

WP 4: Computation

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spectroscopy

expectation-maximization

model equation:

$$y = H \cdot x + b$$

DRM and bremsstrahlung:

$$H = \frac{\bar{n}V}{4\pi R^2} D \cdot Q.$$

likelihood:

$$p(y|x) = \prod_{i \in R} \frac{e^{-(Hx+b)_i}}{y_i!} (Hx + b)_i^{y_i}$$

discrepancy:

$$KL(y, x) = \sum_{i \in R} y_i \log \frac{y_i}{(Hx + b)_i} + (Hx + b)_i - y_i$$

constrained optimization:

$$\min KL(y, x) \quad | \quad x \geq 0$$

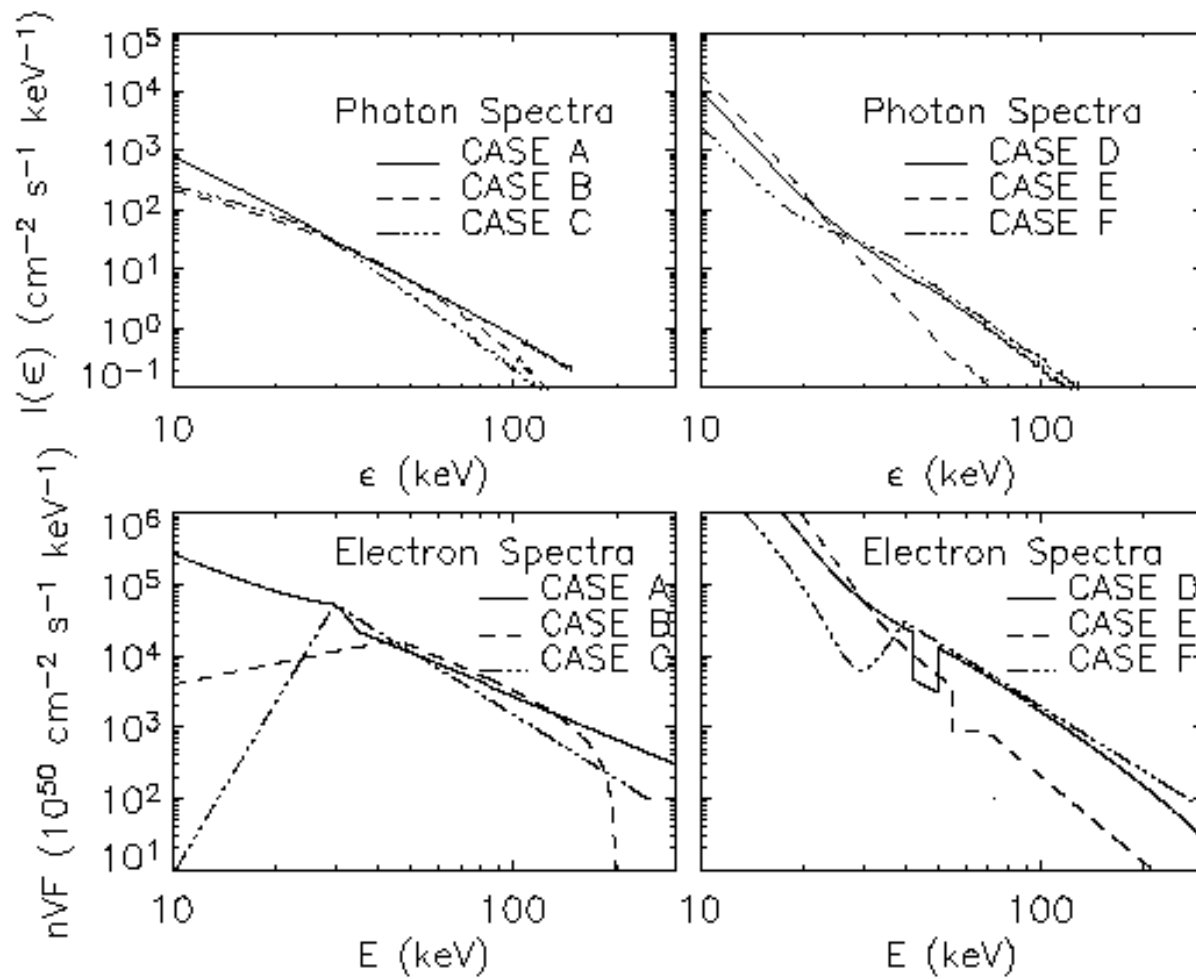
successive approximation:

$$x^{k+1} = \frac{x^k}{H^T \mathbf{I}} H^T \left(\frac{y}{Hx^k + b} \right)$$

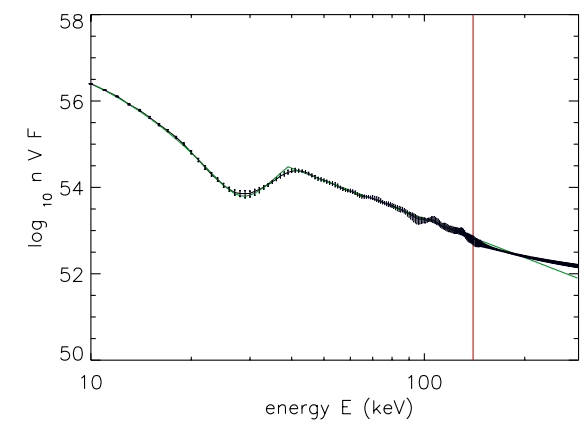
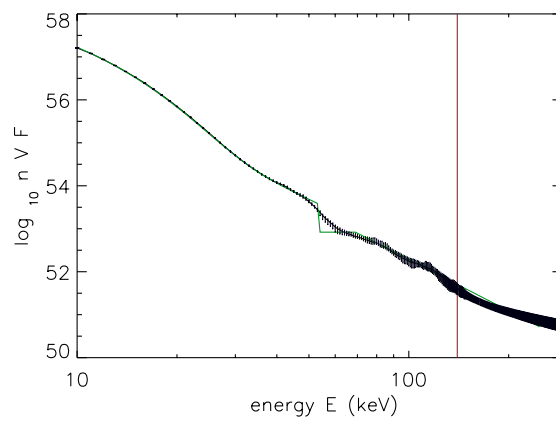
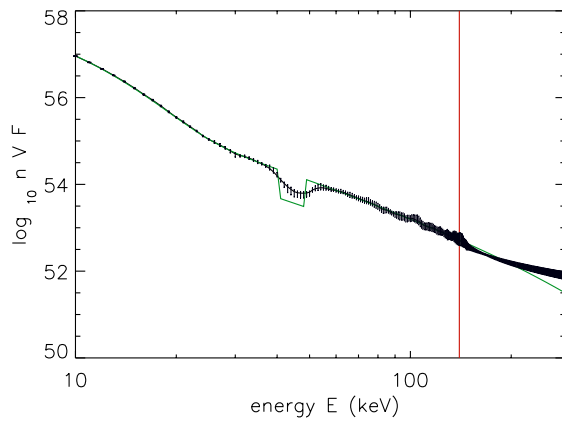
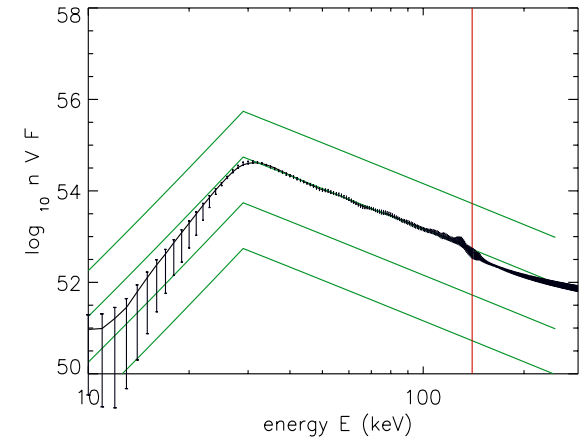
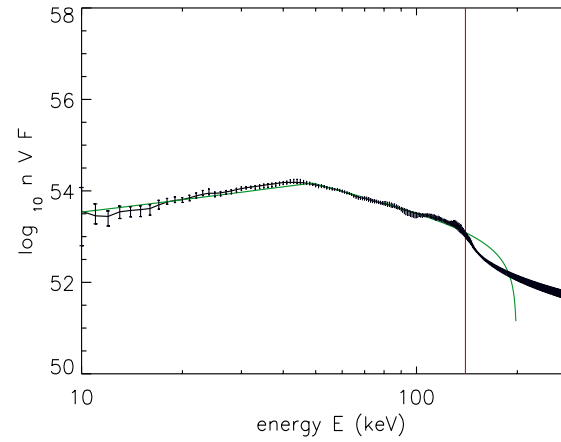
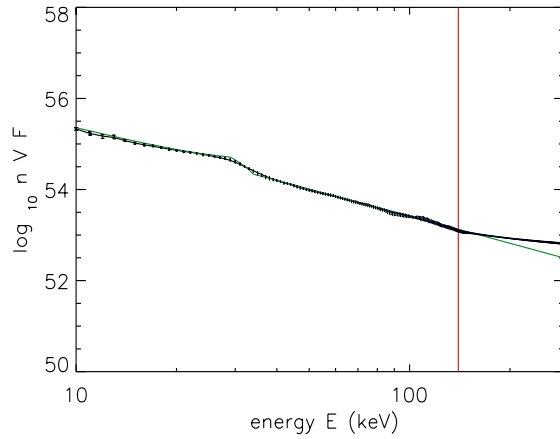
cumulative residuals:

$$S^{(j)}(q) = \frac{1}{j} \sum_{i=1}^j r_i^{(k)}(q),$$

blind test - 1



blind test - 2



uncertainties - 1

- forward-fit of photon spectra by means of parameterized models
- an MCMC method samples the parameter space to compute the posterior distribution
- detailed confidence intervals for the parameters

Ireland J, Tolbert A K, Schwartz R A, Holman G D and Dennis B R, ApJ, 2013

uncertainties 2

Bayesian Confidence Limits of electron spectra

(G. Emslie and A.M Massone)

- The expectation values of the source brightness and its variance in a given photon energy bin are in general not equal to the number n of counts observed in that energy bin but depend on n and on prior knowledge of the overall photon spectrum
- Using Bayesian analysis we computed the posterior probability for an event with n^* counts given an event with n counts and with a given expectation values
- We used this result to determine the confidence strip in a mean electron flux spectrum computed by means of regularization

the ion problem -1

$$I(\varepsilon) = \sum_k \sum_j \int_0^{\infty} F_j(E) A_k j_{kj}(\varepsilon, E) dE$$

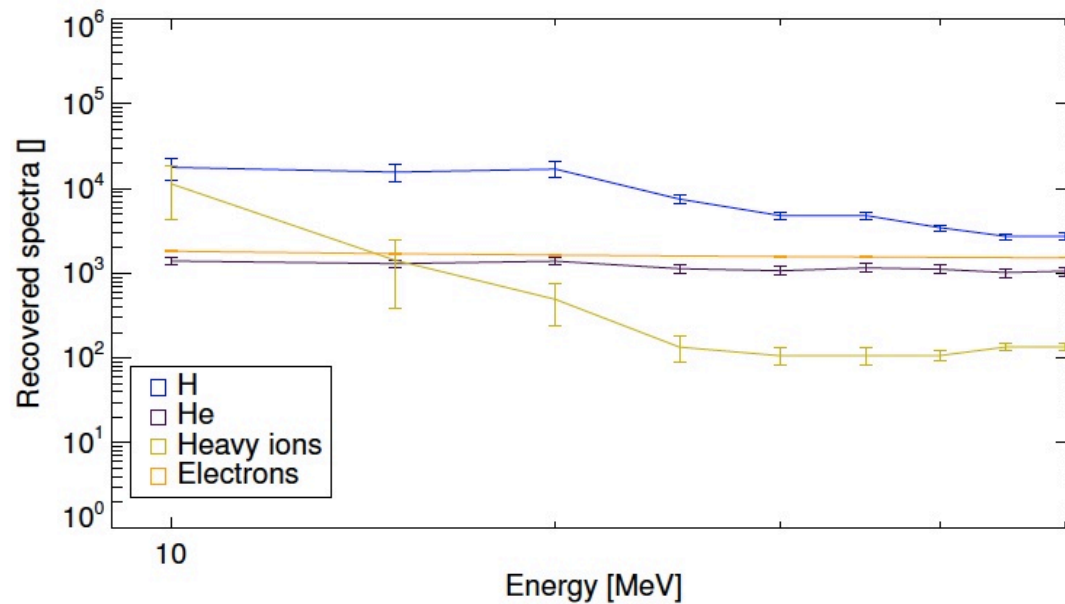
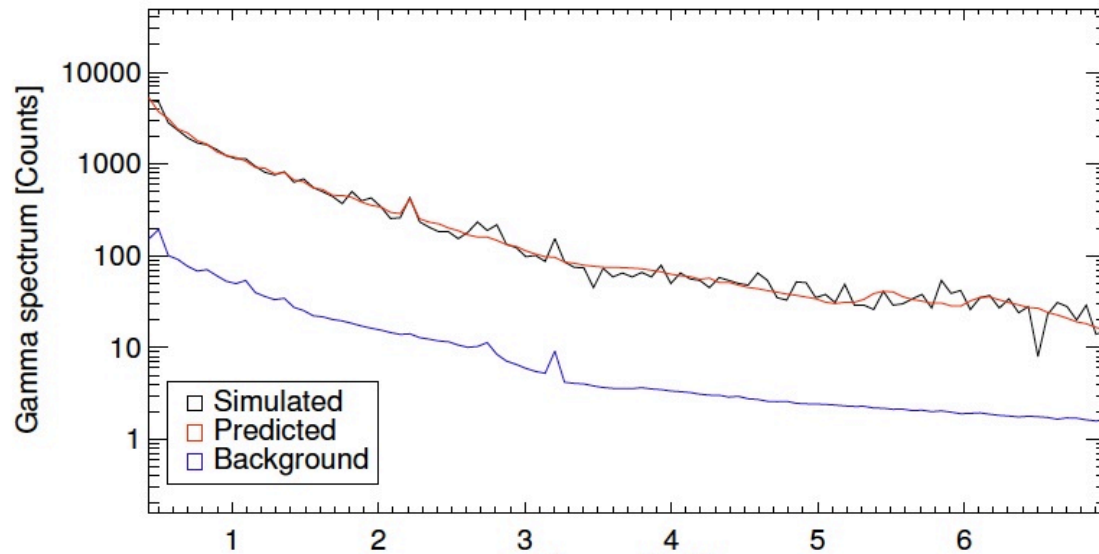
$$y = aHx$$

find (a, x) such that: $KL(x, a), y) = \min$

$$x_j \leftarrow x_j \frac{1}{\sum_{i,p} a_p H_{ij}^p} \sum_{i,p} a_p H_{ij}^p \frac{y_i}{\sum_{q,p} a_p H_{iq}^p x_q} ,$$

$$a_p \leftarrow a_p \frac{1}{\sum_{i,j} H_{ij}^p x_j} \sum_{i,j} H_{ij}^p x_j \frac{y_i}{\sum_{q,p} a_p H_{iq}^p x_q} .$$

the ion problem - 2



the ion problem -3

target

Recovered	Type
1.0 ± 0.0	H
0.0 ± 0.0	He3
0.000093 ± 0.000009	He
0.00006 ± 0.00008	C
0.000008 ± 0.000004	N
0.00011 ± 0.00012	O
0.00003 ± 0.00002	Ne
0.00055 ± 0.00018	Mg
0.0003 ± 0.0007	Al
0.00019 ± 0.00010	Si
0.00185 ± 0.00012	S
$9.0 \times 10^{-12} \pm 2.9 \times 10^{-8}$	Ca
0.000016 ± 0.000004	Fe

accelerated particles

Recovered	Type
1.0 ± 0.4	H
0.14 ± 0.04	He
0.010 ± 0.005	C
0.012 ± 0.005	N
1.8 ± 0.7	O
0.011 ± 0.006	Ne
0.0026 ± 0.0010	Mg
0.024 ± 0.018	Al
0.025 ± 0.011	Si
0.015 ± 0.007	S
0.00003 ± 0.00009	Ca
0.011 ± 0.004	Fe

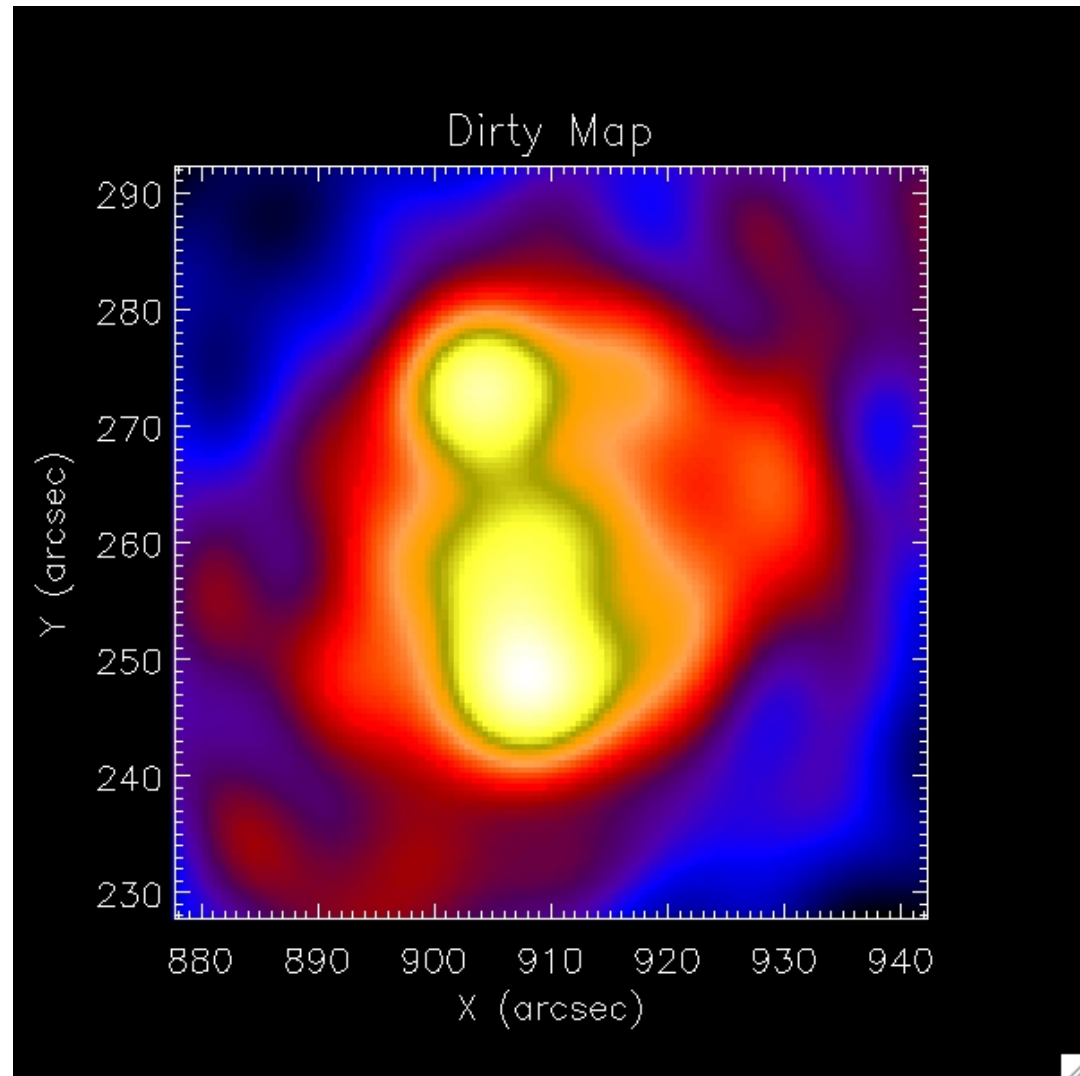
imaging

Back-Projection from visibilities instrument-independent

vis_bpmap

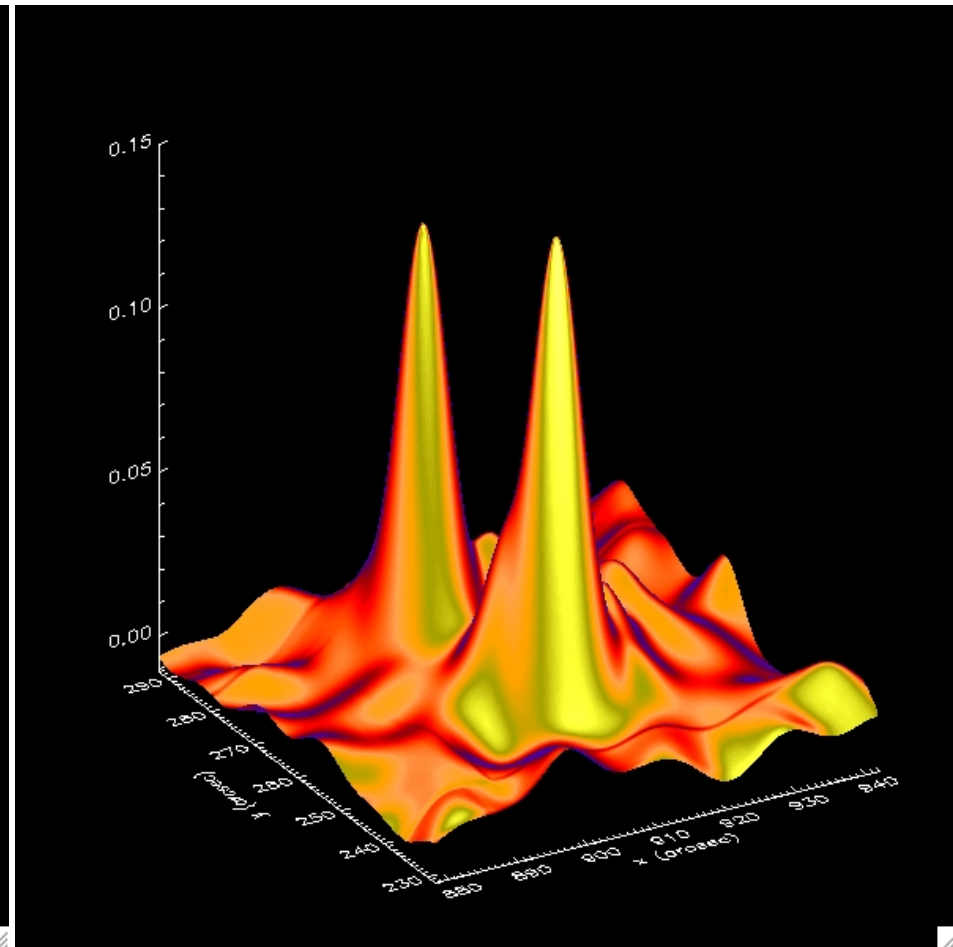
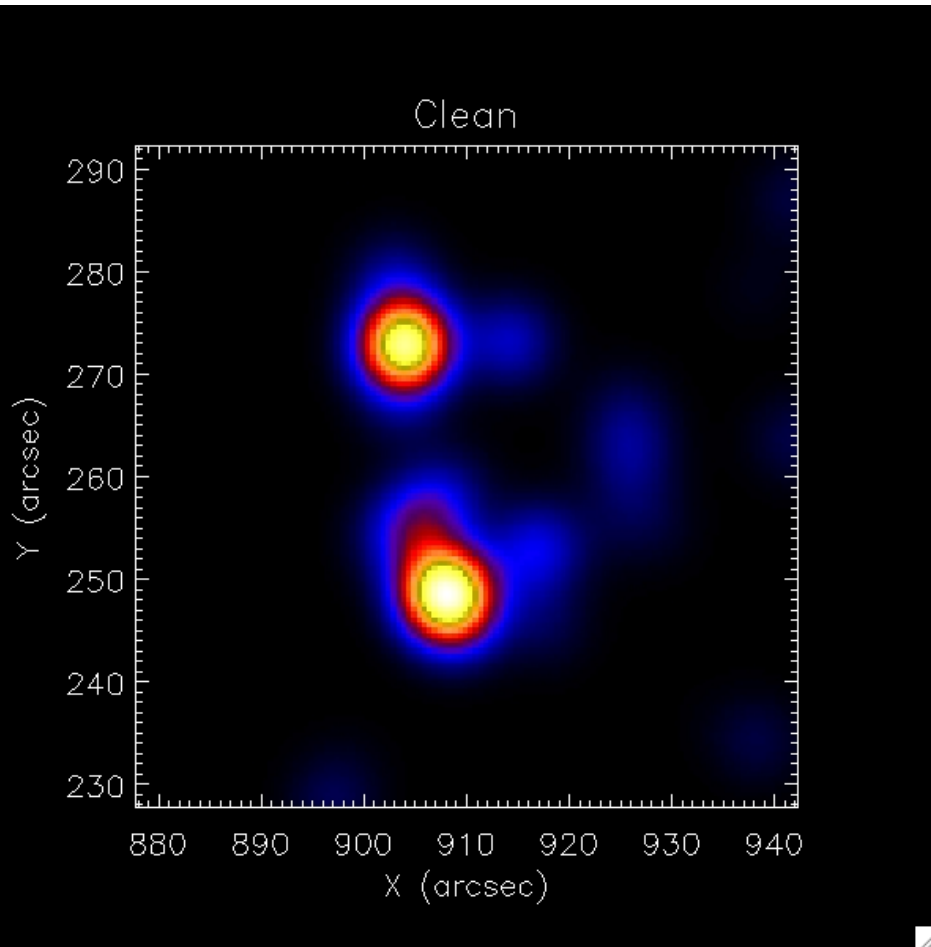
20 February 2002

(11:06:02 - 11:06:34 UT)



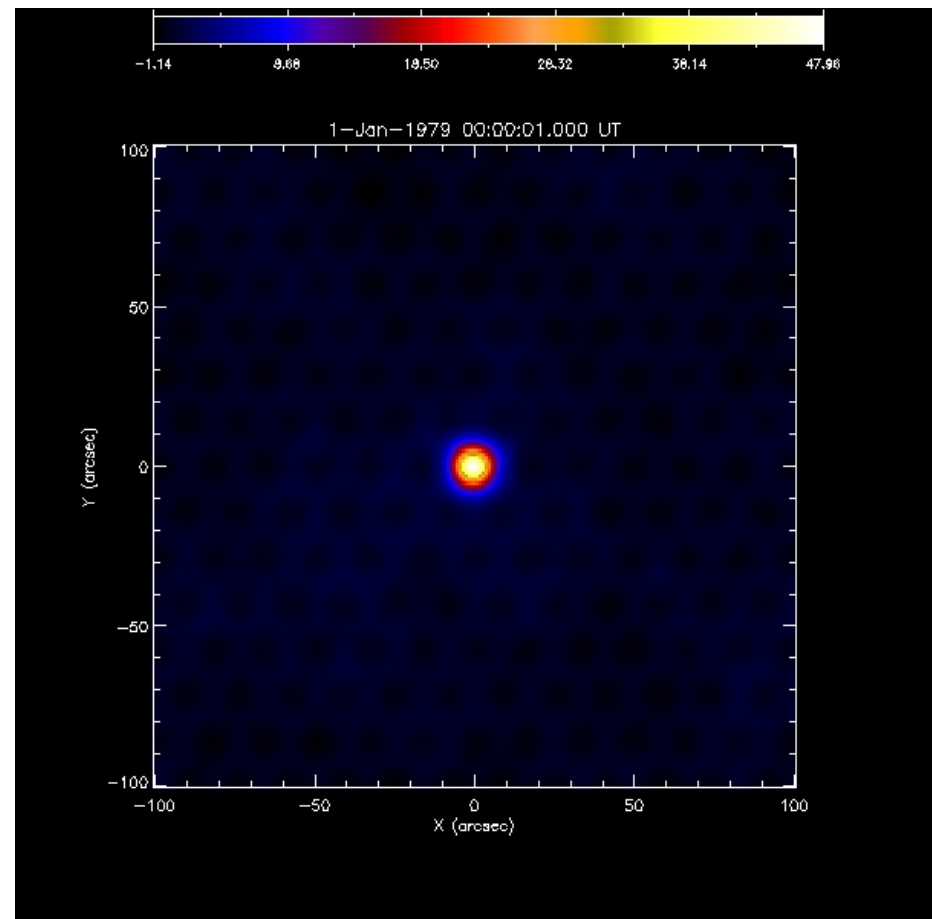
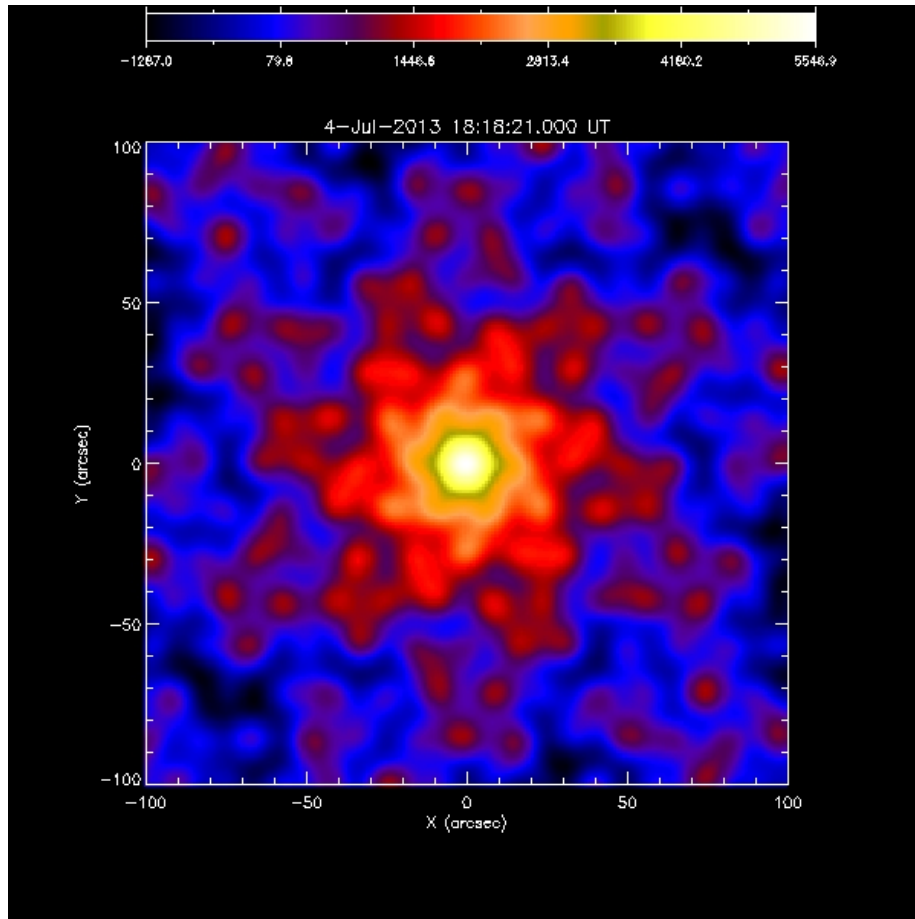
Clean algorithm from visibilities: vis_clean

HESPE developed a new version of the CLEAN algorithm where inputs are visibilities instead of counts

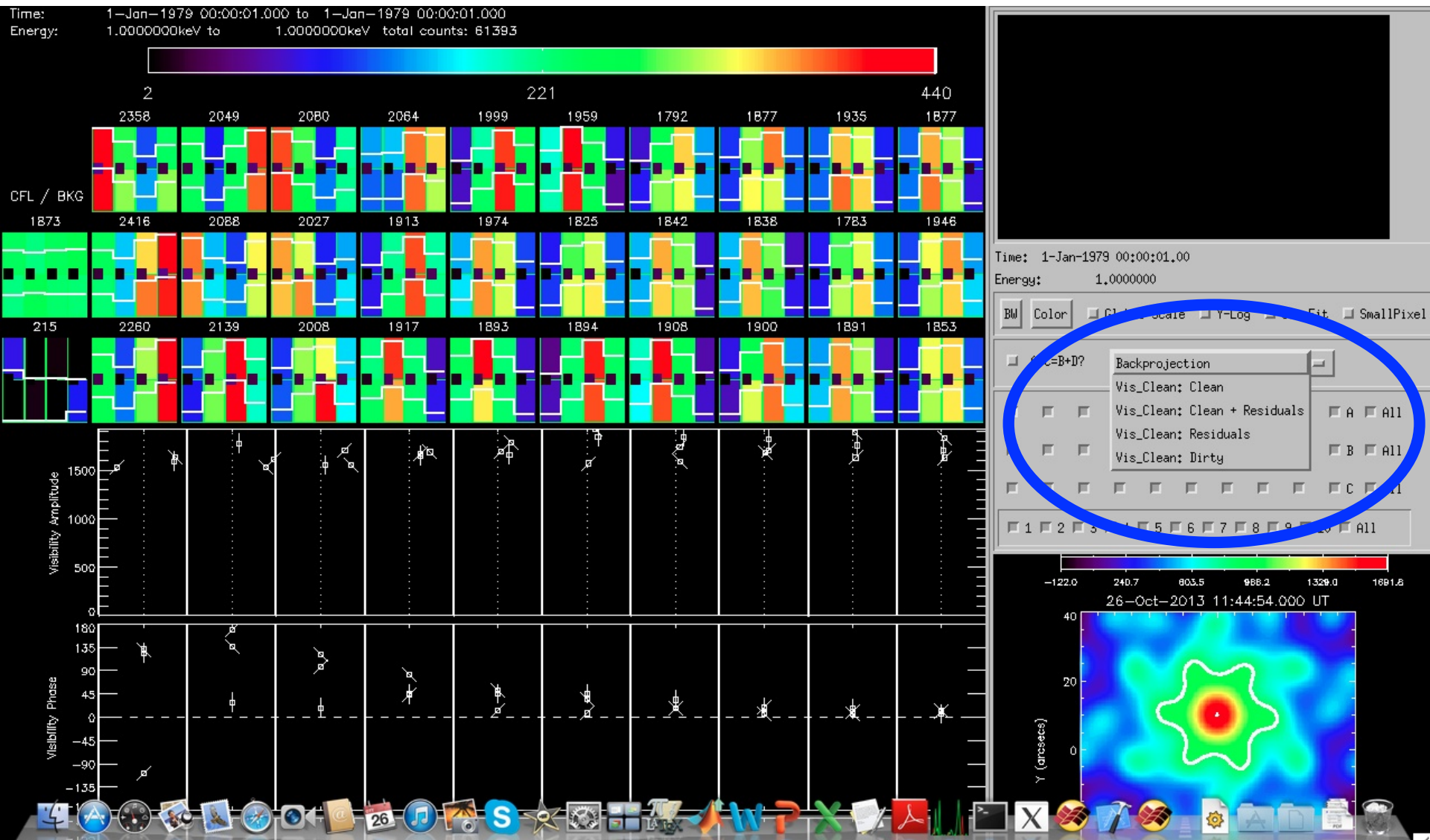


February 20, 2002 11:06:10 - 11:06:40 UT

Clean algorithm from visibilities: instrument independent

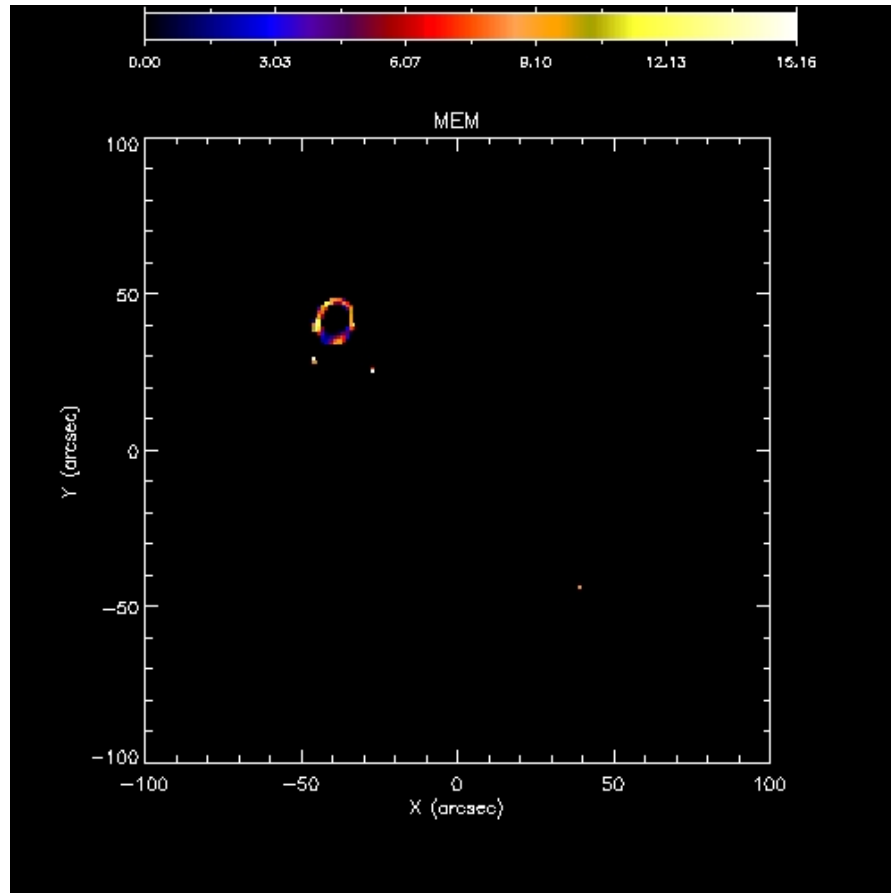


Clean & backprojection algorithms in STIX Software



MEM_NJIT

The maximum entropy routine is not always successful:
sometimes MEM–NJIT iterates too much



MEM_NJIT

MEM objective function $J = H - \alpha\chi^2 - \beta F$

where $H = -\sum_j T_j \ln\left(\frac{T_j}{m_j e}\right)$

$$\chi^2 = \sum_i \frac{|V_i - V'_i|^2}{\sigma_i^2} - n_v$$

$$F = \sum_j T_j - F'$$

T = actual reconstruction

V = visibilities predicted by T

V' = observed visibilities

F' = observed flux

m = guess map

n_v = number of visibilities

$$\varepsilon \approx 0.03 \quad \xi_j = \frac{T_j}{m_j} - \frac{m_j}{T_j}$$

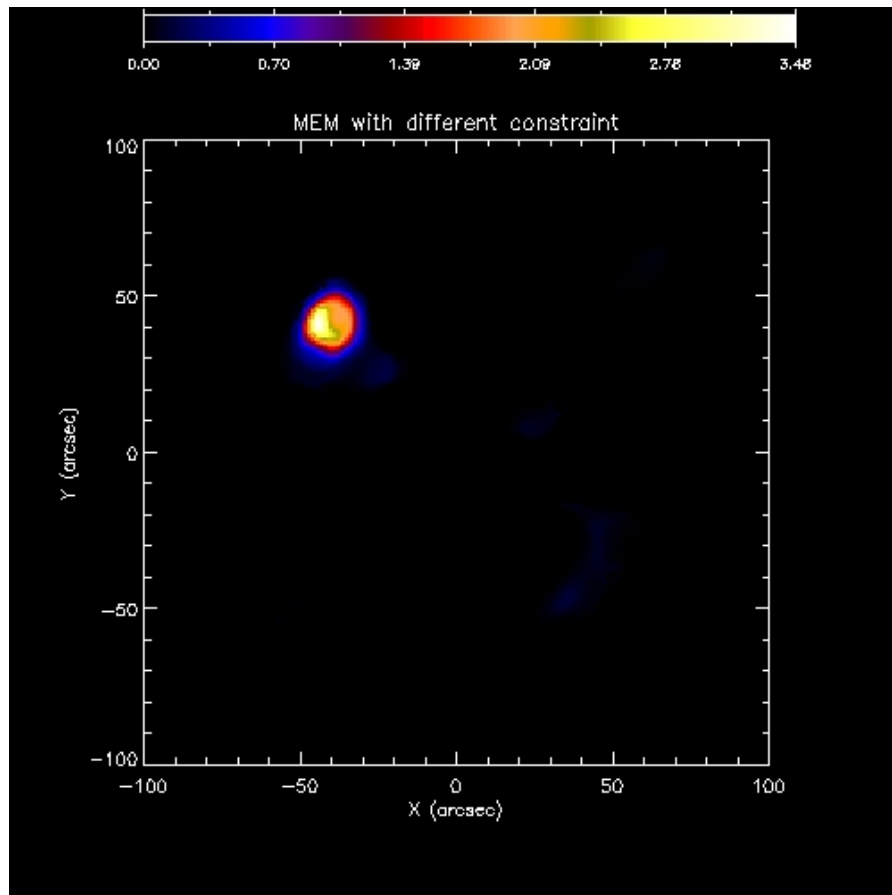
under the constraints:

$$|\chi^2| \leq \max(\varepsilon n_v, \sqrt{2n_v}, |\nabla_{\xi} \chi^2| \sqrt{\sum_j (\delta \xi_j)^2})$$

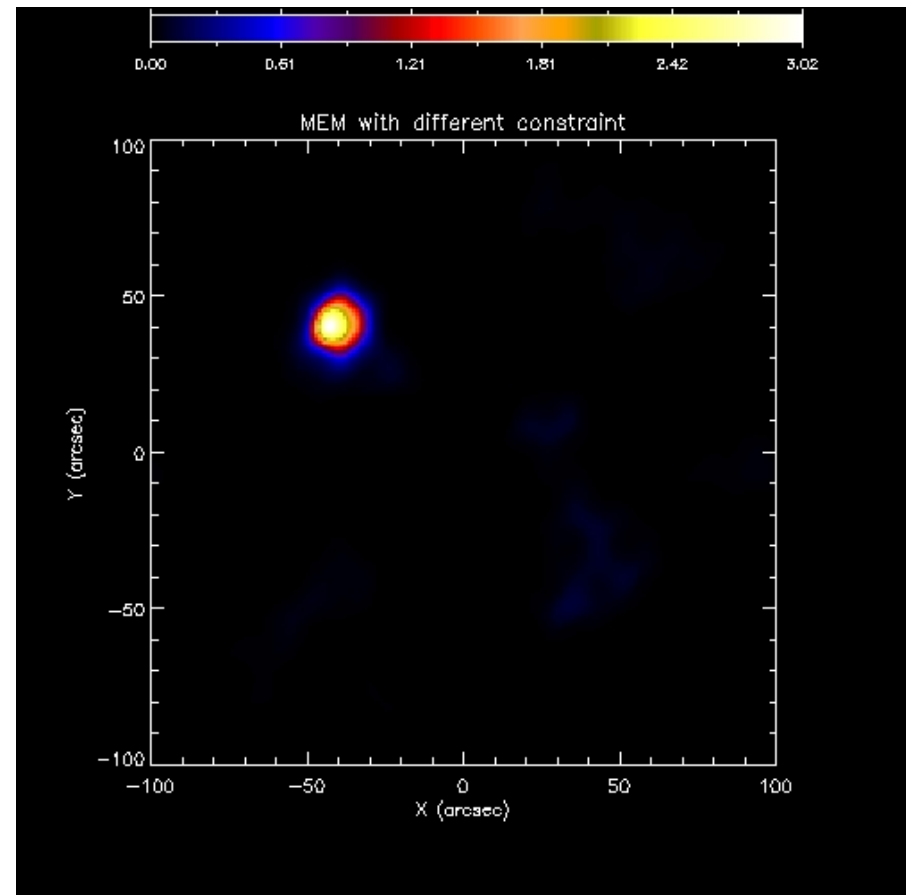
$$|F| \leq \max(\varepsilon n_v, \sqrt{2n_v}, |\nabla_{\xi} F| \sqrt{\sum_j (\delta \xi_j)^2})$$

MEM_NJIT

Just one constraint verified:



The two constraints relaxed:



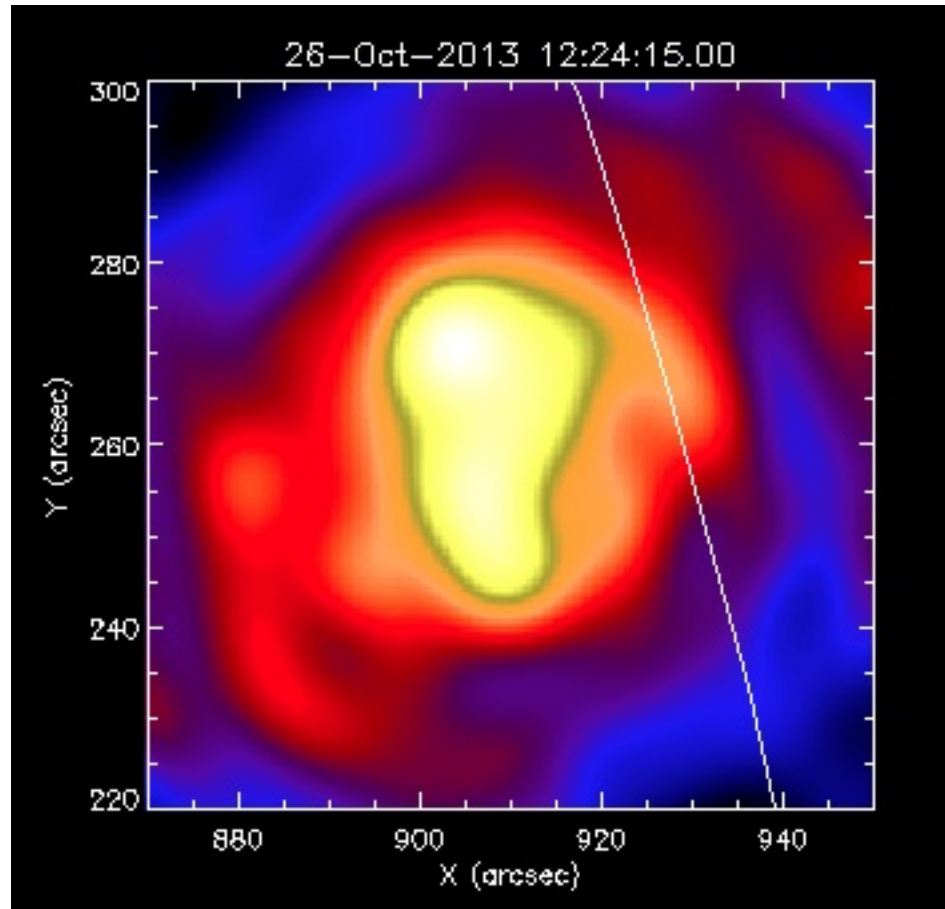
Vis Forward Fit: instrument independent

RHESSI data:

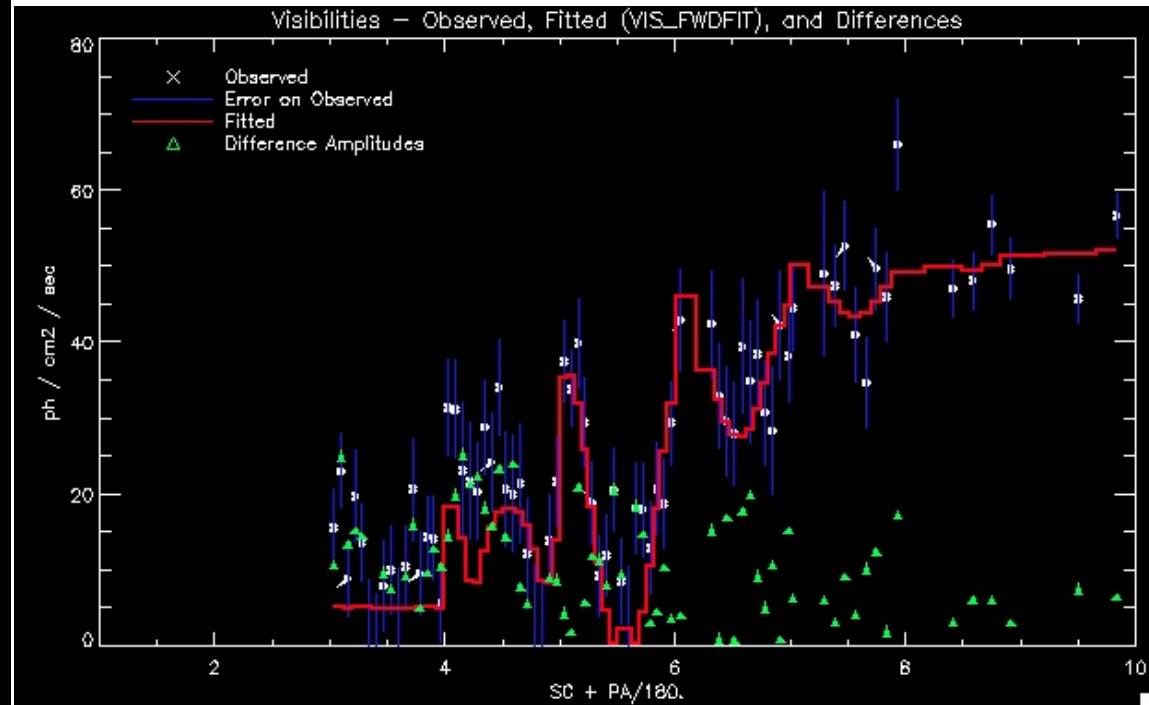
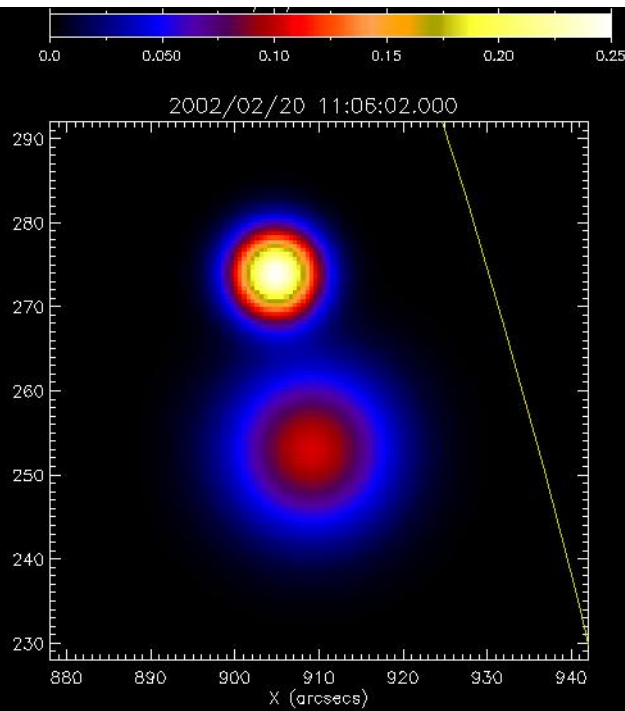
20 February 2002

(11:06:02 - 11:06:34 UT)

18-22 keV

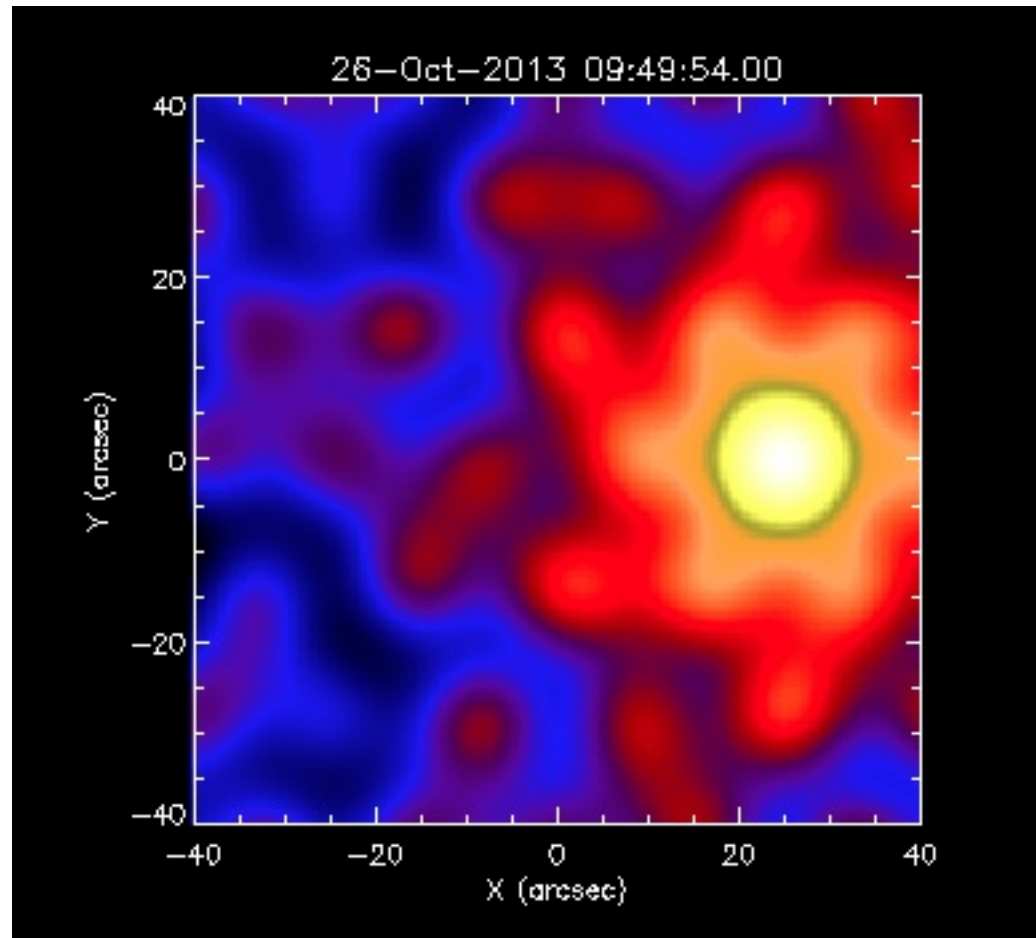


Vis Forward Fit: instrument independent

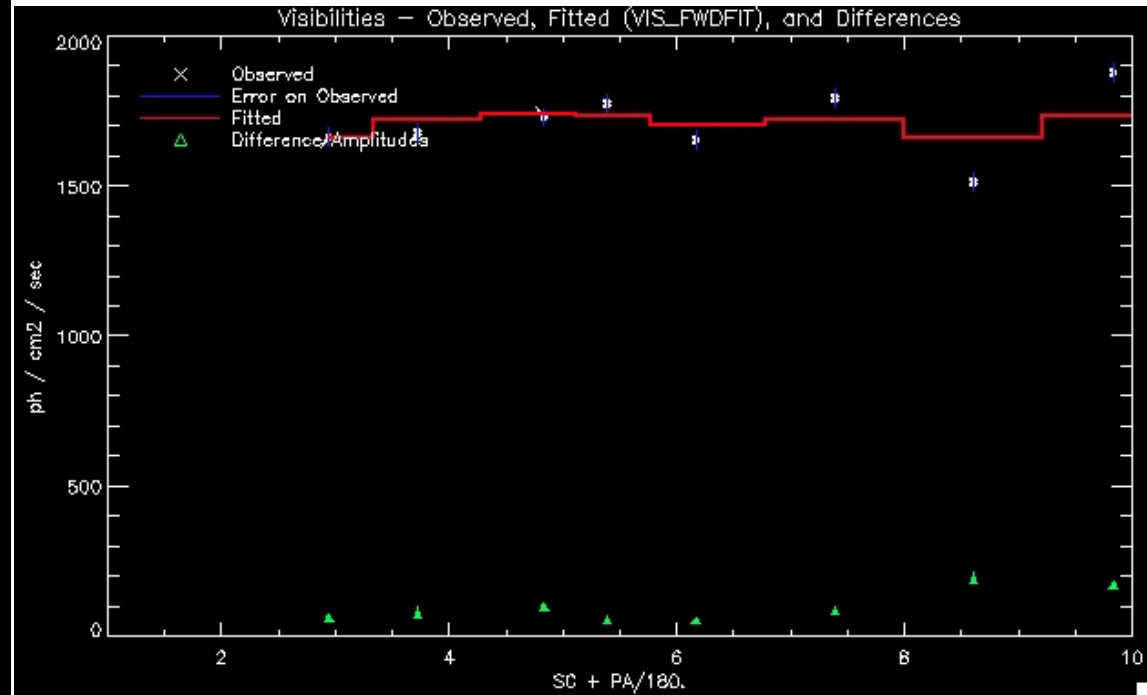
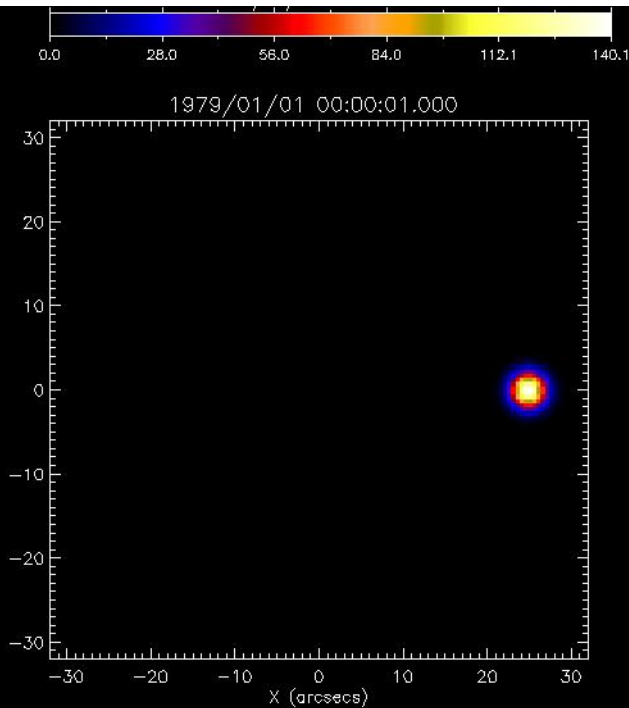


Vis Forward Fit: instrument independent

STIX Simulation



Vis Forward Fit: instrument independent



Imaging test

RHESSI IX --- Genova (*first* test design)

RHESSI X --- Annapolis (Sam analysis)

RHESSI XI --- Glasgow (Anna analysis)

RHESSI XII --- Nanjing (results collected)

Algorithms:

MEM_NJIT & uv_smooth (A.M. Massone)

VIS FORWARD_FIT (M. Battaglia)

Pixon (R. Schwartz)

Clean (A. Gopie)

Making images

For each algorithm and for each count rate:

Default image with subcollimators 1-9, 2-9, 3-9

Best fine-tuned image with subcollimators 1-9, 2-9, 3-9

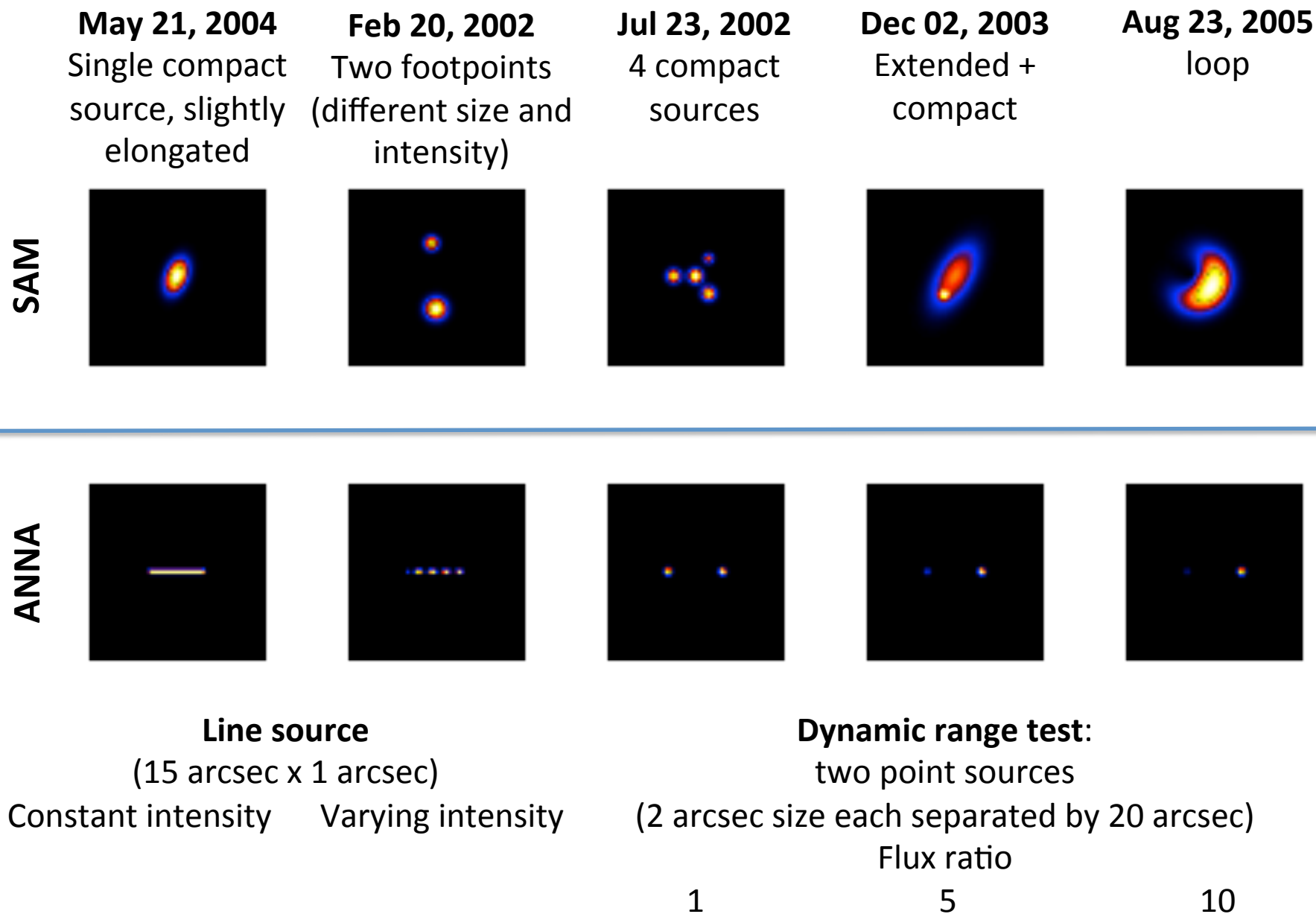
Analysis on about 900 reconstructed maps:

Images 1-> 5 (S. Krucker)

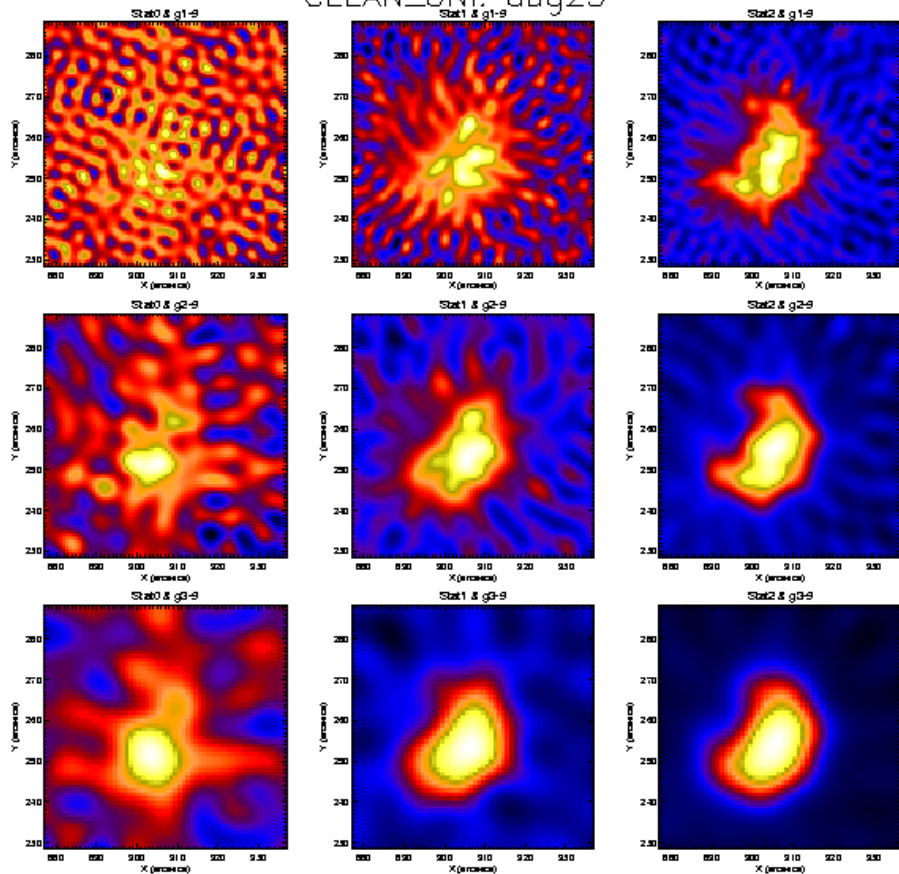
Images 6->10 (A.M. Massone)

- application of routines for the quantitative assessment of the algorithms' performances

Test images

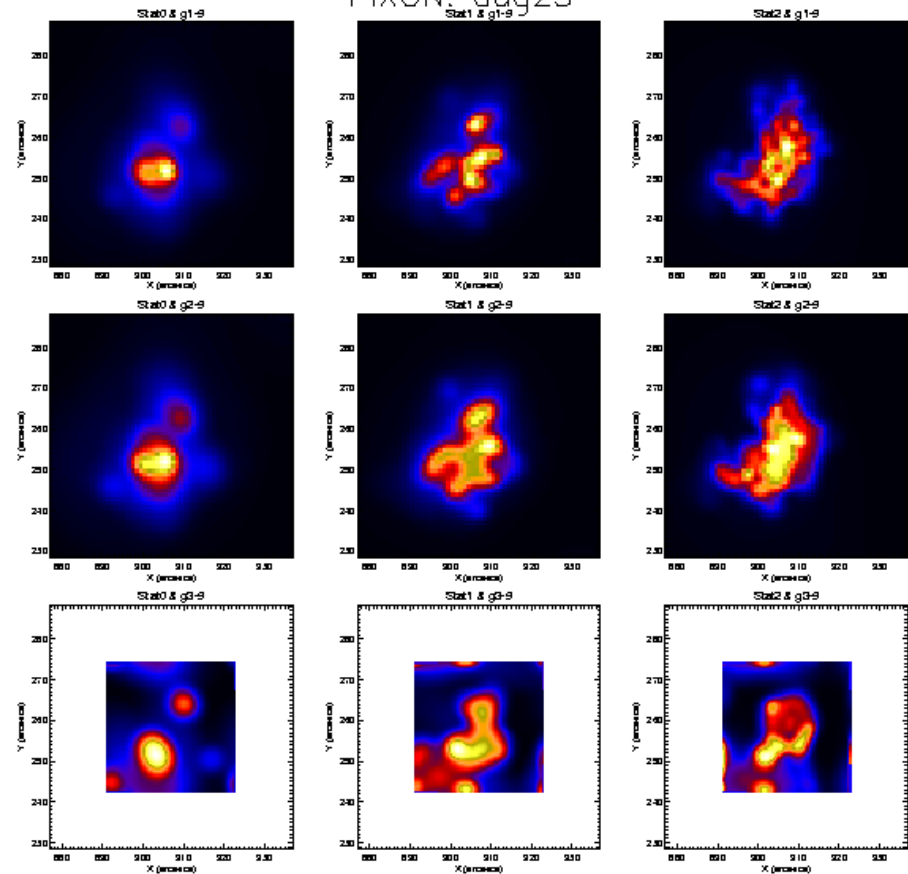


CLEAN_UNI: aug23

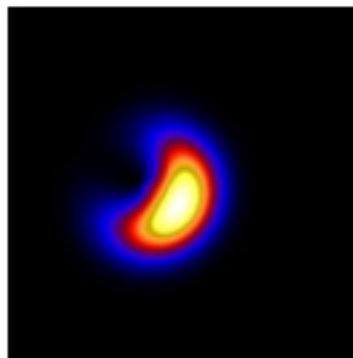


CLEAN

PIXON: aug23



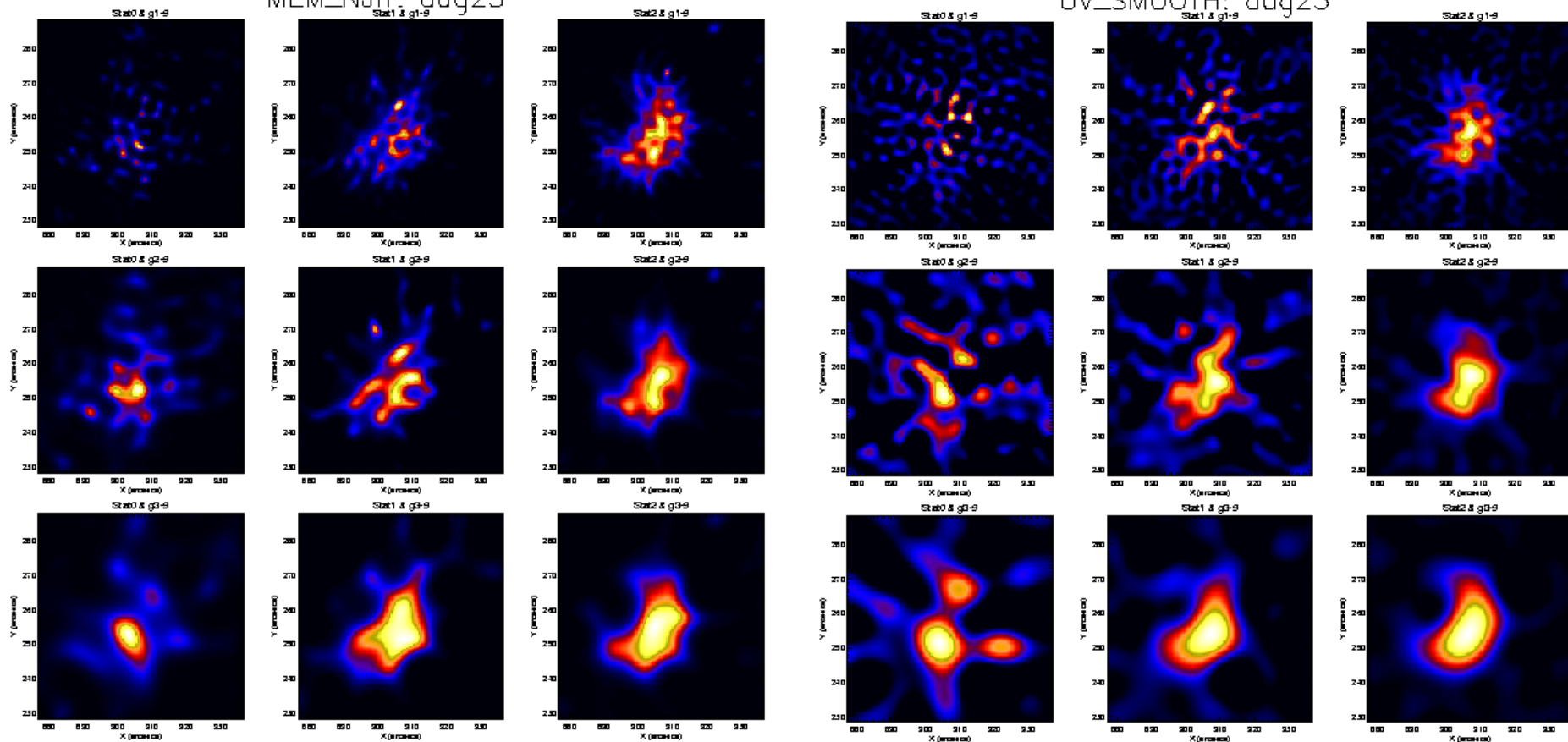
PIXON



Sam analysis: Aug 23

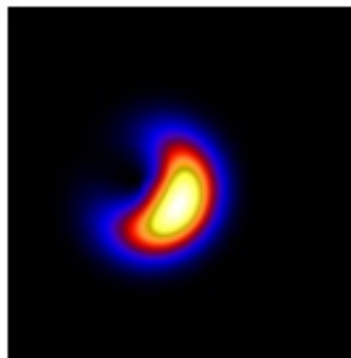
MEM_NJIT: aug23

UV_SMOOTH: aug23



MEM_NJIT

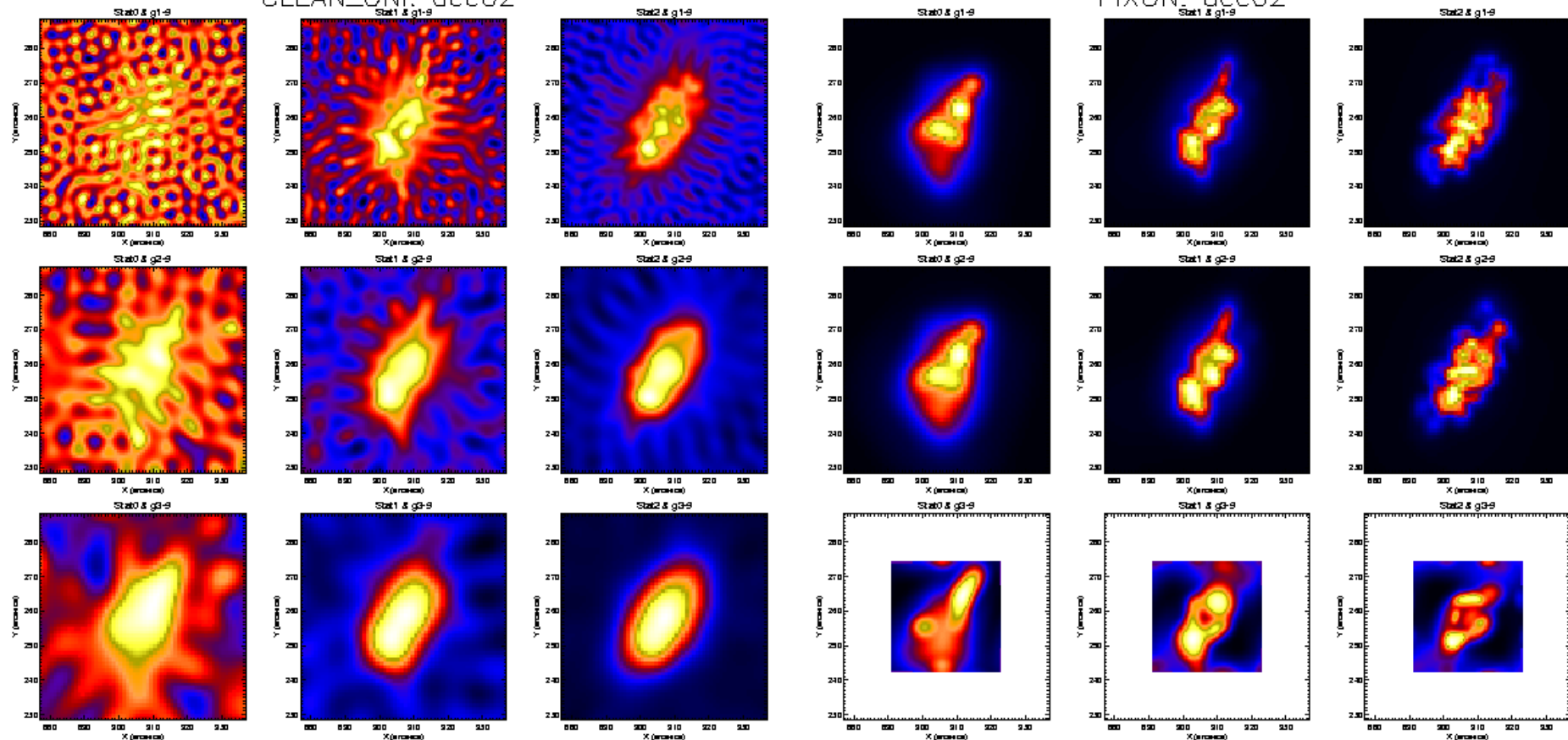
UV_SMOOTH



Sam analysis: Aug 23

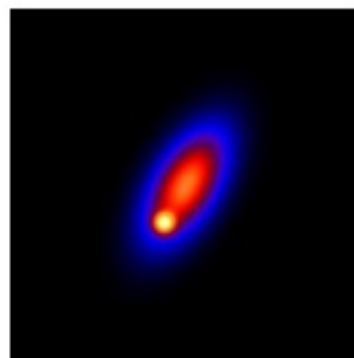
CLEAN_UNI: dec02

PIXON: dec02



CLEAN

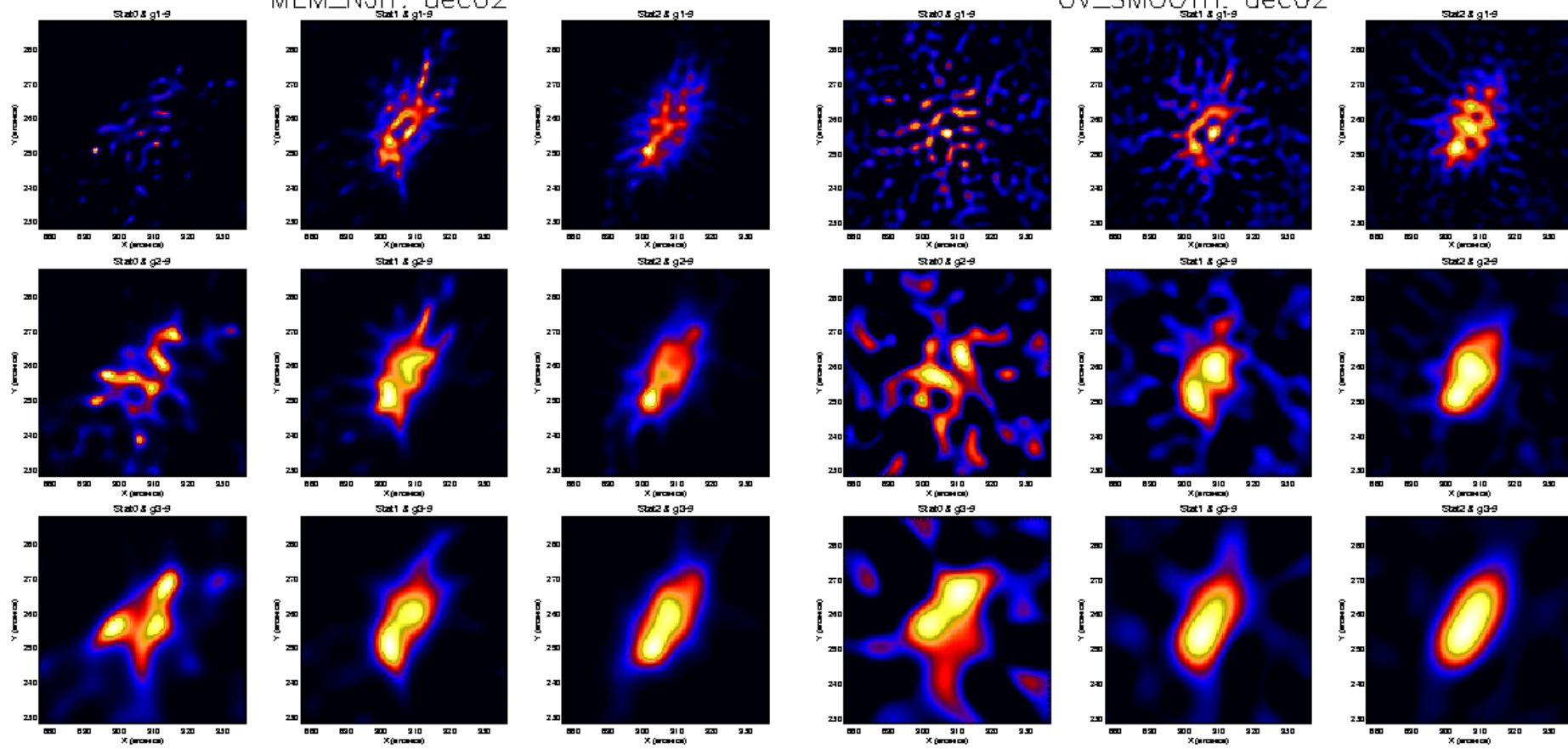
PIXON



Sam analysis: Dec 02

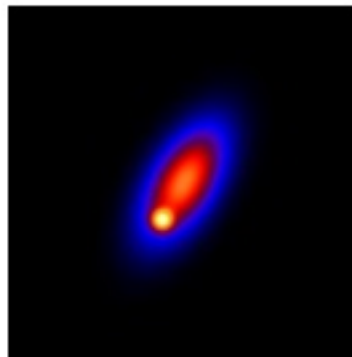
MEM_NJIT: dec02

UV_SMOOTH: dec02



MEM_NJIT

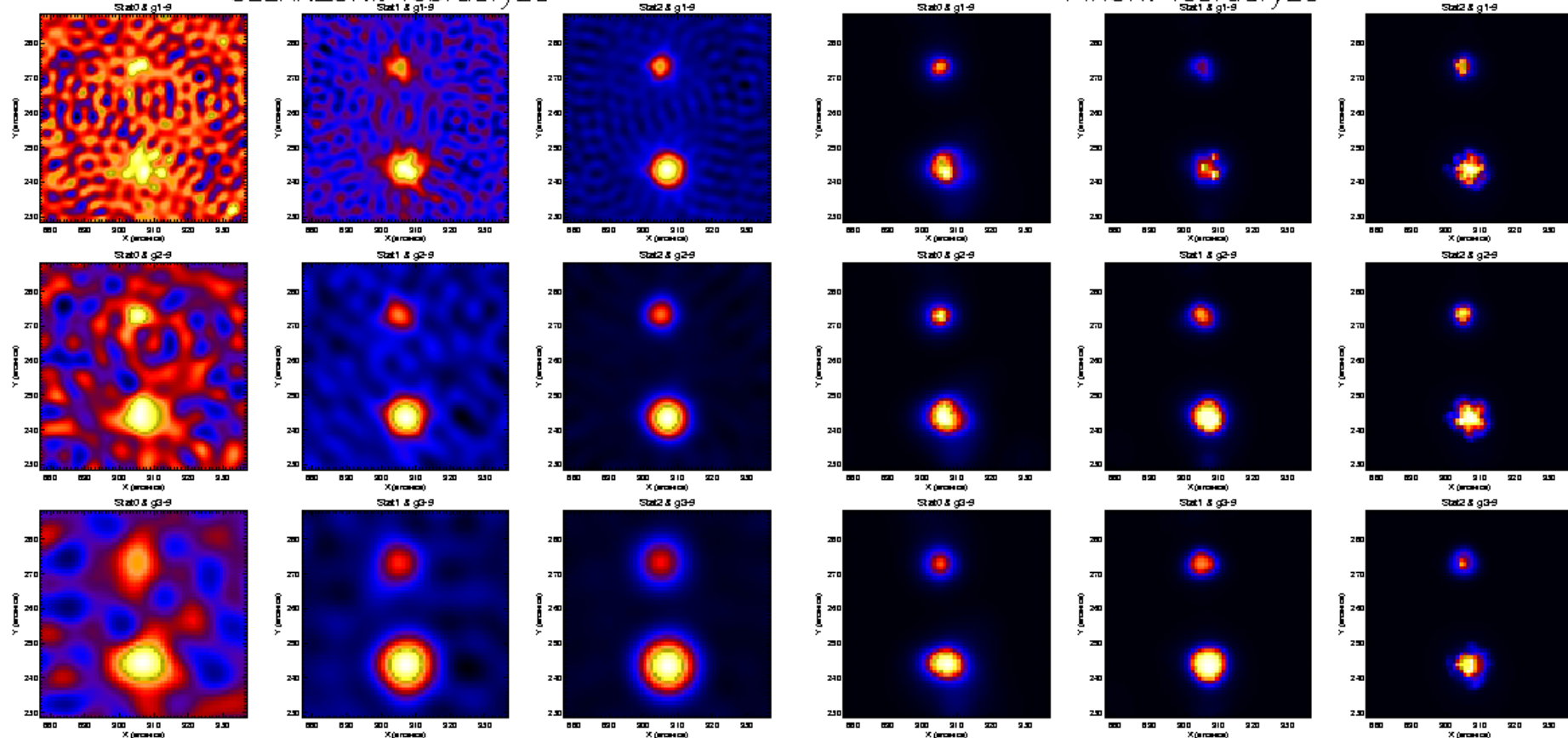
UV_SMOOTH



Sam analysis: Dec 02

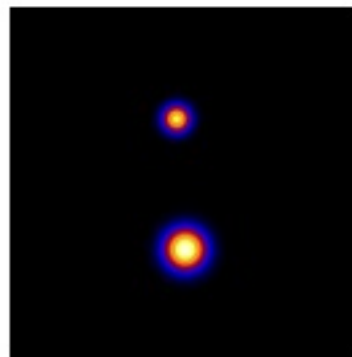
CLEAN_LUN1: february20

PIXON: february20



CLEAN

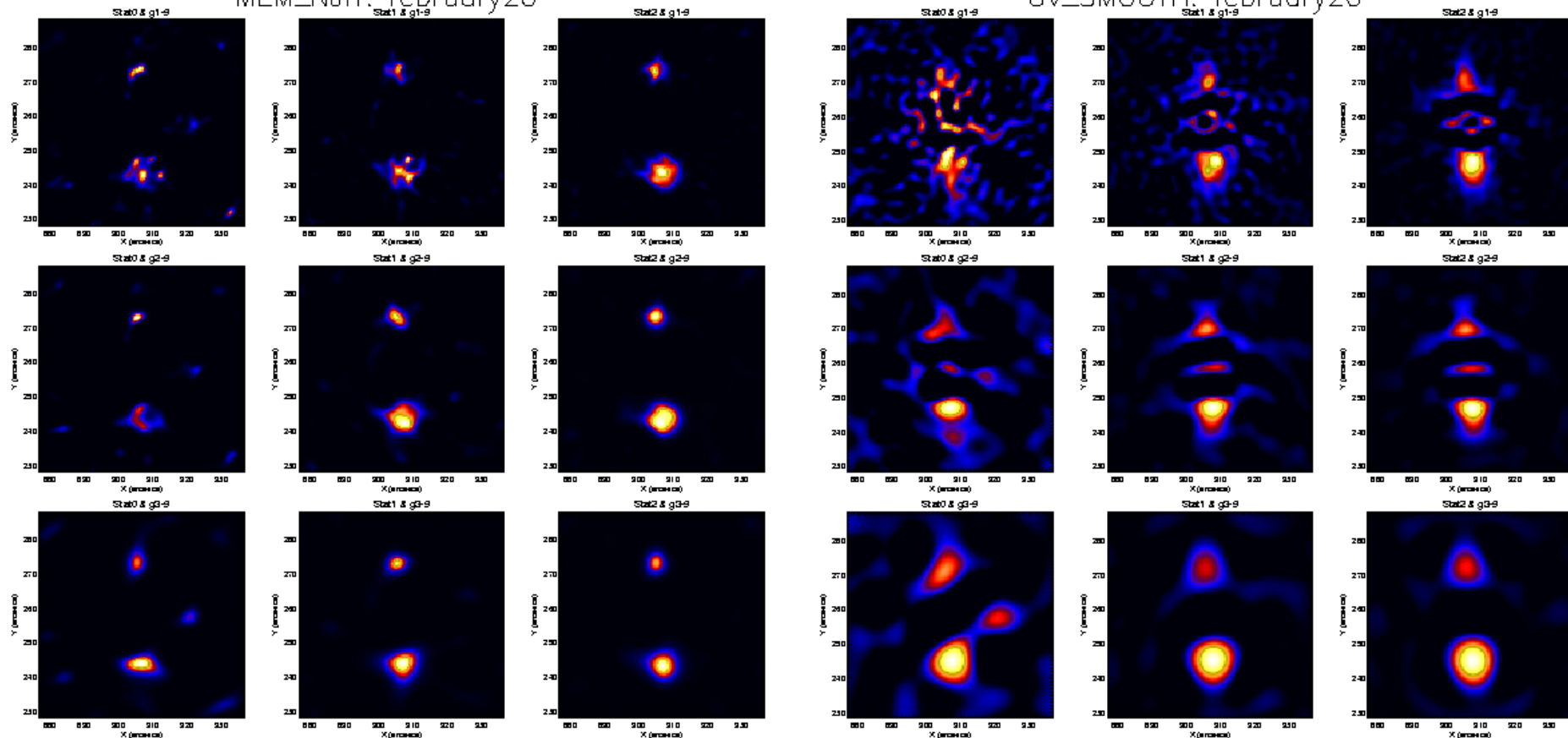
PIXON



Sam analysis: Feb 20

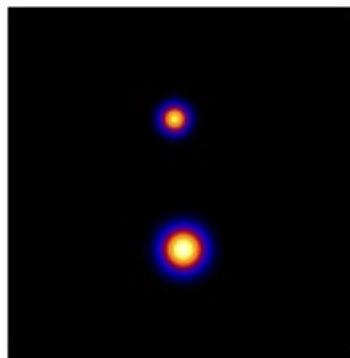
MEM_NJIT: february20

UV_SMOOTH: february20



MEM_NJIT

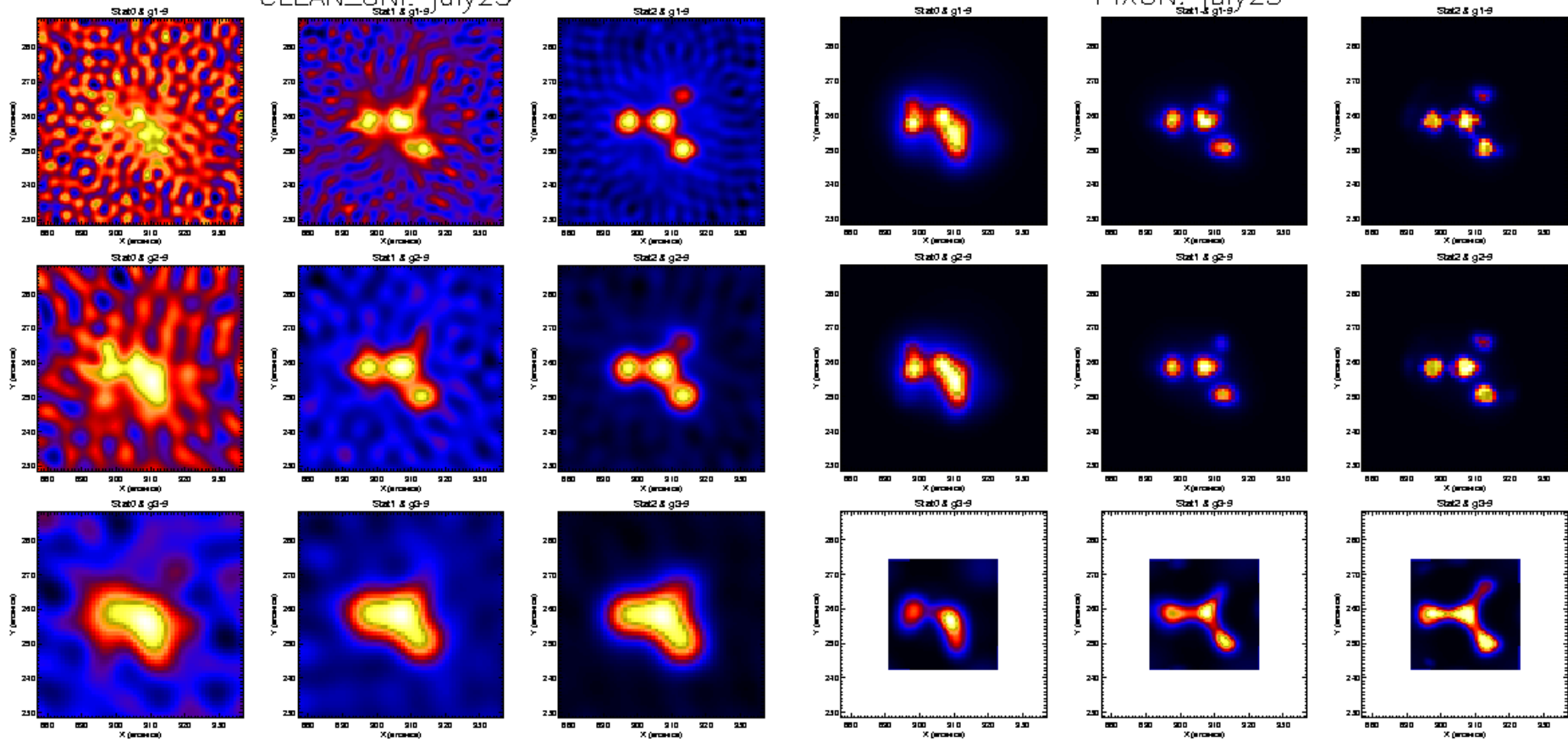
UV_SMOOTH



Sam analysis: Feb 20

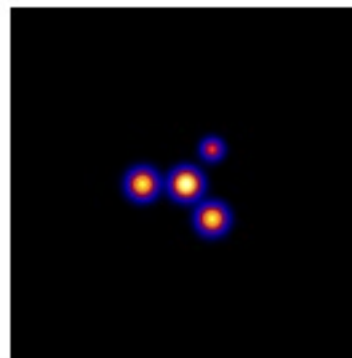
CLEAN_LUN1: july23

PIXON: july23



CLEAN

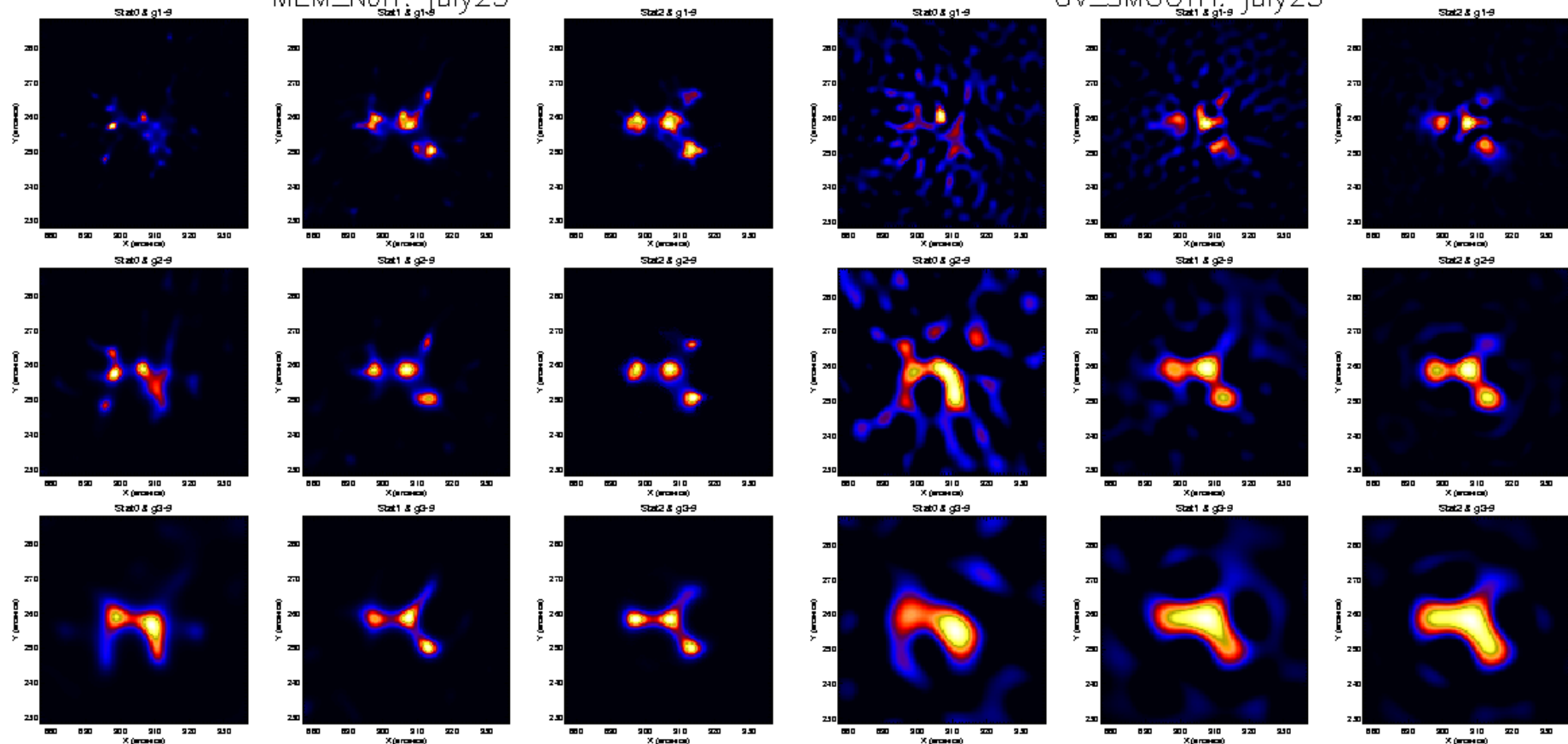
PIXON



Sam analysis: Jul 23

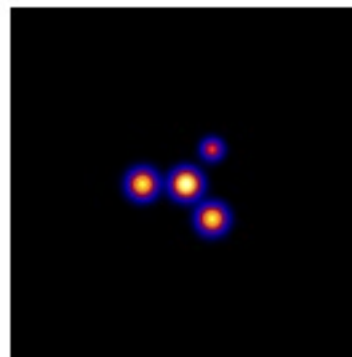
MEM_NJIT: july23

UV_SMOOTH: july23



MEM_NJIT

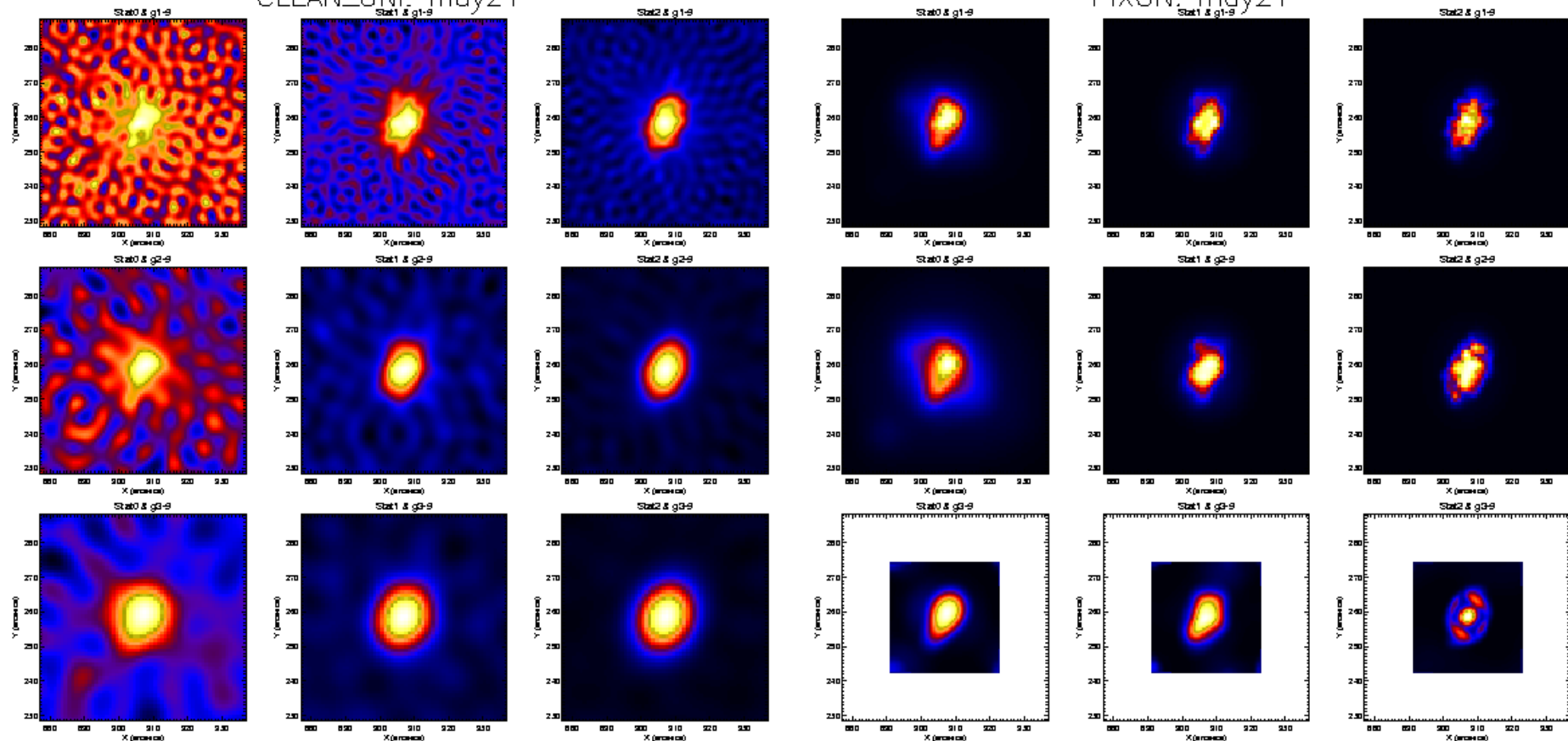
UV_SMOOTH



Sam analysis: Jul 23

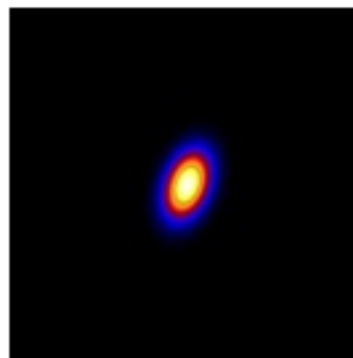
CLEAN_UNI: may21

PIXON: may21



CLEAN

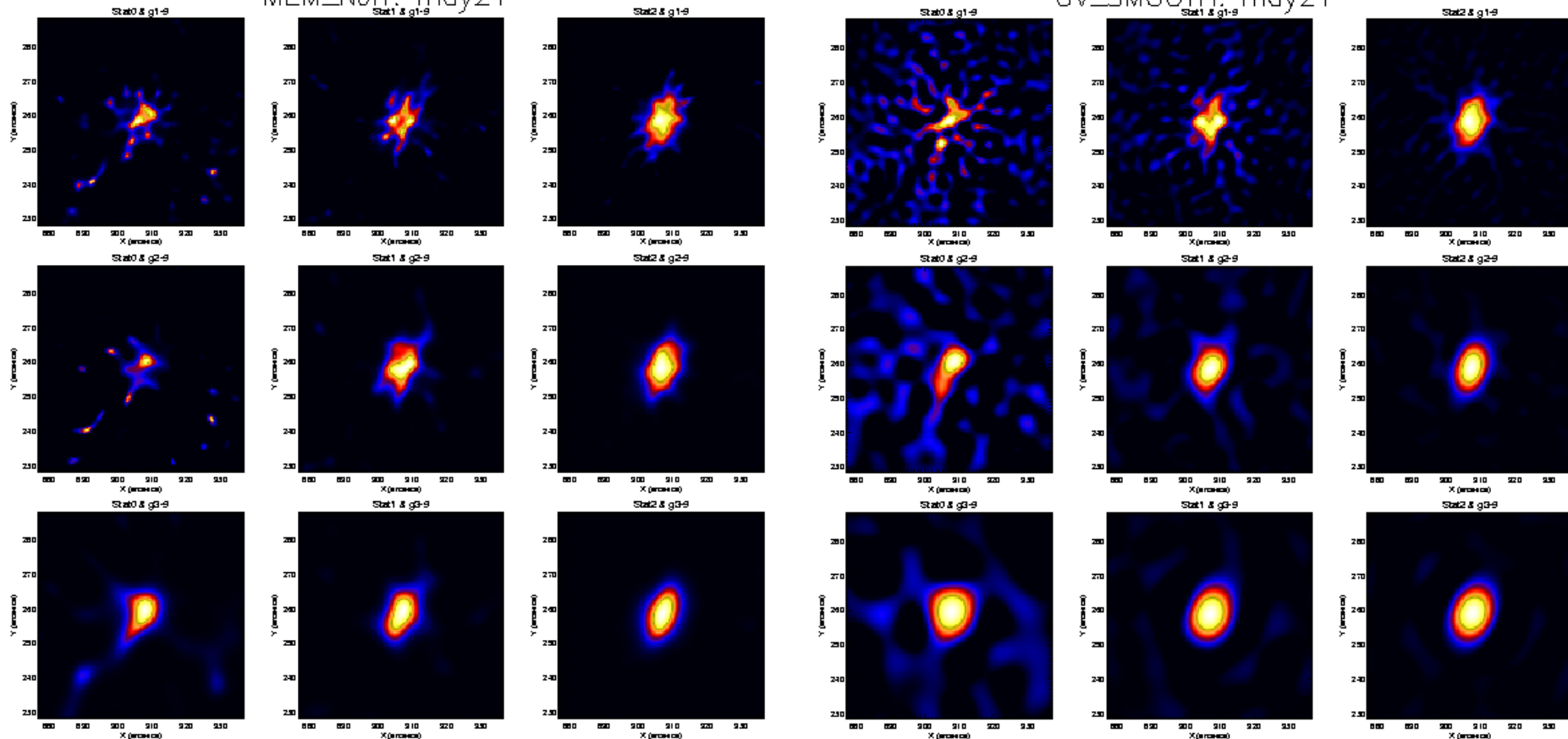
PIXON



Sam analysis: May 21

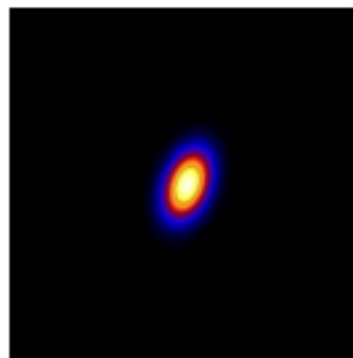
MEM_NJIT: may21

UV_SMOOTH: may21

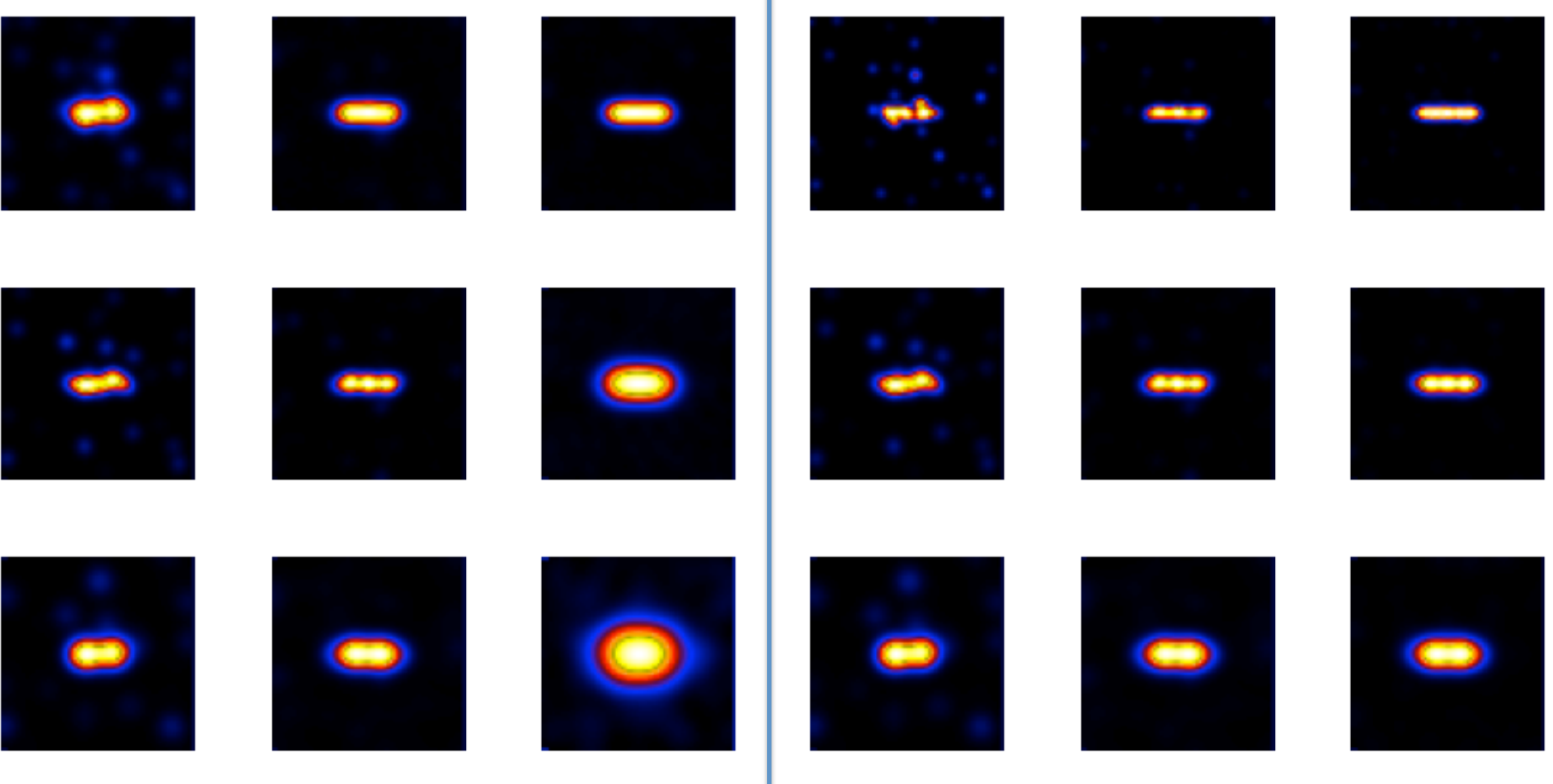


MEM_NJIT

UV_SMOOTH

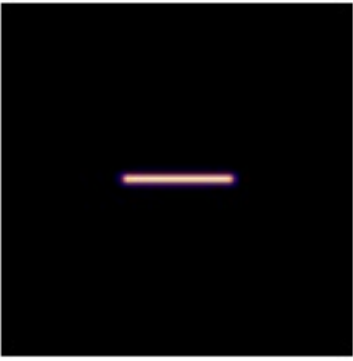


Sam analysis: May 21

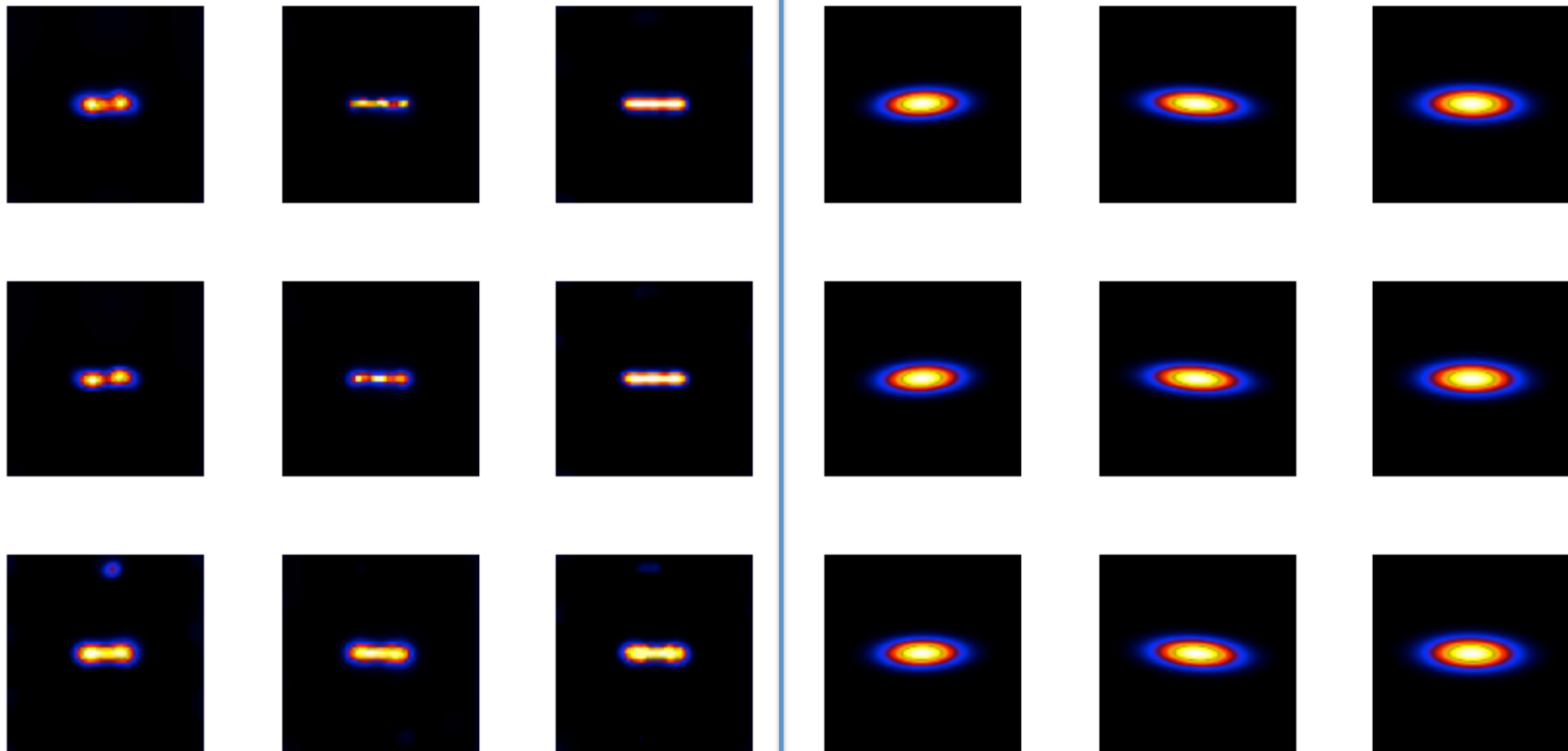


CLEAN DEFAULT

CLEAN ENHANCED



Anna analysis: Line source (constant intensity)

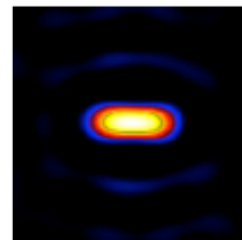
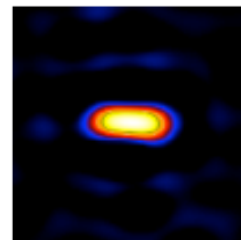
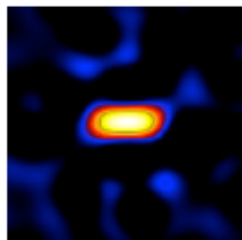
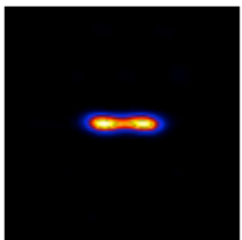
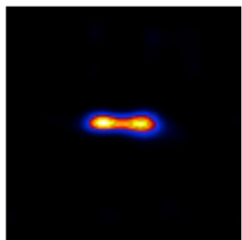
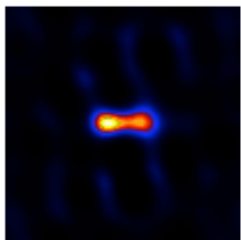
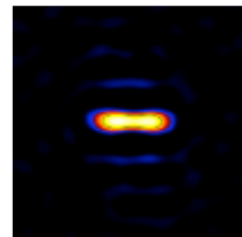
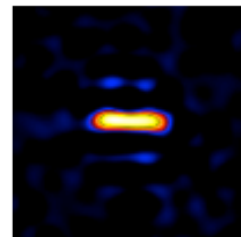
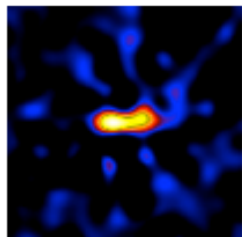
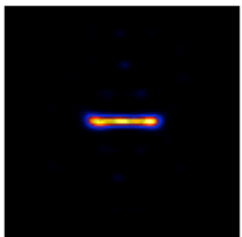
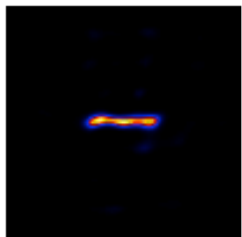
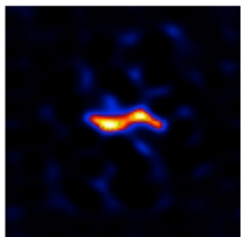
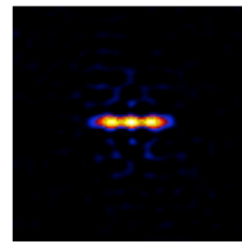
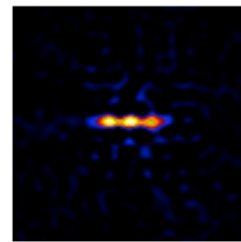
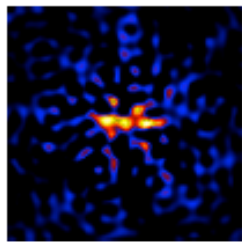
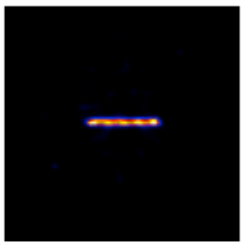
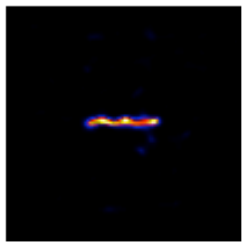
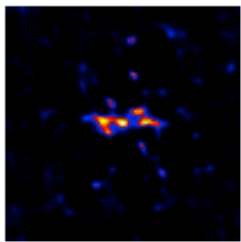


PIXON

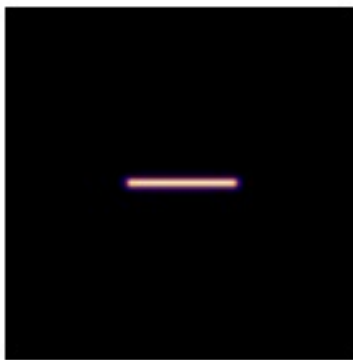
VIS FWD



Anna analysis: Line source (constant intensity)

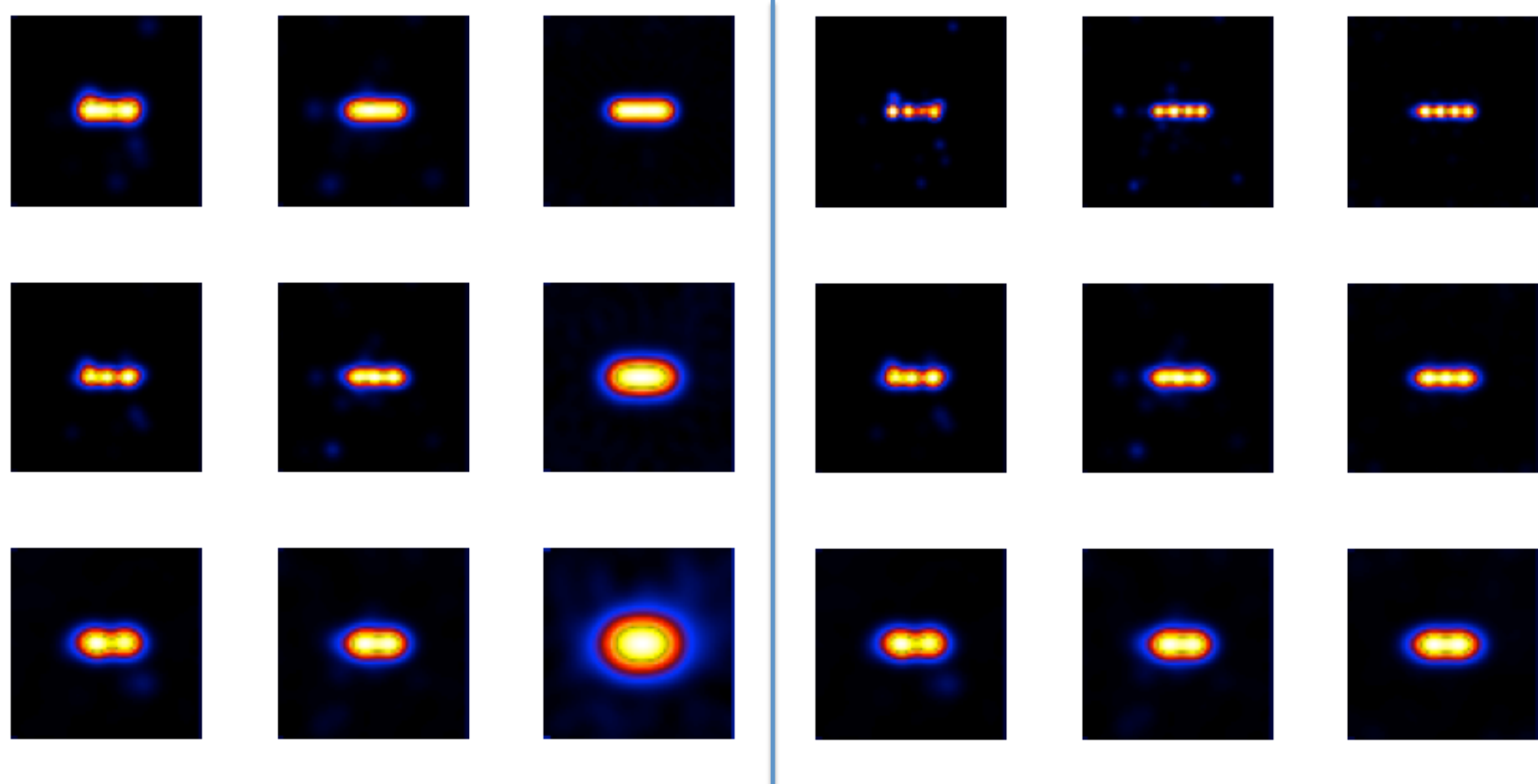


MEM_NJIT



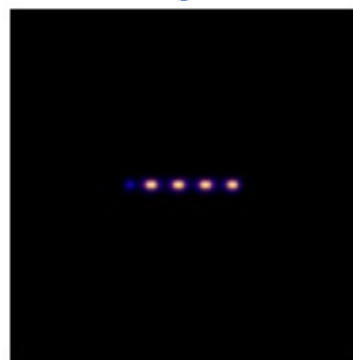
UV_SMOOTH

Anna analysis: Line source (constant intensity)

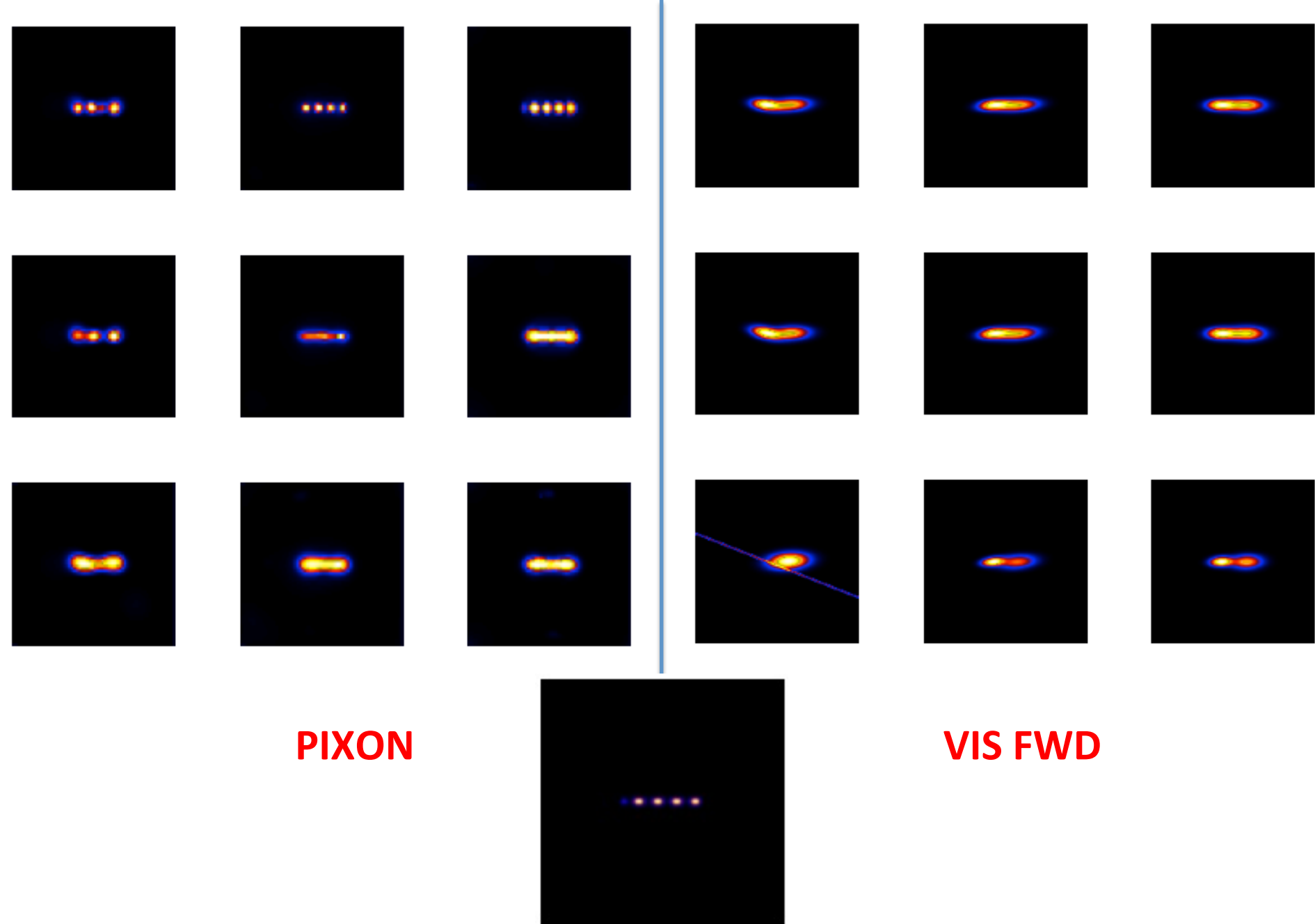


CLEAN DEFAULT

CLEAN ENHANCED



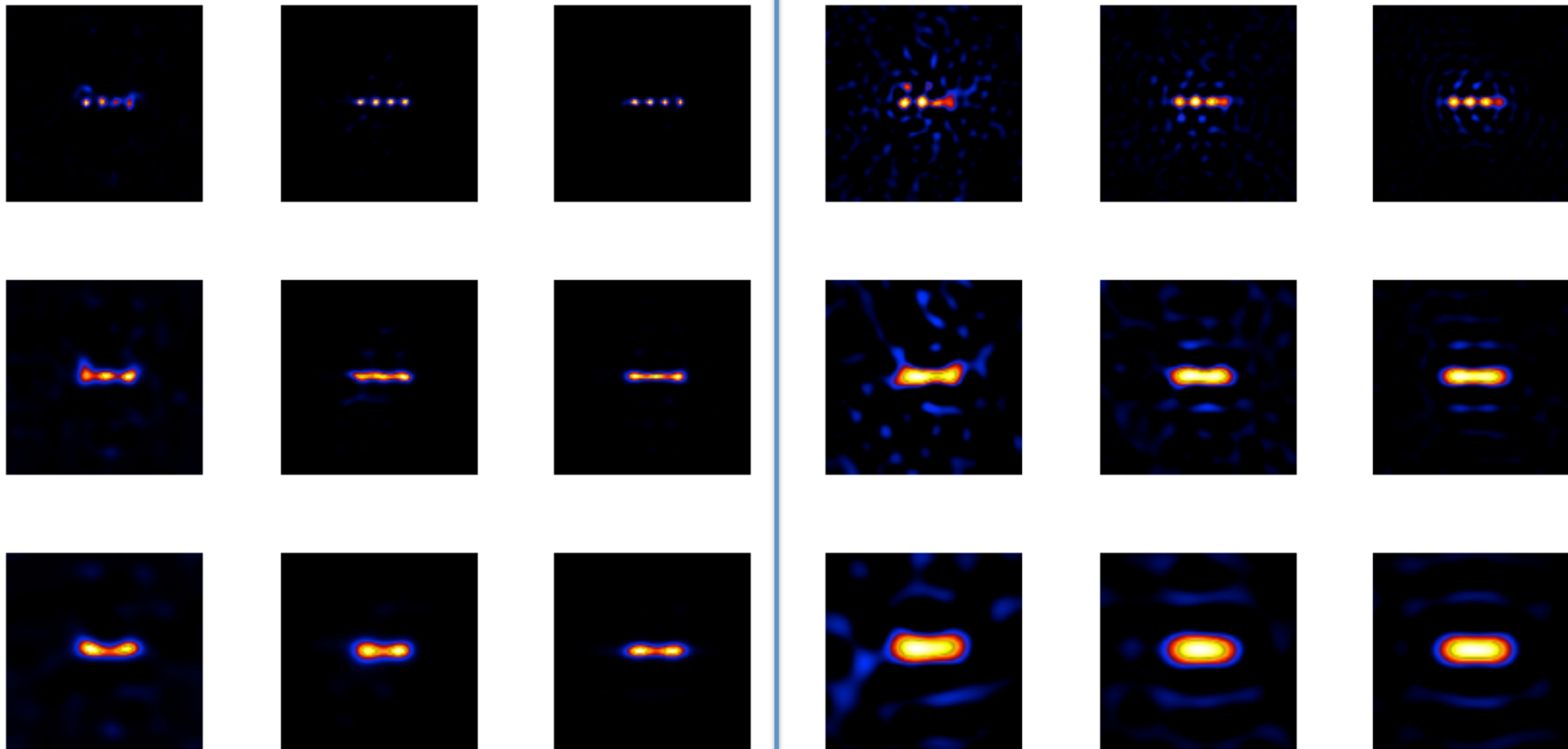
Anna analysis: Line source (varying intensity)



PIXON

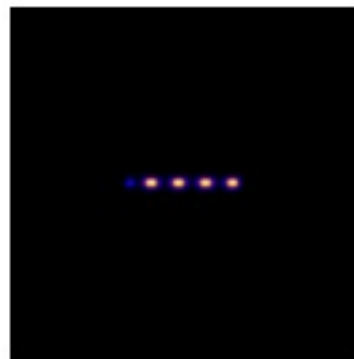
VIS FWD

Anna analysis: Line source (varying intensity)

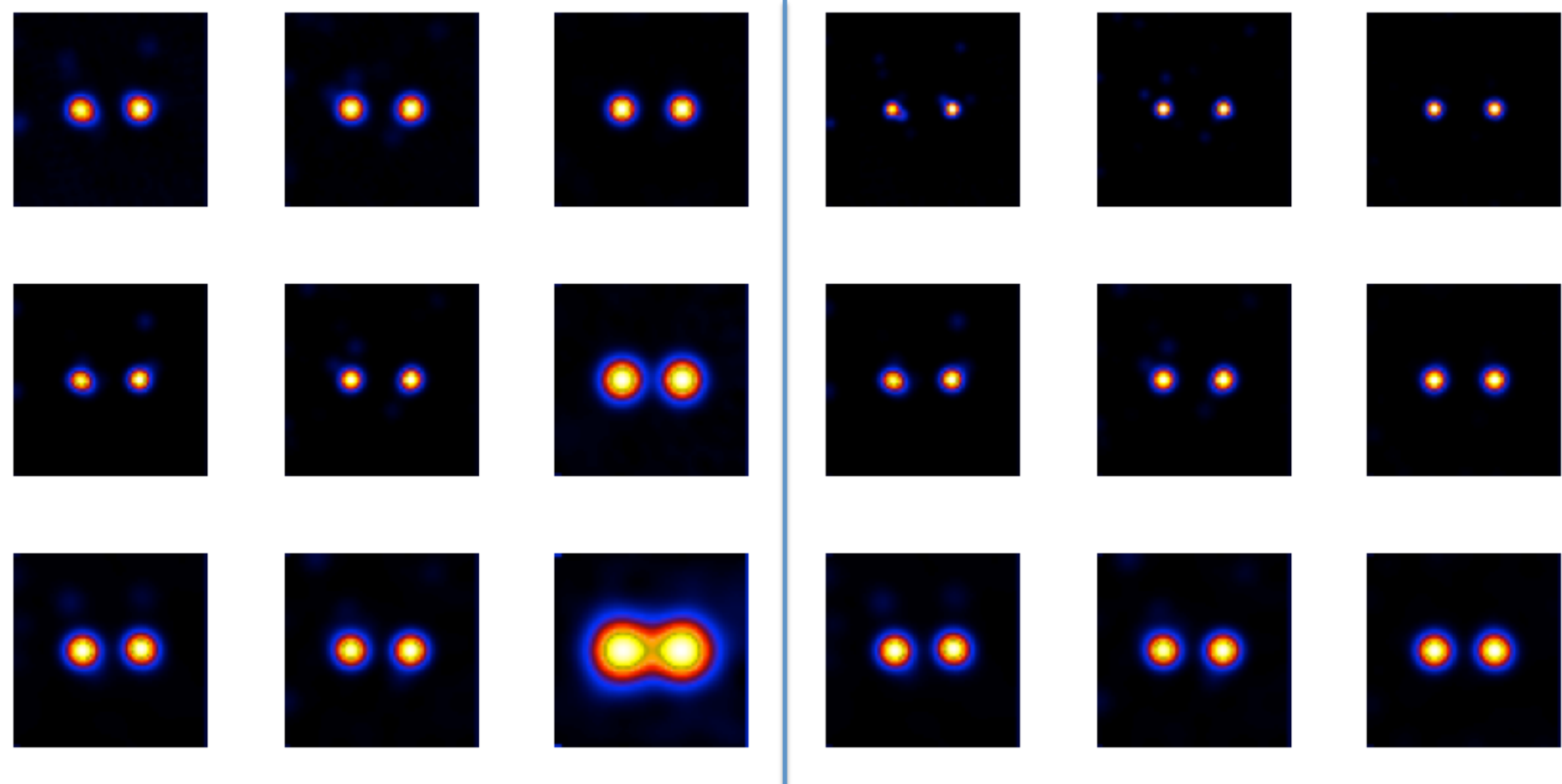


MEM_NJIT

UV_SMOOTH



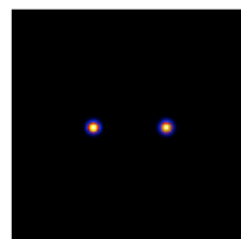
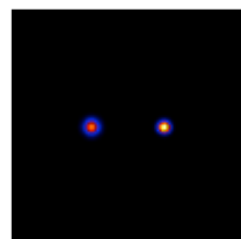
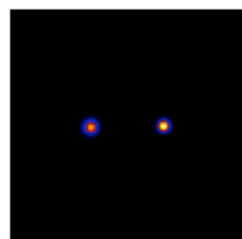
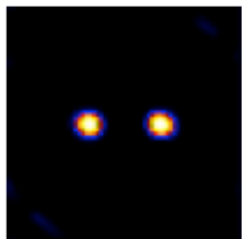
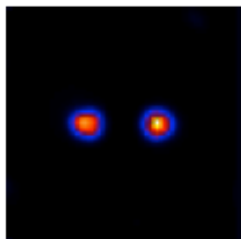
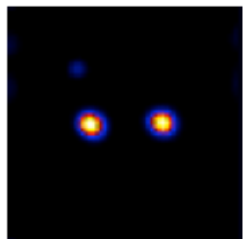
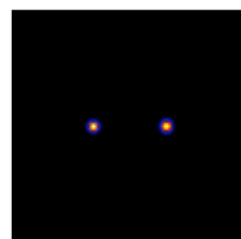
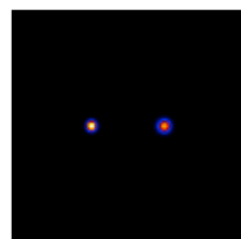
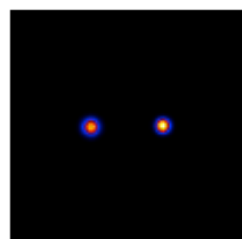
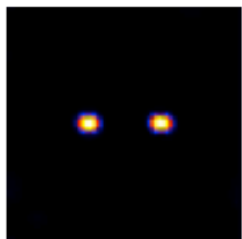
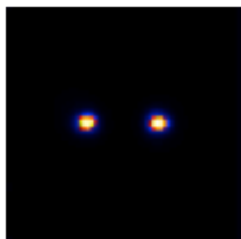
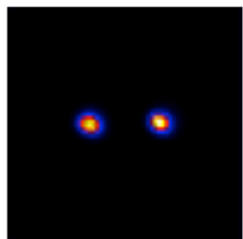
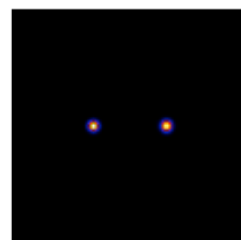
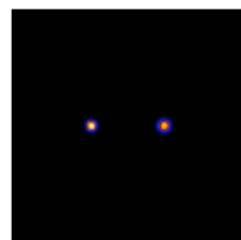
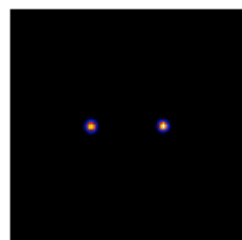
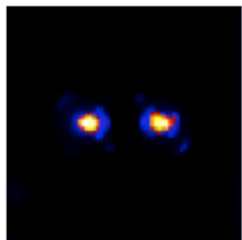
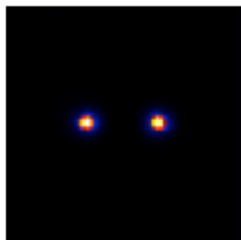
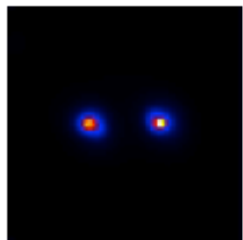
Anna analysis: Line source (varying intensity)



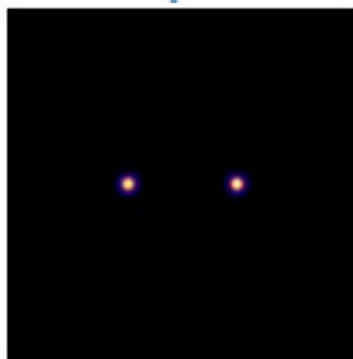
CLEAN DEFAULT

CLEAN ENHANCED

Anna analysis: Dynamic range (Flux ratio=1)

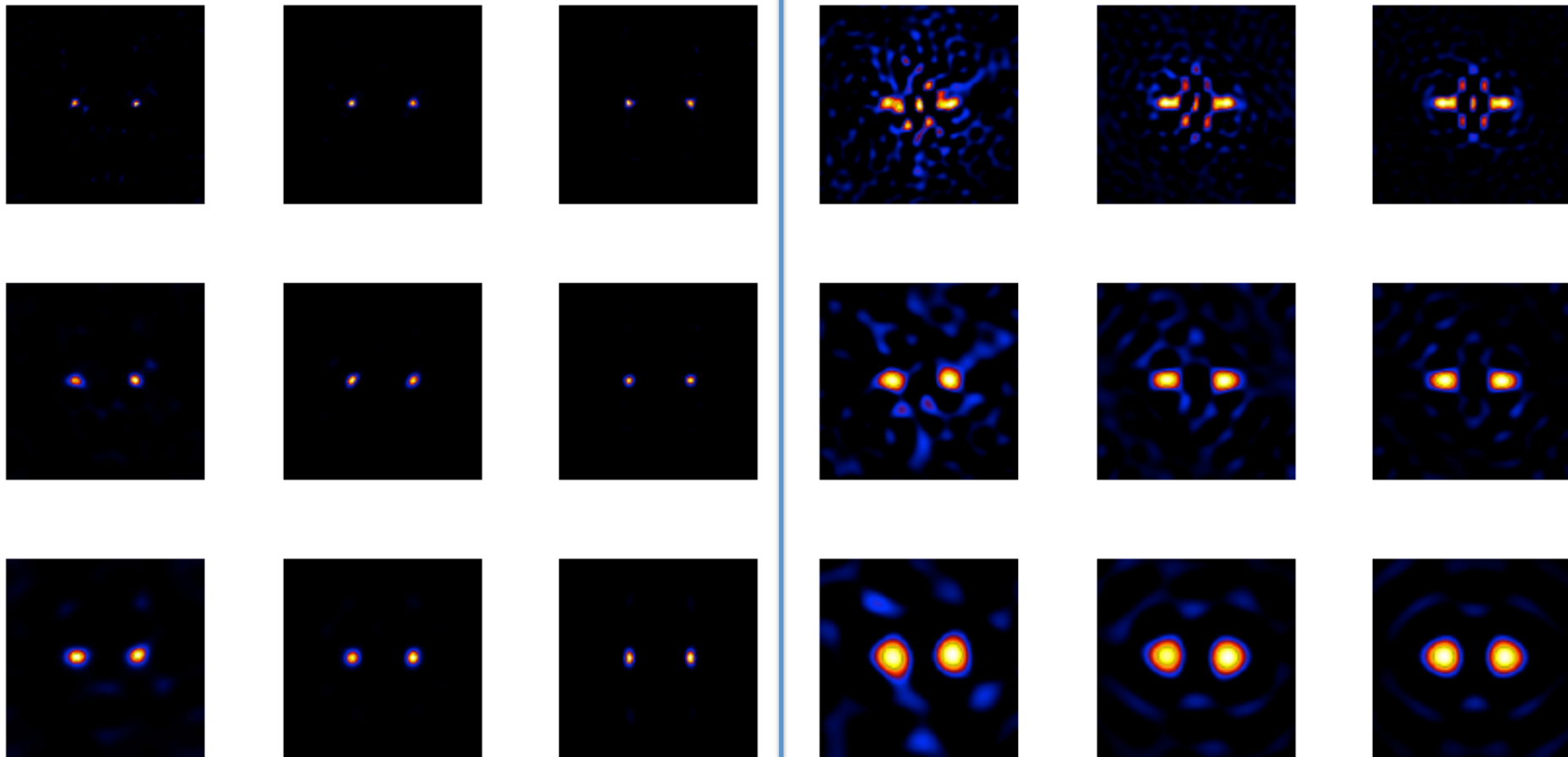


PIXON



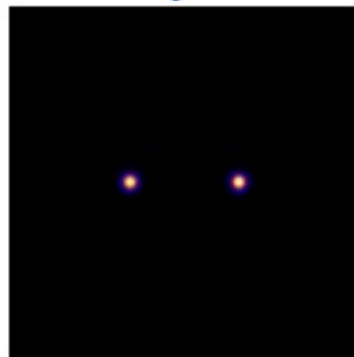
VIS FWD

Anna analysis: Dynamic range (Flux ratio = 1)

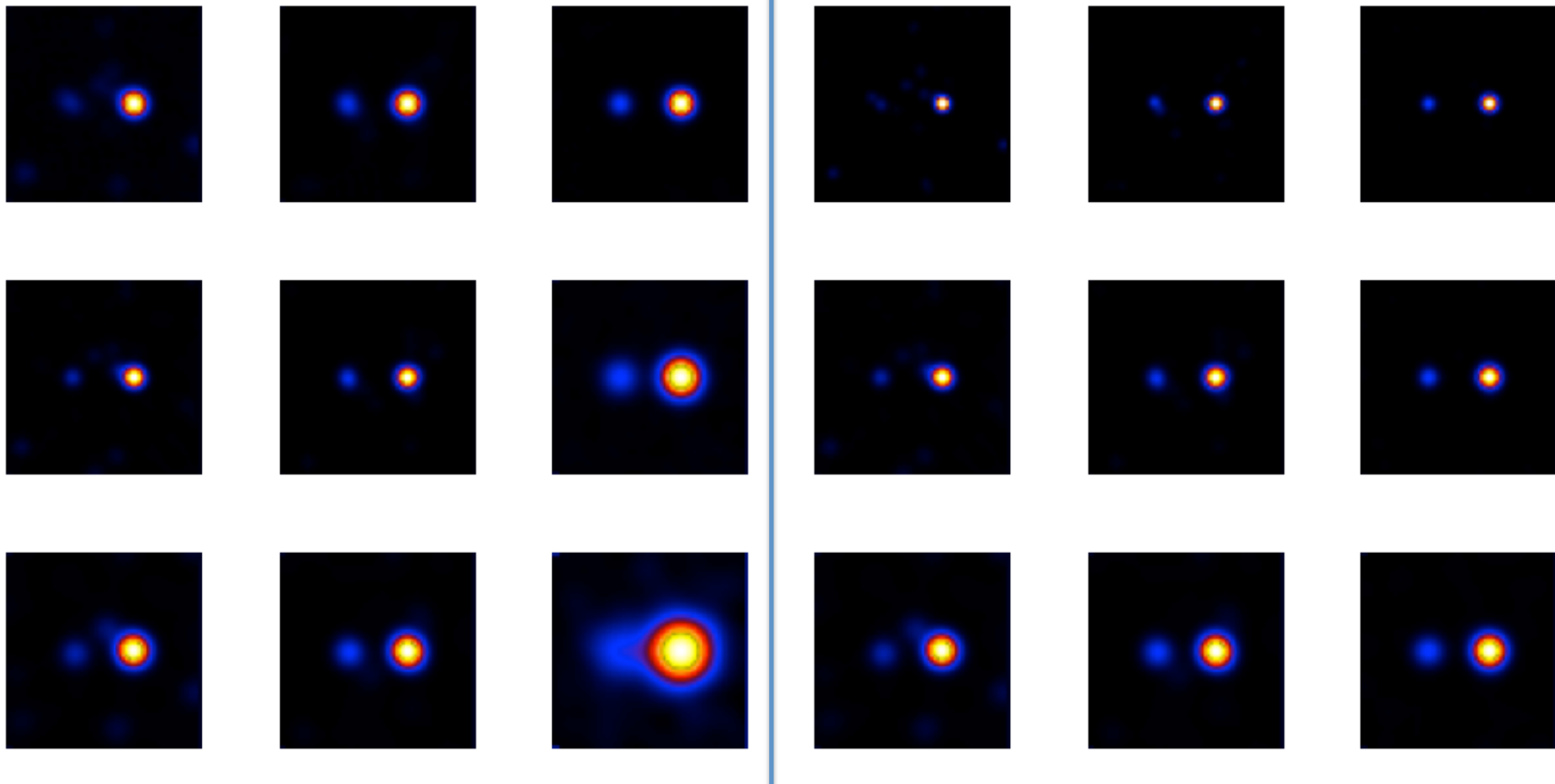


MEM_NJIT

UV_SMOOTH

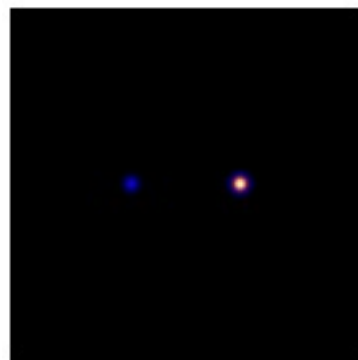


Anna analysis: Dynamic range (Flux ratio = 1)

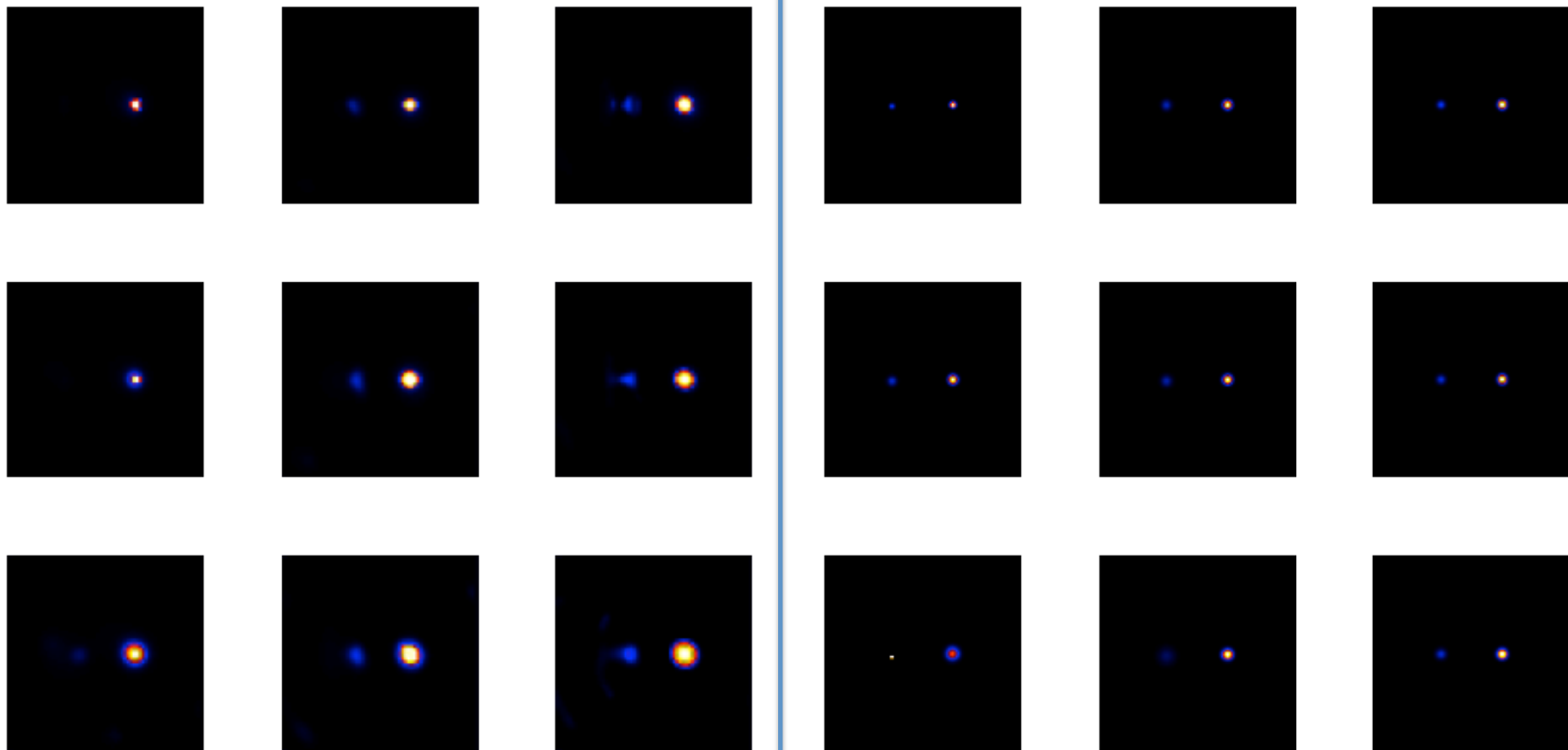


CLEAN DEFAULT

CLEAN ENHANCED

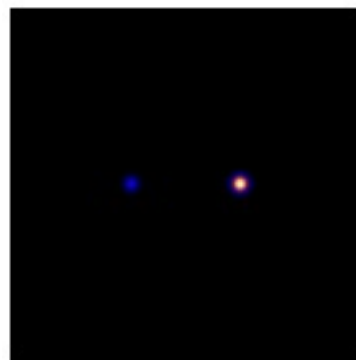


Anna analysis: Dynamic range (Flux ratio = 5)

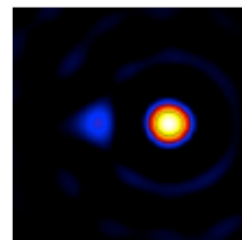
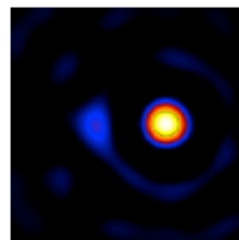
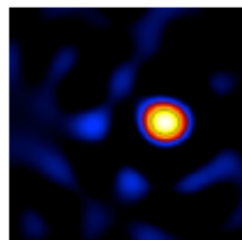
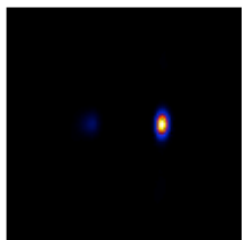
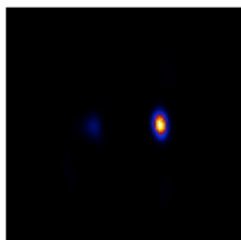
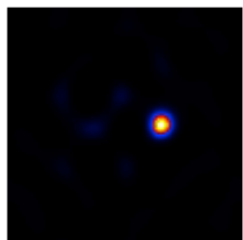
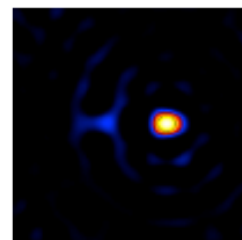
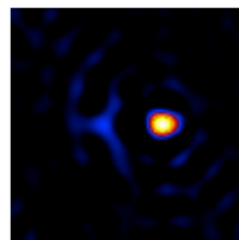
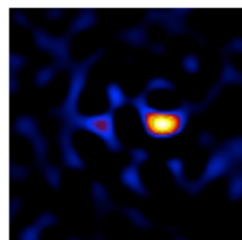
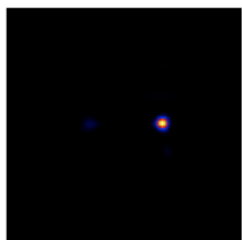
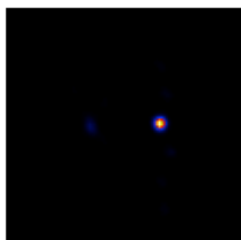
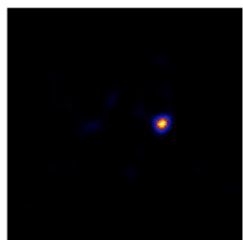
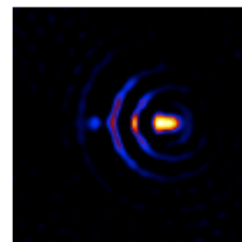
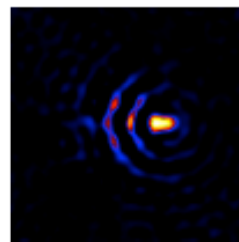
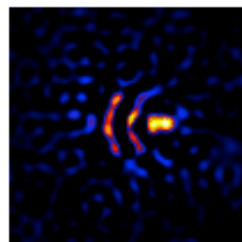
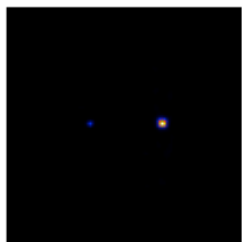
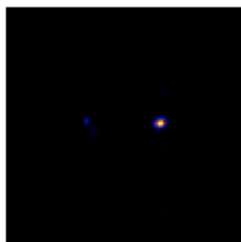
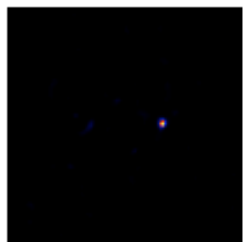


PIXON

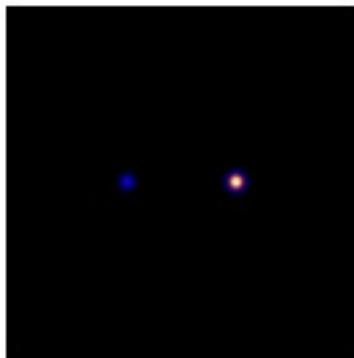
VIS FWD



Anna analysis: Dynamic range (Flux ratio = 5)

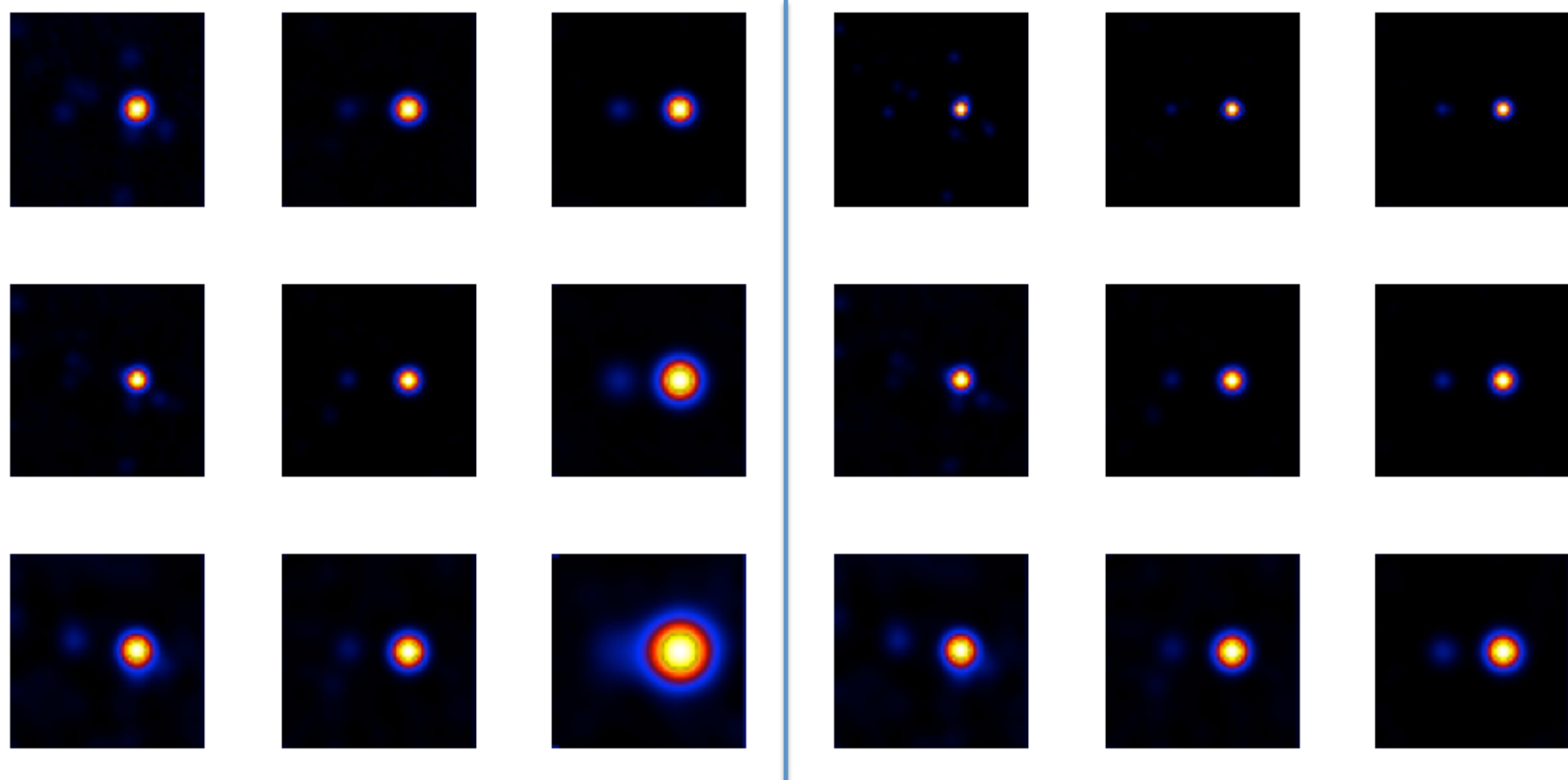


MEM_NJIT



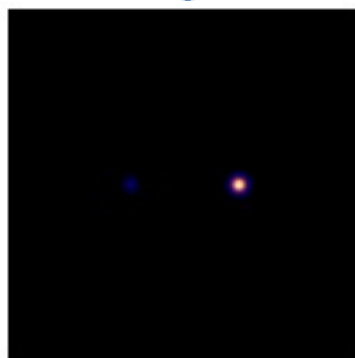
UV_SMOOTH

Anna analysis: Dynamic range (Flux ratio = 5)

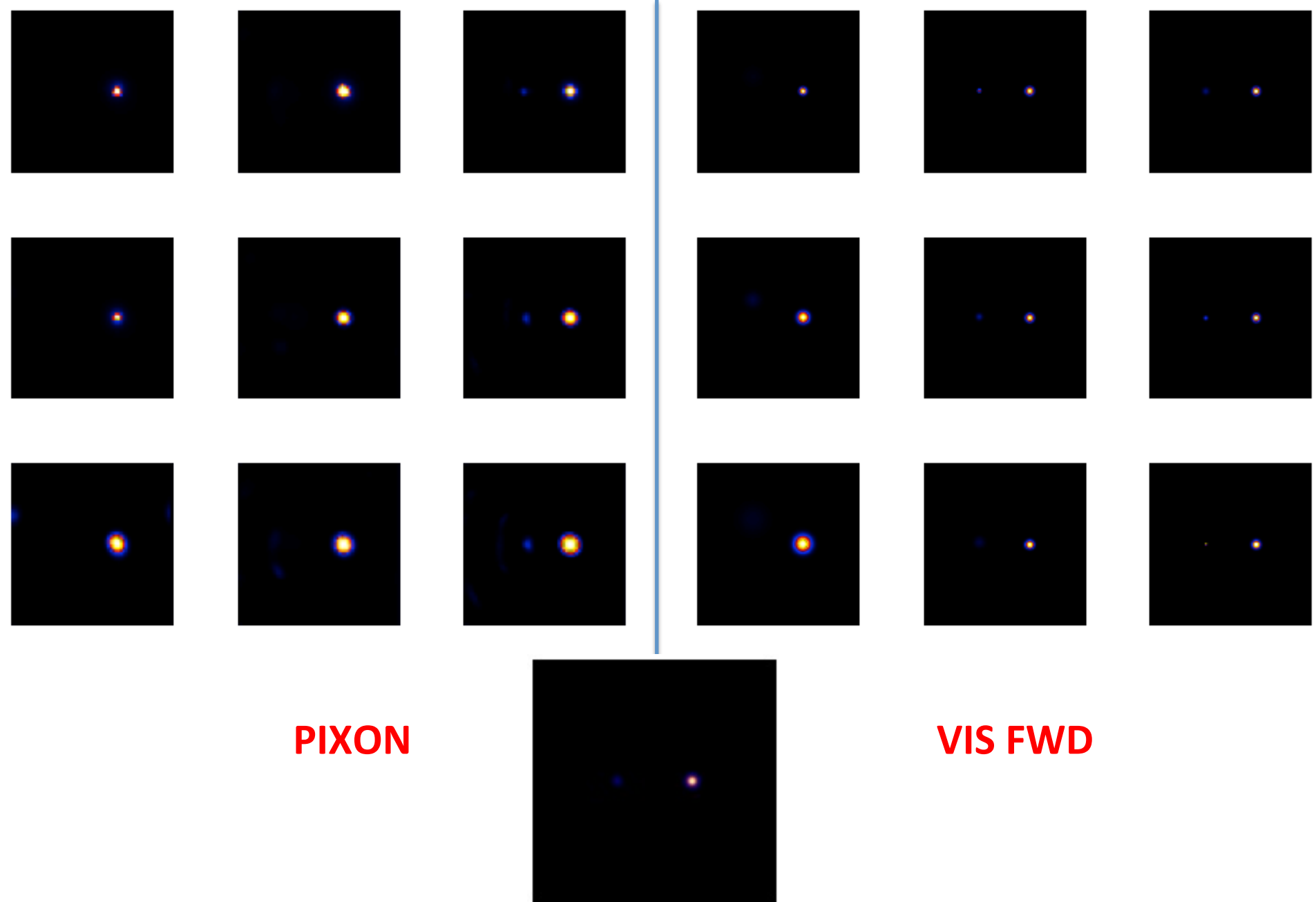


CLEAN DEFAULT

CLEAN ENHANCED



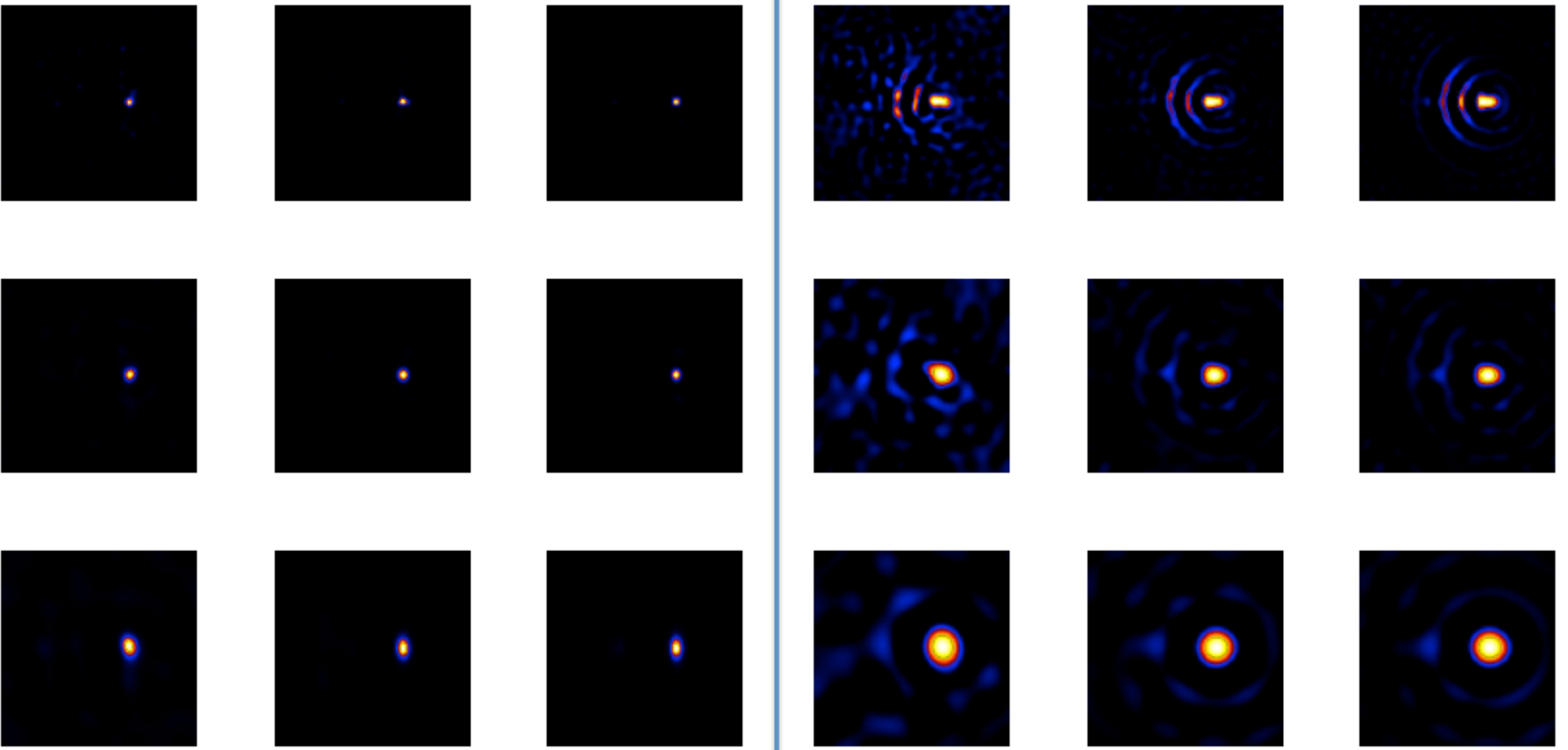
Anna analysis: Dynamic range (Flux ratio = 10)



PIXON

VIS FWD

Anna analysis: Dynamic range (Flux ratio = 10)



MEM_NJIT

UV_SMOOTH

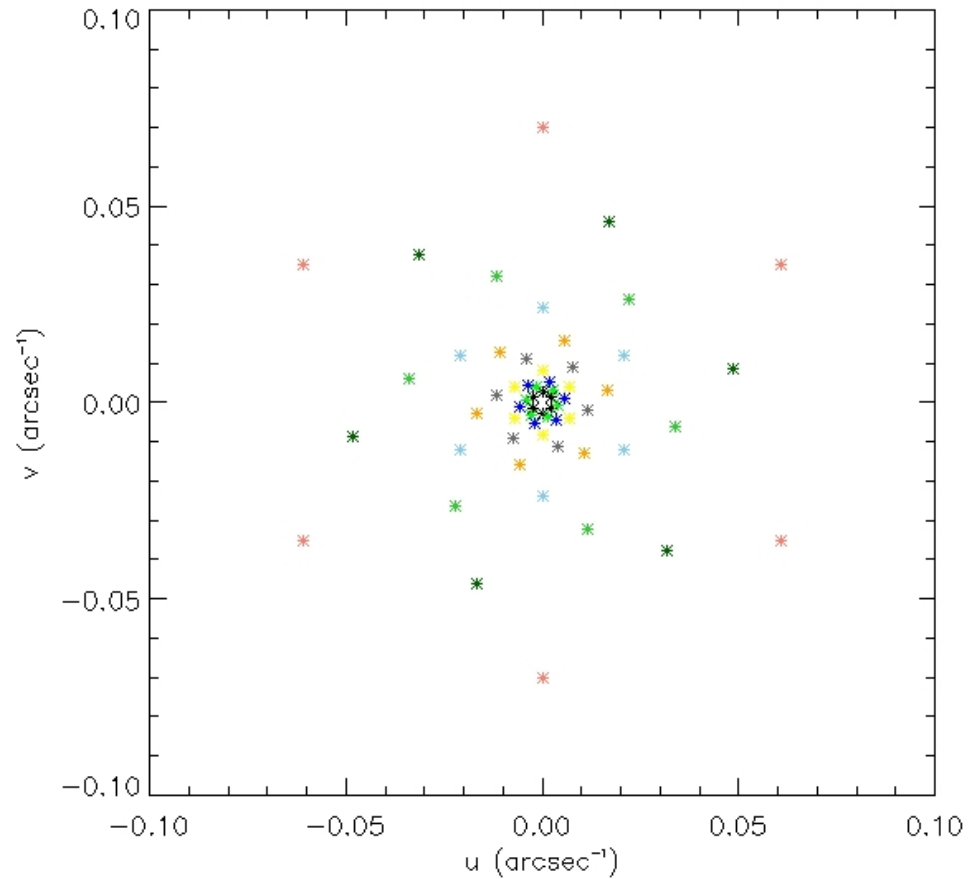
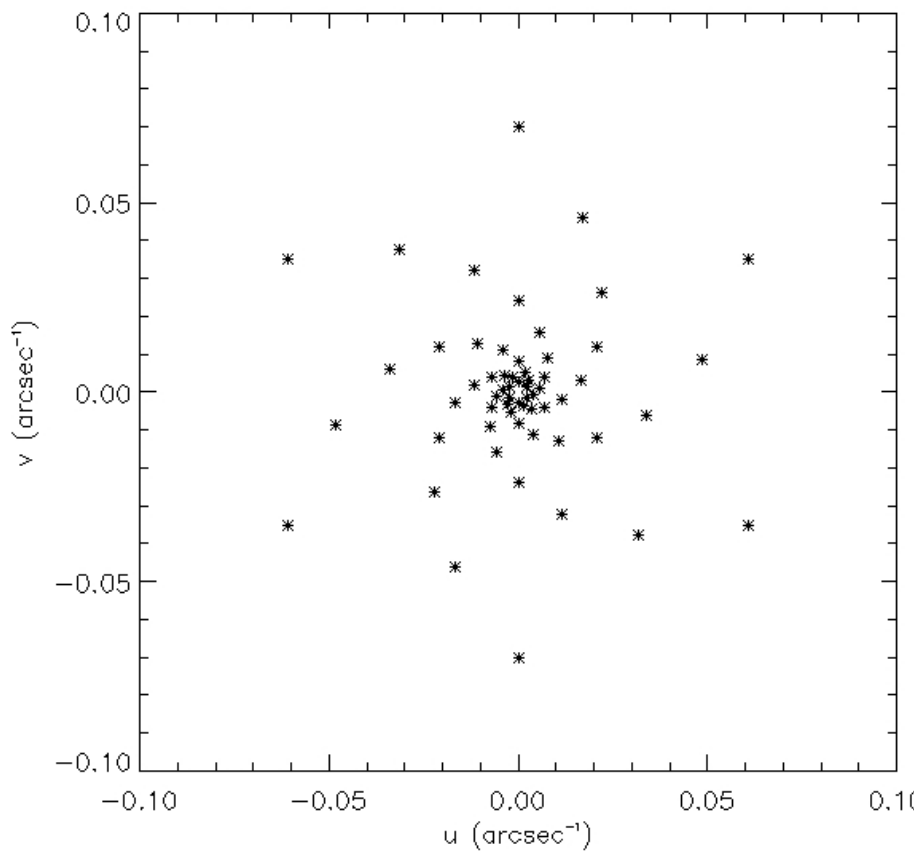
Anna analysis: Dynamic range (Flux ratio = 10)

Lookup table connecting each algorithm to a specific imaging parameter

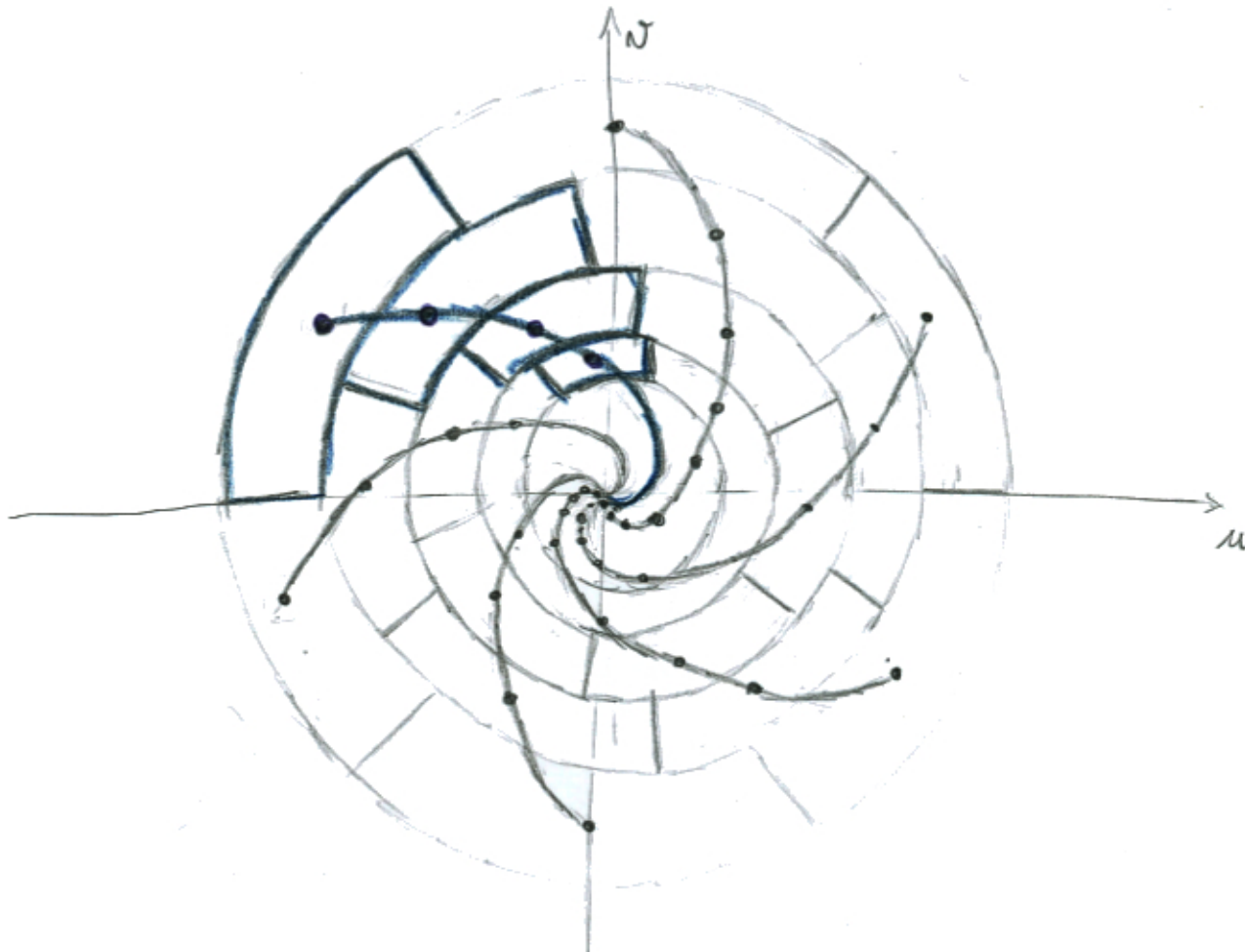
RHESSI Imaging Algorithm Evaluation Framework							
Criteria		Back projection	CLEAN	PIXON	MEM-NJIT	VIS_FWDFIT	UV_SMOOTH
General Characteristics							
	Robustness			slower	can crash		
	Need for parameter optimization						
	Error estimation	Only for single sources					
	Relative Speed						

Different colors correspond to different reliability degrees for each imaging property
 (green: high degree;
 yellow: handle with care;
 red: low degree).

Polar Back-Projection

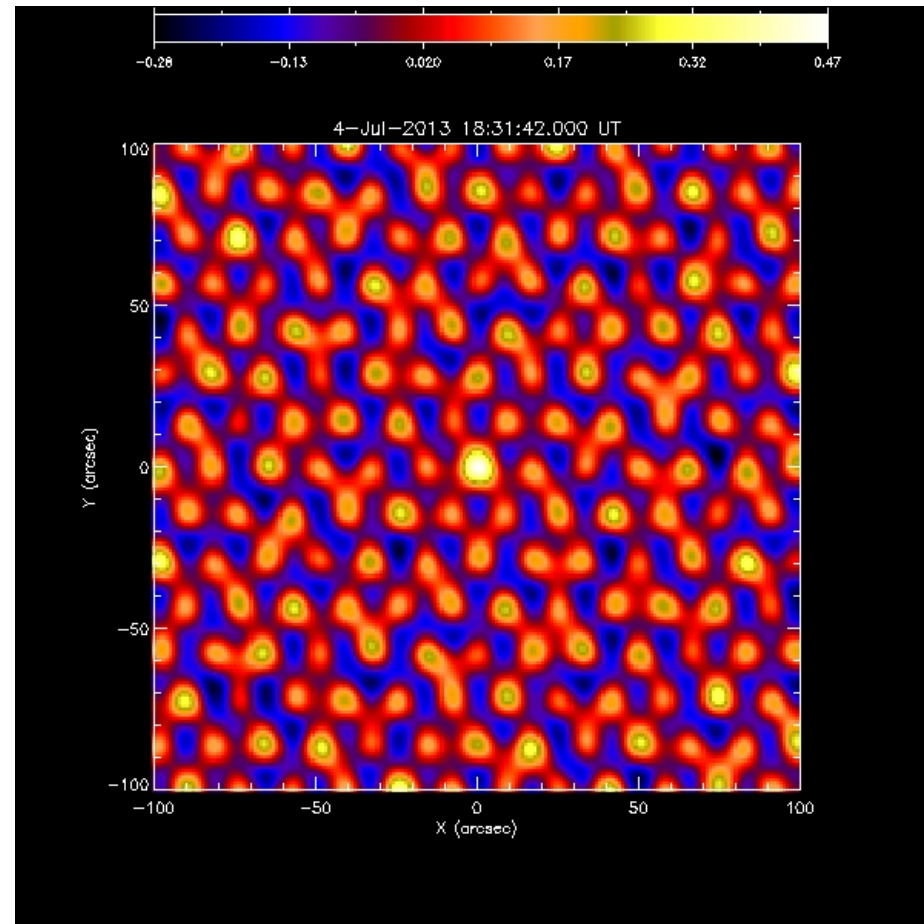
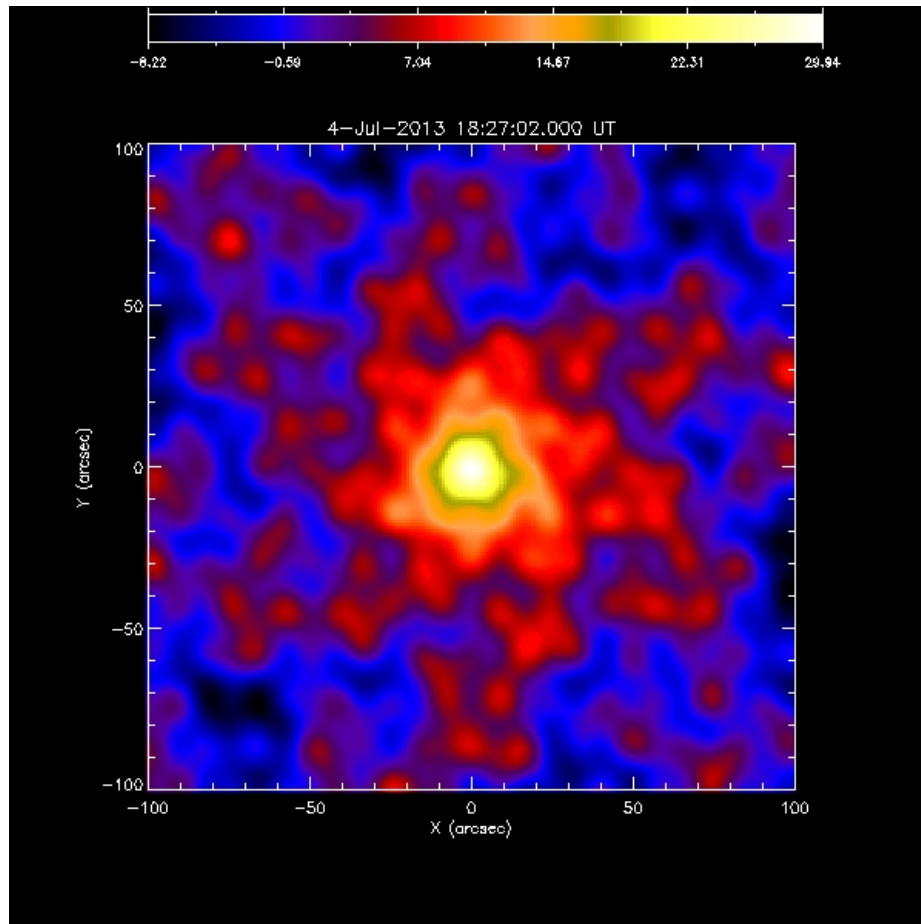


Polar Back-Projection (Gordon Emslie)



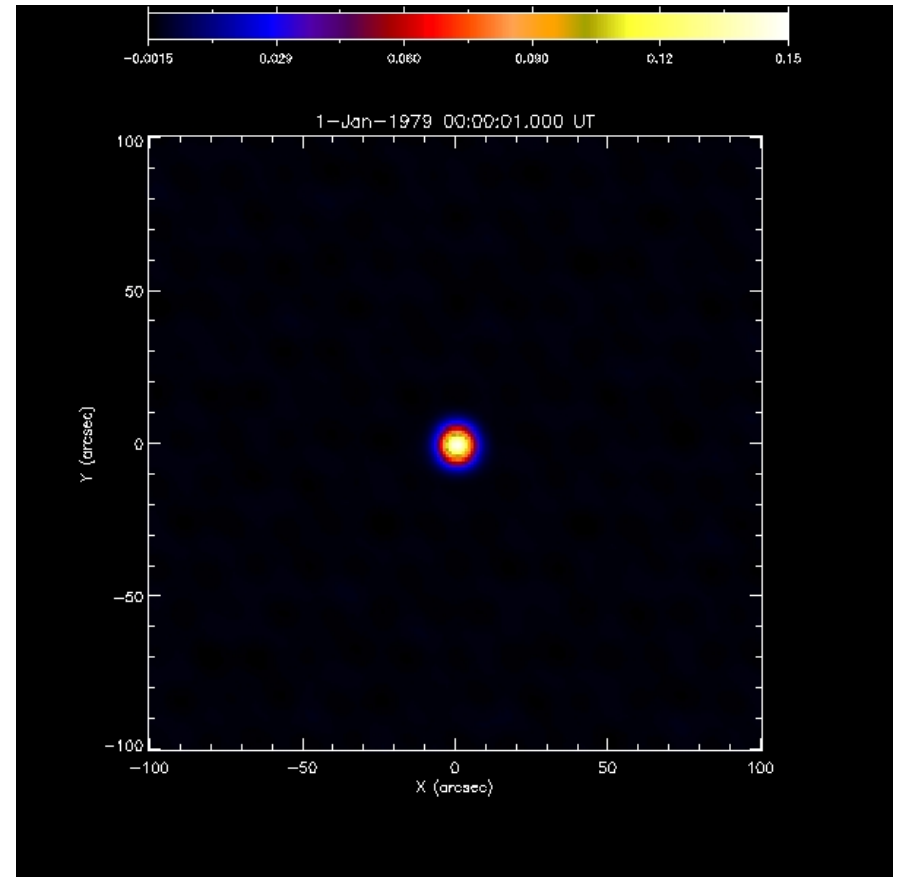
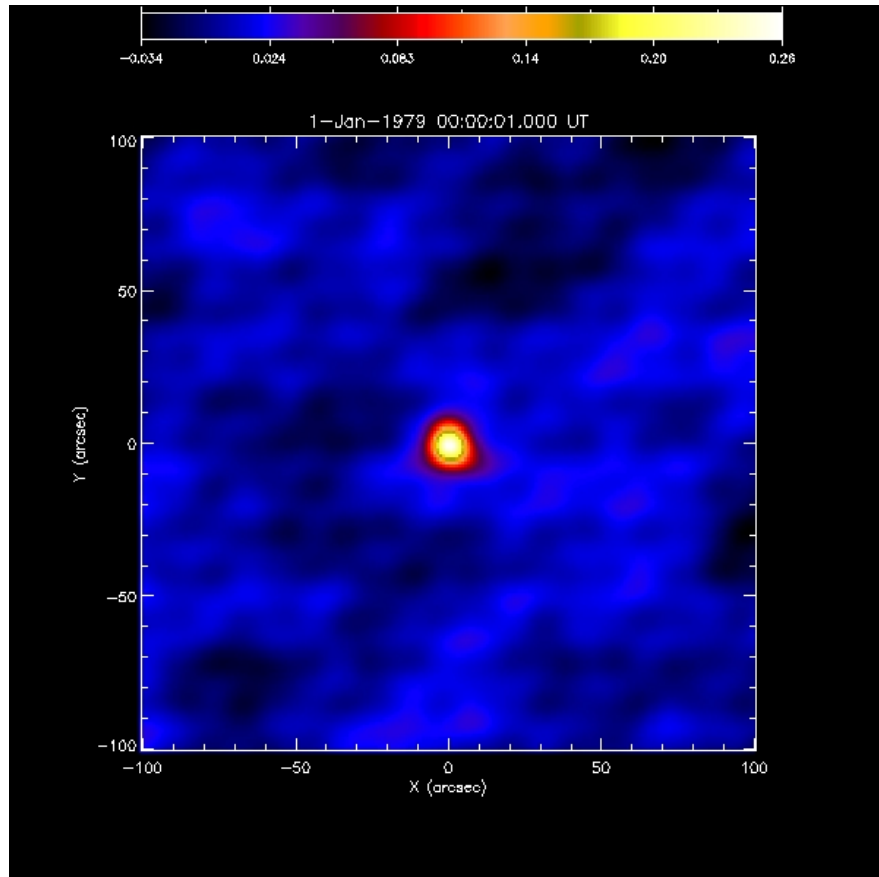
Bpj vs Polar Bpj

Low statistics



Vis-Clean results

Low statistics



expectation-maximization

$$f_{k+1} = f_k \frac{A^T \left(\frac{g}{Af_k} \right)}{A^T \mathbf{1}}$$

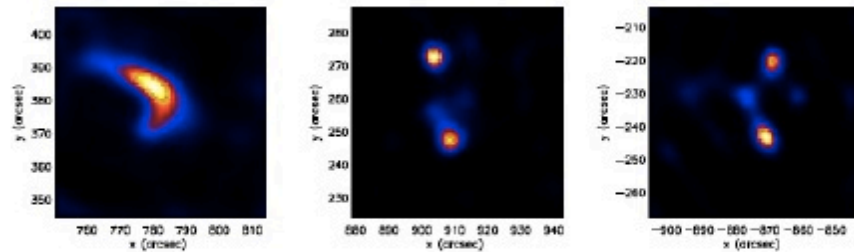
solves a fixed point problem

either $f_k \rightarrow 0$ or $\alpha_k = \frac{A^T \left(\frac{g}{Af_k} \right)}{A^T \mathbf{1}} \rightarrow 1$

therefore $z_k := \left\| f_k A^T \left(1 - \frac{g}{Af_k} \right) \right\|^2 \rightarrow 0$

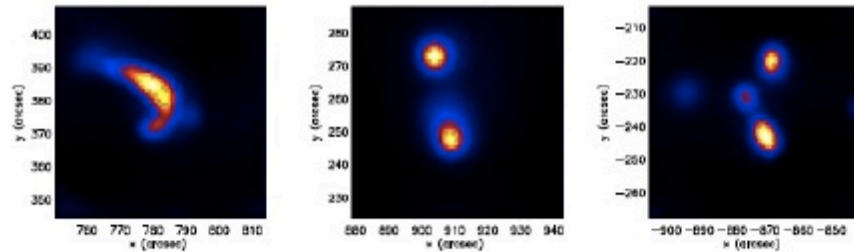
therefore $z_k = E(z_k)$ is a stopping rule

benvenuto, schwartz, piana and massone, AA, 2013



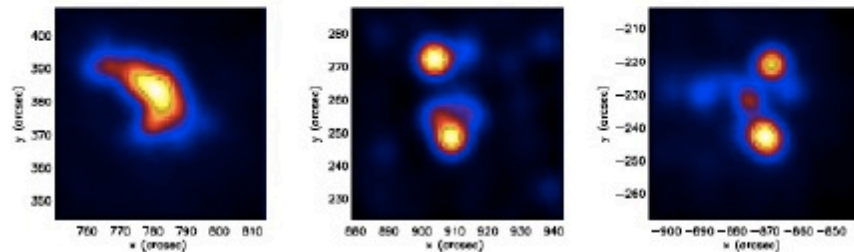
April 15 2002

	3	4	5	6	7	8
EM	1.360	1.384	1.207	2.207	2.312	5.302
Pixon	1.459	1.507	1.486	2.302	2.464	5.111
CLEAN	14.50	13.98	13.86	27.65	41.07	77.79



February 20 2002

	3	4	5	6	7	8
EM	1.093	1.104	1.037	1.157	1.056	0.882
Pixon	1.212	1.250	1.154	1.167	1.302	1.106
CLEAN	1.725	1.695	1.721	2.194	2.505	3.249



July 23 2002

	3	4	5	6	7	8
EM	1.080	0.973	1.220	1.341	1.690	1.837
Pixon	1.176	1.007	1.224	1.340	1.702	1.797
CLEAN	5.302	5.326	4.711	5.756	3.908	9.878

SMC

$$p(I | V_{obs}) = \frac{p_{pr}(I)p(V_{obs} | I)}{p(V_{obs})}$$

V_{obs} observed visibilities

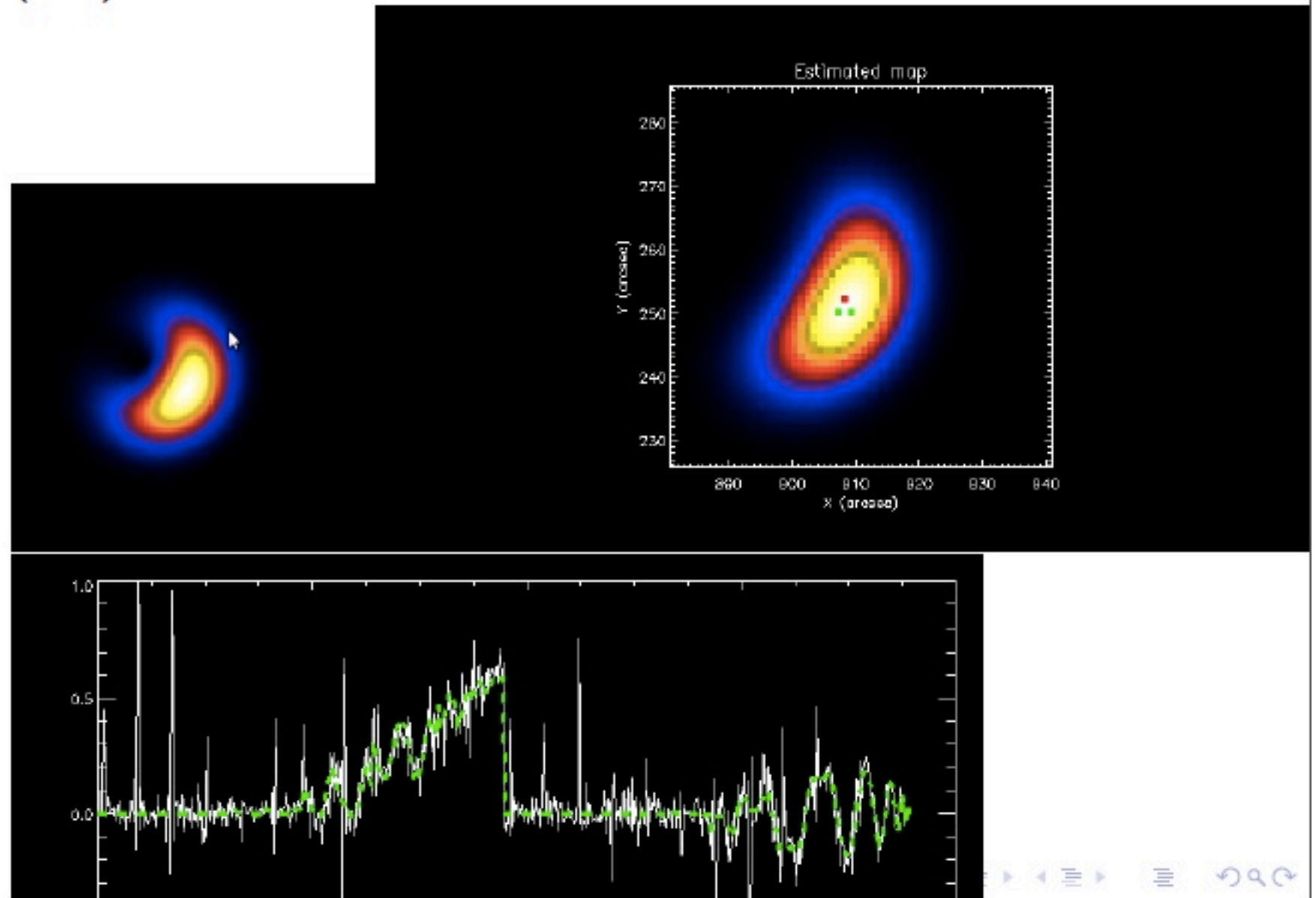
I set of N sources (ellipses or loops)

SMC

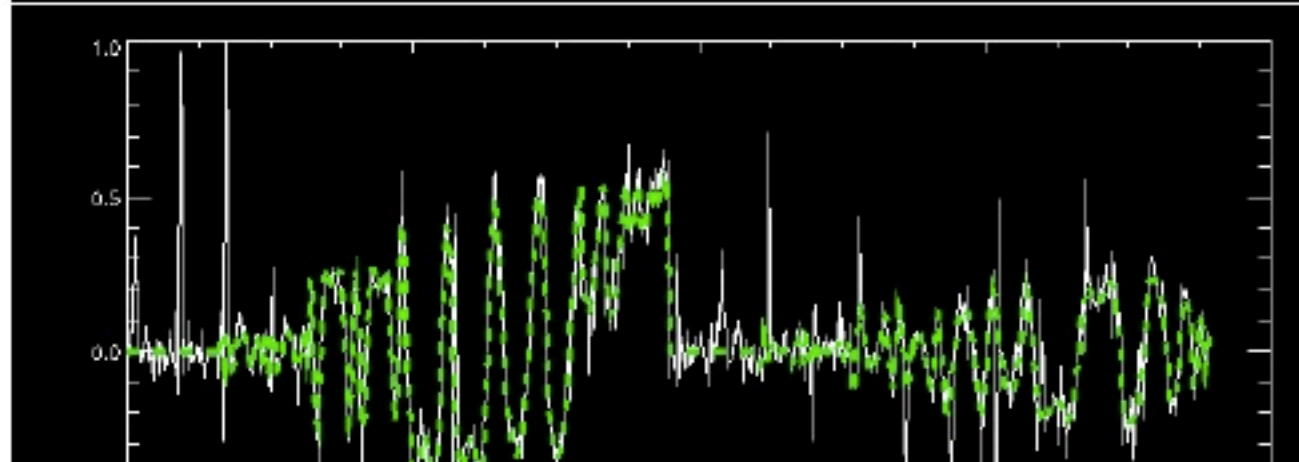
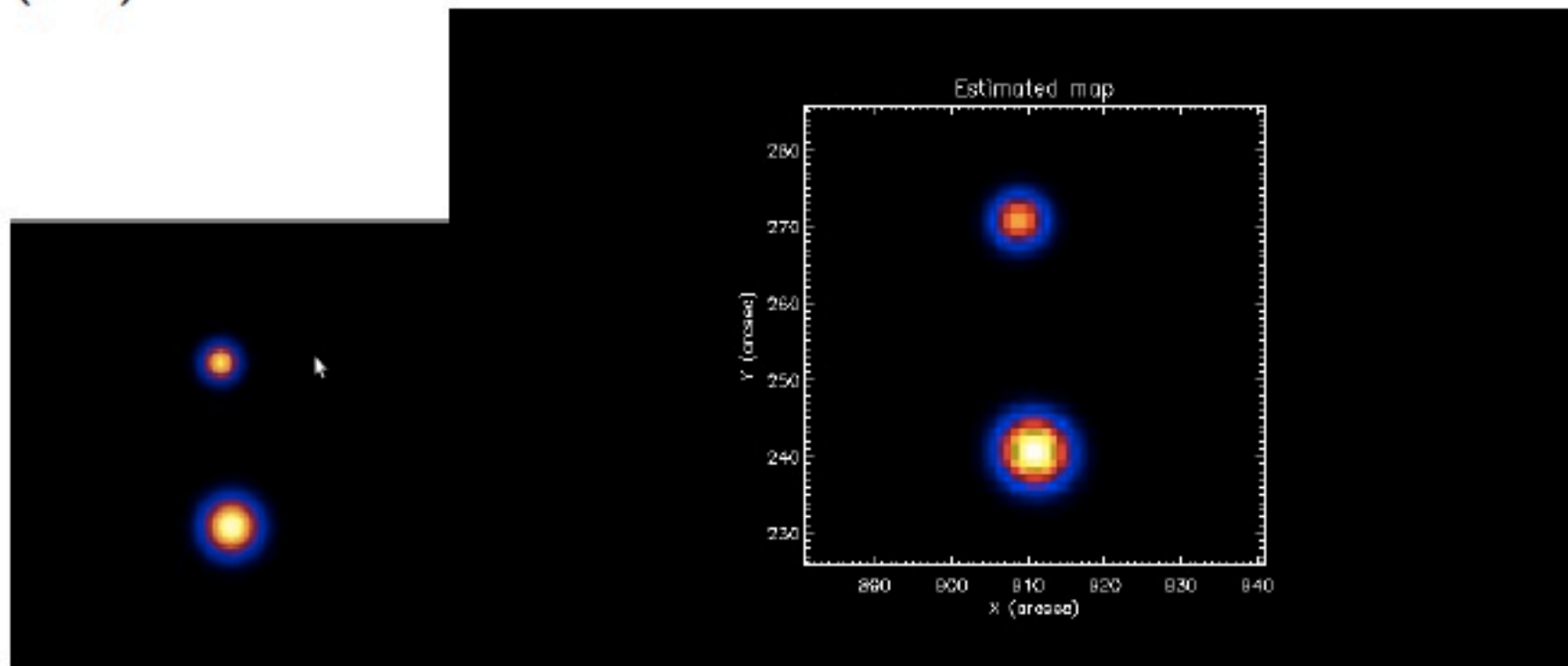
algorithm:

1. generate P random candidate solutions using the prior
2. compute the visibilities corresponding to each candidate
3. assign a weight to each candidate according to the likelihood
4. discard low-weight candidates and reproduce high weight ones
5. let the candidate evolve according to an MCMC process

(true)



(true)



imaging spectroscopy

electron maps

IMAGING (* - changing these parameters forces reprocessing and takes longer)

Select Input:

Selected Time Range: 20-Feb-2002 11:00:00 to 20-Feb-2002 11:22:00
Flare 2022003: 20-Feb-2002 11:04:16.000 to 11:15:08.000 Peak: 11:06:58.000, 480.000 c/s

* 1 Image Time Interval:

* 1 Energy Band (keV): Binning Code:

Collimators and Detector Front/Rear Segments Selected:
1FR, 2FR, 3FR, 4FR, 5FR, 6FR, 7FR, 8FR, 9FR

Automatic Time Bin Calculation: Enabled Digital Quality: 0.95

Pixel Size (asec): 4.0 x 4.0 Image Dimensions (pixels): 64 x 64
Offset of Map Center from Sun Center (asec): X: 910.87 Y: 255.76
Image Size = 256 x 256 asec Range: X: 783 to 1039 asec Y: 128 to 384 asec

* Image Algorithm:

Visibility Type:

Flatfield: Enabled Modpat_skip: 1 Phase Stacker: Disabled Cull: Enabled (Fraction: 0.50)
Weighting: Natural Tapering Width (asec): 0.00 Local Average: Disabled
Variable Flux Correction: Enabled Decimation Correction: Front Rate-based BProj: Enabled

Send Image(s) to: ☒ GUI ☐ FITS File Show: ☒ Progress Bar ☐ Verbose ☐ Images ☐ Profiles

new release in progress:
non-uniform photon
energy channels allowed

source dimension

$$\frac{L(E)}{2} = \frac{L_0}{2} + \frac{1}{Kn} \sqrt{\frac{2}{(\delta - 3)(\delta - 5)}} E^2.$$

volume: $V_0 = \frac{\pi W^2 L_0}{4}$

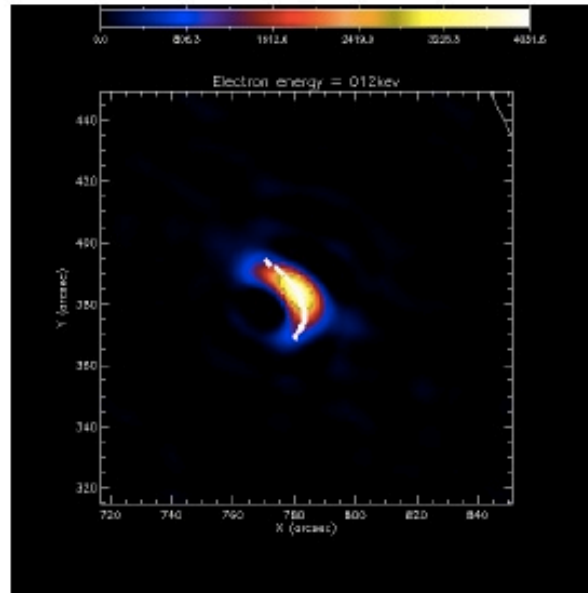
number of particles: $\mathcal{N} = n V_0$

acceleration rate: $\eta(E_0) = \frac{1}{\mathcal{N}} \frac{d\mathcal{N}}{dt}(\geq E_0)$

filling factor: $f = \frac{V_{\text{emit}}}{V} = \frac{\text{EM}}{n^2 V}$

Event No.	L_0 (arcsec)	W (arcsec)	V_0 (100 arcsec ³)	n (10 ¹¹ cm ⁻³)	\mathcal{N} (10 ³⁷)	η (20 keV) (10 ⁻³ s ⁻¹)	f
1	18.6	7.0	7.2	1.5	4.1	6.5	0.45
2	16.3	6.9	6.2	1.4	3.2	14.5	0.83
3	16.7	7.3	7.0	4.4	11.7	4.0	0.04
4	16.6	7.3	7.0	4.8	12.8	7.3	0.11
5	16.6	8.2	8.7	10.5	34.9	3.3	0.03
6	11.9	5.9	3.3	4.9	6.0	0.6	0.02
7	10.4	6.0	3.0	1.8	2.0	12.1	0.44
8	17.8	6.9	6.4	2.6	7.1	24.1	0.90
9	18.8	6.6	6.5	2.9	7.7	23.1	1.05
10	15.1	6.0	4.2	2.9	5.4	13.8	0.72
11	16.0	5.7	4.1	1.9	3.1	27.8	1.95
12	10.3	6.6	3.5	5.1	6.7	4.9	0.08
13	9.9	6.5	3.3	4.6	5.7	4.1	0.18
14	21.5	5.3	4.8	1.5	2.8	1.4	0.13
15	17.4	6.3	5.4	0.8	1.7	1.7	1.03
16	17.8	6.4	5.8	2.3	5.1	0.3	0.18
17	11.0	6.2	3.3	3.9	5.0	2.9	0.05
18	9.9	6.3	3.1	3.2	3.8	7.0	0.22
19	19.9	6.2	6.1	11.1	25.7	13.6	0.02
20	14.5	6.1	4.2	5.2	8.3	23.4	0.10
21	9.9	6.1	2.9	2.2	2.4	16.5	0.53
22	12.4	6.0	3.6	1.7	2.3	5.2	0.26

continuity equation



$$F(s, E) = F_+(s, E) + F_-(s, E)$$
$$\pm \frac{\partial}{\partial s} F_{\pm}(s, E) - \frac{\partial}{\partial E} \left(-\frac{dE}{ds} F_{\pm}(s, E) \right) = S(s, E)$$

the physics is in:

- the energy loss rate $\frac{dE}{ds}(n, T)$
- the injection term $S(s, E)$

continuity equation

some definitions:

$$\Phi_{\pm}(s, E) := -\frac{dE}{ds} F_{\pm}(s, E) \quad A(s, E) := -\frac{dE}{ds} \frac{S(s, E)}{\sqrt{2mE}}$$

$$x(E) := x_0 - \int_{E_0}^E \frac{1}{\frac{dE}{ds}} dE' \quad \Phi(s, E) = \Phi_+(s, E) + \Phi_-(s, E)$$

wave equation for the electron maps:

$$-\frac{\partial^2}{\partial x^2} \Phi(s, E(x)) + \frac{\partial^2}{\partial s^2} \Phi(s, E(x)) = 2 \frac{\partial}{\partial x} A(s, E(x))$$

initial conditions:

$$\lim_{x \rightarrow \infty} \Phi(x, s \pm x) = 0 \quad \lim_{x \rightarrow \infty} \partial_x \Phi(x, s \pm x) = 0 \quad \forall s$$

continuity equation

$$\square G(s, s'; x, x') = -\partial_x \delta(s, s'; x, x')$$

two solutions:

$$G(s, s'; x, x') = \delta_{\pm}(x - x' \mp |s - s'|)$$

but just one (δ_-) is coherent with the initial conditions. therefore:

$$\Phi(s, E(x)) = (\delta_- * 2A)(x, s)$$

$$F(s, E(x)) = \sqrt{\frac{2}{m}} \frac{1}{\frac{dE}{ds}} \int \delta(x - x' + |s - s'|) \frac{dE}{ds} \frac{S(s', E(x'))}{\sqrt{E(x')}} dx' ds'$$

lhs: measurements

rhs: models (for the energy loss rate and for the injection term)

continuity equation

15–Apr–2002

Time[UT]	Mod	$E_0[keV]$	$n[cm^{-3}]$	$h_s[cm^{-5}keV^{-1}s^{-1}]$	χ^2
00:03:00-00:06:00	Mod 1	7.7 ± 0.2	$(8.0 \pm 0.8) \times 10^{10}$	$(1.6 \pm 0.3) \times 10^{30}$	0.80
	Mod 2	2.5 ± 0.1	$(1.81 \pm 0.16) \times 10^{11}$	$(4.7 \pm 0.8) \times 10^{30}$	1.04
	Mod 3	7.9 ± 1.9	$(7.86 \pm 0.12) \times 10^9$	$(4.2 \pm 0.5) \times 10^{30}$	0.82
	Mod 4	0	$(3.4 \pm 0.3) \times 10^{10}$	$(9.52 \pm 1.19) \times 10^{30}$	3.58
00:06:00-00:09:00	Mod 1	8.2 ± 0.3	$(6.3 \pm 0.6) \times 10^{10}$	$(1.04 \pm 0.21) \times 10^{30}$	0.33
	Mod 2	3.2 ± 0.1	$(1.42 \pm 0.11) \times 10^{11}$	$(3.3 \pm 0.5) \times 10^{30}$	0.55
	Mod 3	3.5 ± 0.8	$(6.99 \pm 0.12) \times 10^9$	$(3.5 \pm 0.4) \times 10^{30}$	0.43
	Mod 4	0	$(2.30 \pm 0.19) \times 10^{10}$	$(7.2 \pm 0.8) \times 10^{30}$	2.45
00:09:00-00:12:00	Mod 1	4.9 ± 1.2	$(1.6 \pm 0.3) \times 10^{10}$	$(5.1 \pm 2.3) \times 10^{28}$	0.65
	Mod 2	3.2 ± 0.6	$(4.5 \pm 0.7) \times 10^{10}$	$(3.45 \pm 1.11) \times 10^{29}$	0.70
	Mod 3	0.2 ± 0.1	$(4.47 \pm 0.11) \times 10^9$	$(1.00 \pm 0.24) \times 10^{30}$	0.79
	Mod 4	0	$(6.2 \pm 0.5) \times 10^9$	$(1.4 \pm 0.3) \times 10^{30}$	1.21