



The large inverse problem of modern radio interferometry

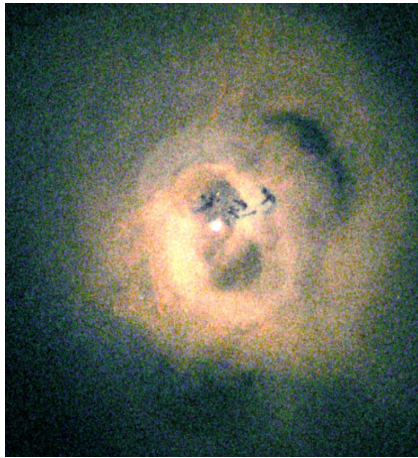
Cyril Tasse

SKA - South Africa (Oleg Smirnov group)
GEPI - Observatoire de Paris

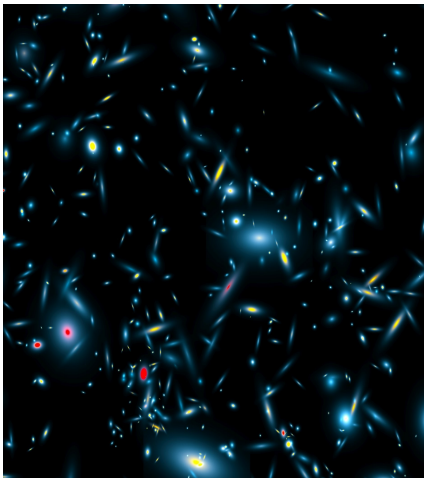
Outline

- 1- Inverting the measurement equation: challenges for the SKA**
- 2- Imaging and deconvolution**
- 3- Calibration and ill-conditioning (solvers and filters)**

Galaxy formation and AGN evolution

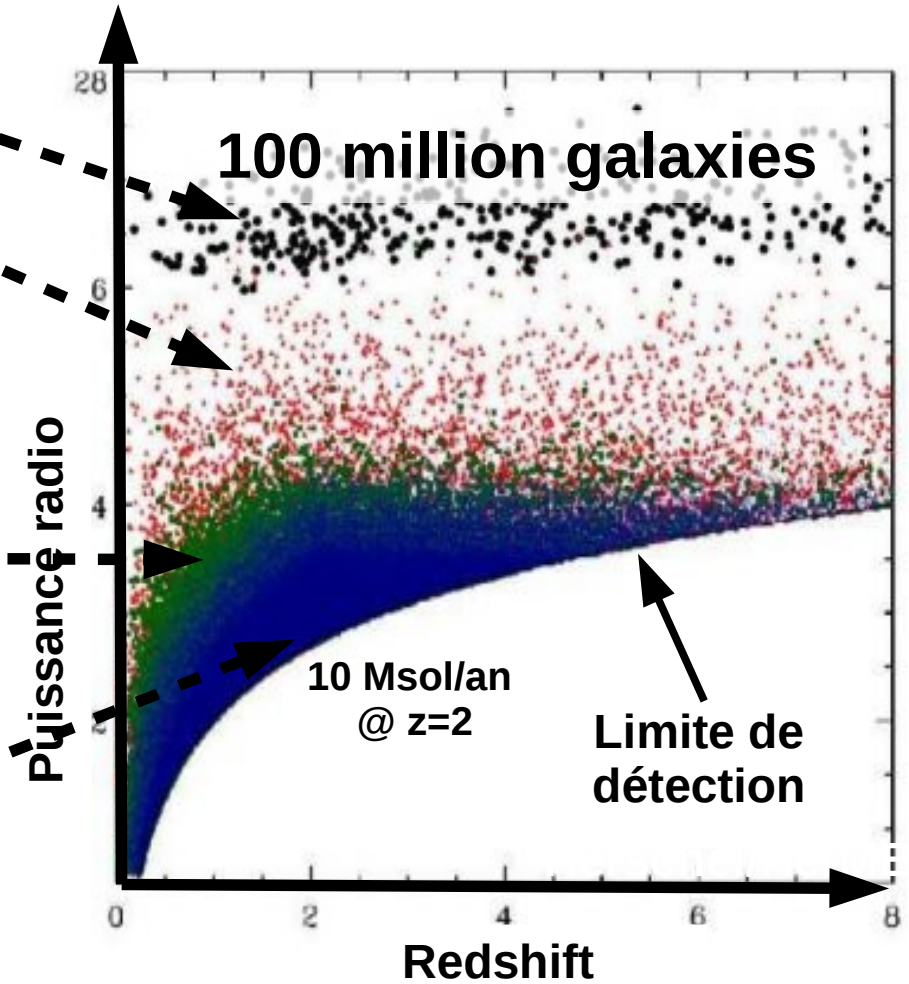


Radio loud
AGN

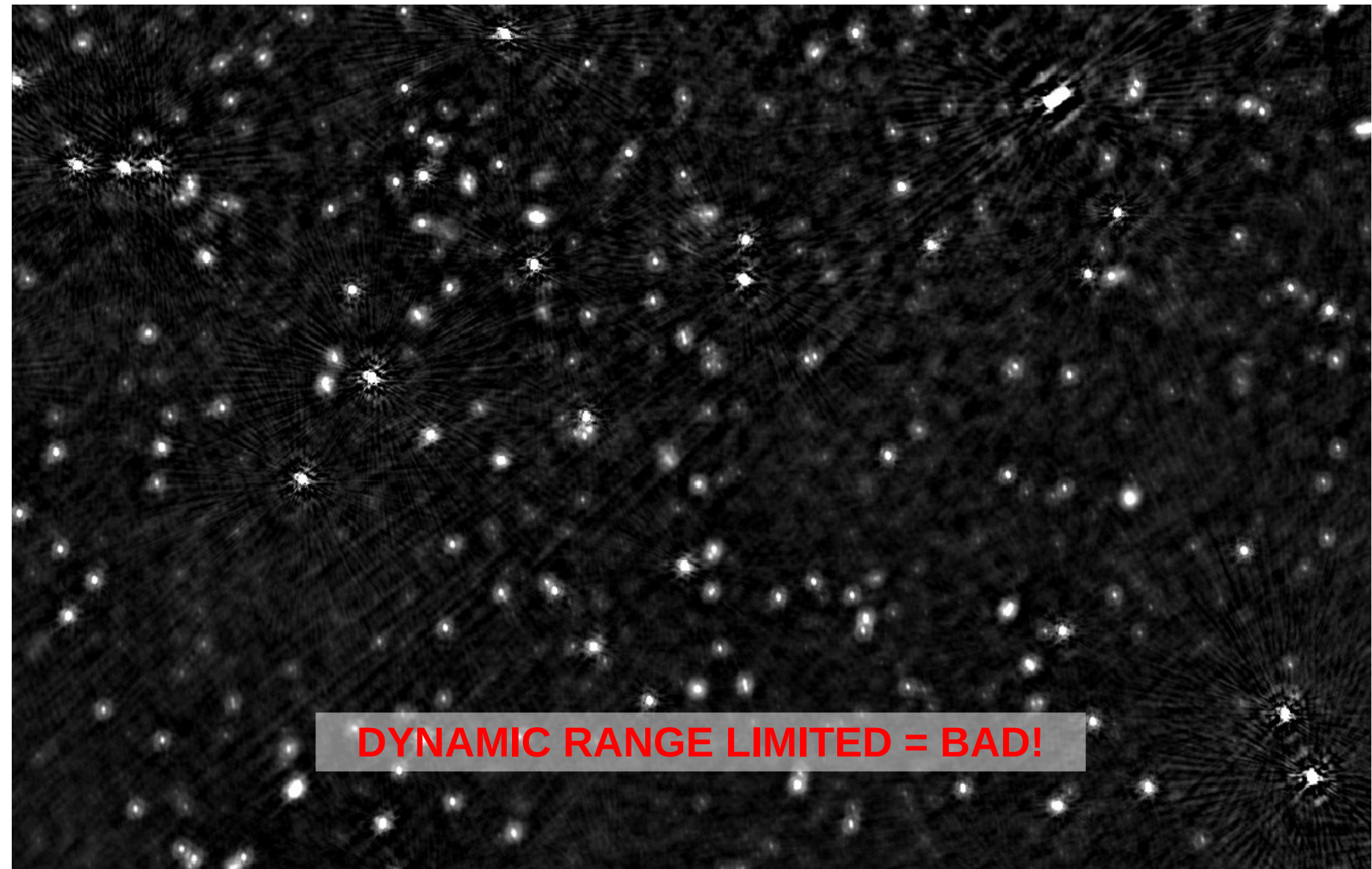


Radio quiet
AGN

Star forming
galaxies



3C295 Observation (110-190 MHz)



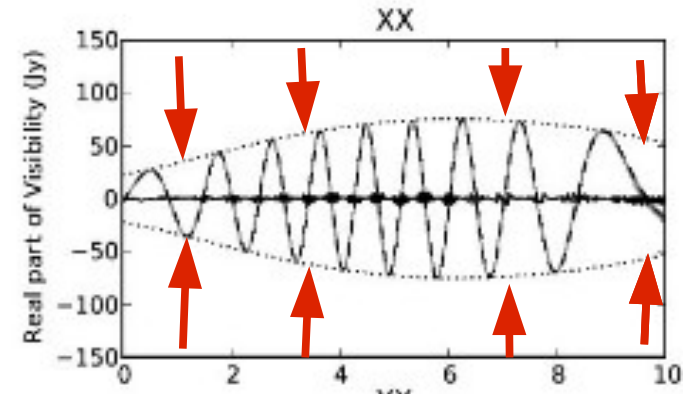
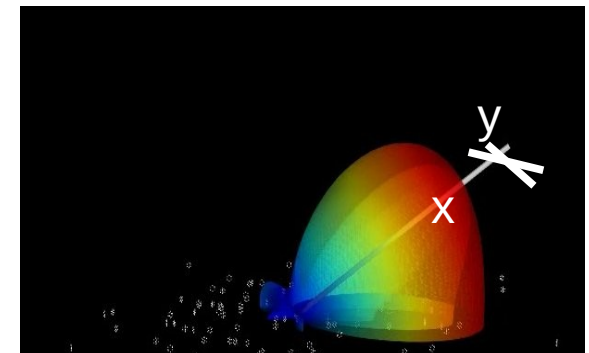
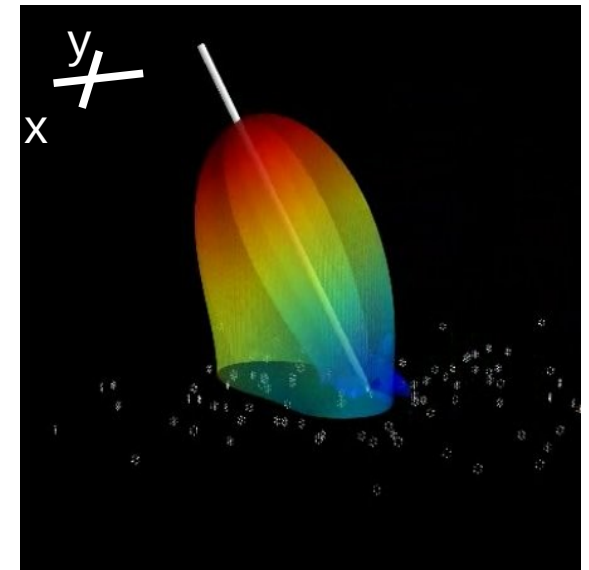
... When Direction Dependent Effects (DDE) become a problem : Beam



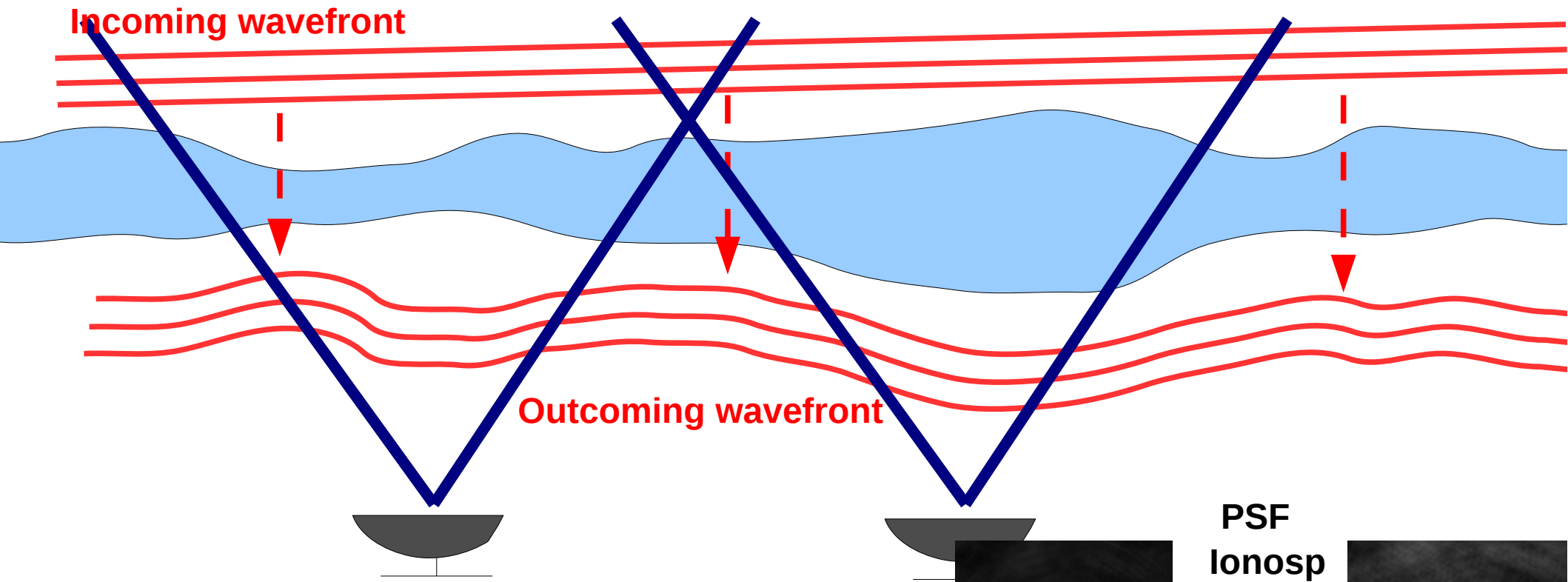
LOFAR stations are phased arrays

- Beam is variable in frequency and time
- Projection of the dipoles in the sky is non trivial
- Beam can be station-dependent
- Individual clock effects

--> Strong effects on polarisation



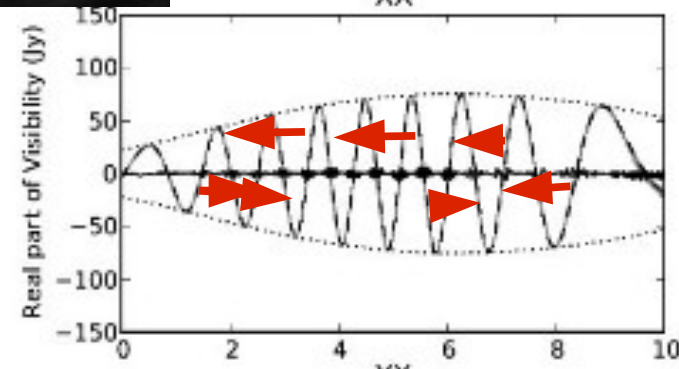
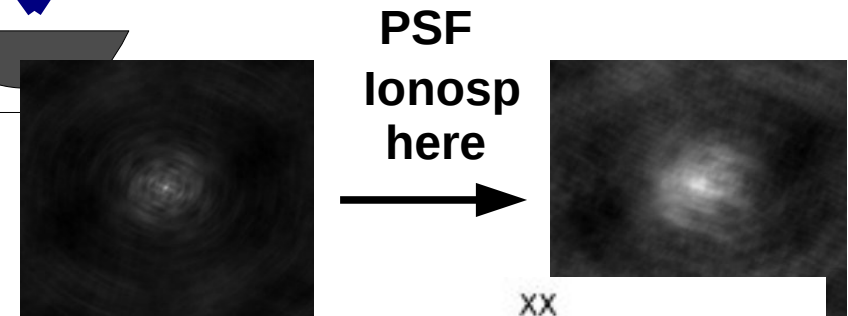
... When Direction Dependent Effects (DDE) become a problem : Ionosphere/troposphere



Big field of view : station, direction, time and frequency dependent

Other direction dependent effects :

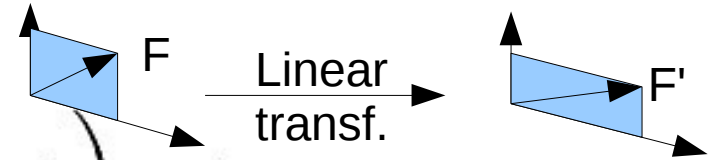
- Faraday rotation
- + **Effect on the polarisation**



The Measurement Equation

Hamaker 1996

$$V_{pq} = \overbrace{G_p}^{\text{Direction independent}} \left(\sum_{i=1}^N \overbrace{B_{pi} K_{pi} I_{pi} F_i}^{\text{Direction dependent}} \cdot \overbrace{F_i^+ I_{qi}^+ K_{qi}^+ B_{qi}^+}^{\text{Source coherency}} \right) \overbrace{G_q^+}^{\text{Direction independent}}$$



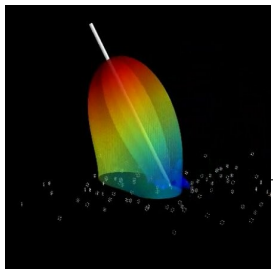
[Voltage antenna p] x [Voltage antenna q]*

Beam

Geometrical delay
+Correlator

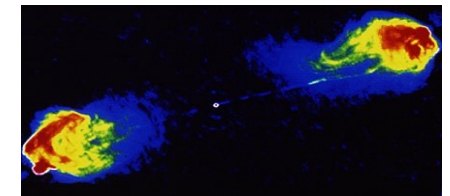
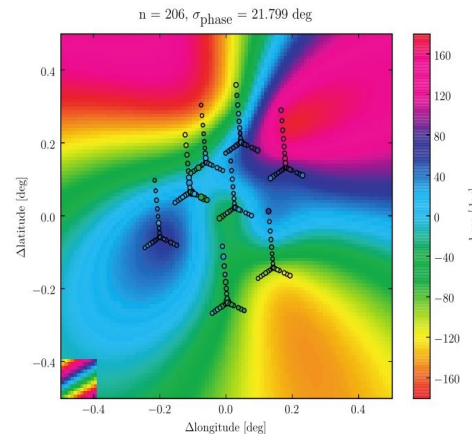
Ionosphere

Electric field



$$K_p K_q^+ = \exp(-2i\pi\phi_{pq}) \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

$$\phi_{pq} = u_{pq}l + v_{pq}m + w_{pq}(\sqrt{1-l^2-m^2}-1)$$



Van der Tol thesis

Challenge for the SKA

$$\begin{array}{c}
 V_{pq} = G_p \left(\sum_{i=1}^N B_{pi} K_{pi} I_{pi} F_i \cdot F_i^+ I_{qi}^+ K_{qi}^+ B_{qi}^+ \right) G_q^+ \\
 \bullet \\
 \bullet \\
 \bullet \\
 \bullet \\
 V_{pq} = G_p \left(\sum_{i=1}^N B_{pi} K_{pi} I_{pi} F_i \cdot F_i^+ I_{qi}^+ K_{qi}^+ B_{qi}^+ \right) G_q^+ \\
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 \bullet \\
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 V_{pq} = G_p \left(\sum_{i=1}^N B_{pi} K_{pi} I_{pi} F_i \cdot F_i^+ I_{qi}^+ K_{qi}^+ B_{qi}^+ \right) G_q^+
 \end{array}$$

Unknown physical process
Unknown Sky
Unknown physical process

Those processes include:

- Antenna beams
- Ionosphere
- Troposphere
- Electronics

But also:

- Solar wind scintillation
- ISM scintillation
- and many more...

Measurement
(10¹³++ points)



Very (very) large inverse (non-linear) problem

Interferometry

TRUTH domain

- Ionosphere
- Troposphere
- Beam
- Sky
- Faraday rotation
- Electronics
- etc

baseline

Direction

time

freq

?

Measurement domain

baseline

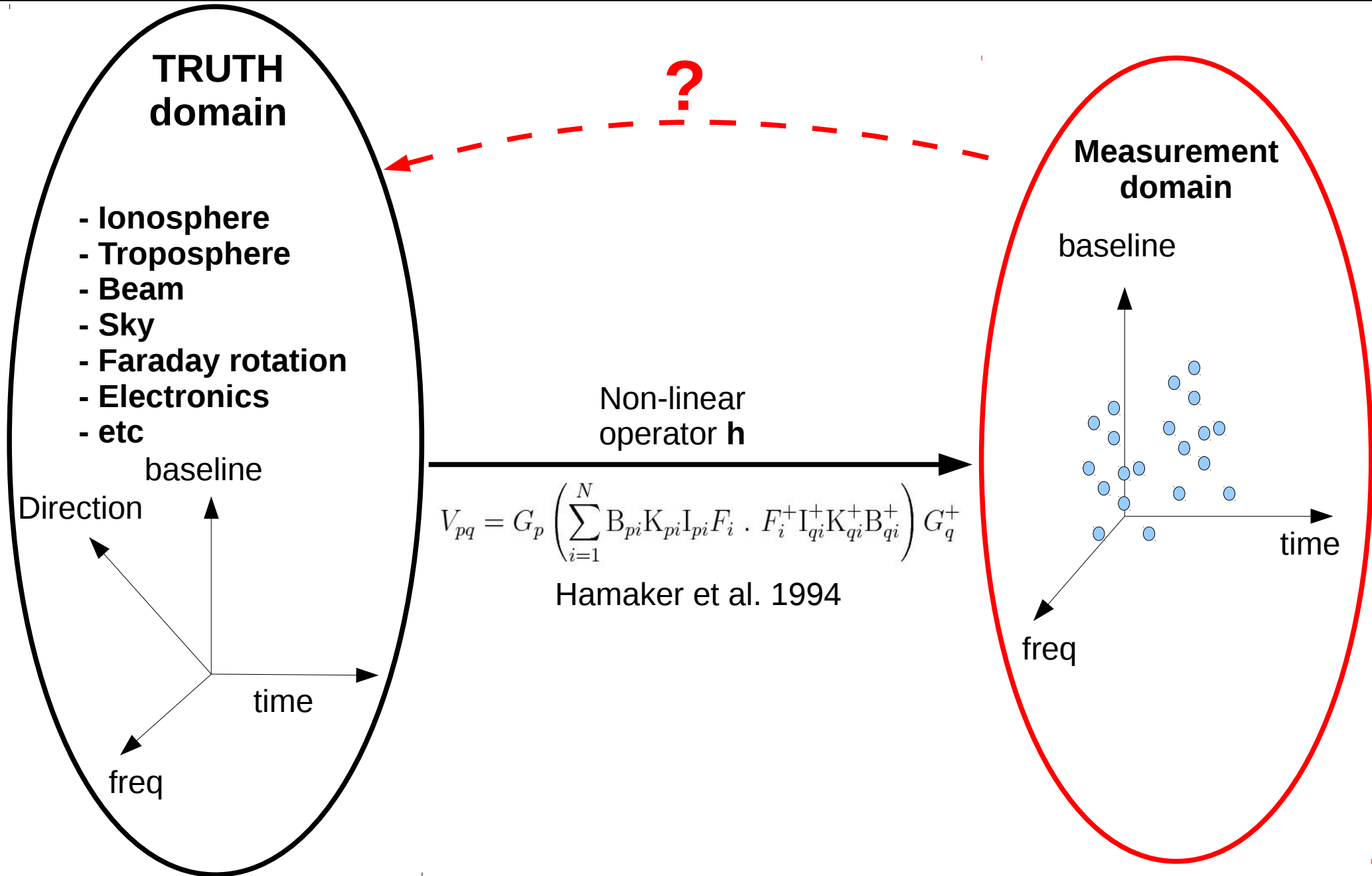
time

freq

Non-linear operator h

$$V_{pq} = G_p \left(\sum_{i=1}^N B_{pi} K_{pi} I_{pi} F_i \cdot F_i^+ I_{qi}^+ K_{qi}^+ B_{qi}^+ \right) G_q^+$$

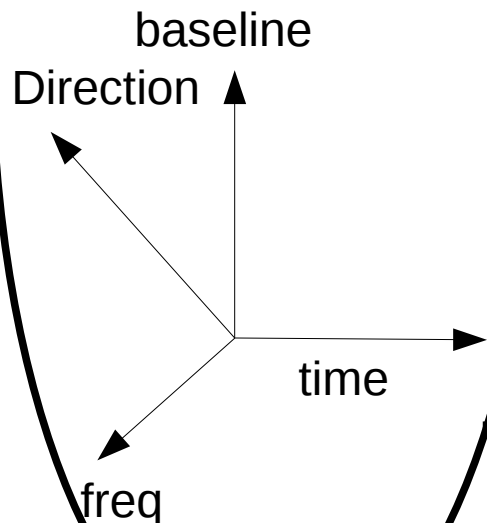
Hamaker et al. 1994



Imaging and deconvolution

TRUTH domain

- Ionosphere
- Troposphere
- Beam
- **Sky**
- Faraday rotation
- Electronics
- etc



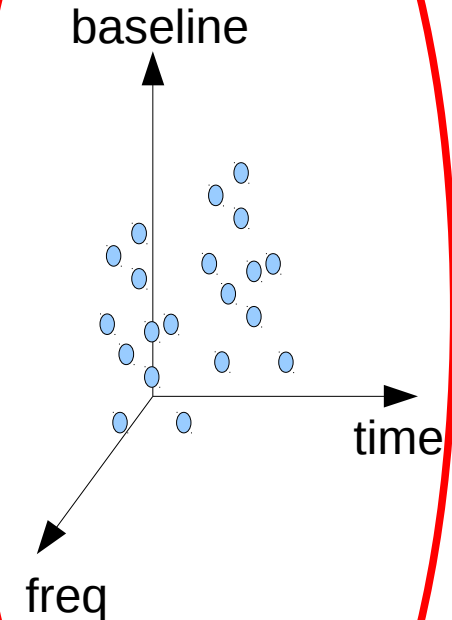
Most algorithms use CLEAN:

- Multi-scale (resolved objects)
- Multi-Term (large fractional bandwidth)

New deconvolution algorithms:

- Compressive sensing (understanding and generalisation of CLEAN, see Jean-Luc Starck, and Arwa Dabbech talk)
- Bayesian inference techniques

Measurement domain



Non-linear operator \mathbf{h}

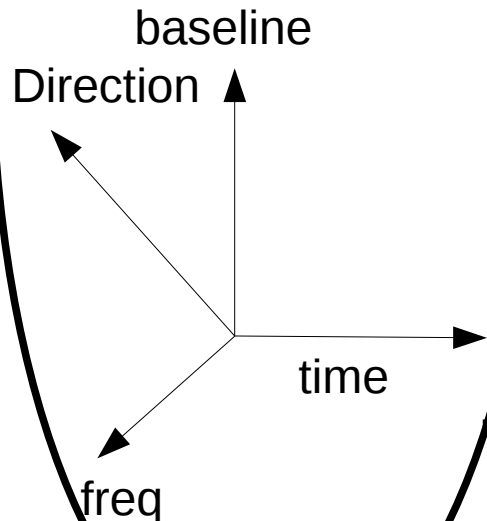
$$V_{pq} = G_p \left(\sum_{i=1}^N B_{pi} K_{pi} I_{pi} F_i \cdot F_i^+ I_{qi}^+ K_{qi}^+ B_{qi}^+ \right) G_q^+$$

Hamaker et al. 1994

Imaging and deconvolution

TRUTH domain

- Ionosphere
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- etc



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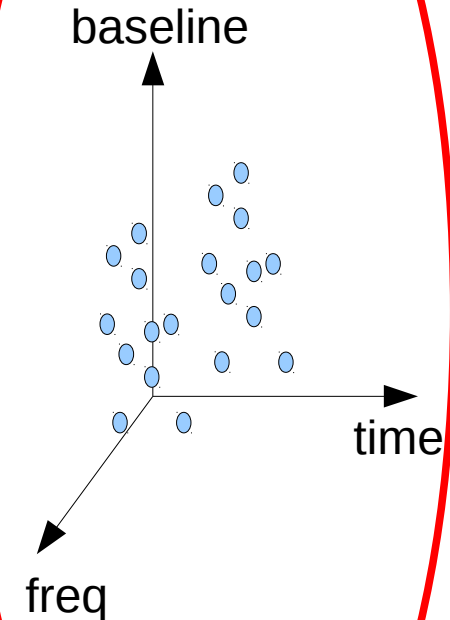
New deconvolution algorithms:

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- Bayesian inference techniques

Full polarisation AW-projection
(Tasse 12): AWimager for LOFAR

$$V = A \cdot I$$

Measurement domain



Non-linear operator \mathbf{h}

$$V_{pq} = G_p \left(\sum_{i=1}^N B_{pi} K_{pi} I_{pi} F_i \cdot F_i^+ I_{qi}^+ K_{qi}^+ B_{qi}^+ \right) G_q^+$$

Hamaker et al. 1994

Full Polarisation A-Projection

Tasse 2013

DDE are in general smooth on the sky --> small support in the uv-domain

$$\text{Vec}(V_{pq}^{corr}) = \int_S (D_{q,s}^{t\nu,*} \otimes D_{p,s}^{t\nu}) \cdot \text{Vec}(I_s) \cdot \exp(-2i\pi\phi(u, v, w, s)) ds$$

Convolution theorem

$$V_{pq}^{t,\nu}(u, v, w, i) = \mathcal{F} \left(\sum_{j=1}^4 D_{pq}^{t,\nu}(i, j, s) W(w, s) I_j(l, m) \right)$$

DDE (4*4)

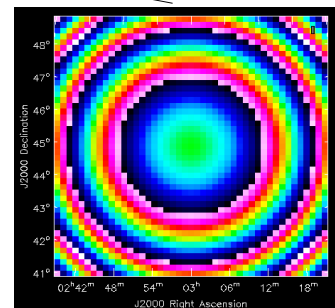
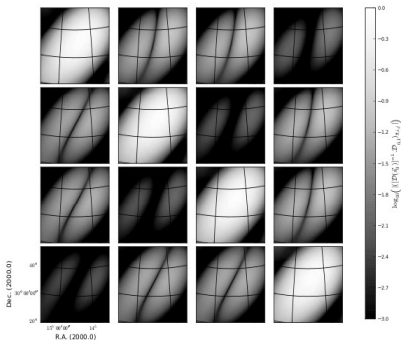
W term (scalar)

$$= \sum_{j=1}^4 \left[\mathcal{F} (D_{pq}^{t,\nu}(i, j, s) W(w, s)) \right]$$

$$* \mathcal{F} (I_j(l, m))$$

Convolution

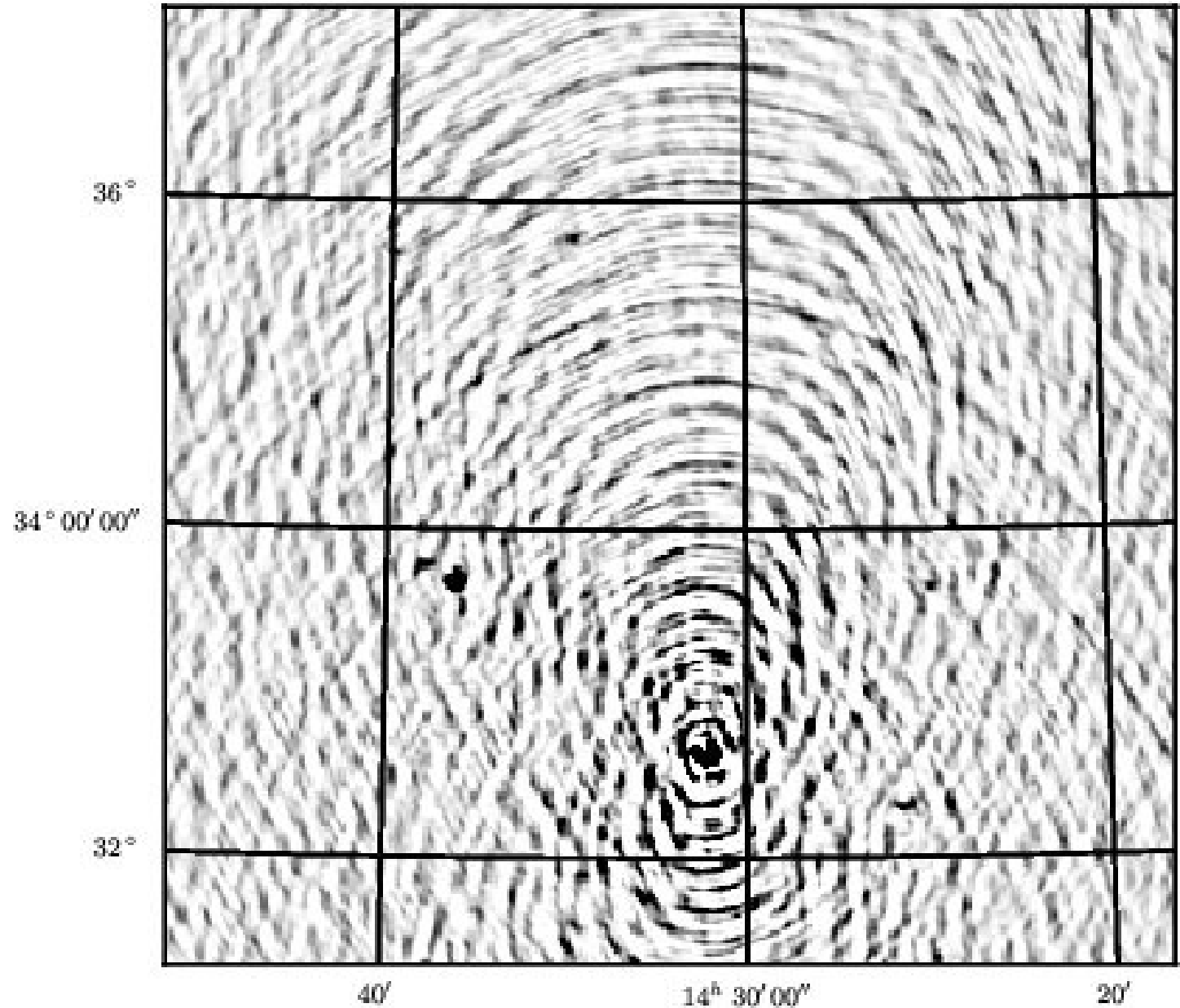
2D FFT



Tests on simulated data

W-term only (Casa)

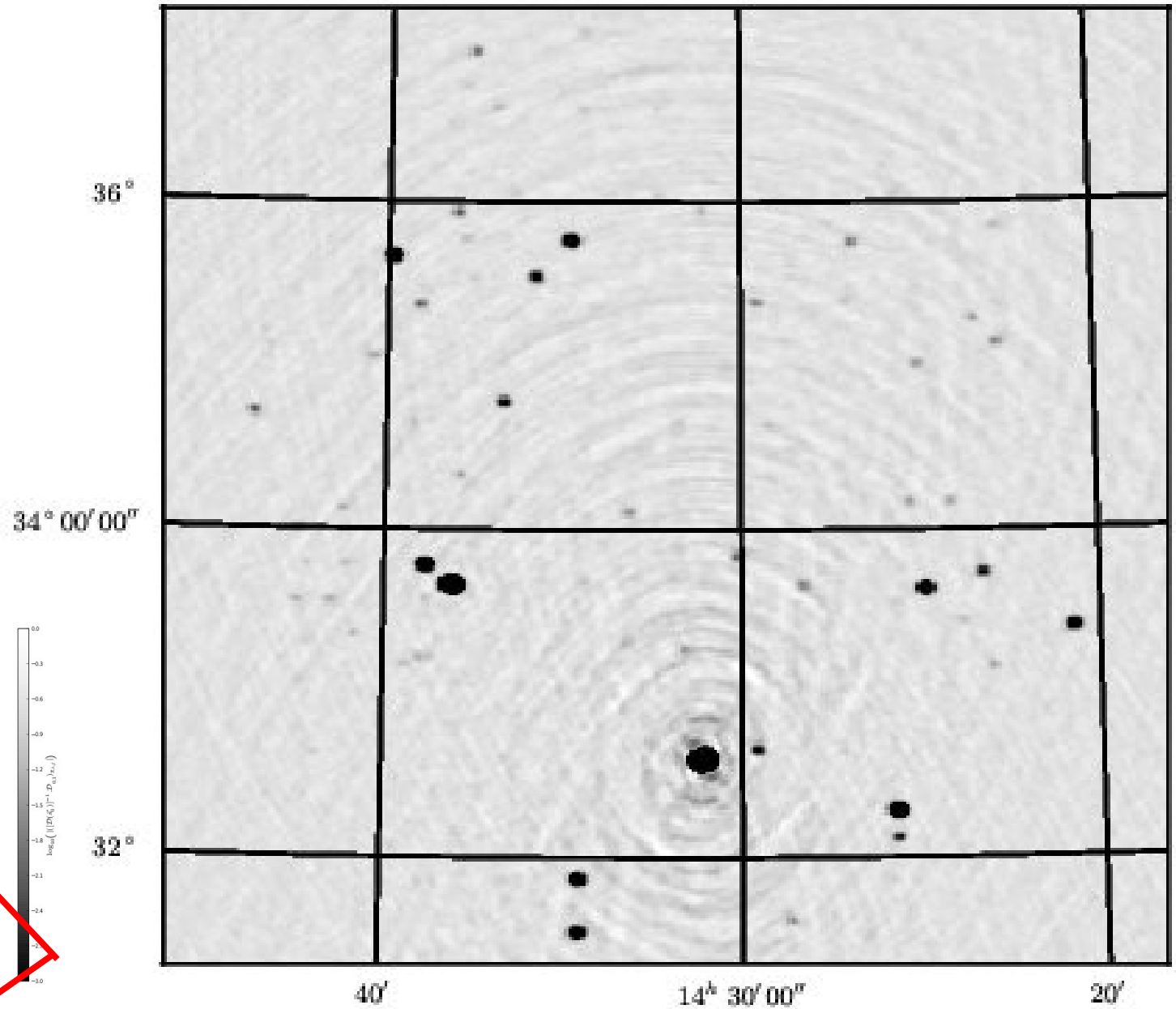
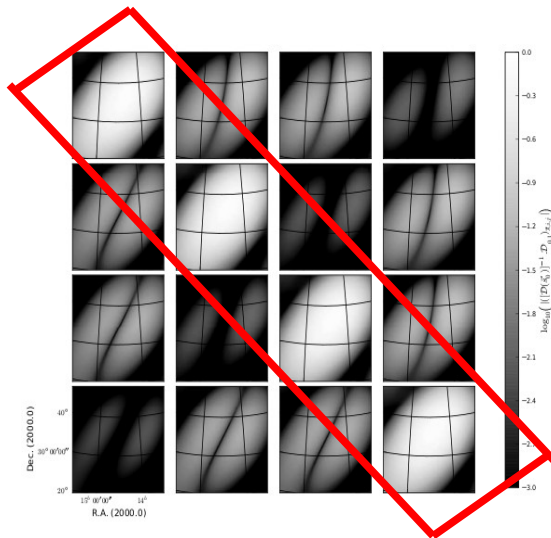
100 sources with
flux density
following NVSS
1.4 GHz source
counts



Tests on simulated data

Wterm + Scalar /diagonal beam

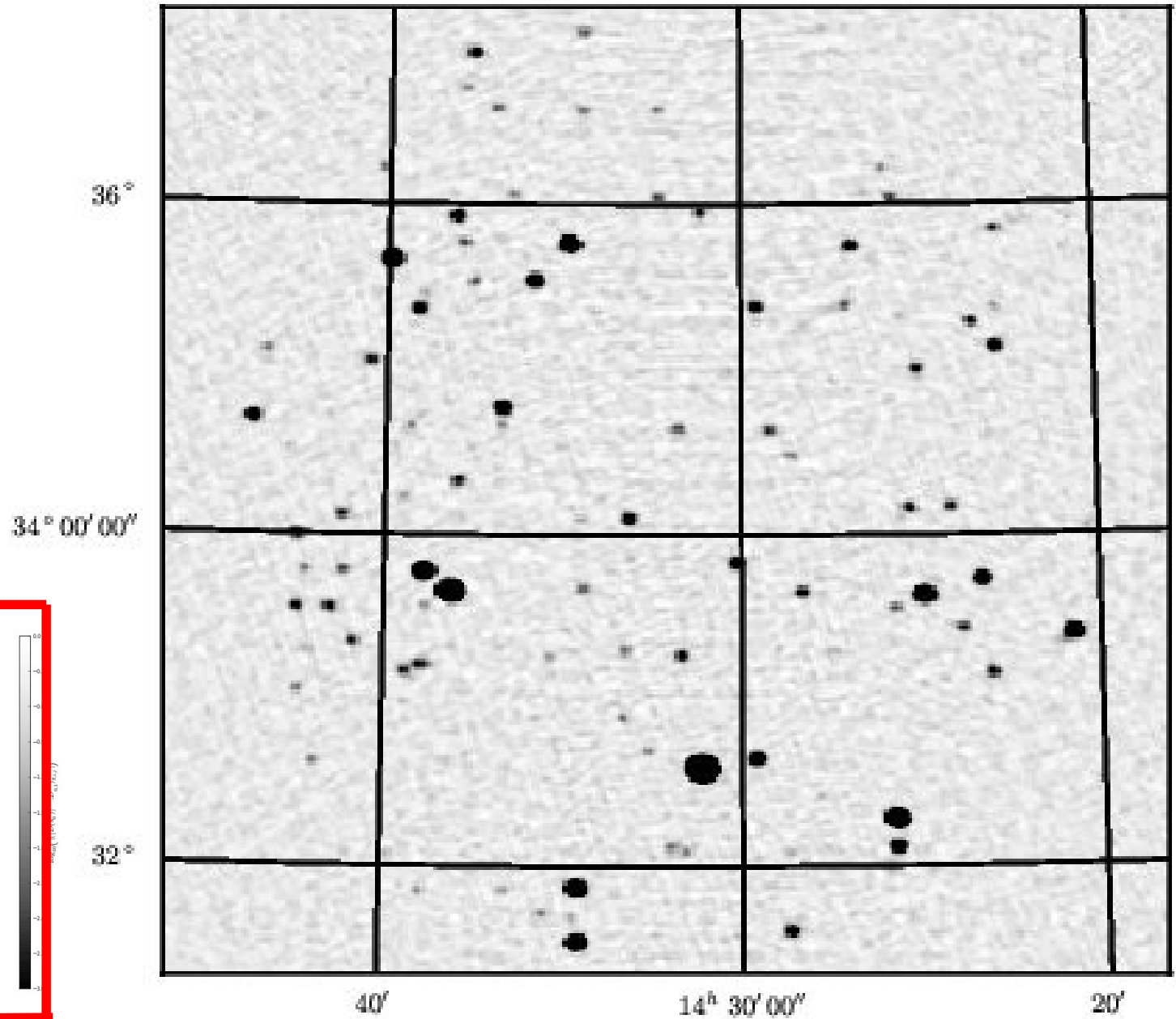
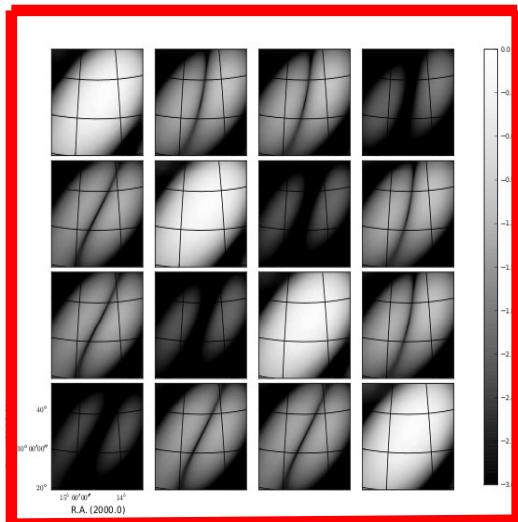
100 sources with
flux density
following NVSS
1.4 GHz source
counts



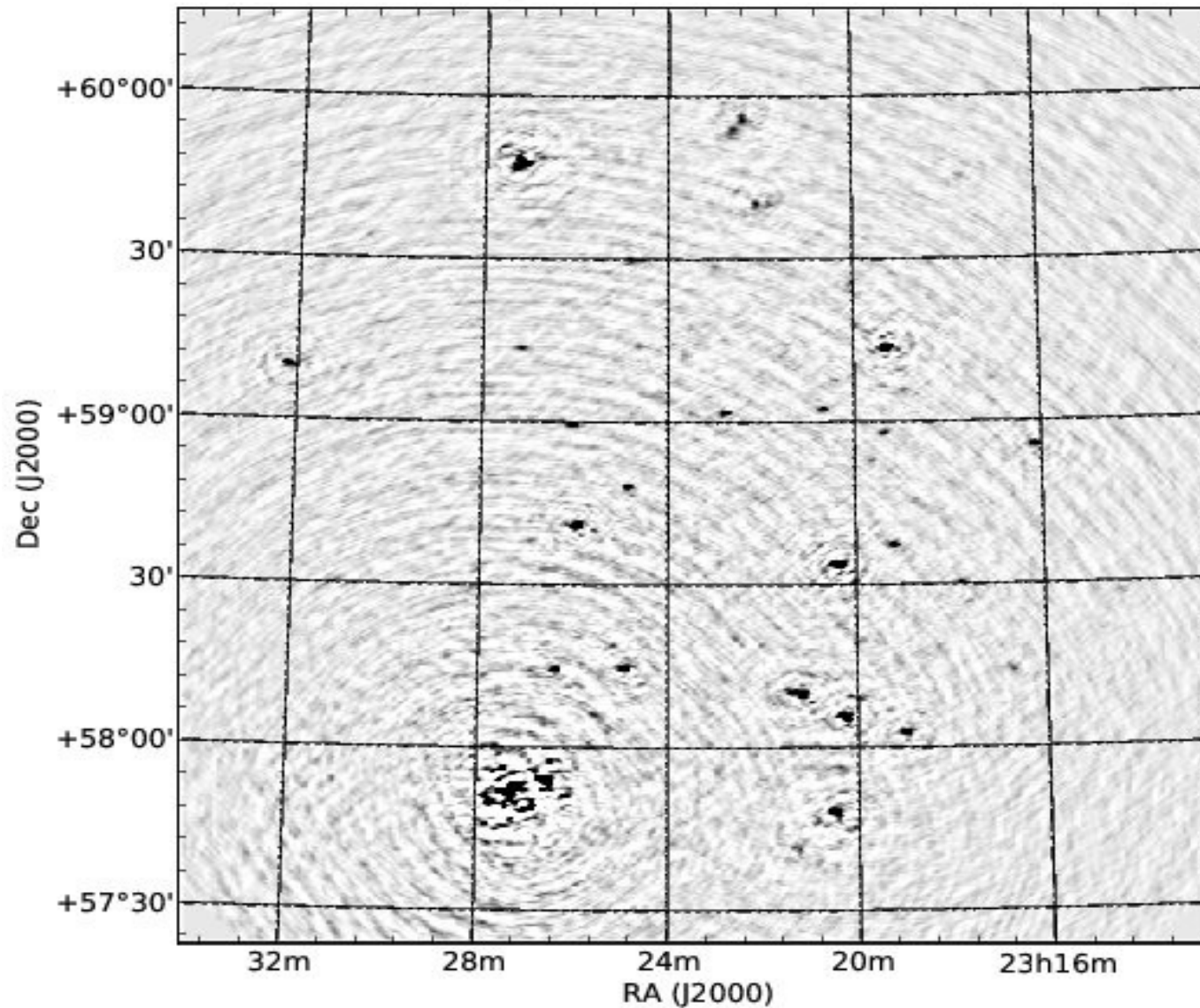
Tests on simulated data

Wterm + full beam Mueller matrix

100 sources with
flux density
following NVSS
1.4 GHz source
counts

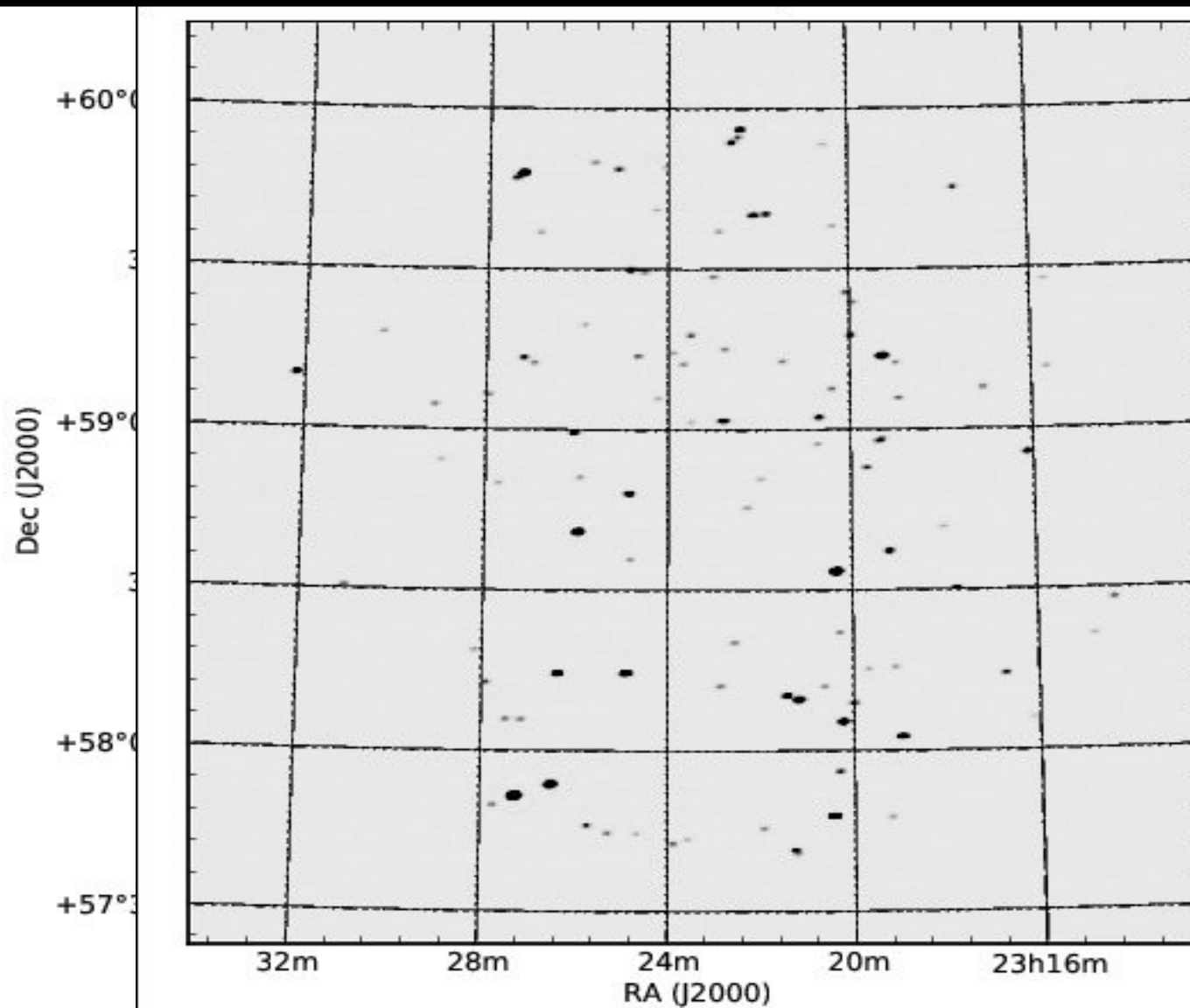


Putting ionosphere in the imager...



**Simulated dataset with many point sources
(LOFAR's beam+ionosphere)**

Putting ionosphere in the imager...



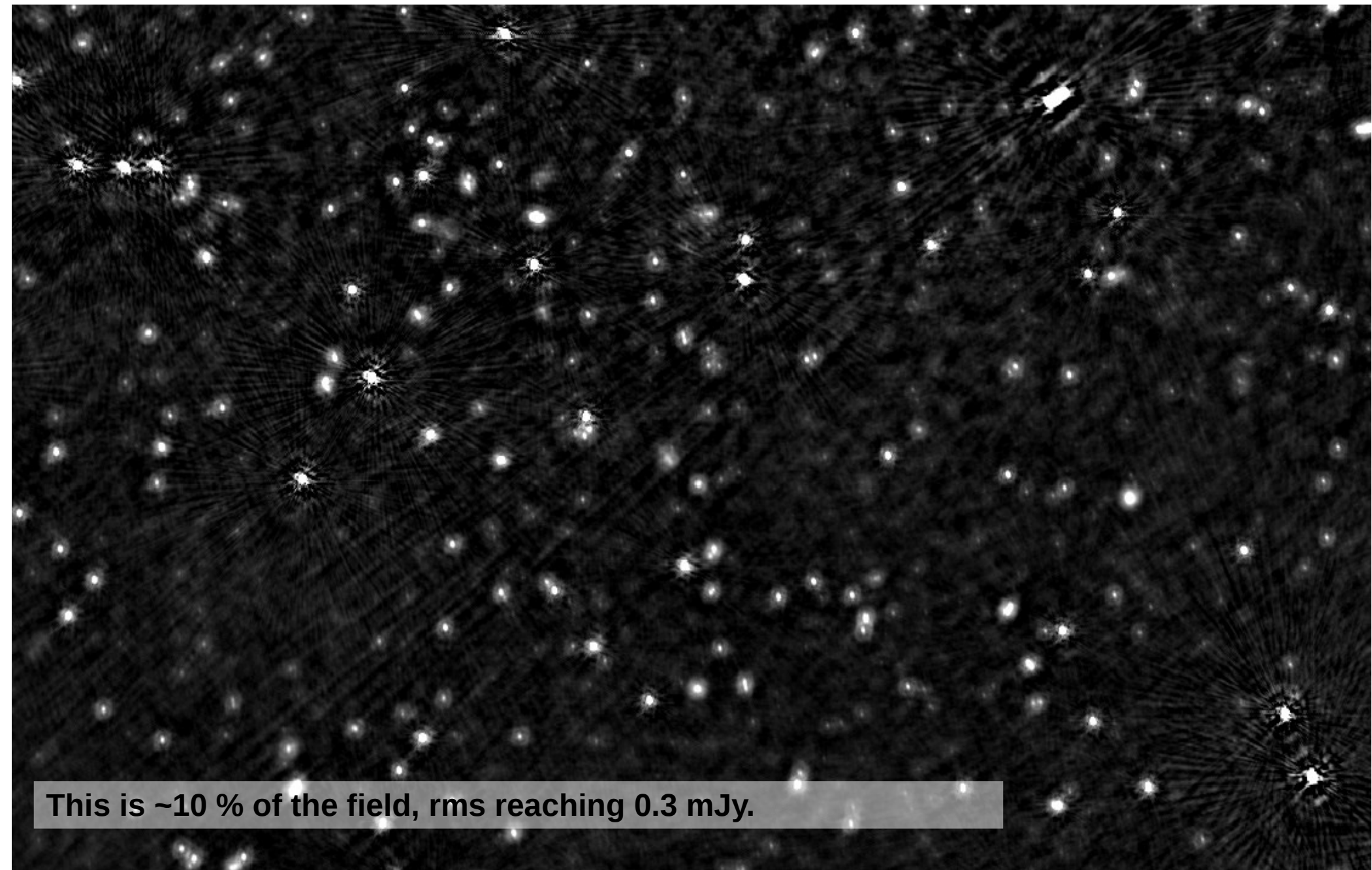
**Simulated dataset with many point sources
(LOFAR's beam+ionosphere)**

3C295 Observation (110-190 MHz)

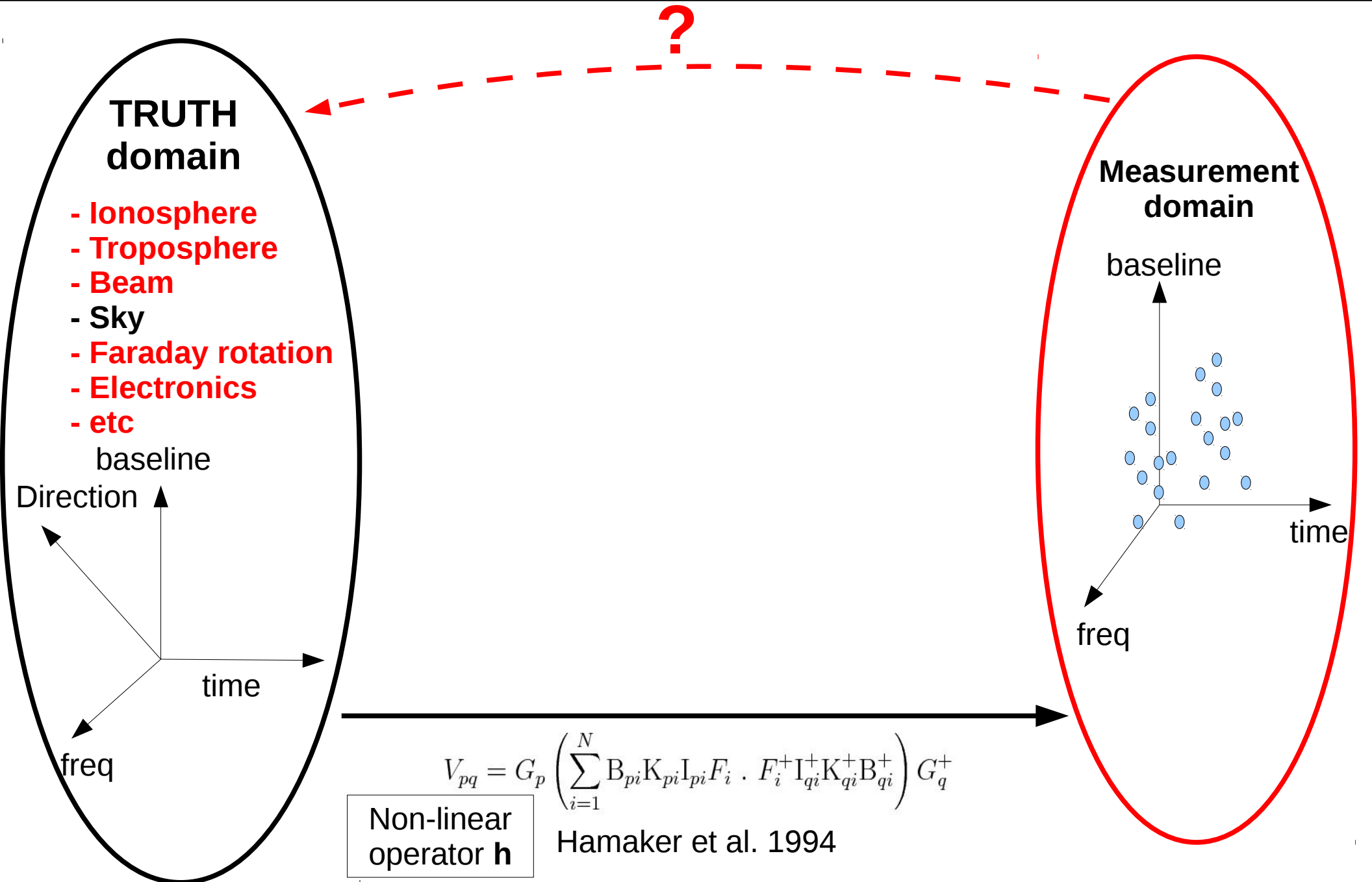


Calibration on 3C295 only (2 point model): bad calibration, but still...

3C295 Observation (110-190 MHz)



Calibration

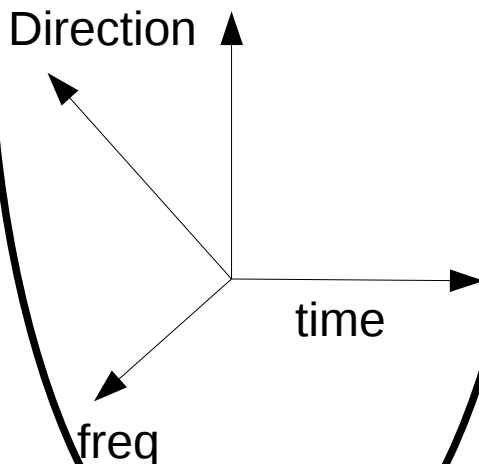


Calibration

TRUTH domain

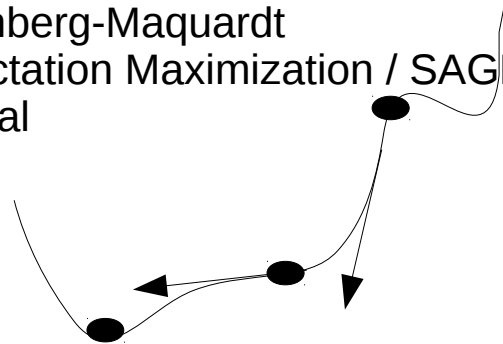
- Ionosphere
- Troposphere
- Beam
- Sky
- Faraday rotation
- Electronics
- etc

baseline



Only iterative solvers have been used so far:

- Levenberg-Maquardt
- Expectation Maximization / SAGE
- StefCal



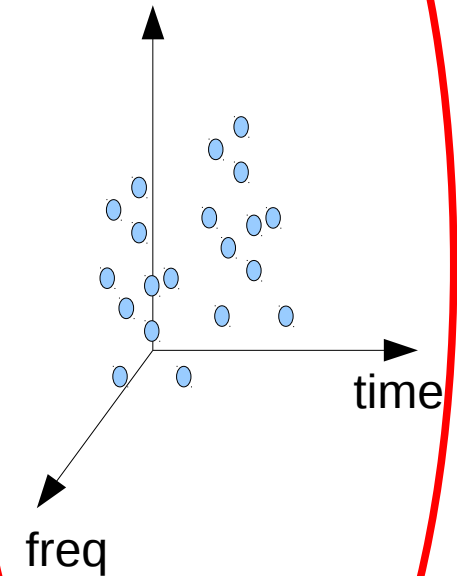
Assuming a sky, find the maximum likelihood

- each time/frequency bin independent
- each variable independent

When solving for a large number of free parameters, give rise to **ill-conditioning**

Measurement domain

baseline

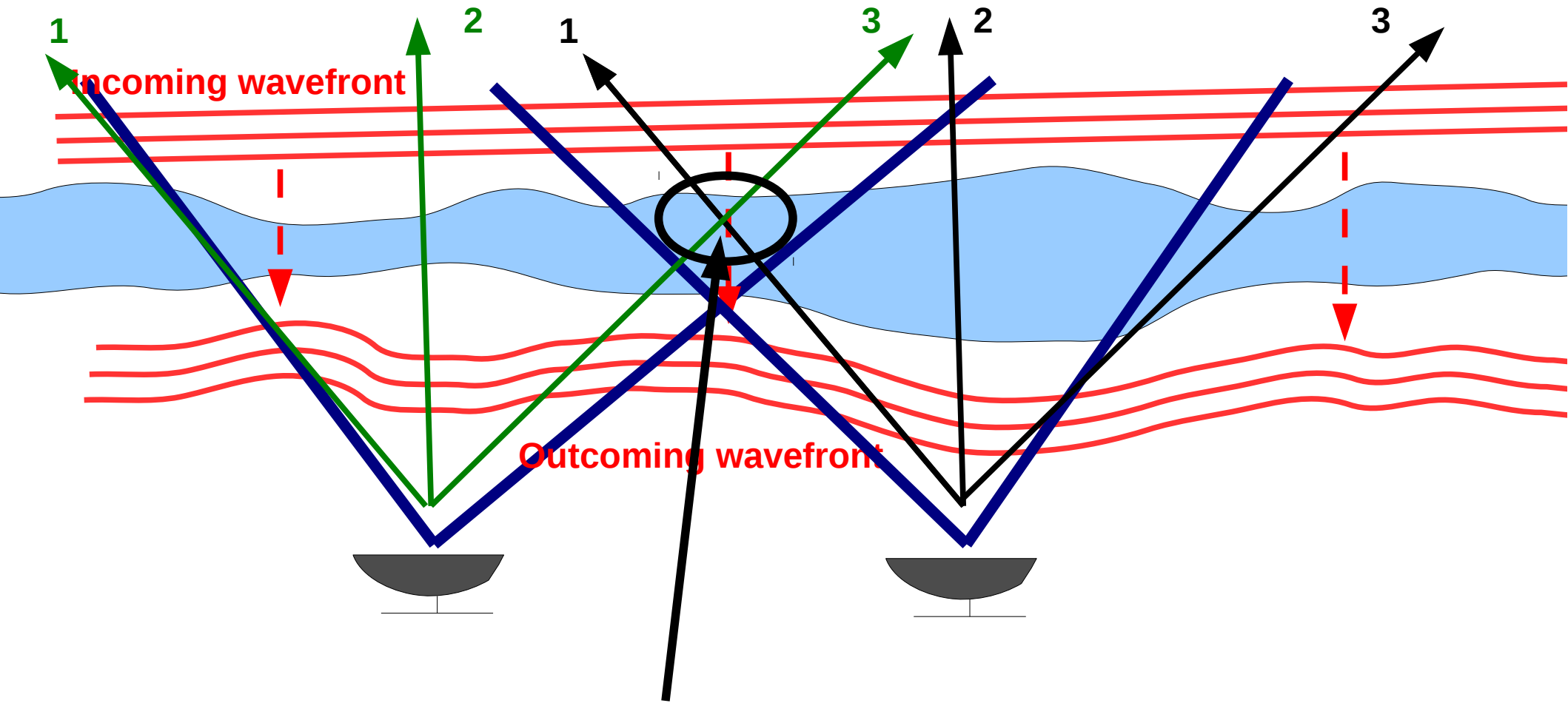


Non-linear operator \mathbf{h}

$$V_{pq} = G_p \left(\sum_{i=1}^N B_{pi} K_{pi} I_{pi} F_i \cdot F_i^+ I_{qi}^+ K_{qi}^+ B_{qi}^+ \right) G_q^+$$

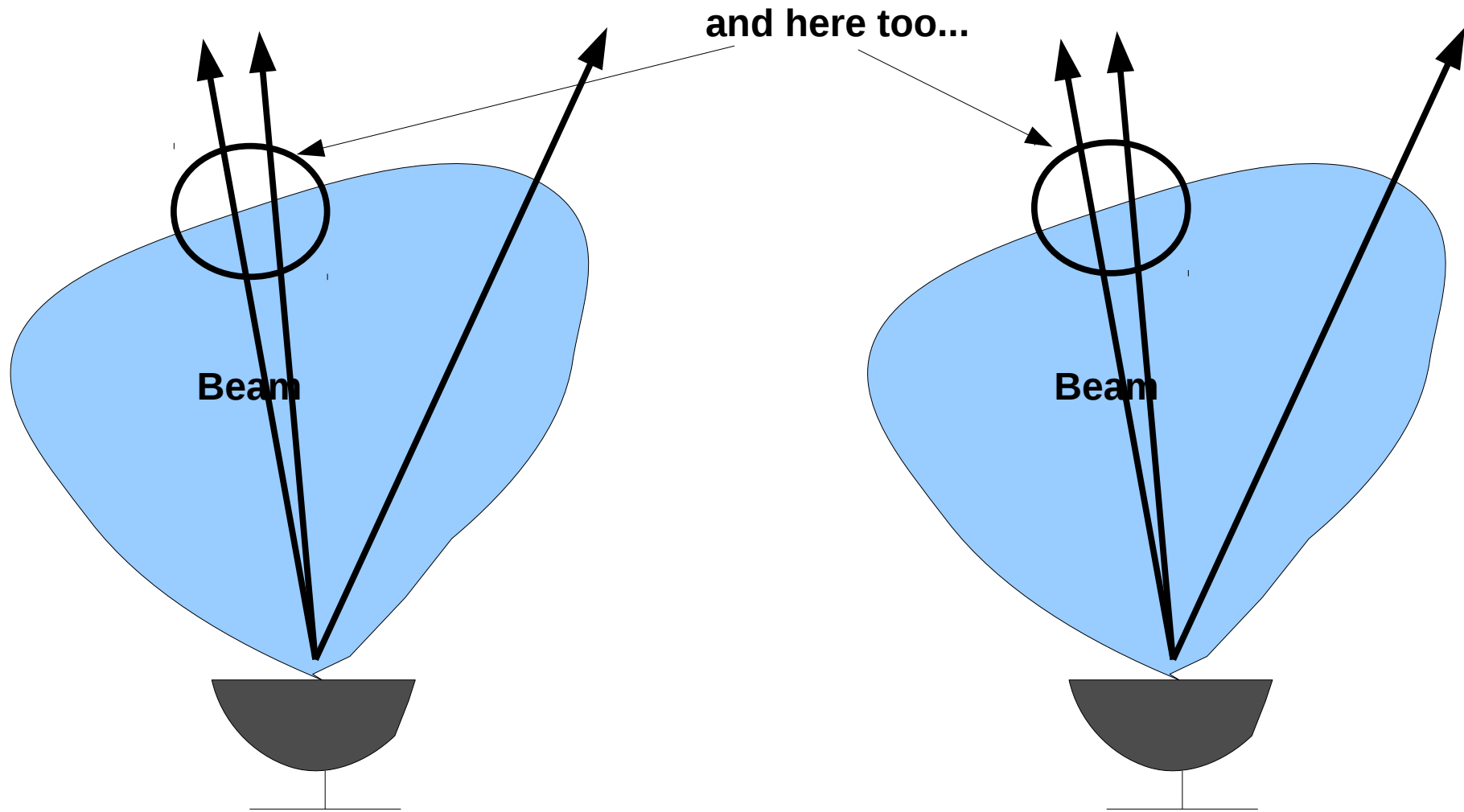
Hamaker et al. 1994

While... there is regularity in the process

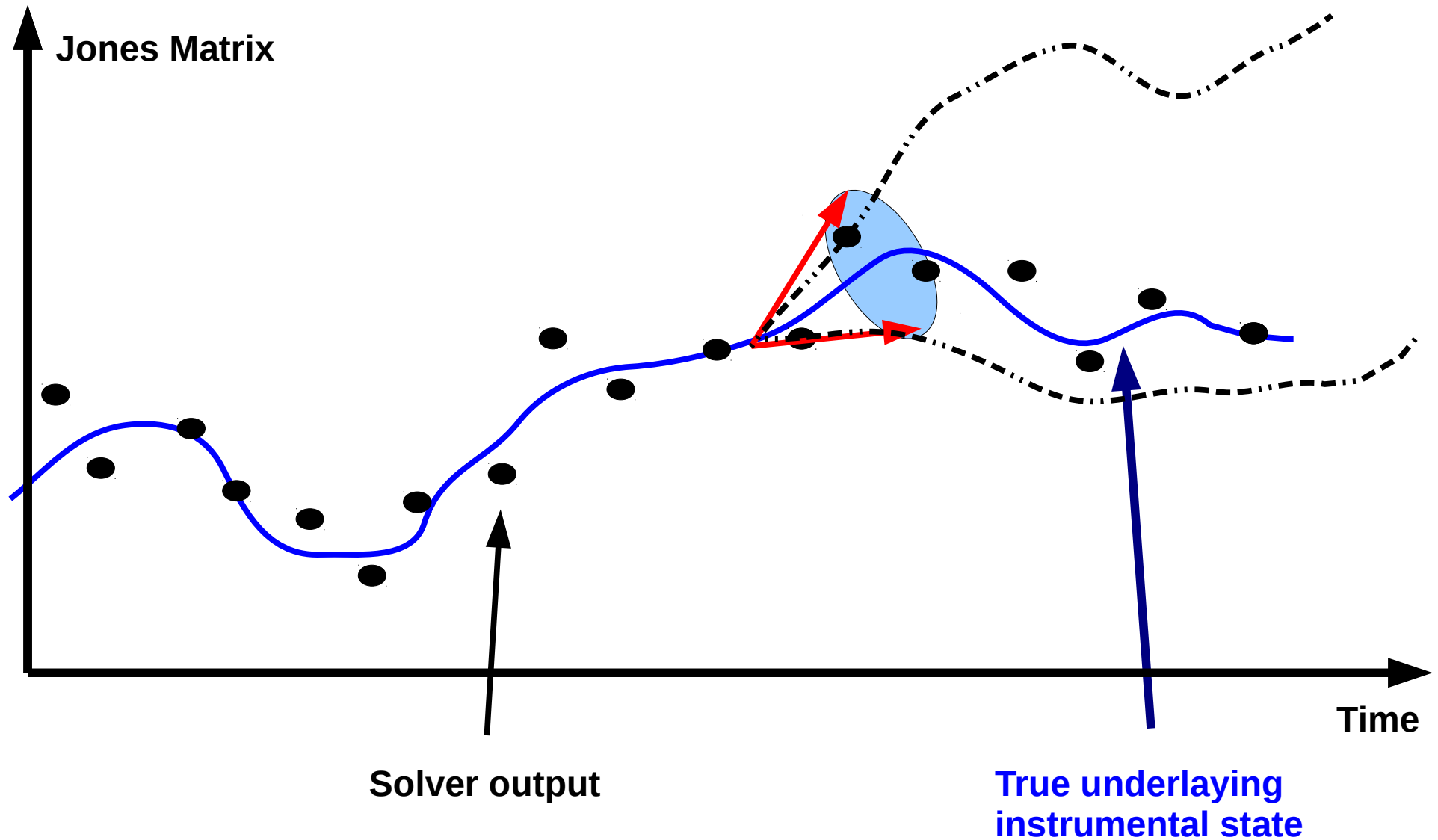


Process parameters are correlated here

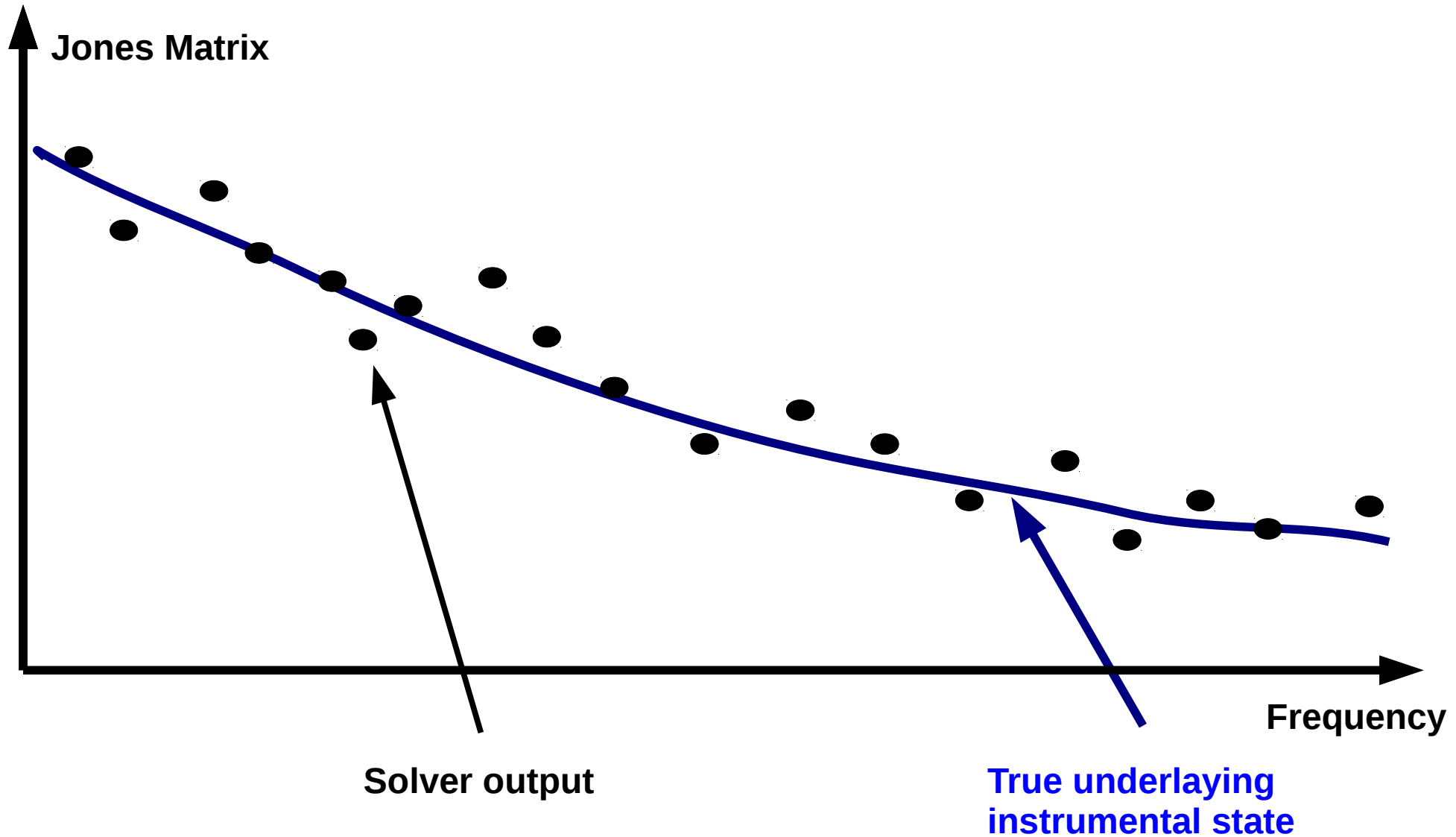
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While... there is regularity in the process



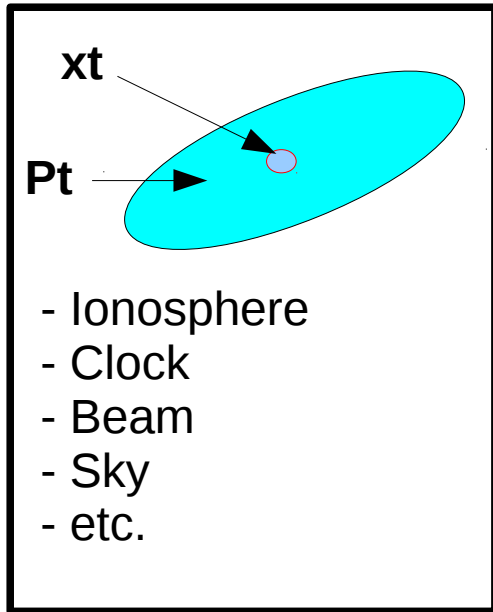
While... there is regularity in the process



Non-linear Kalman Filters....

Process domain:

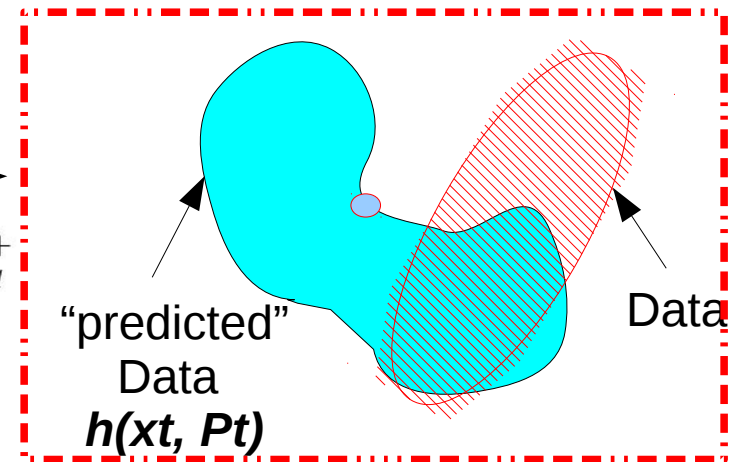
$Dim=10^2-10^4$



$$V_{pq} = G_p \left(\sum_{i=1}^N B_{pi} K_{pi} I_{pi} F_i \cdot F_i^+ I_{qi}^+ K_{qi}^+ B_{qi}^+ \right) G_q^+$$

Data domain:

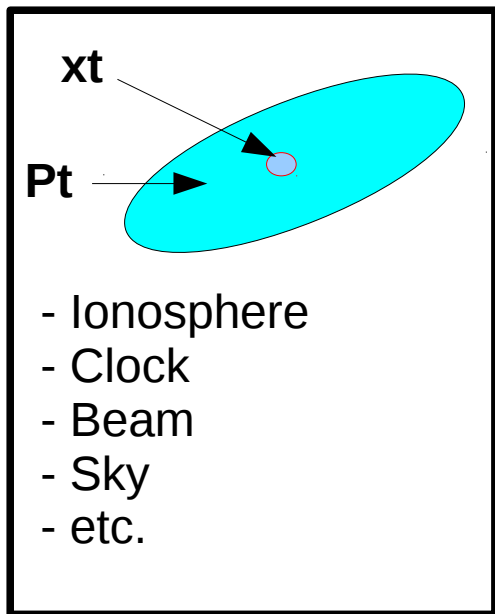
$Dim=10^4-10^6$



Non-linear Kalman Filters....

Process domain:

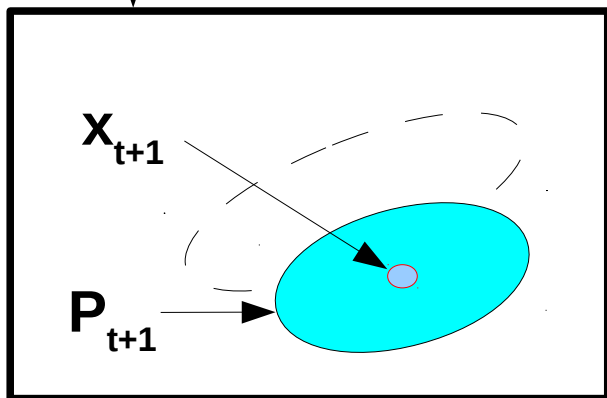
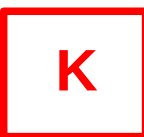
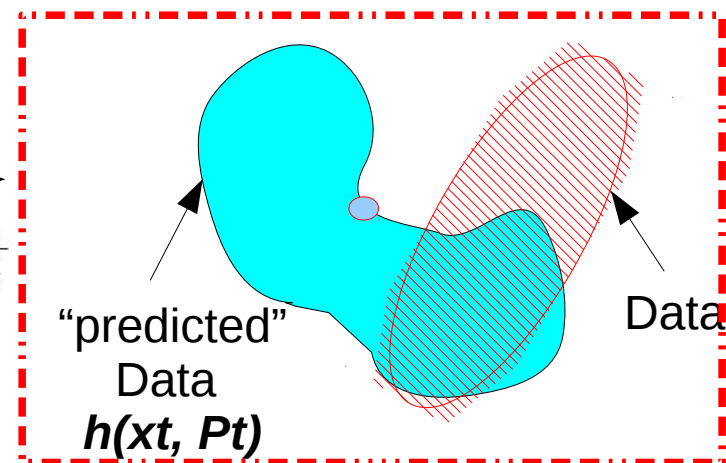
$Dim=10^2-10^4$



$$V_{pq} = G_p \left(\sum_{i=1}^N B_{pi} K_{pi} I_{pi} F_i \cdot F_i^+ I_{qi}^+ K_{qi}^+ B_{qi}^+ \right) G_q^+$$

Data domain:

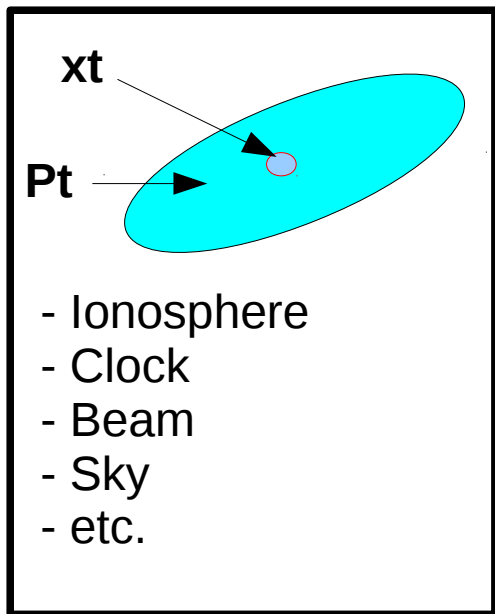
$Dim=10^4-10^6$



Non-linear Kalman Filters....

Process domain:

$Dim=10^2-10^4$

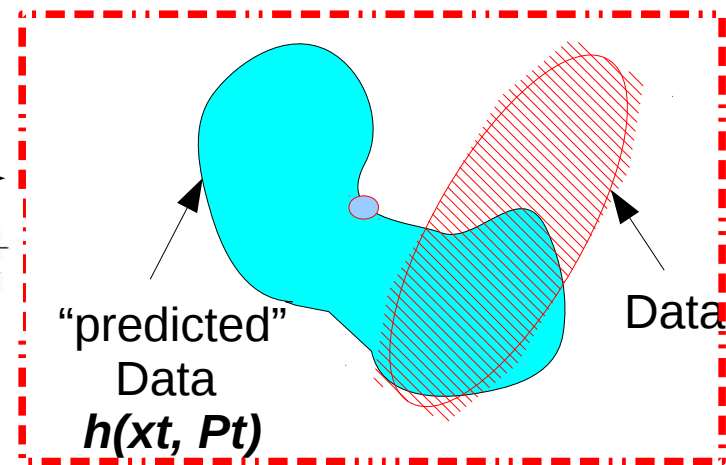


$$V_{pq} = G_p \left(\sum_{i=1}^N B_{pi} K_{pi} I_{pi} F_i \cdot F_i^+ I_{qi}^+ K_{qi}^+ B_{qi}^+ \right) G_q^+$$

h

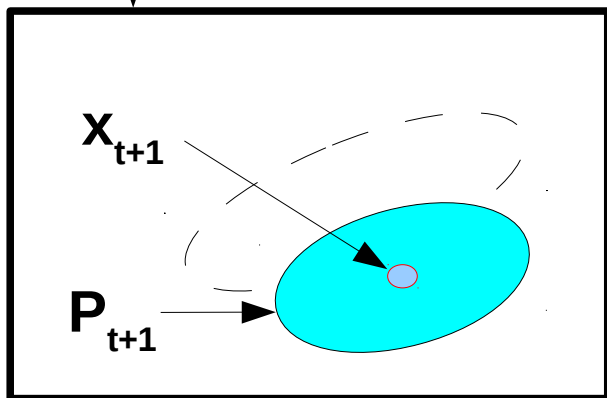
Data domain:

$Dim=10^4-10^6$

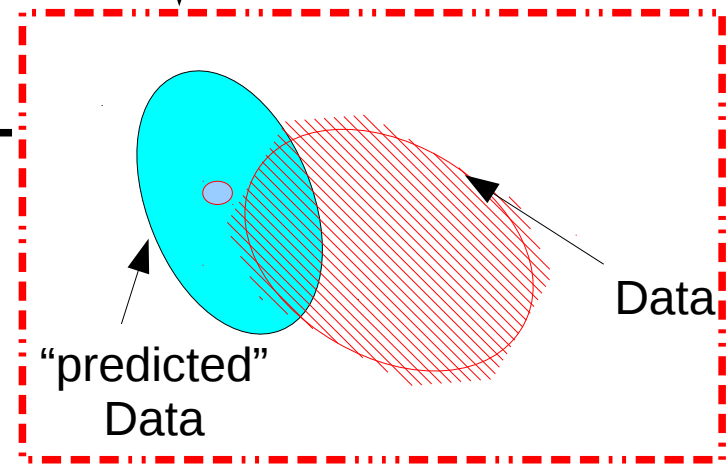


R

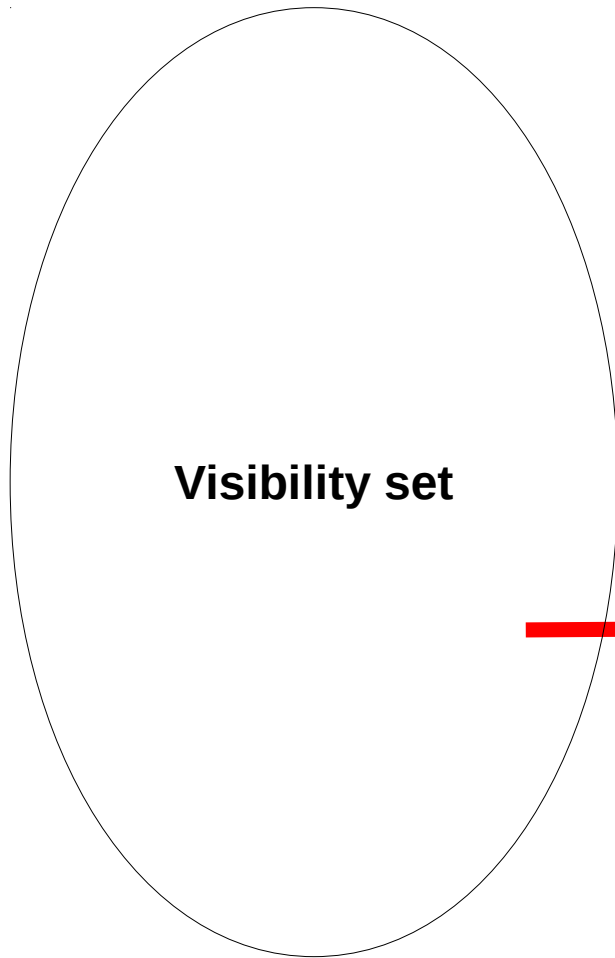
Data projected to a different representation



K

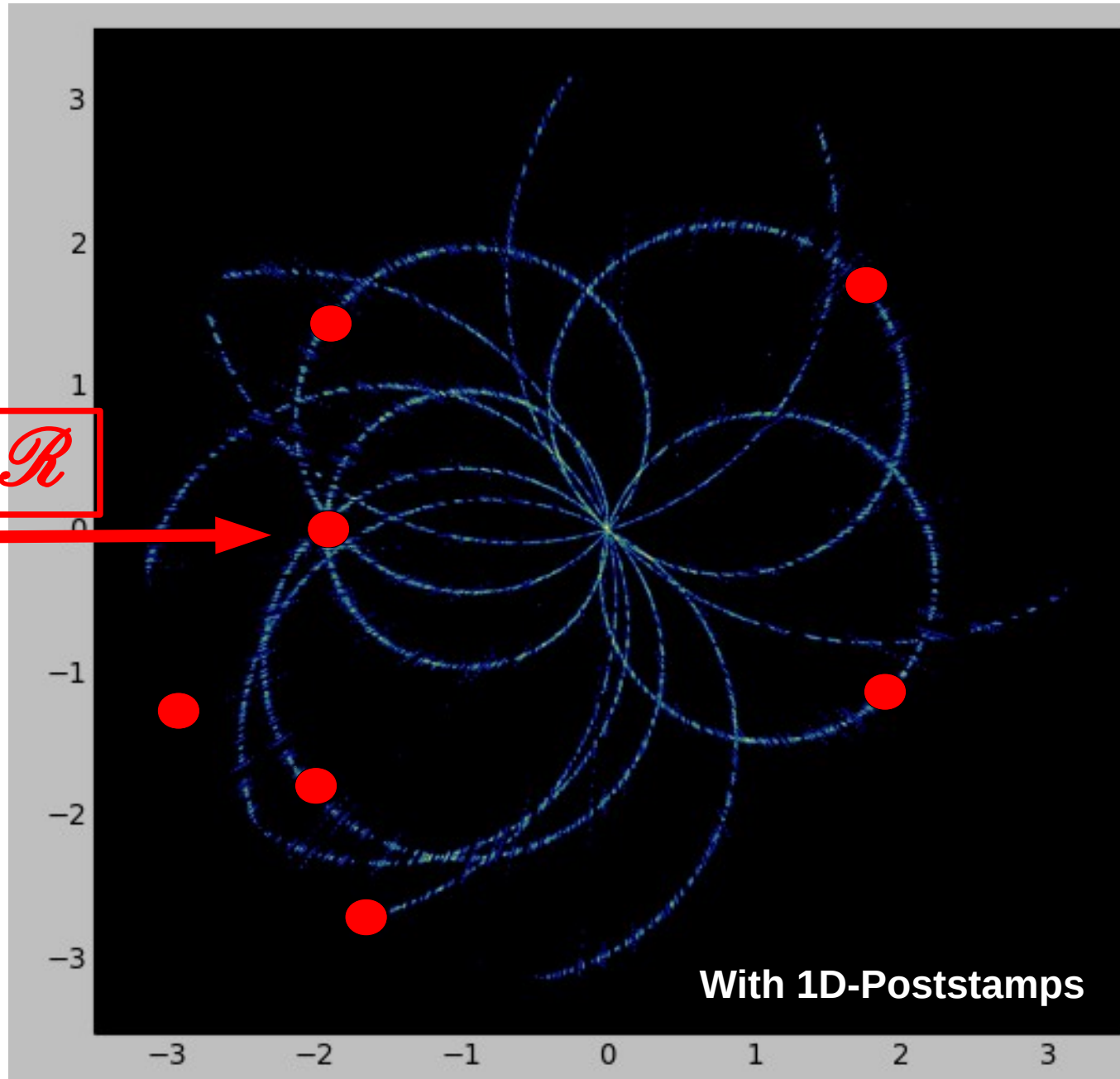


Representation issues....



Equivalent to
1-Dimensional raw-data
images

\mathcal{R}



Simulated dataset (ionosphere TEC-screen + Clock drift)

Measurement
Equation

$$\mathbf{V}_{(pq)tv} = \mathbf{h}(\mathbf{x}) = \mathbf{G}_{ptv}(\mathbf{x}) \left(\sum_s \mathbf{V}_{(pq)tv}^s(\mathbf{x}) k_{(pq)tv}^s \right) \mathbf{G}_{qtv}^H(\mathbf{x}) \quad (1)$$

$$\mathbf{V}_{(pq)tv}^s(\mathbf{x}) = \mathbf{D}_{pstv}(\mathbf{x}) \mathbf{X}_s \mathbf{D}_{qstv}^H(\mathbf{x}) \quad (2)$$

Clocks
Ionosphere

$$\mathbf{G}_{ptv}(\mathbf{x}) := \exp(2\pi i \nu \delta \mathbf{t}_p(\mathbf{x})) \mathbf{I}$$
$$\mathbf{D}_{ptv}^d(\mathbf{x}) := \exp\left(ik\nu^{-1} \mathbf{T}_p^d(\mathbf{x})\right) \mathbf{I}$$

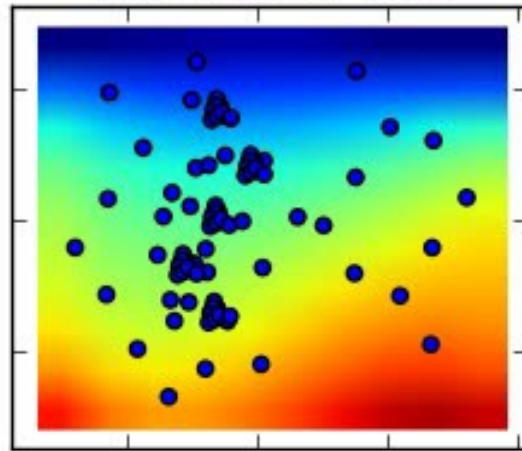
- LOFAR HBA (1500 baselines, 4 pols)
- 30 subbands from 100 to 150 MHz
- S/N=5

- variable ionosphere TEC-screen
- Variable clock drift

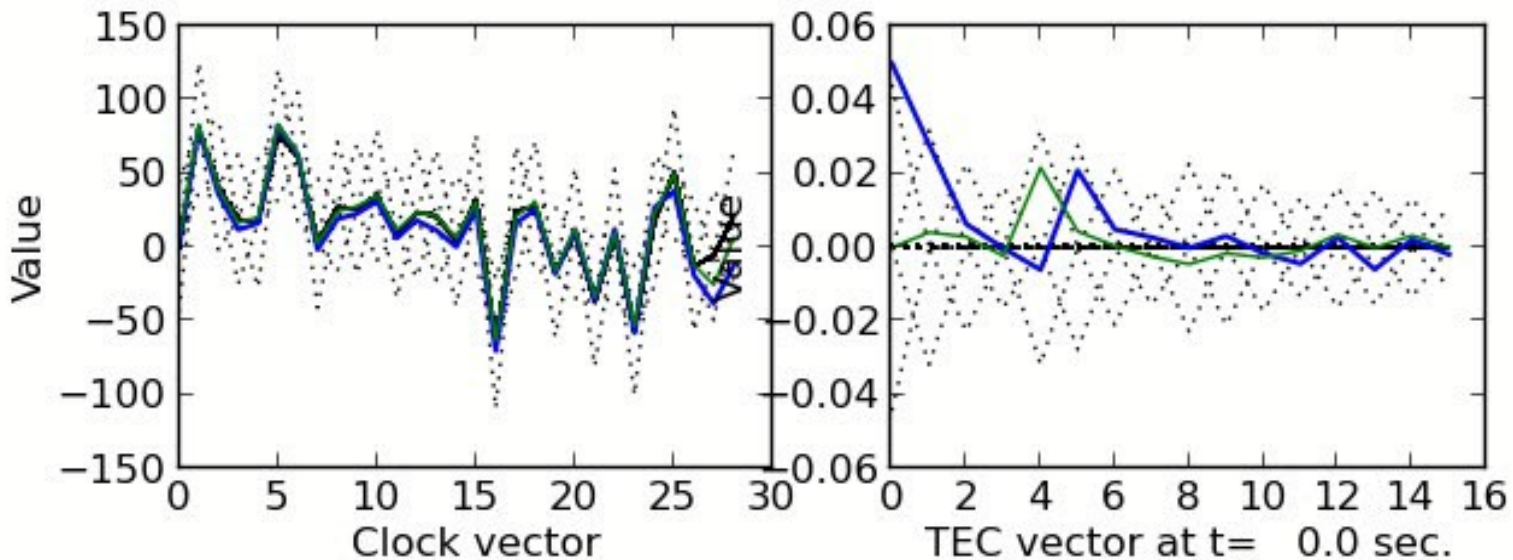
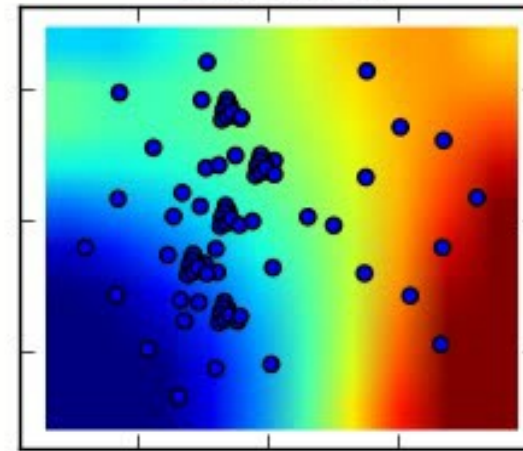
- 18.000 data points / [time bin (30s)]

Simulated dataset (ionosphere TEC-screen + Clock drift)

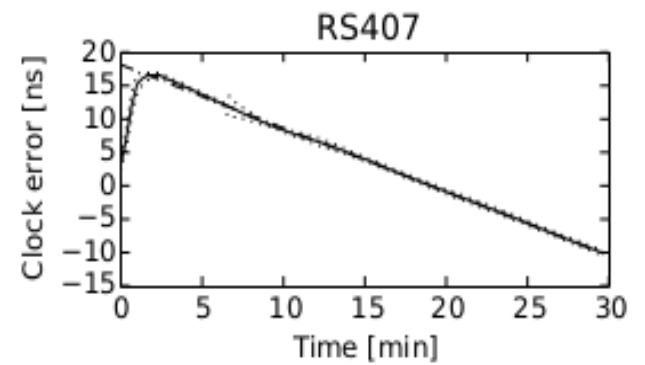
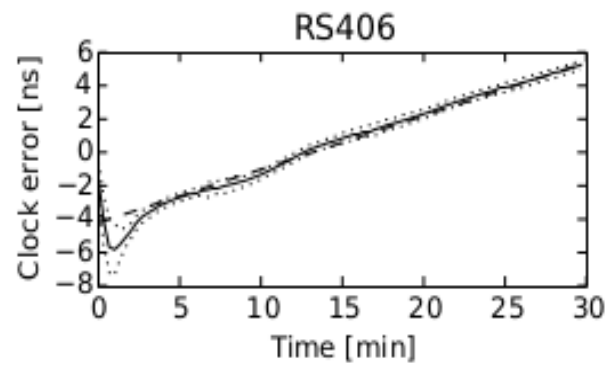
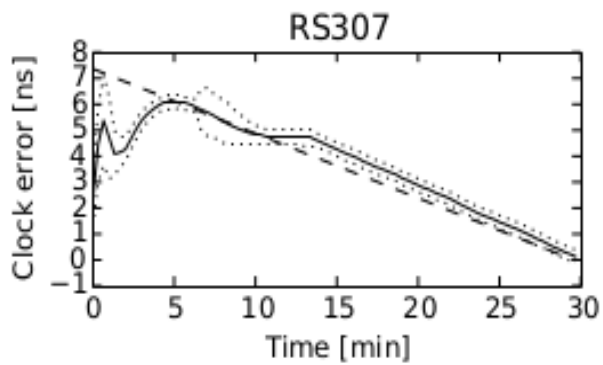
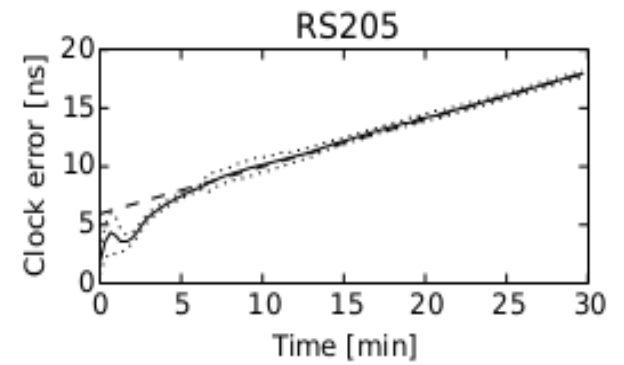
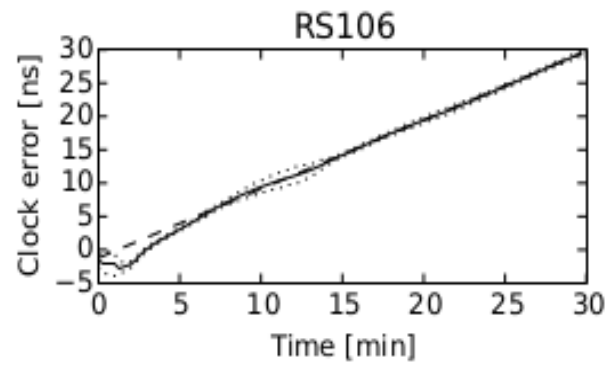
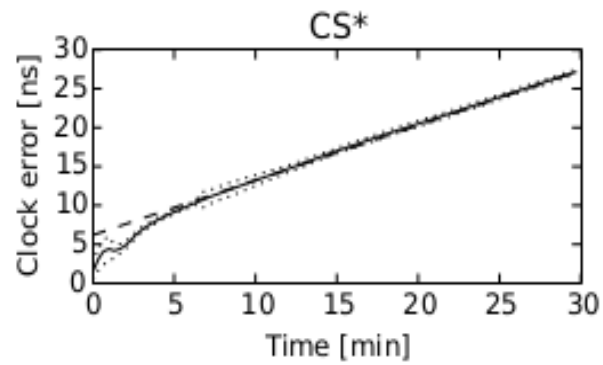
Estimated TEC



True TEC



Clocks solutions

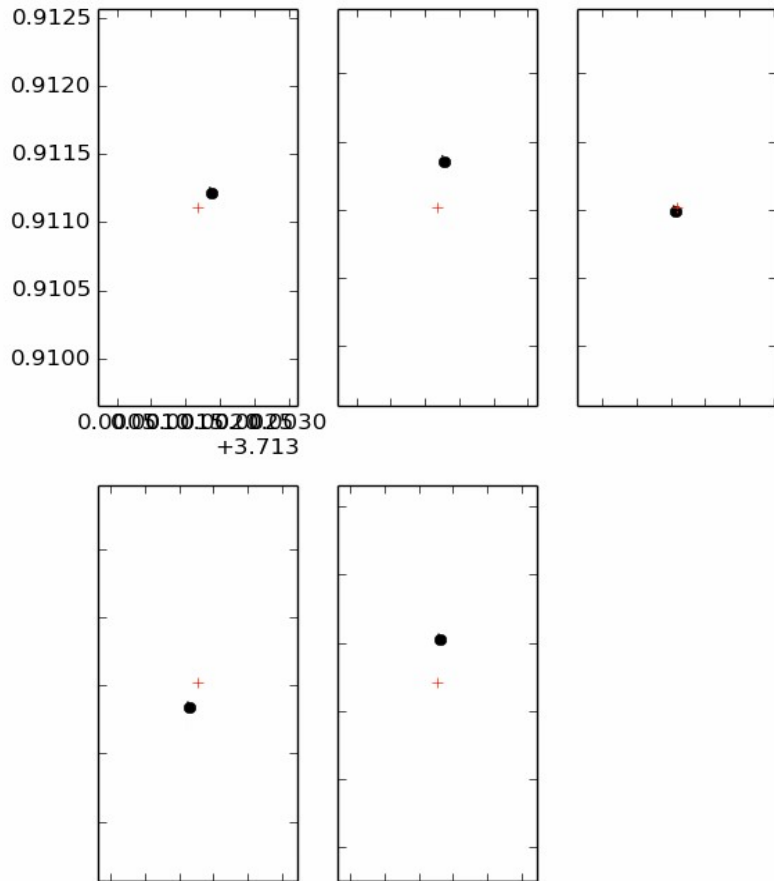


Solving for the sky term ?!

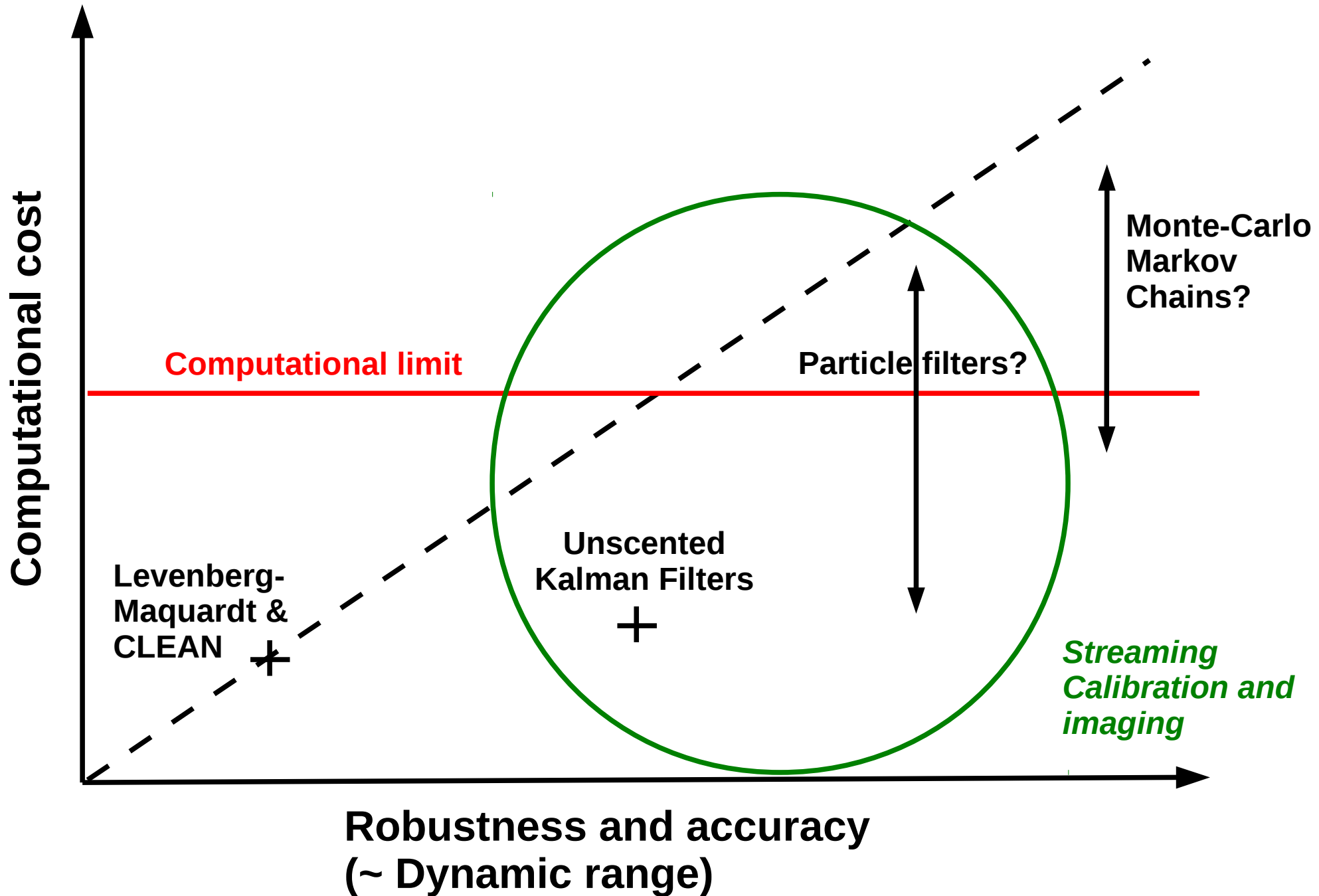
- Robustness seems high enough now....
- Let's simulate a grid of sources, and solve for their position

Solving for the sky term ?!

- Robustness seems high enough now....
- Let's simulate a grid of sources, and solve for their position



Conclusion - I



Conclusion - II

Towards a (streaming) adaptative optics (wide field, full polarization)

- Radio interferometry is **simple to formalize** and **hard to solve**
- A LOT of things happening in the domain:
 - Imaging and deconvolution
 - New calibration techniques
- **ill conditioning** is a real issue

Filters:

- **We calibrate the raw data in the image plane**
- We **decrease** number of parameters **by orders of magnitude** (seem to properly address ill-conditioning)
- The **solutions are physical & valid everywhere** (not only at discrete locations)
- This robustness allows to add additional degrees of freedom such as sky itself !
- Fairly fast given what it does
- Works only if instrument and sky are analytically describable (need for a “analytically stable” instrument)
- Need to access all the frequency data simultaneously (that's why it takes time to get to test the algorithm on real data)