

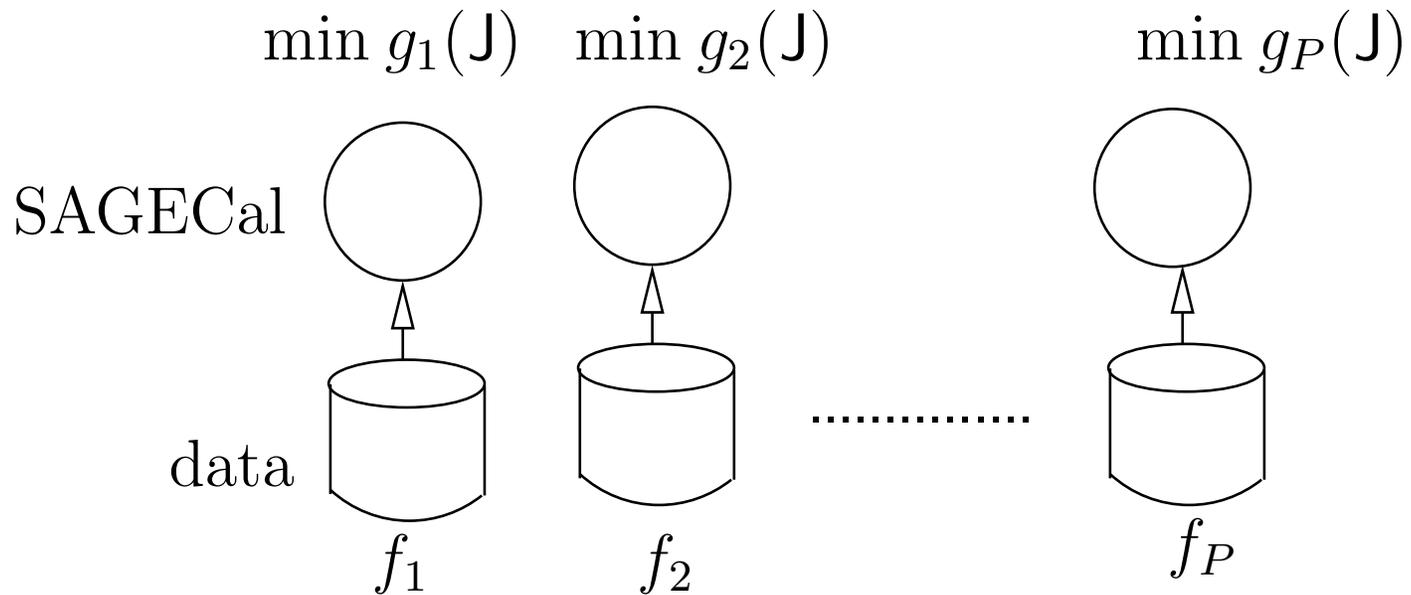
Distributed SAGECal

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ASTRON

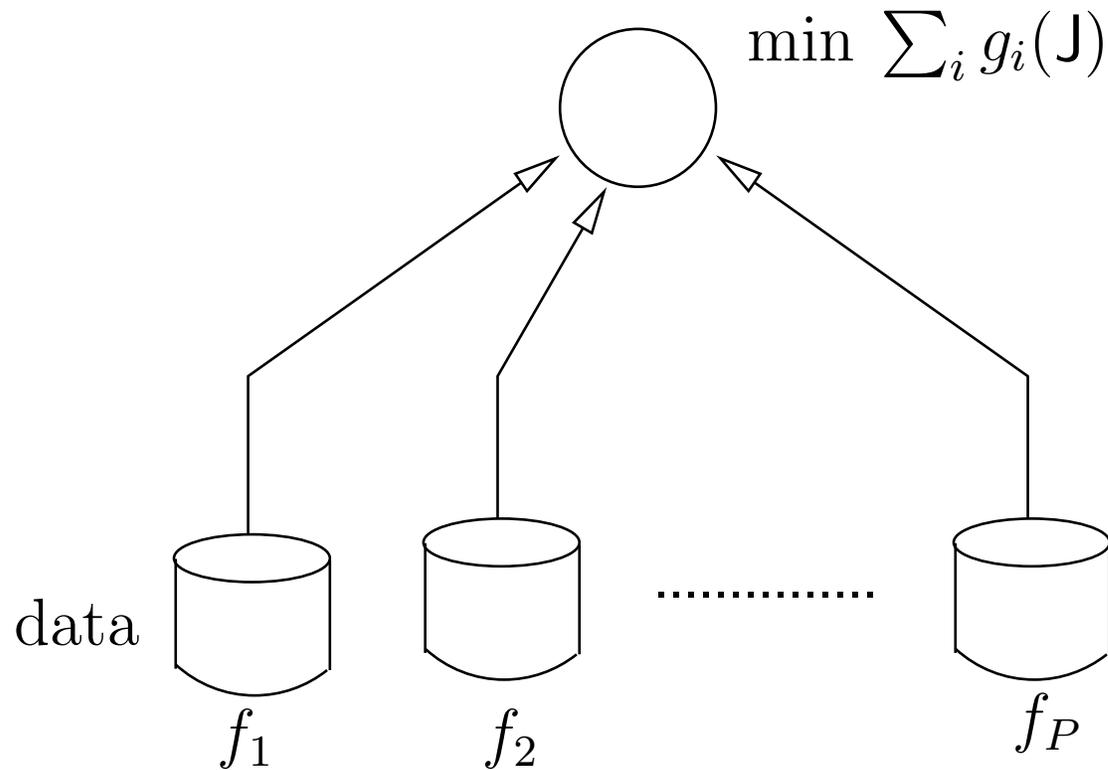
The Netherlands

Normal Calibration



Each SAGECal operates independently on data at different frequencies f_i . Solutions are only later interpolated.

What We Want



We want a unified solution exploiting smoothness in frequency.
But this **does not work** in practice: too much data, not enough memory, no accurate model to parametrize.

Distributed Calibration

- Normal calibration: each SAGECal works independently

$$J_{f_i} = \arg \min_J g_{f_i}(J)$$

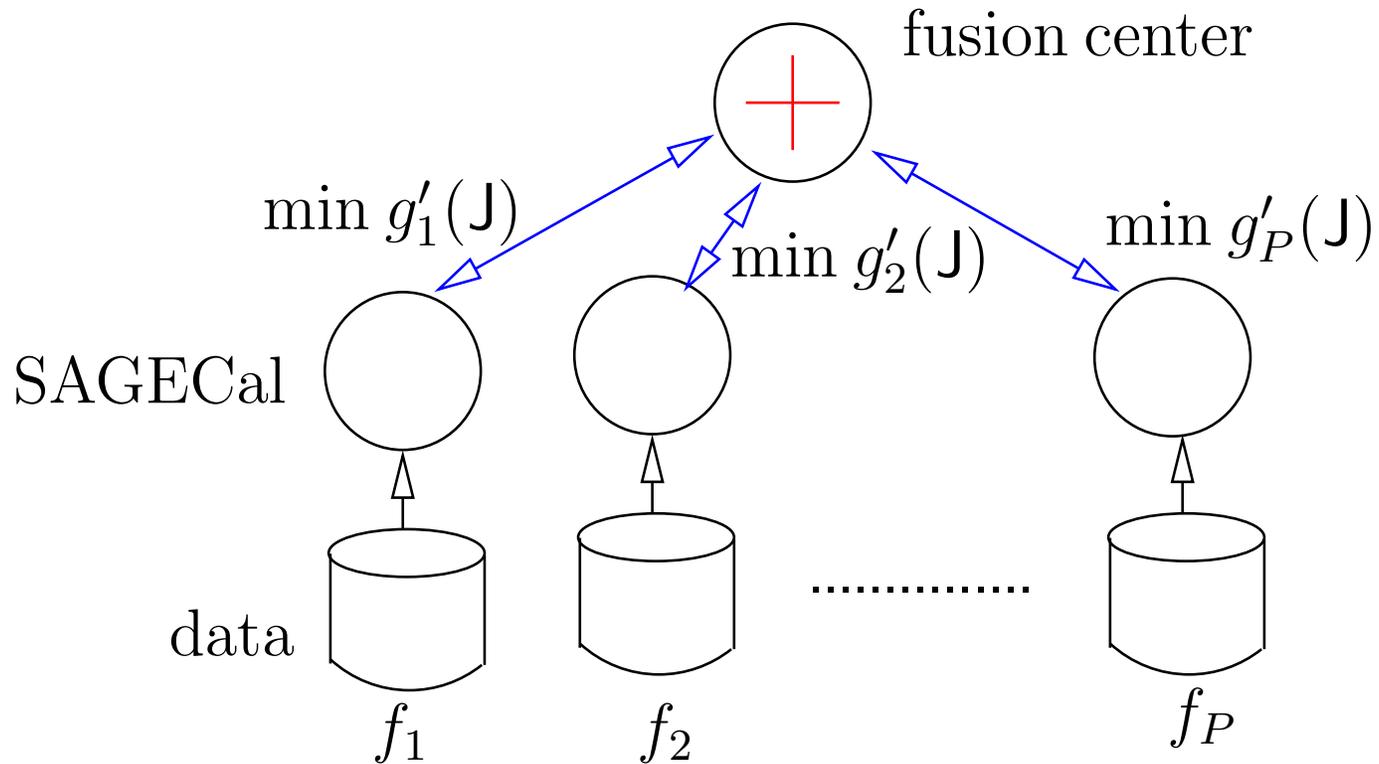
- Distributed calibration: each SAGECal **appears** to work independently, but actually solves

$$\{J_{f_1}, J_{f_2}, \dots, Z\} = \arg \min_{J_{f_i}, \dots, Z} \sum_i g_{f_i}(J_{f_i})$$

$$\text{subject to } J_{f_i} = B_{f_i}Z, \quad i \in [1, P]$$

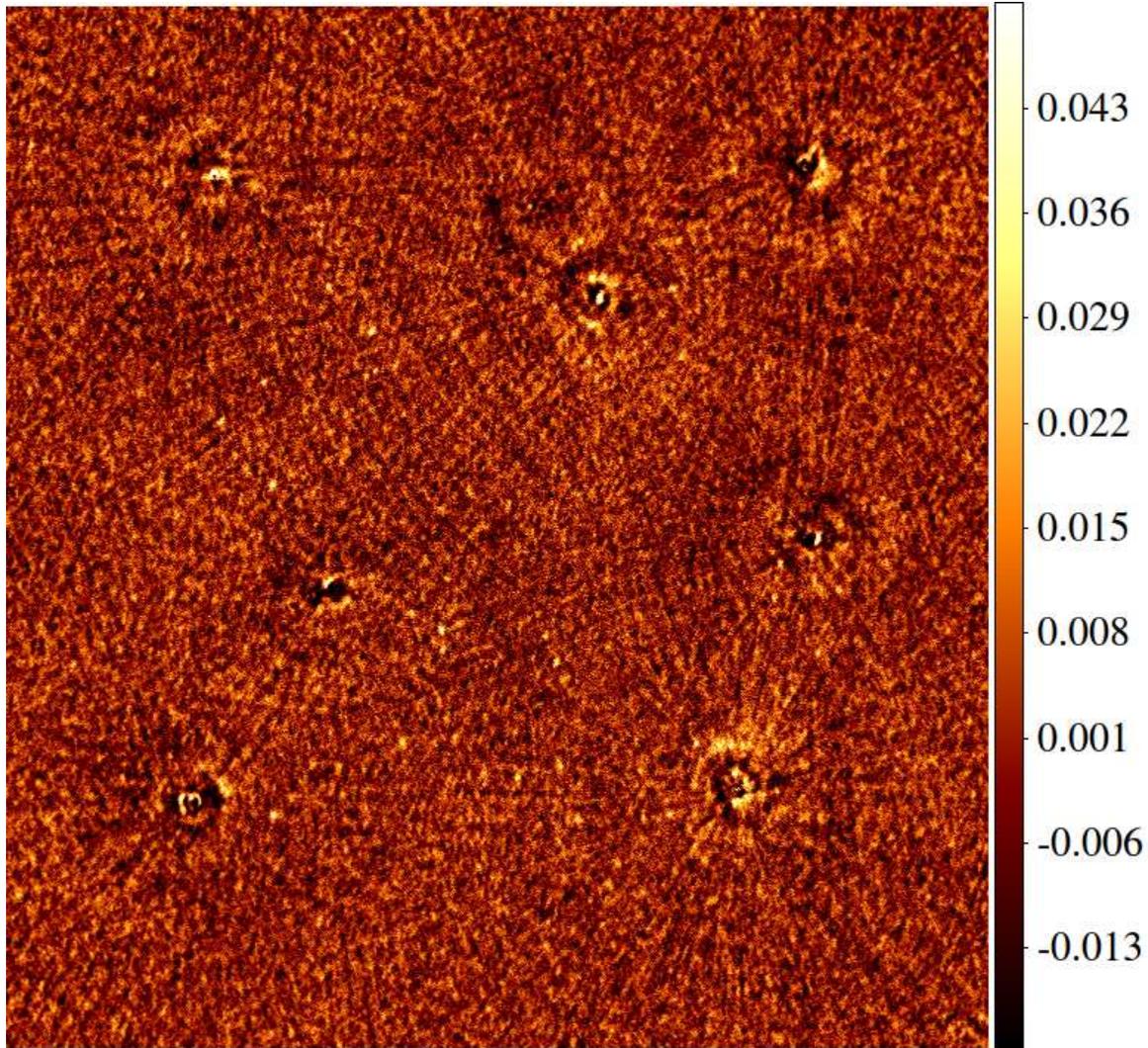
- Basic principle is **consensus optimization** : details [Tsitsiklis,1984], [Boyd et al.,2011], [Yatawatta,2015]

Distributed Calibration



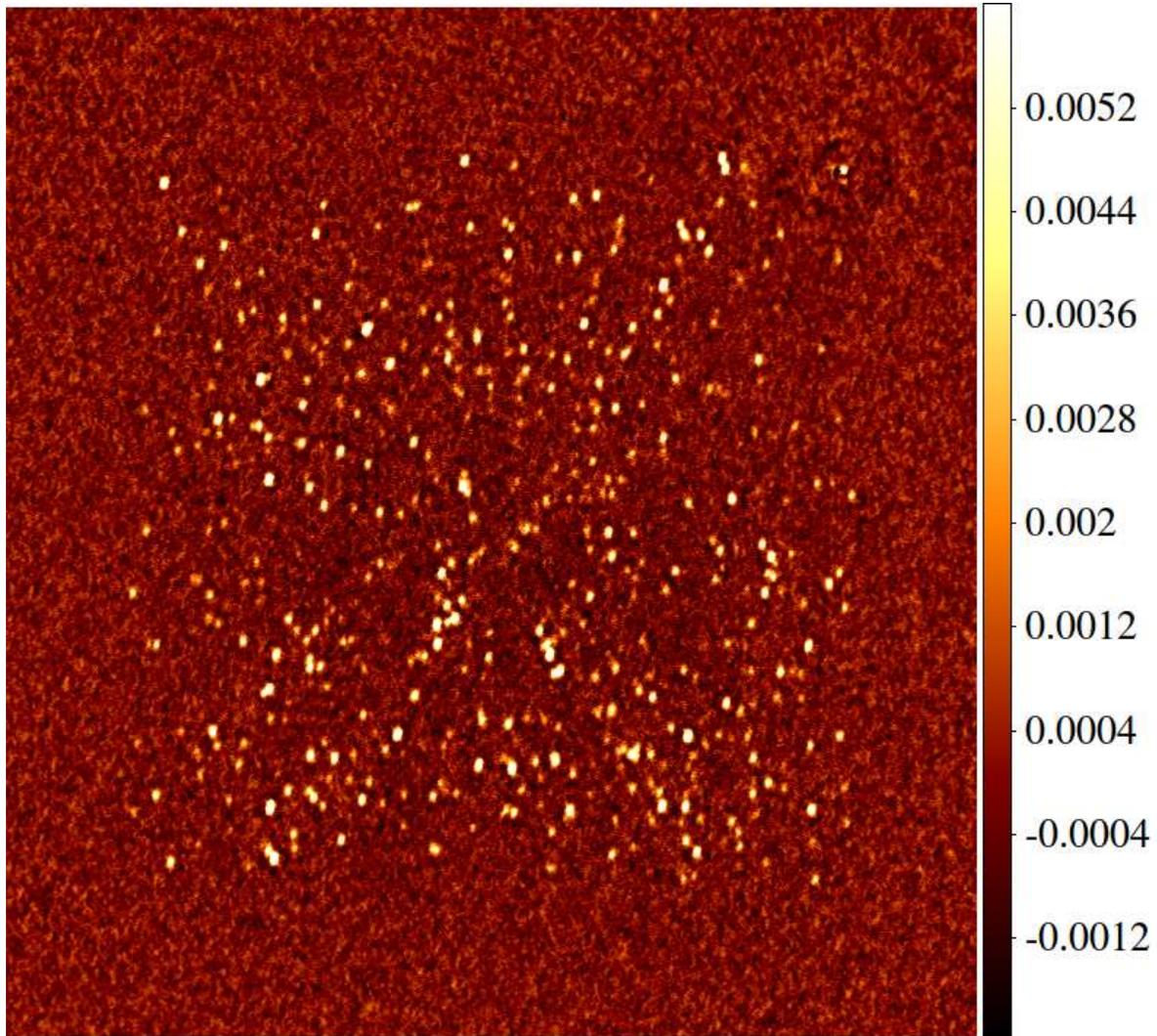
Information passed **is much less** than actual data calibrated. Only minor modifications to software needed, not running out of memory.

Example I



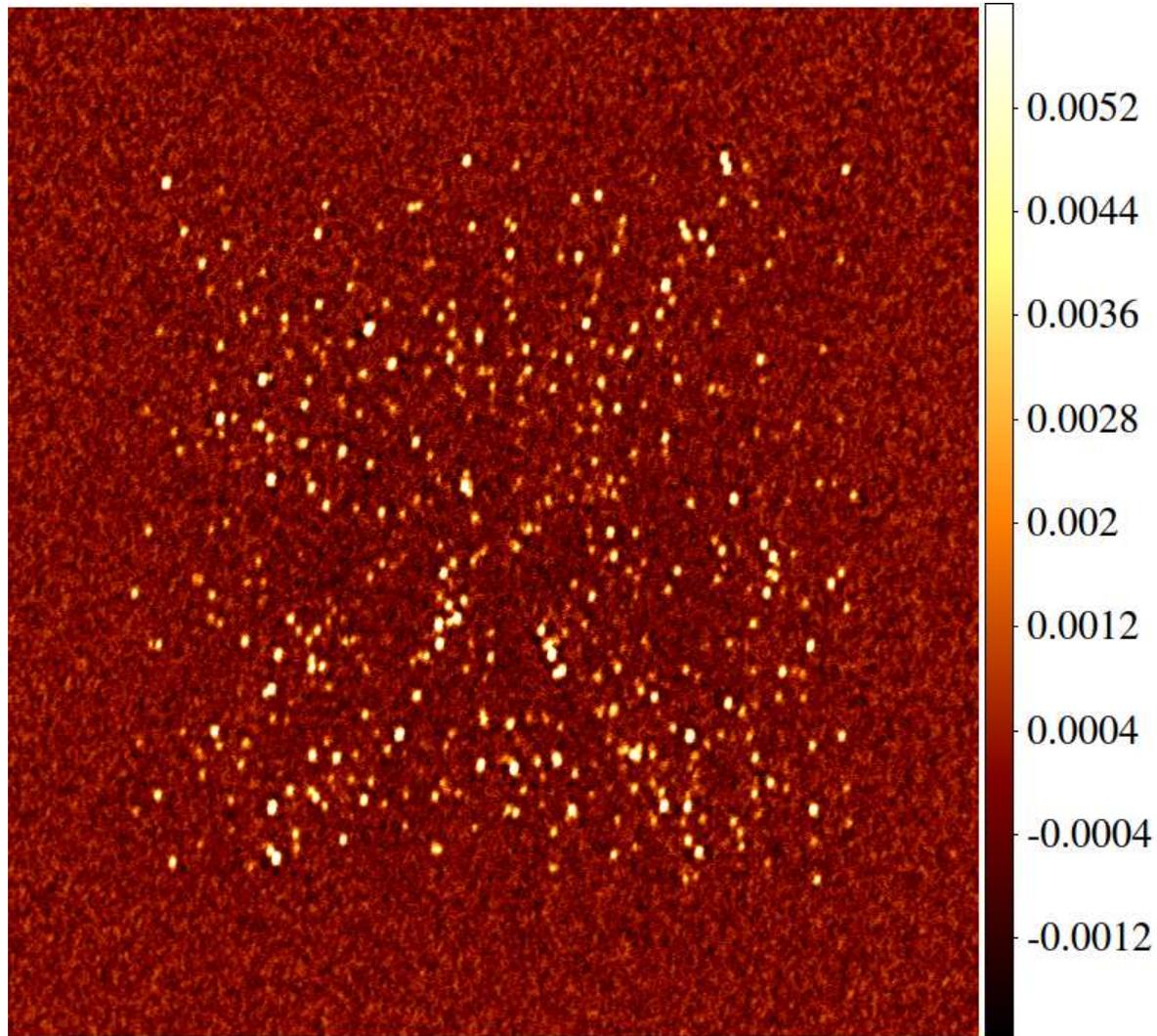
16 clusters calibrated using 1 time sample, 16 subbands 115-185 MHz

Normal Calibration



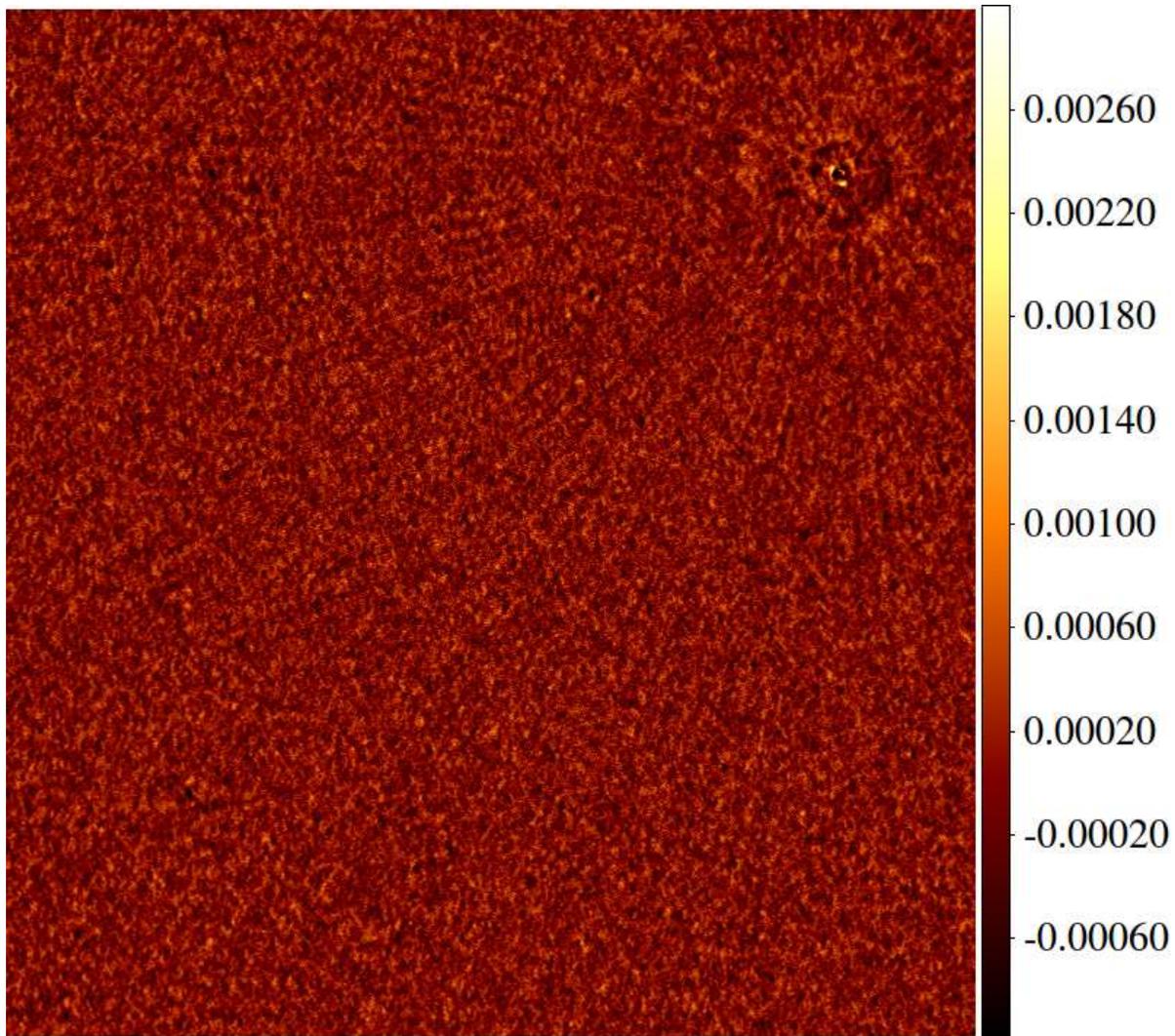
residual continuum image 16 subbands

Distributed Calibration



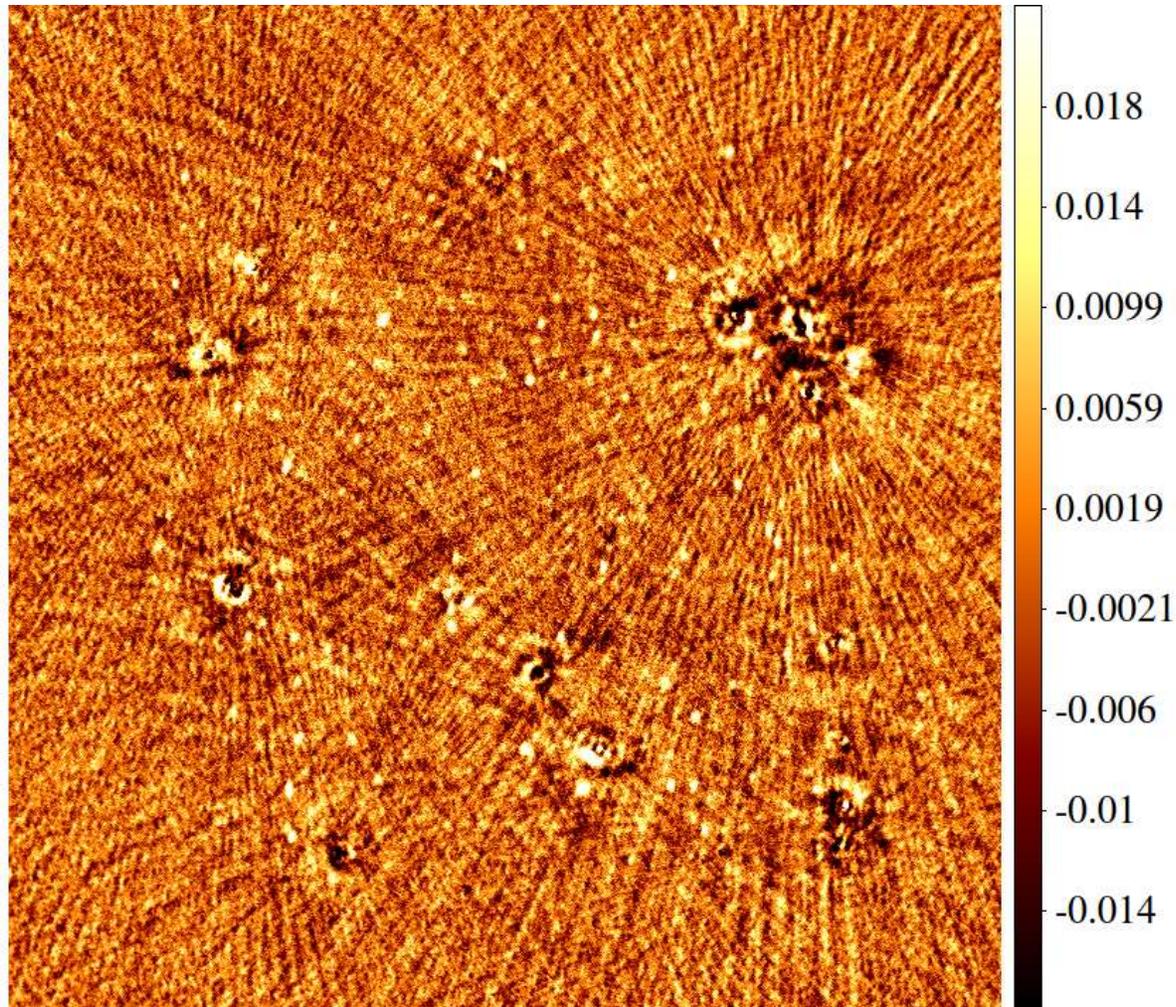
residual continuum image 16 subbands (noise lower)

Difference



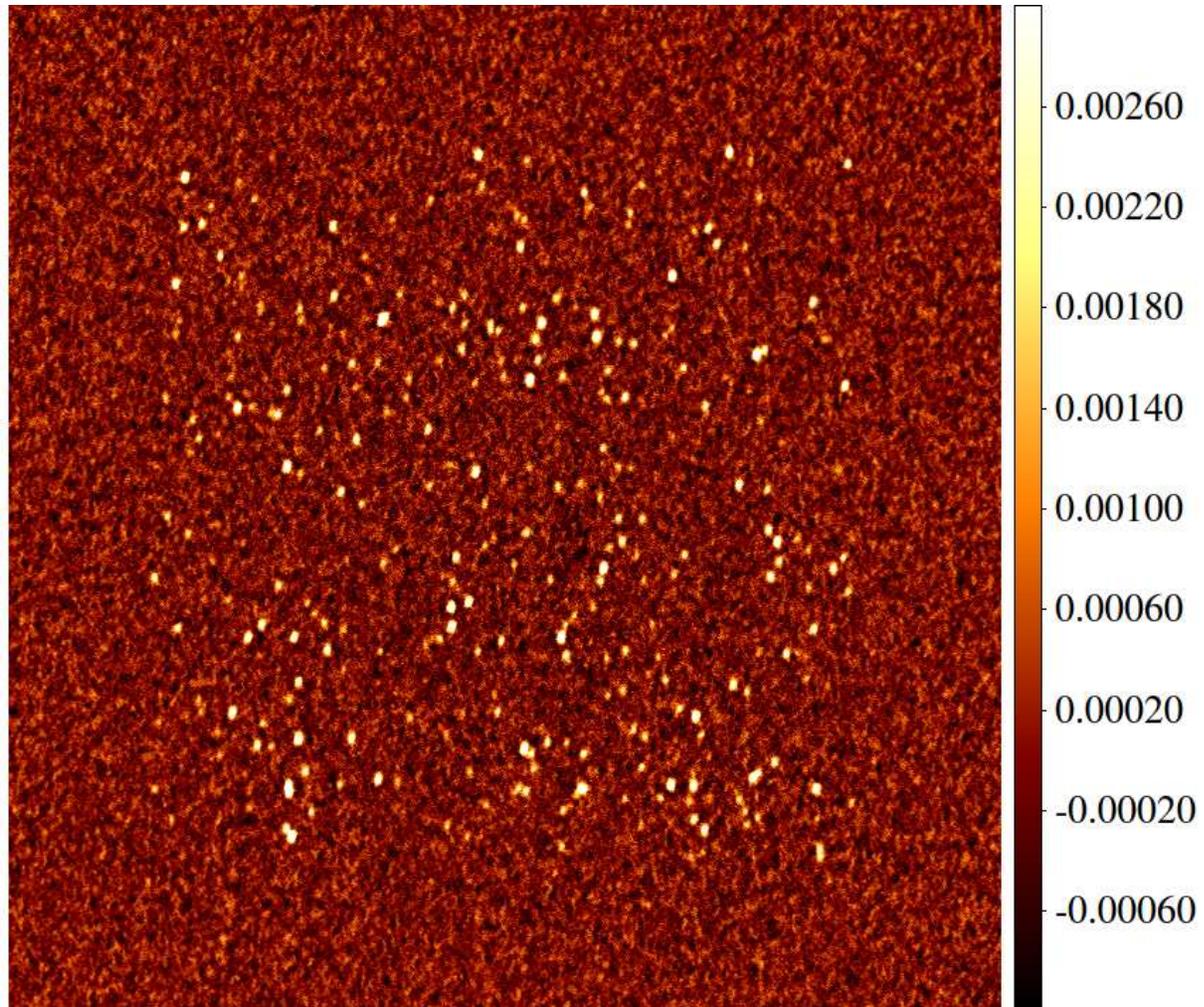
normal-distributed: calibration failed for one (weak) cluster with normal calibration

Example II



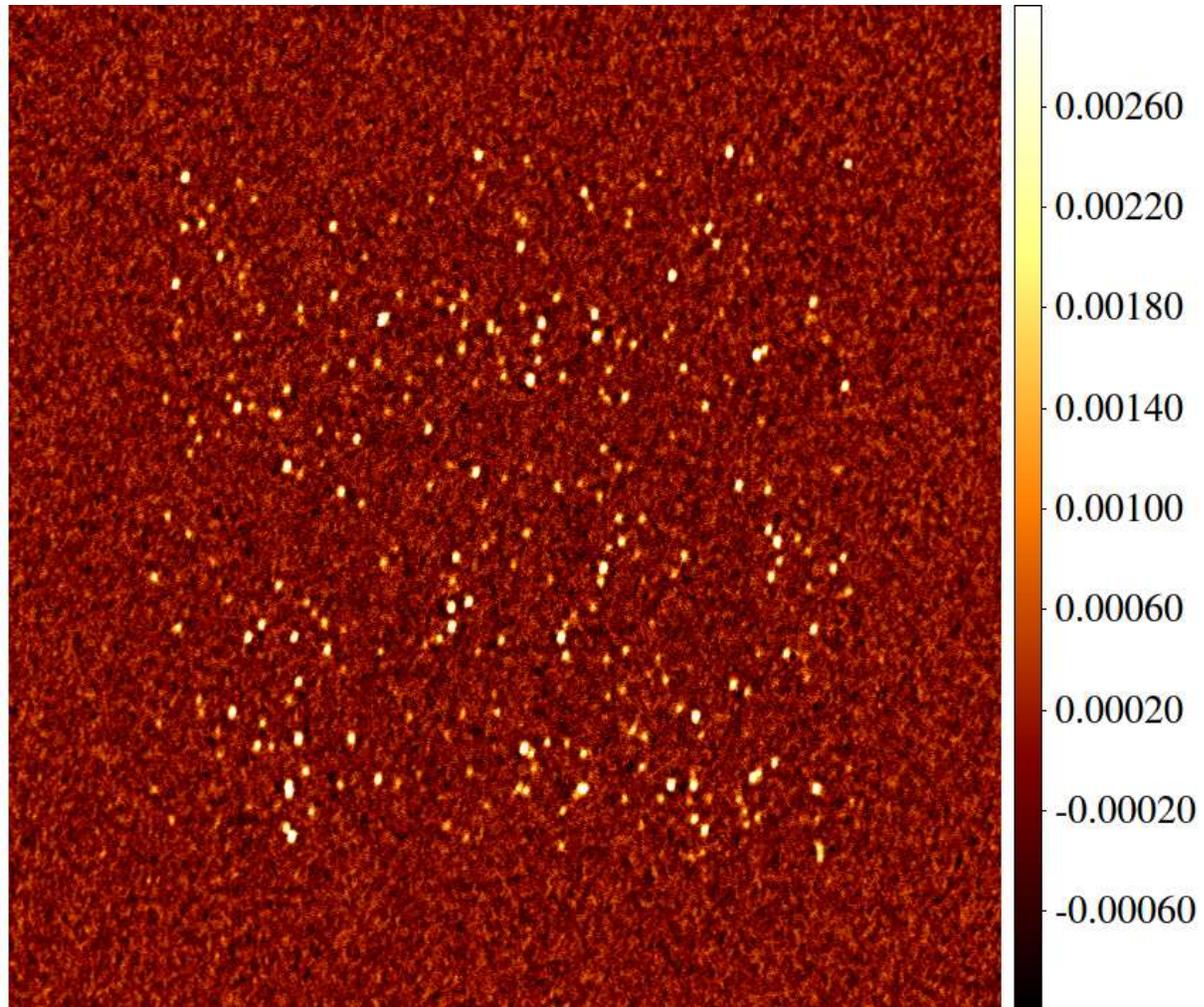
25 clusters calibrated using 1 time sample, 32 subbands 115-185 MHz,
unknowns > constraints

Normal Calibration



noise is higher

Distributed Calibration



noise is lower

Conclusions

Distributed calibration

- Not much additional computational cost.
- No need to access full dataset at each node.
- Robust to model errors.
- Calibration less likely to get stuck in local minima.
- Cons?? Theory is more complex.