



Non-linear Kalman filters for calibration in radio interferometry

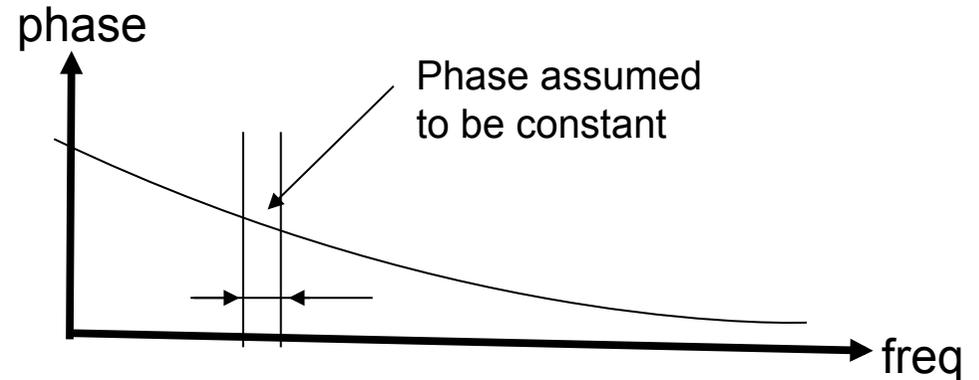
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Tasse 2014 [2014A&A...566A.127T]

Main issue: available information, ill conditioning

Traditional Calibration algorithms:

- Fit for Jones matrices
 - But Jones matrices vary over time/freq:
 - Limit usable bandwidth
 - Limit usable time chunk
 - Need for ~10 to ~100 discrete directions
 - /antenna /polarisation /datachunk
 - tens of thousands of *free* parameters



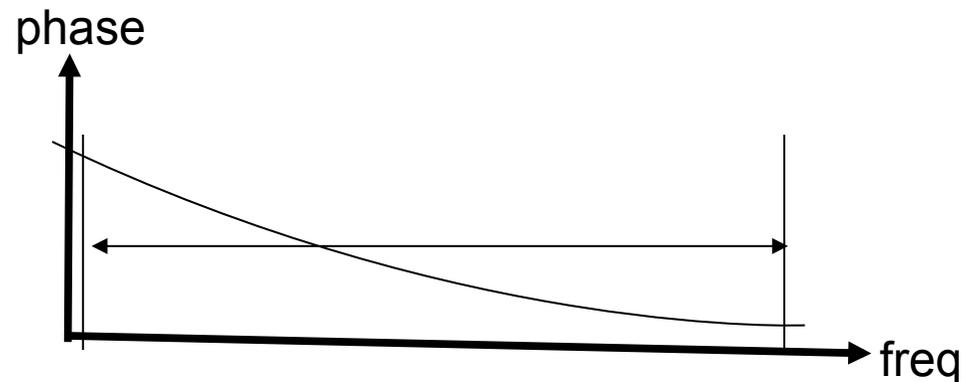
**Ill conditioning is dangerous
(source suppression, fake high dynamic range, etc)**

Our goal: addressing ill-conditioning

Two independent aspects:

- **Physics-based** approach (opposed to **Jones-based** approach)
 - constraining Clocks, TEC-screen – **directly from visibilities**
 - All physical effects have a smooth frequency behaviour
 - We are **not limited in bandpass** anymore!

- **Much less** degrees of freedom
(~100-1000x)
- **Much more** usable data
(~10-100x)

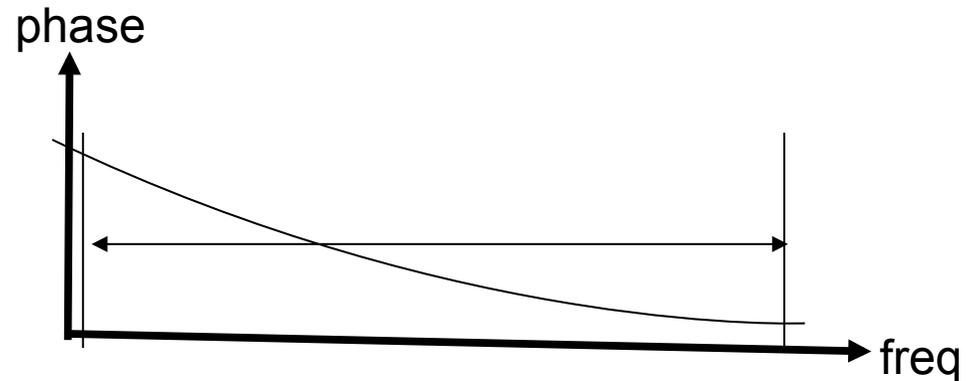


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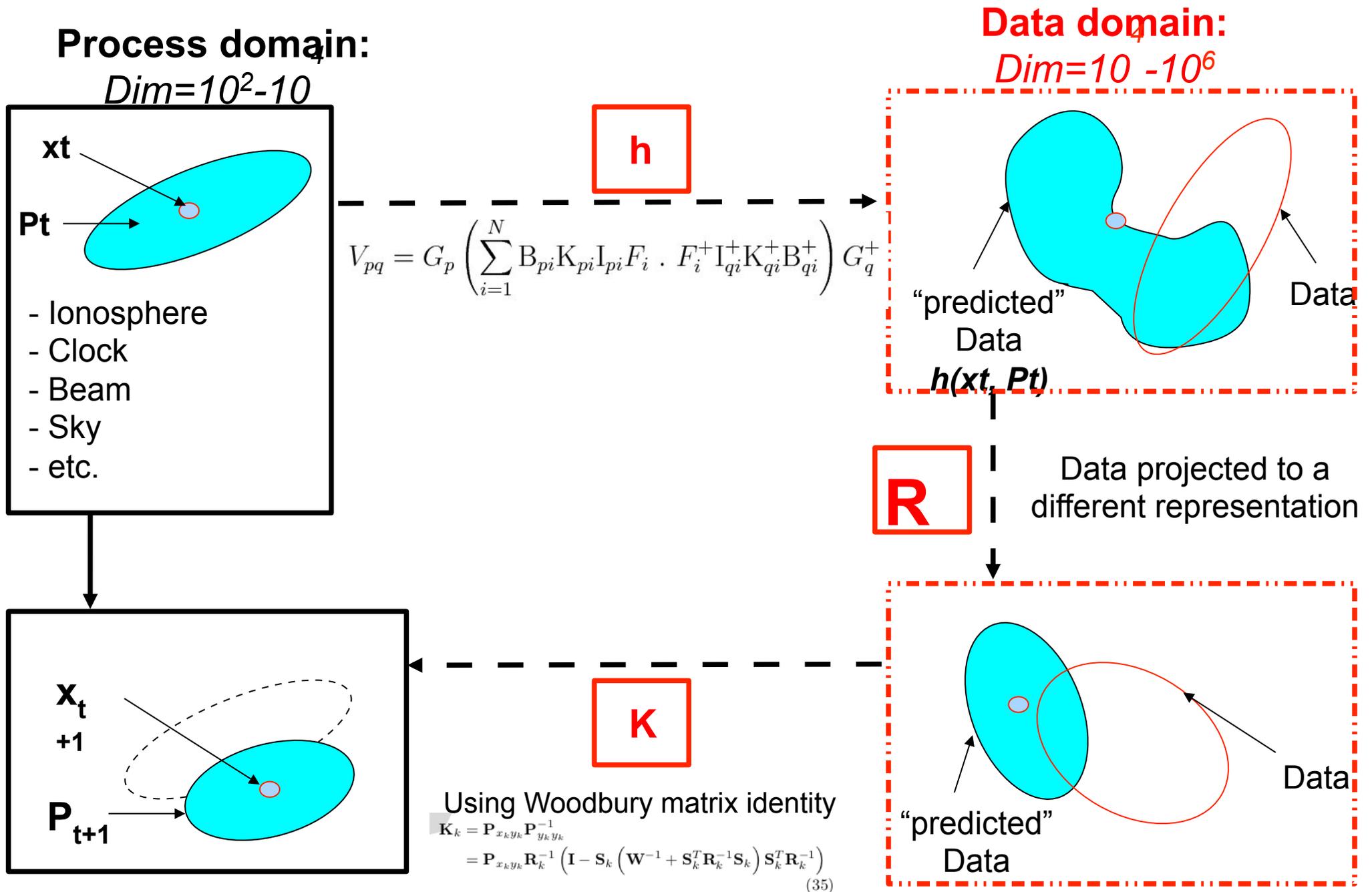
- Physical effects are **not stable** in time though....
 - Kalman filter!

Kalman Filters ? (iterative versus recursive)

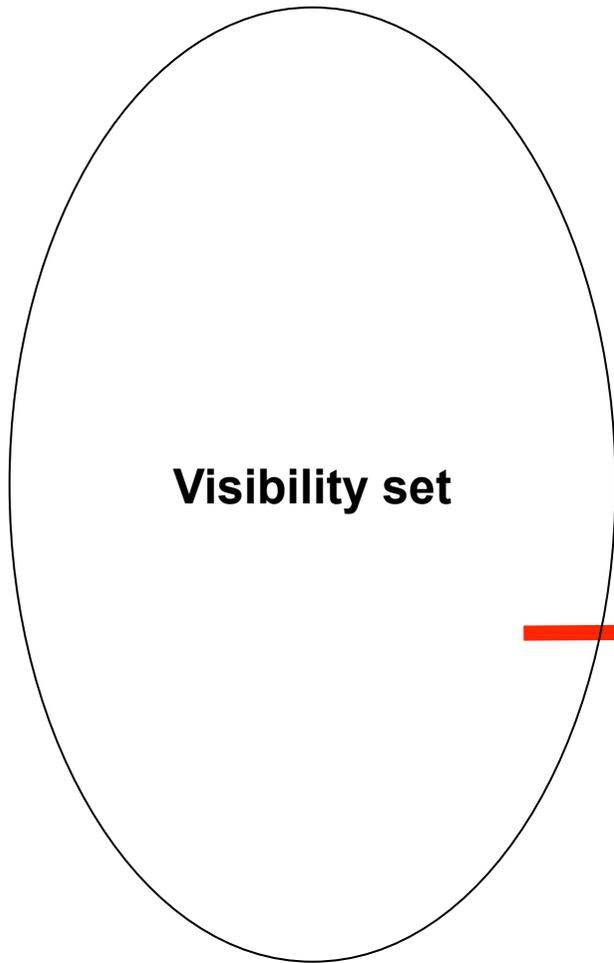
- Different from *Levenberg-Maquardt* or *EM*:
 - Kalman filters **do not try** to fit the data at best
 - “Minimum mean square estimators”: fit the ***data given information on the expected state***
 - They “**track** more than they solve”
 - Kalman filters use a **recursive** sequence (as opposed to ***iterative*** for LM, EM etc)

Problem: the **process to measurement** equation is non-linear

Non-linear Kalman Filters....

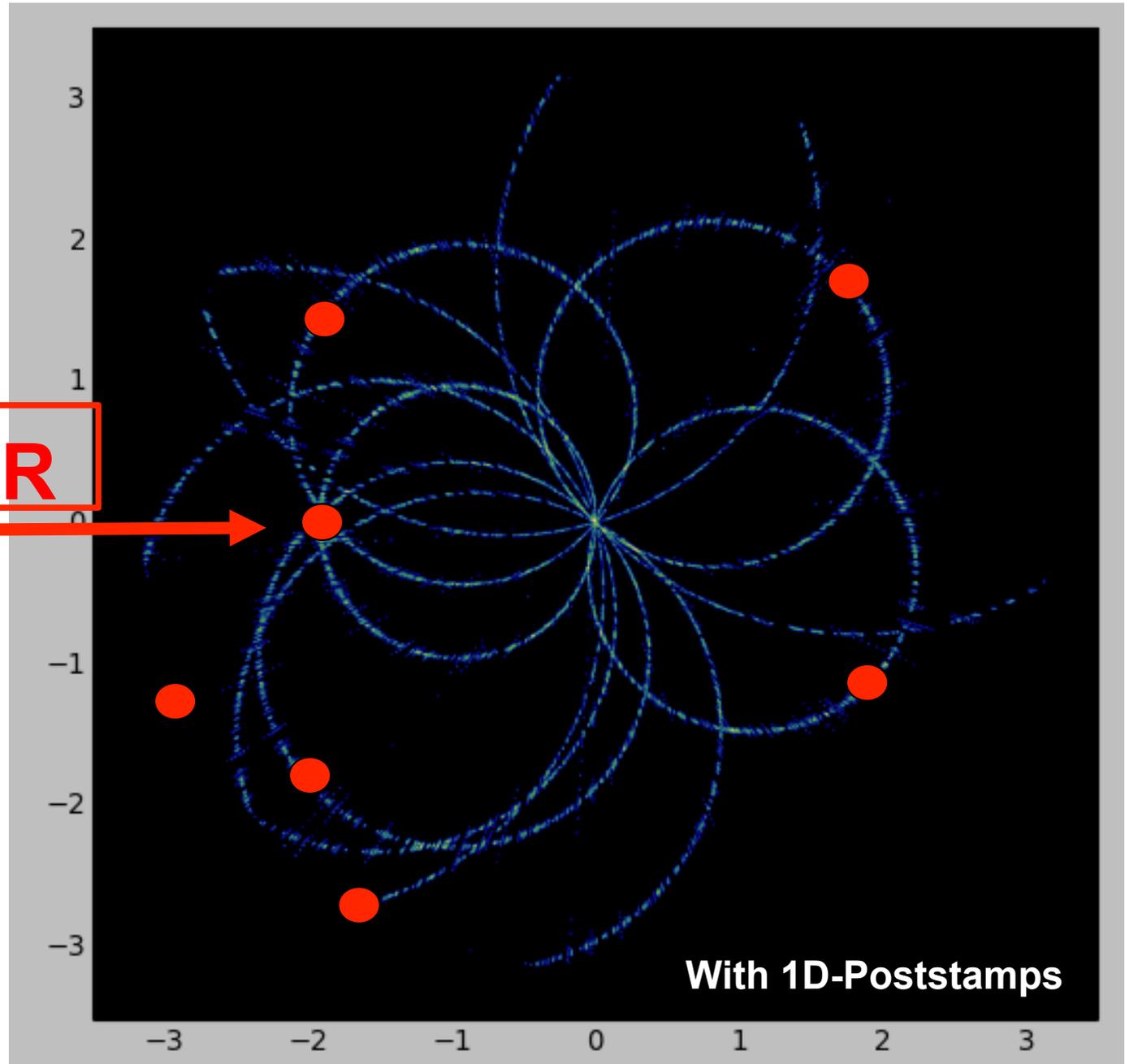
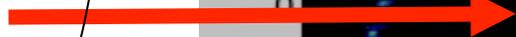


Representation issues....

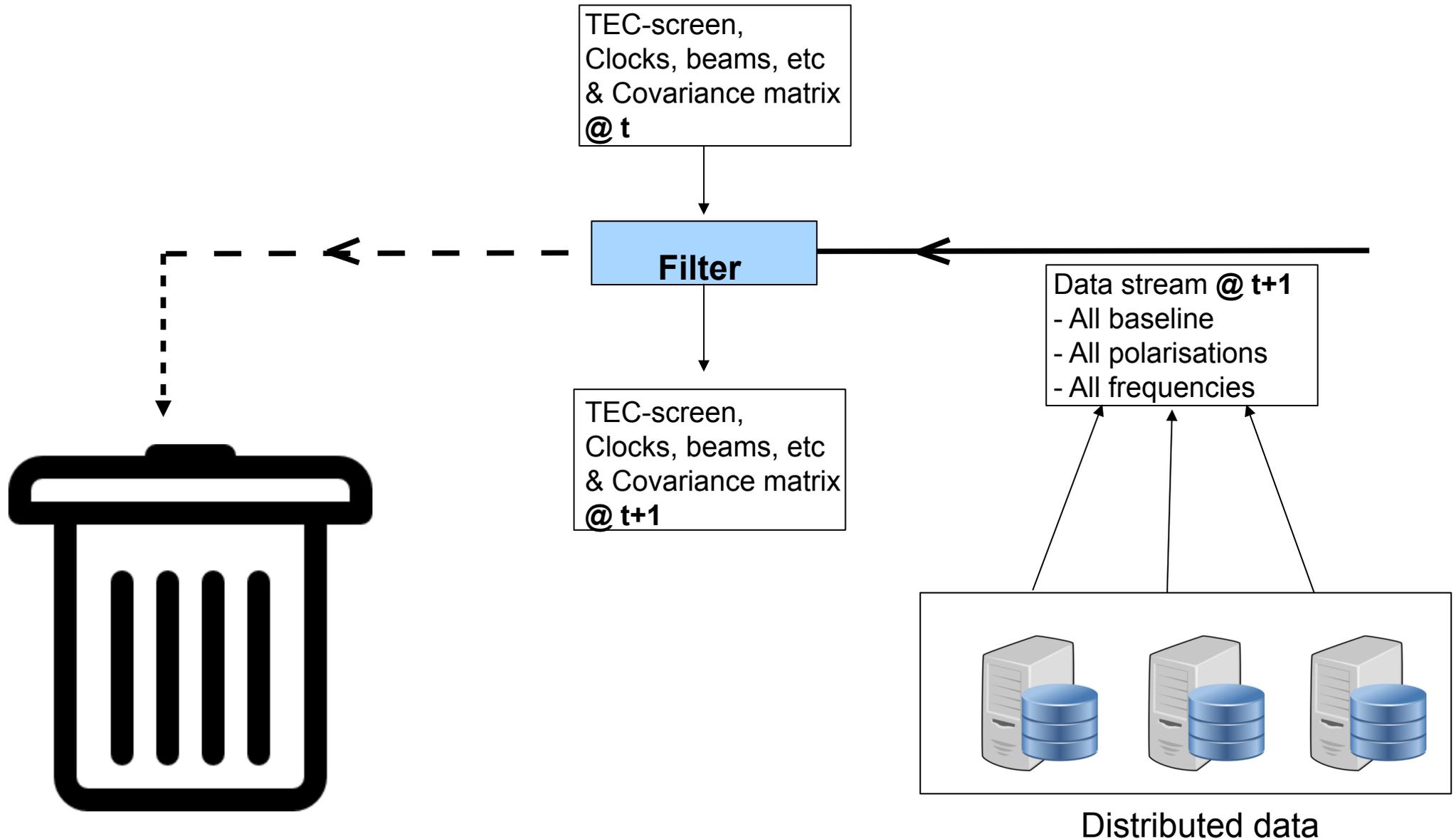


Equivalent to 1-
Dimensional raw-data
images

R



KaFCa (Kalman Filter for Calibration)



Simulated dataset (ionosphere TEC-screen + Clock drift)

- 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)]

Measurement
Equation

$$\mathbf{V}_{(pq)t\nu} = \mathbf{h}(\mathbf{x}) = \mathbf{G}_{pt\nu}(\mathbf{x}) \left(\sum_s \mathbf{V}_{(pq)t\nu}^s(\mathbf{x}) k_{(pq)t\nu}^s \right) \mathbf{G}_{qt\nu}^H(\mathbf{x}) \quad (1)$$

$$\mathbf{V}_{(pq)t\nu}^s(\mathbf{x}) = \mathbf{D}_{pst\nu}(\mathbf{x}) \mathbf{X}_s \mathbf{D}_{qst\nu}^H(\mathbf{x}) \quad (2)$$

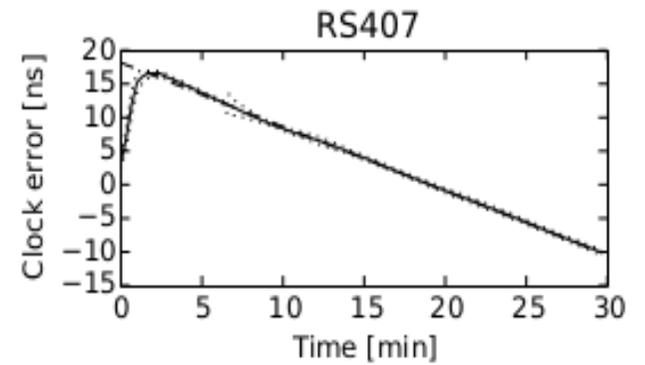
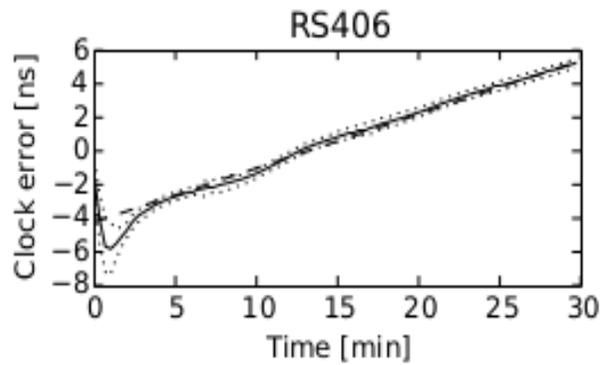
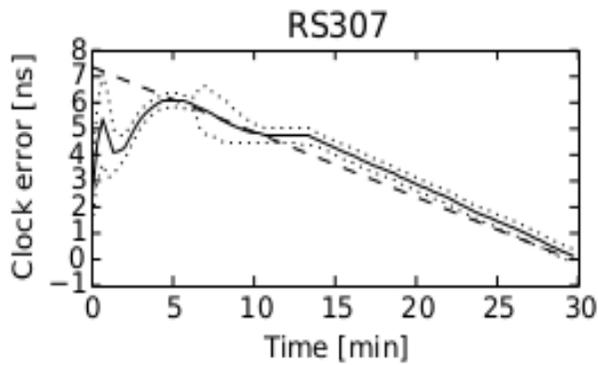
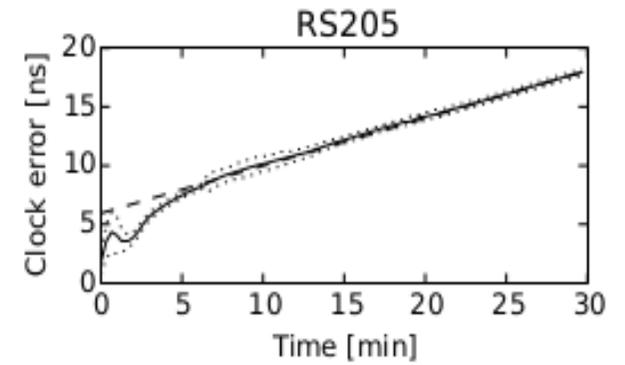
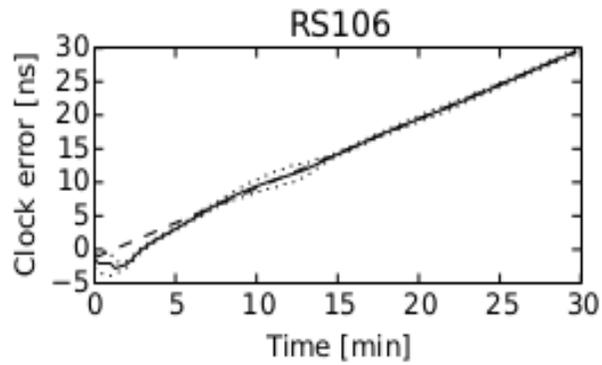
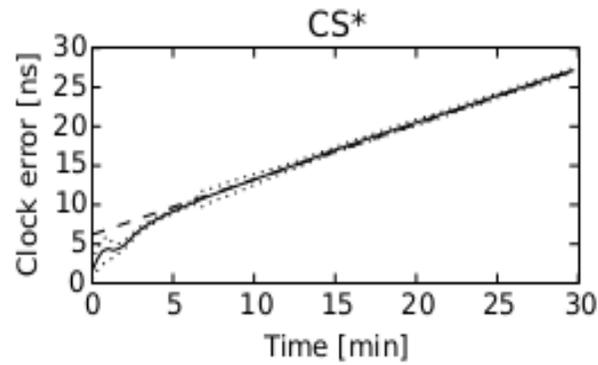
Clocks
Ionosphere

$$\mathbf{G}_{pt\nu}(\mathbf{x}) := \exp(2\pi i\nu \delta\mathbf{t}_p(\mathbf{x})) \mathbf{I}$$

$$\mathbf{D}_{pt\nu}^d(\mathbf{x}) := \exp\left(ik\nu^{-1} \mathbf{T}_p^d(\mathbf{x})\right) \mathbf{I}$$

Simulated dataset (ionosphere TEC-screen + Clock drift)

Clocks solutions



KaFCa on Real data!

- 50 subbands
- ME: TEC-screen, clocks, constant offset