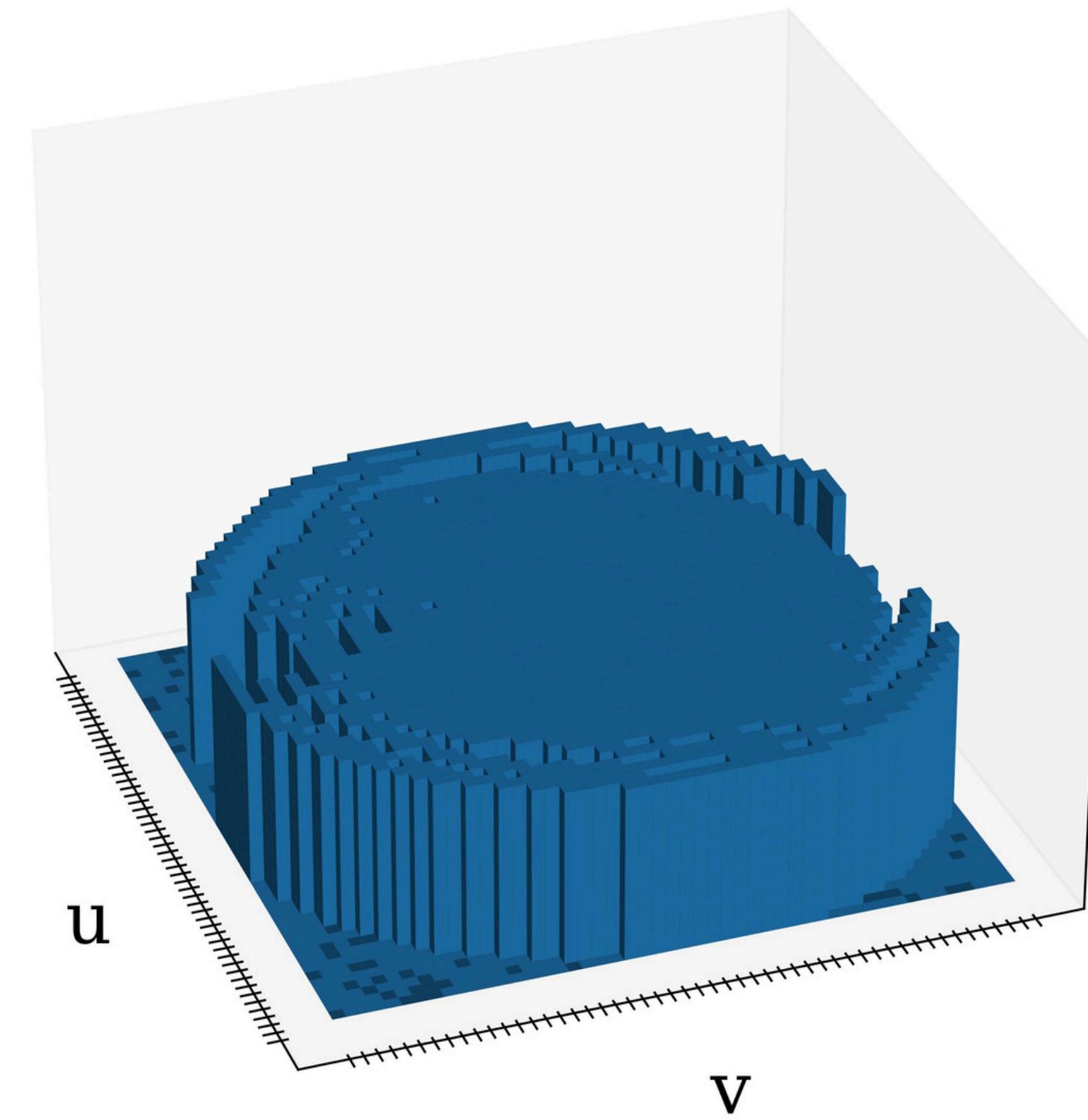


W_{sum}

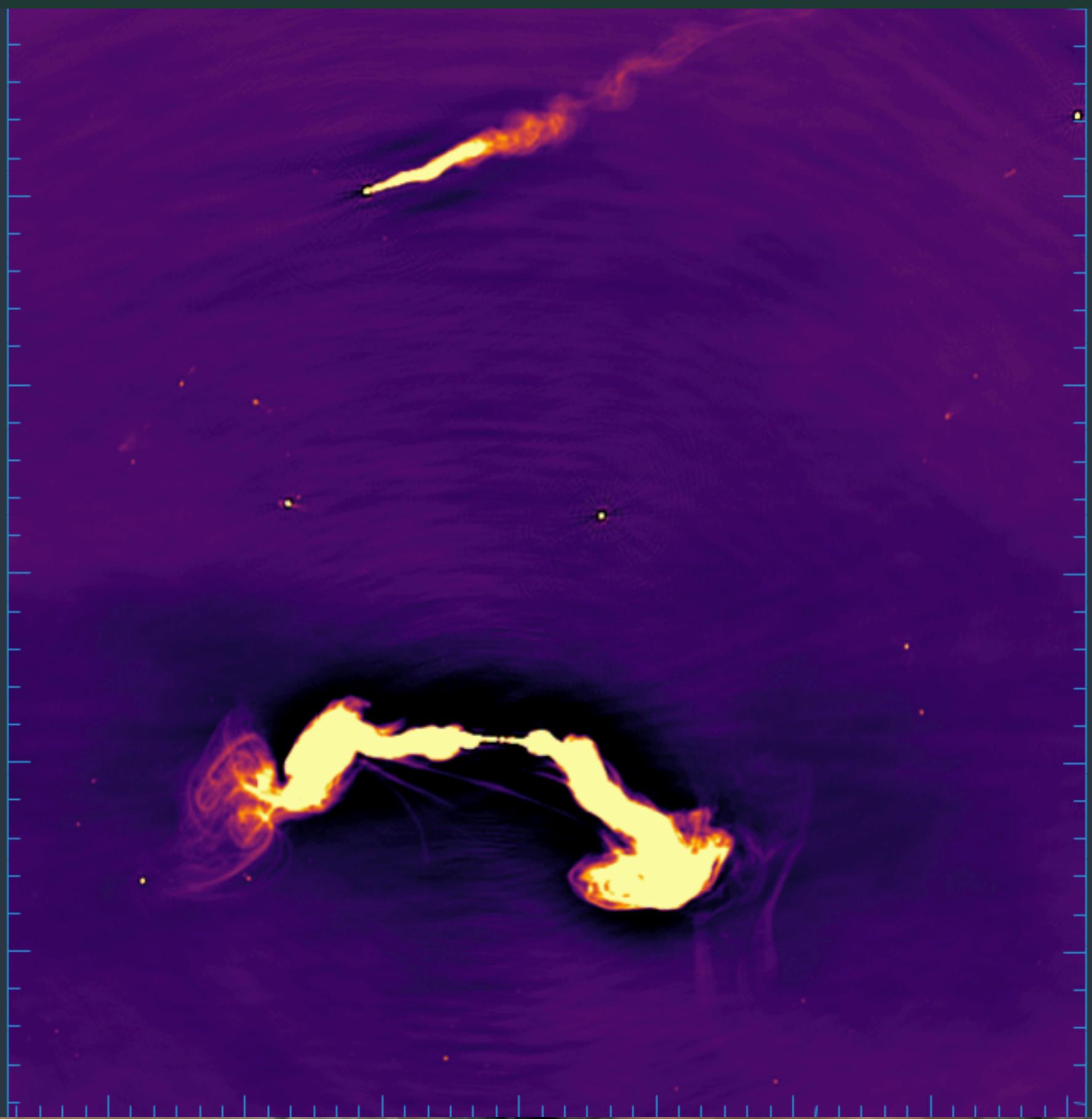
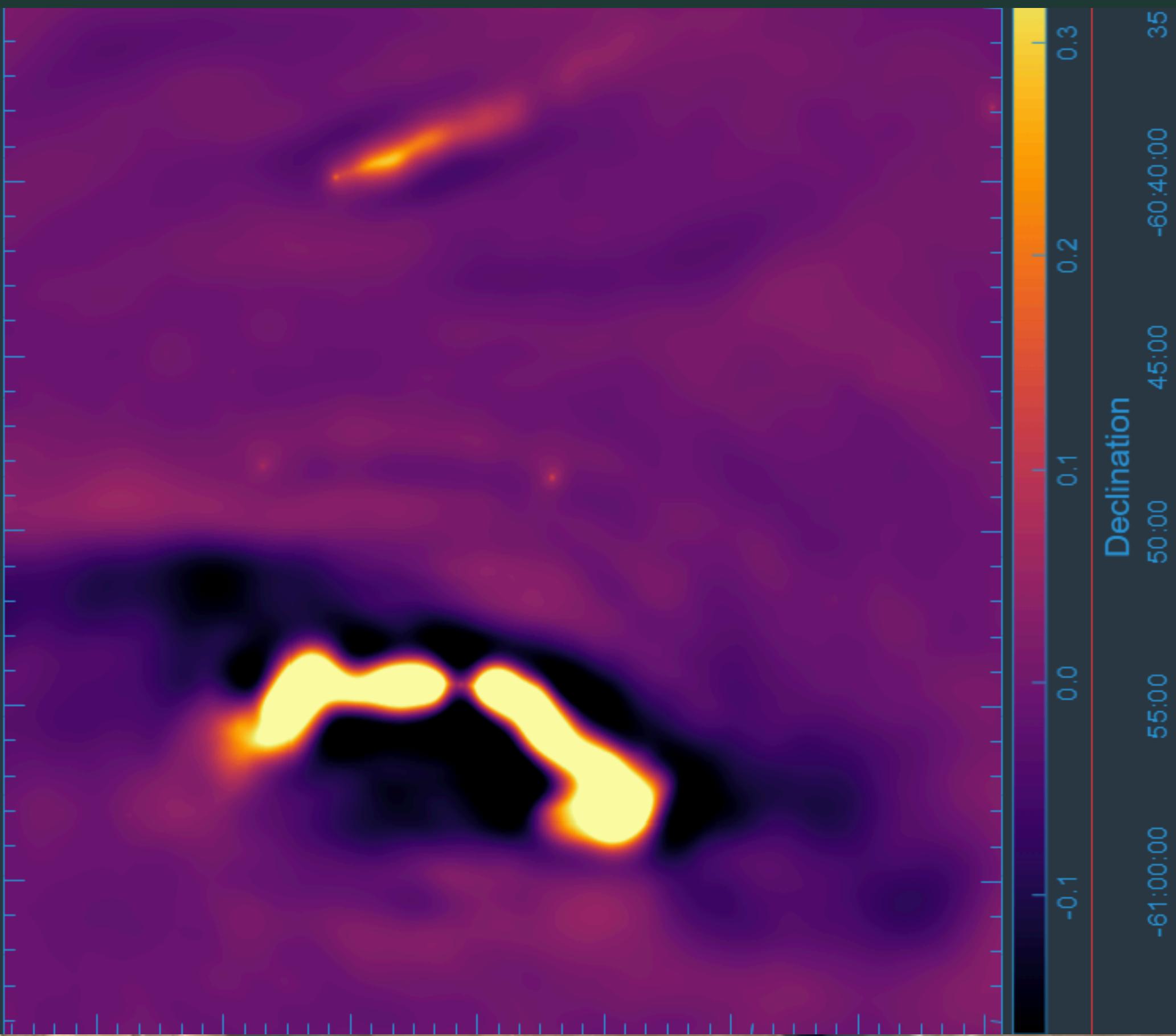
NATURAL WEIGHTING



UNIFORM WEIGHTING

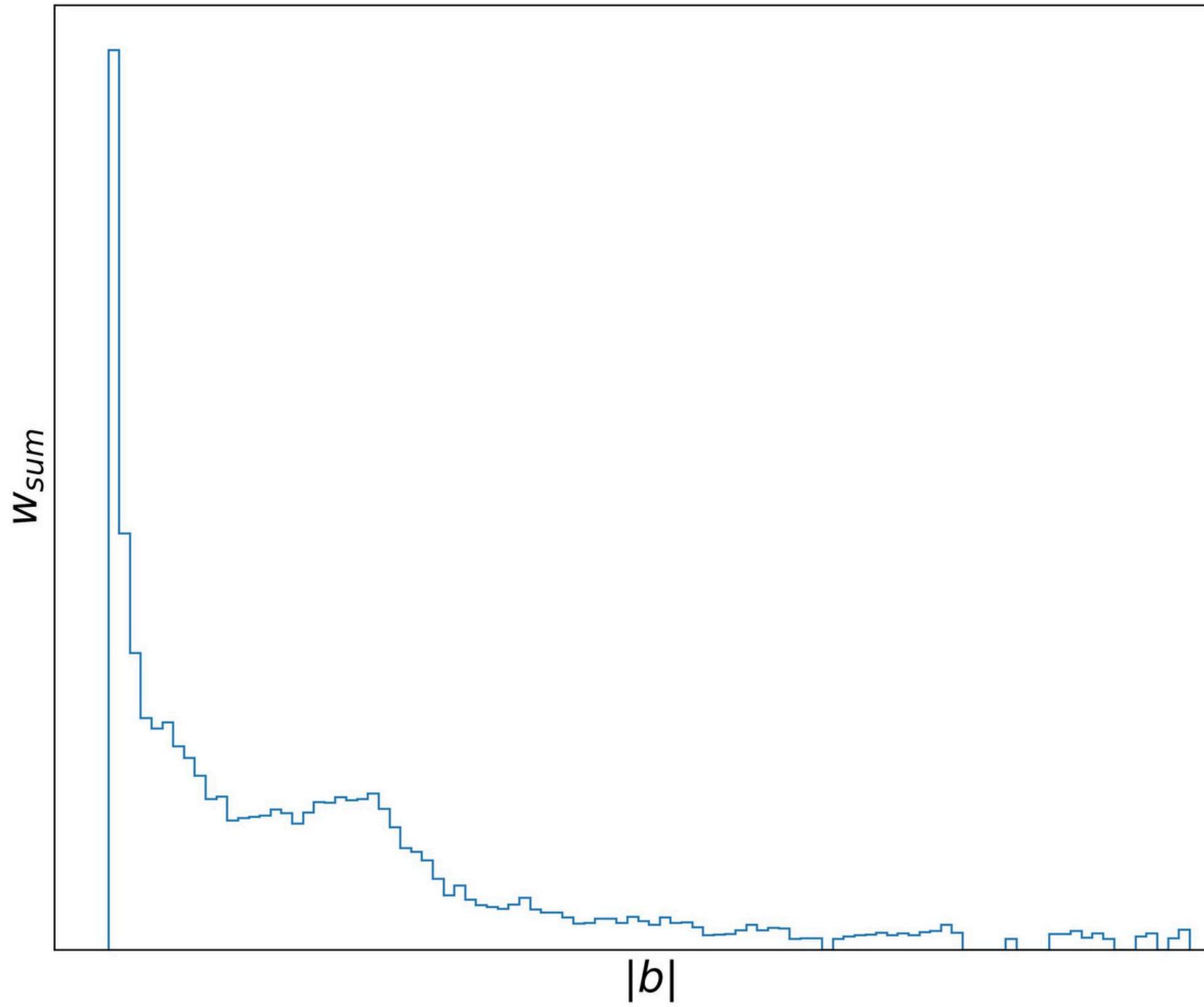


NATURAL VS UNIFORM

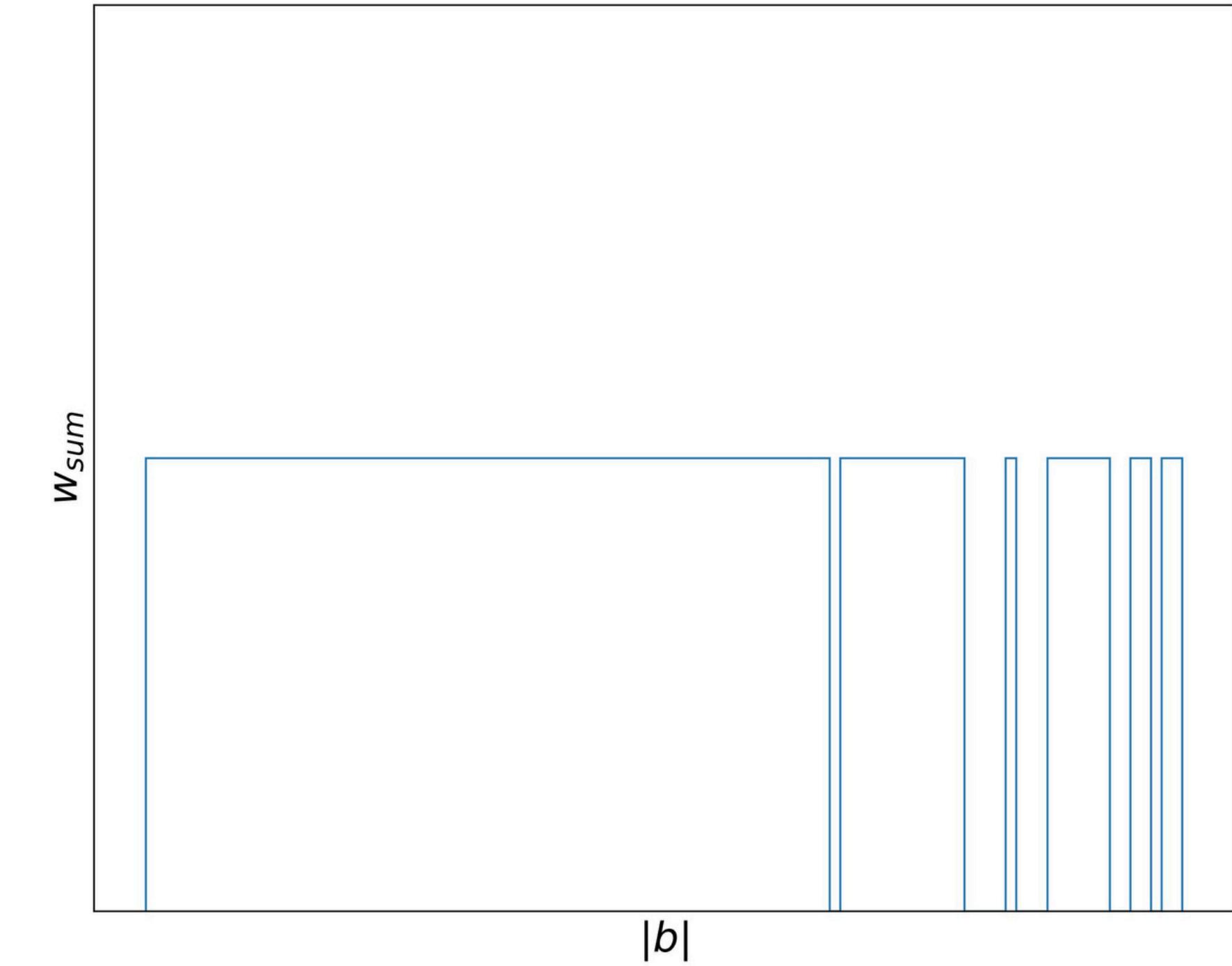


WHY DO WE LOSE SENSITIVITY?

NATURAL

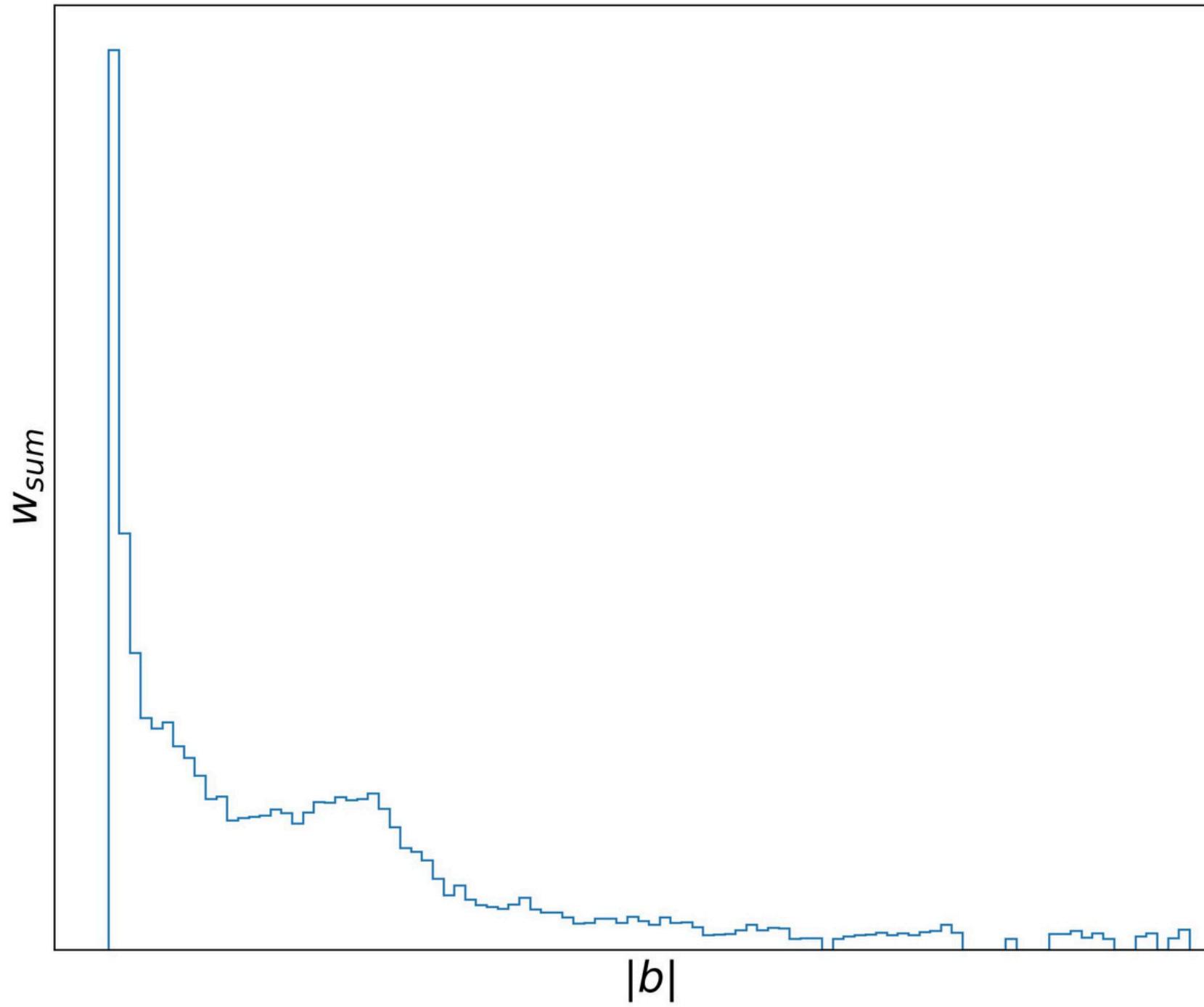


UNIFORM

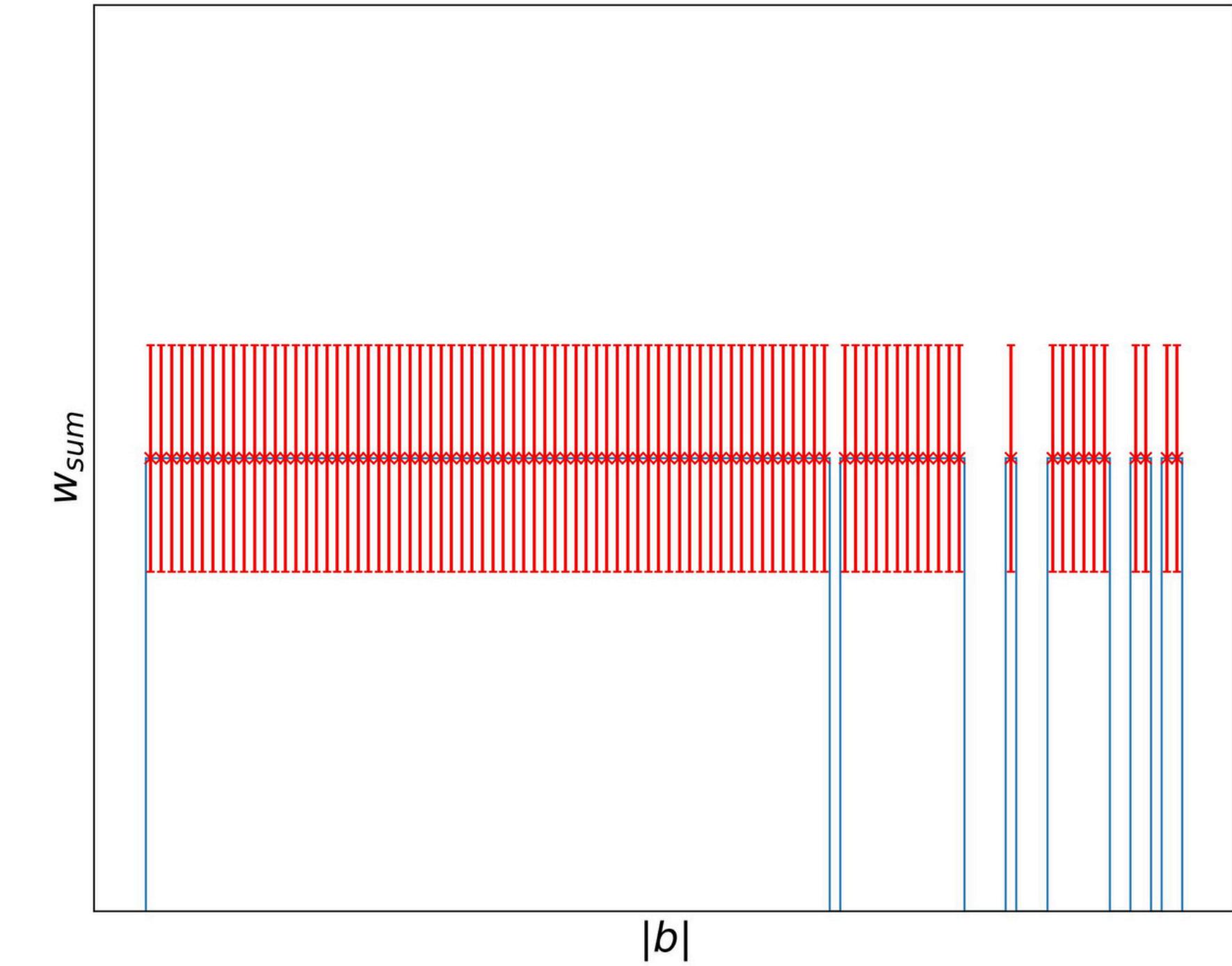


WHY DO WE LOSE SENSITIVITY?

NATURAL

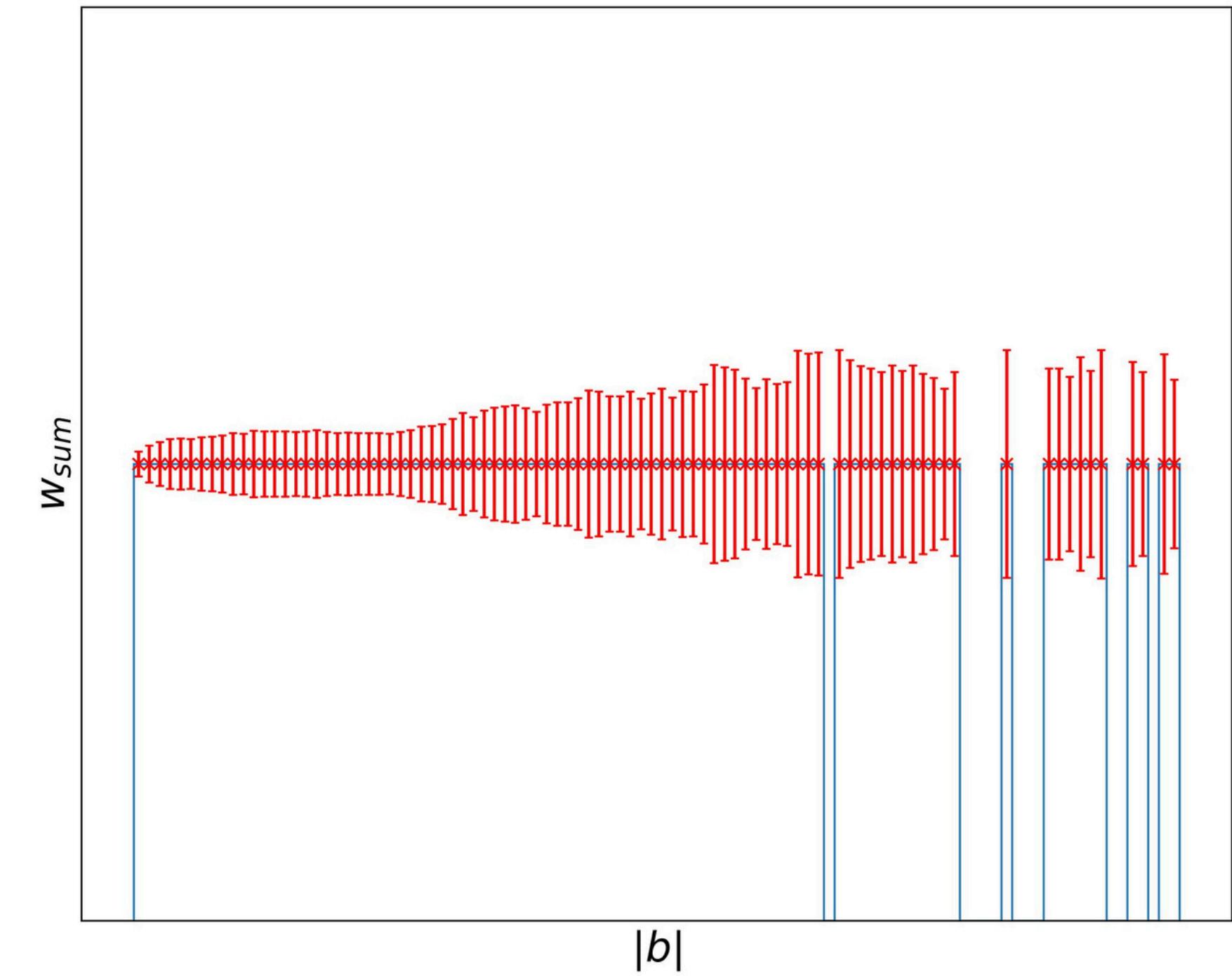
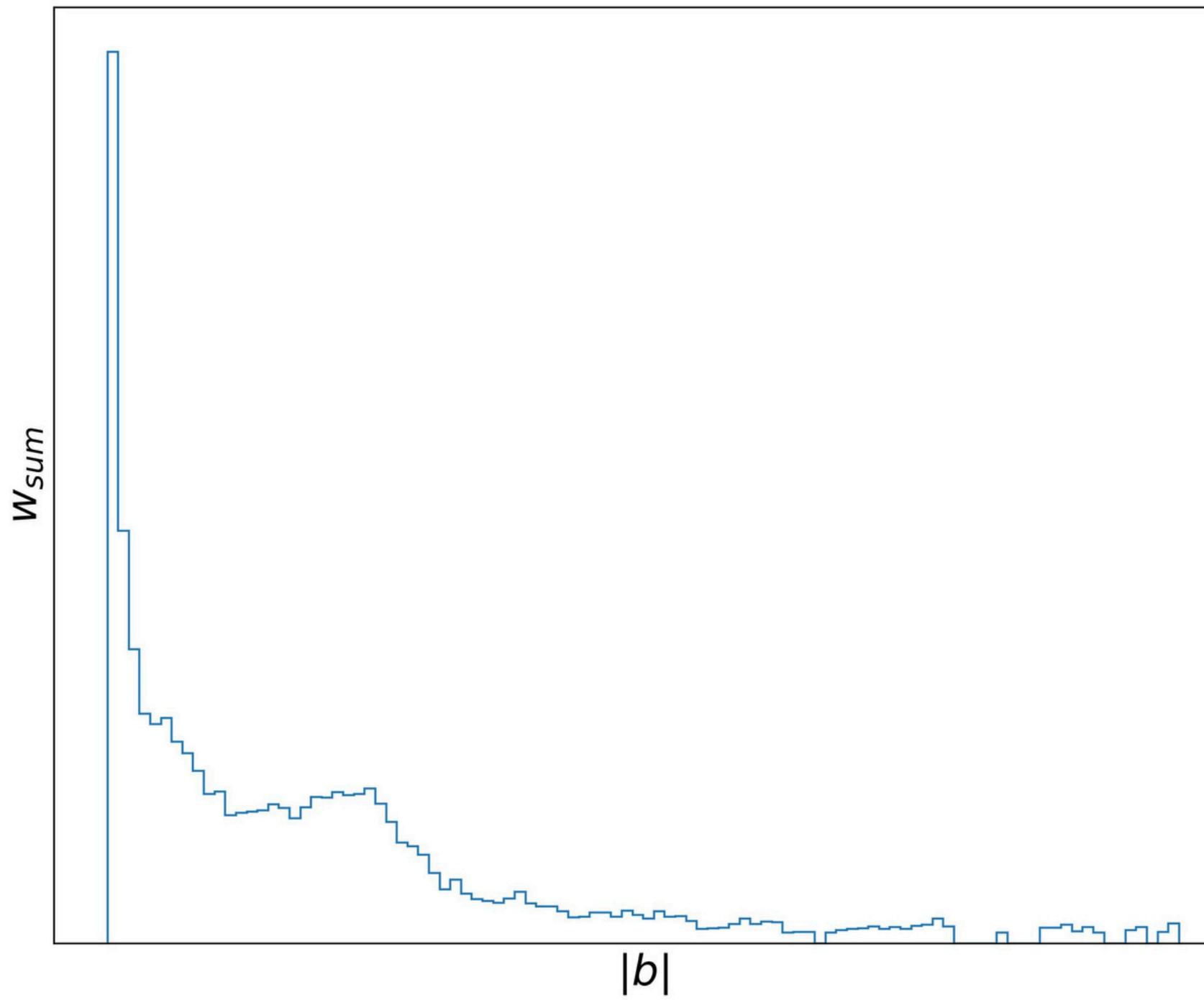


UNIFORM



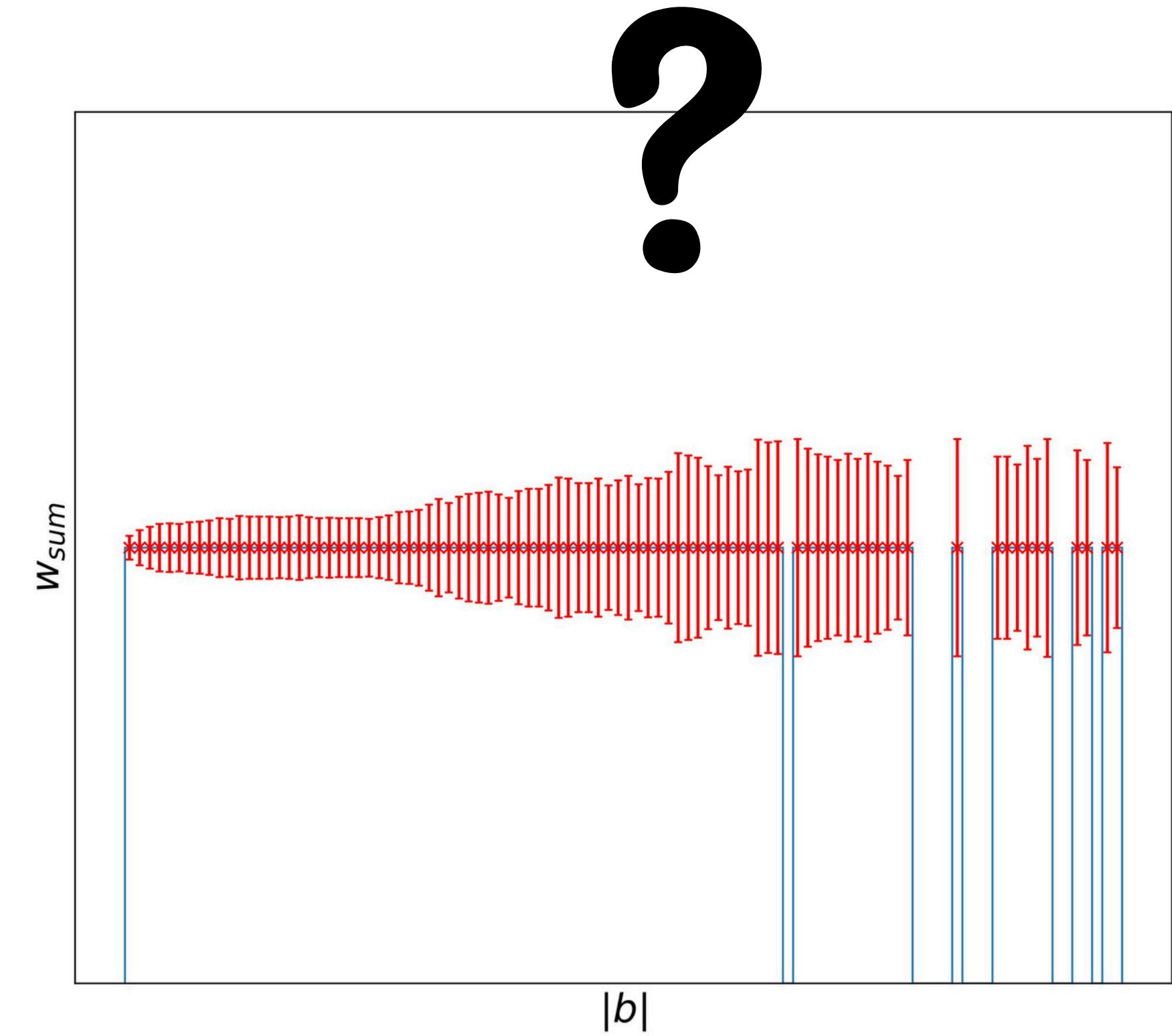
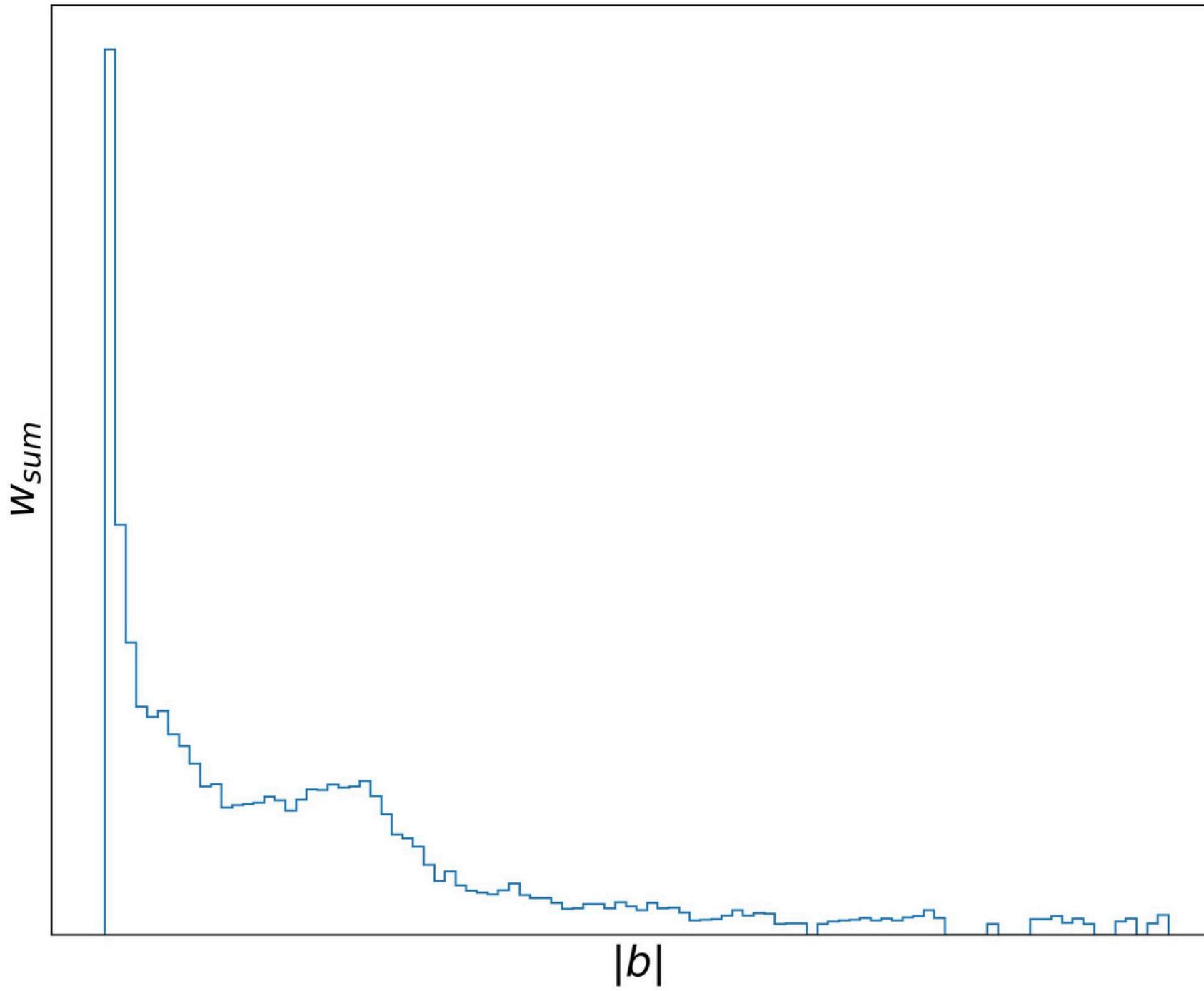
WHY DO WE LOSE SENSITIVITY?

NATURAL



WHY DO WE LOSE SENSITIVITY?

NATURAL



DECONVOLUTION AS A LINEAR MODEL

$$I^D = I^{PSF} * x = F^\dagger \hat{I}^{PSF} F^{\text{micro}} x$$

DECONVOLUTION AS A LINEAR MODEL

$$I^D = I^{PSF} * x = F^\dagger \hat{I}^{PSF} F x$$

$$x \approx (F^\dagger \hat{I}^{PSF} F)^{-1} I^D = F^\dagger (\hat{I}^{PSF})^{-1} F I^D$$

DECONVOLUTION AS A LINEAR MODEL

$$I^D = I^{PSF} * x = F^\dagger \hat{I}^{PSF} \text{ WEIGHTED AND SUMMED DATA}$$

$$x \approx (F^\dagger \hat{I}^{PSF} F)^{-1} I^D = F^\dagger (\hat{I}^{PSF})^{-1} \underbrace{FI^D}_{\text{miro}}$$

DECONVOLUTION AS A LINEAR MODEL

$$I^D = I^{PSF} * x = F^\dagger \hat{I}^{PSF} F x$$

$$x \approx (F^\dagger \hat{I}^{PSF} F)^{-1} I^D = F^\dagger (\hat{I}^{PSF})^{-1} F I^D$$

SUM OF
WEIGHTS

DECONVOLUTION AS A LINEAR MODEL

$$I^D = I^{PSF} * x = F^\dagger \hat{I}^{PSF} F x$$

$$x \approx (F^\dagger \hat{I}^{PSF} F)^{-1} I^D = F^\dagger (\hat{I}^{PSF})^{-1} F I^D$$

Unfortunately not that simple:

- Ill-posed problem (gaps in uv-coverage)
- Non-periodic boundary conditions
- Want physics informed regularisation
- Not just a deconvolution problem

SOLVING THE INVERSE PROBLEM

- USE BAYESIAN PRINCIPLES TO FORMULATE OPTIMISATION PROBLEM
- NATURAL GRADIENT DETERMINES SCALE EMPHASIS (WEIGHT)
- STRENGTH OF PRIOR RELATIVE TO LIKELIHOOD INCORPORATES SCALE UNCERTAINTY (SETTING HYPERPARAMETERS)
- CAN BE DONE VERY EFFICIENTLY USING THE APPROXIMATION

$$I^D \approx I^{PSF} * x$$

A practical preconditioner for wide-field continuum imaging of radio interferometric data

Hertzog L. Bester,^{1,2} Audrey Repetti,³ Simon Perkins,^{1,2} Oleg M. Smirnov,^{2,1} and Jonathan S. Kenyon^{2,1}

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²*Rhodes University, Makhanda (Grahamstown), Eastern Cape, South Africa*

³*Institute of Sensors, Signals and Systems, Heriot-Watt University, Edinburgh, United Kingdom*

Abstract. The celebrated CLEAN algorithm has been the cornerstone of deconvolution algorithms in radio interferometry almost since its conception in the 1970s. For all its faults, CLEAN is remarkably fast, robust to calibration artefacts and in its ability to model point sources. We demonstrate how the same assumptions that afford CLEAN its speed can be used to accelerate more sophisticated deconvolution algorithms.

ARVIX:2101.08072

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Africanus III. pfb-imaging - a flexible radio interferometric imaging suite

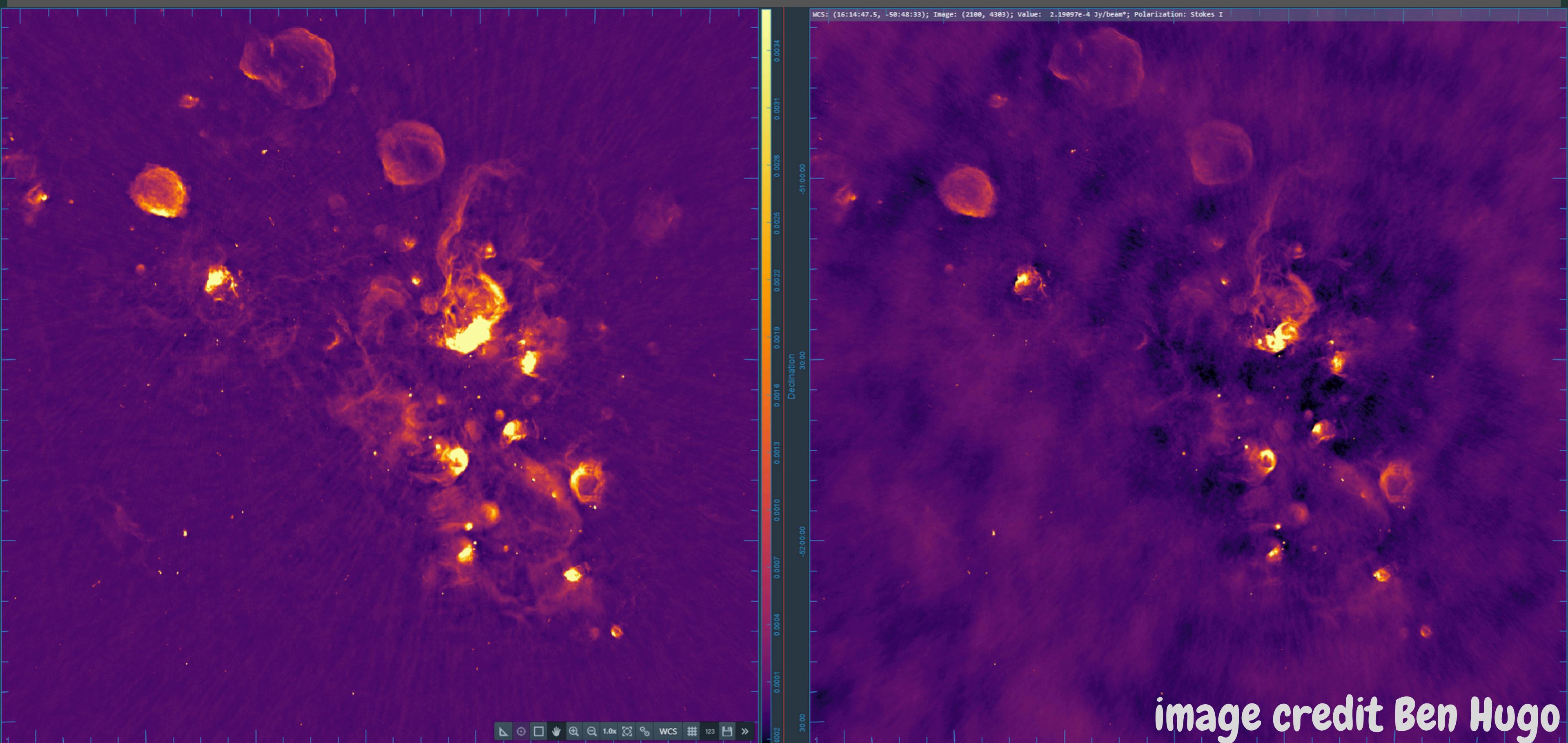
H. L. Bester^{a,b}, J. S. Kenyon^b, S. J. Perkins^a, O. M. Smirnov^{b,a,c}

^a*South African Radio Astronomy Observatory (SARAO), Cape Town, WC, South Africa*

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^c*Institute for Radioastronomy, National Institute of Astrophysics (INAF IRA), Bologna, Italy*

SPARSITY PRIOR VS MULTI-SCALE CLEAN



SPARSITY PRIOR VS MULTI-SCALE CLEAN

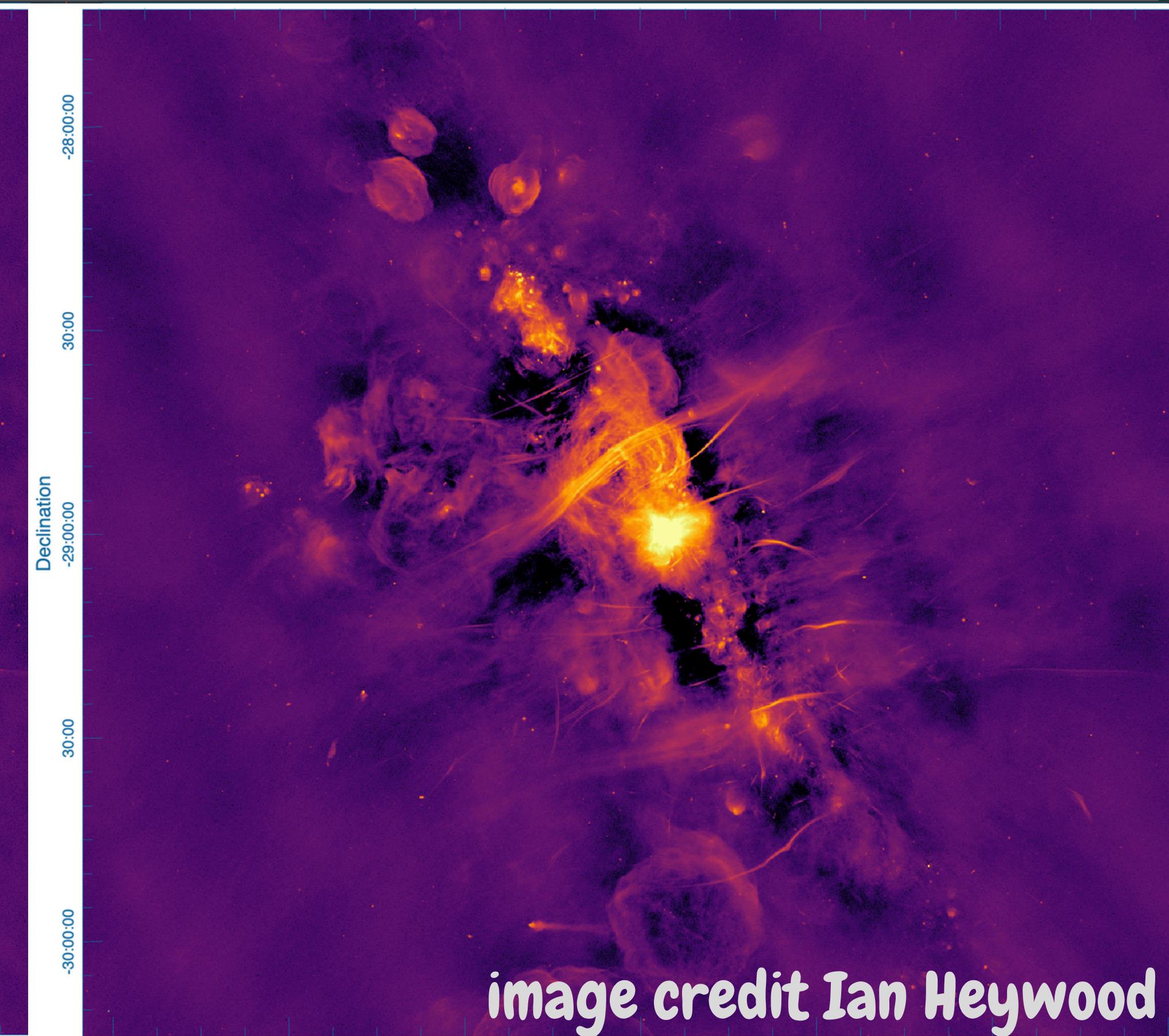
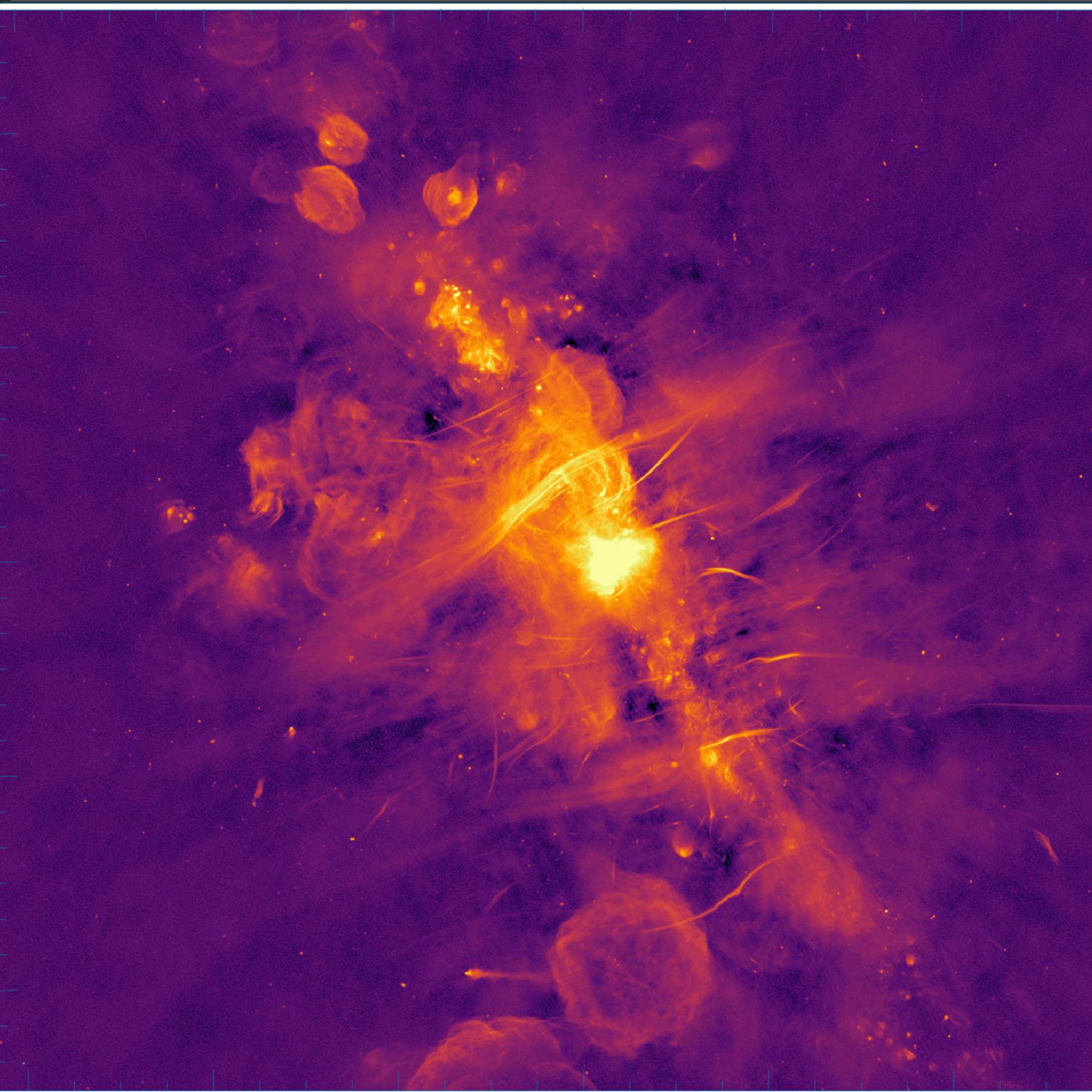


image credit Ian Heywood