

An important lesson learned from redundancy calibration

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AA calibration and calibrability meeting, 12th- 13th July 2012
Amsterdam- The Netherlands



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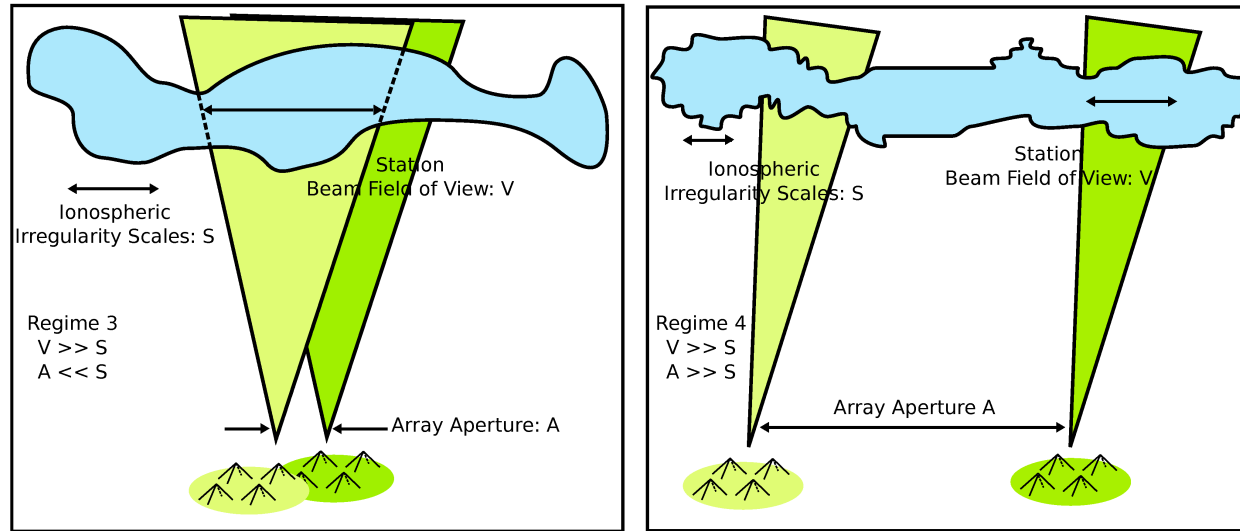
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The lesson is

Self-cal can be optimized “one step” using the constraints learned from the redundancy calibration for an array with arbitrary geometry.

The noise floor reduces accordingly.

Problem statement: ionosphere as a DDE



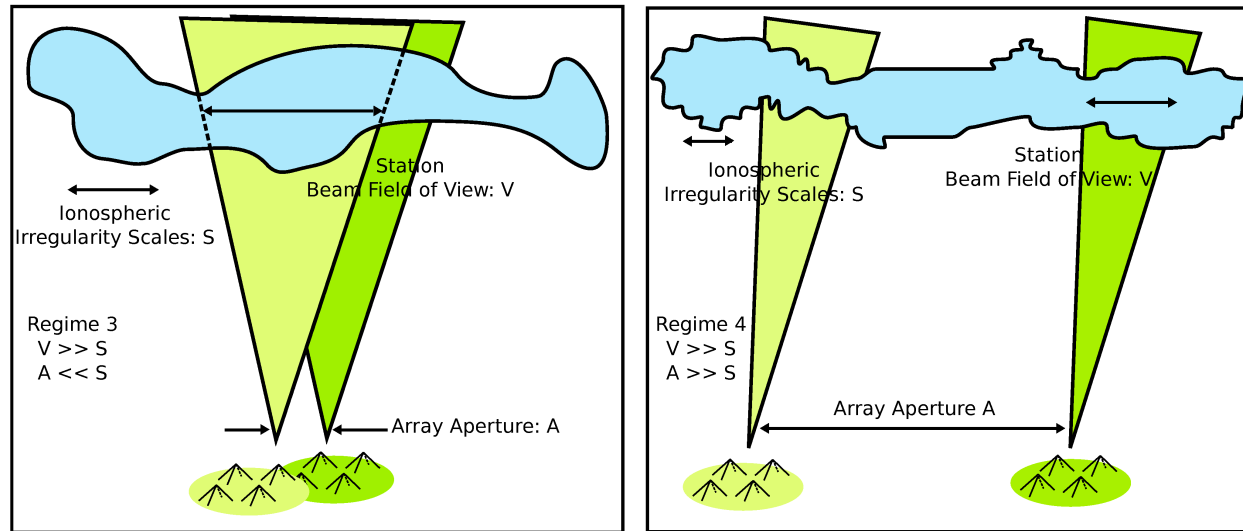
$$\theta = \operatorname{argmin}_{\theta} \left\| \hat{\mathbf{R}} - \mathbf{R}(\theta) \right\|_F^2$$

$$\theta = [\gamma_1, \gamma_1, \dots, \gamma_p, \phi_2, \phi_3, \dots, \phi_p, \sigma_1^2, \sigma_2^2, \dots, \sigma_p^2]^T$$

$$\mathbf{R}(\theta) = \mathbf{G}\mathbf{A}\Sigma_s\mathbf{A}^H\mathbf{G}^H + \Sigma_n$$

Acknowledged by Boonstra & van der Veen (2003), van der Tol et al. (2007), Yatawatta et al. (2008), Wijnholds & van der Veen (2009), Smirnov (2011), Kazemi et al. (2011), and many more.

Ionosphere: source position shift & a phase slope



$$\mathbf{A}' = \begin{bmatrix} e^{-j2\pi \frac{\mathbf{r}_1 \cdot (s_1 + \Delta s_{11})}{\lambda}} & e^{-j2\pi \frac{\mathbf{r}_1 \cdot (s_2 + \Delta s_{21})}{\lambda}} & \dots & e^{-j2\pi \frac{\mathbf{r}_1 \cdot (s_q + \Delta s_{q1})}{\lambda}} \\ e^{-j2\pi \frac{\mathbf{r}_2 \cdot (s_1 + \Delta s_{12})}{\lambda}} & e^{-j2\pi \frac{\mathbf{r}_2 \cdot (s_2 + \Delta s_{22})}{\lambda}} & \dots & e^{-j2\pi \frac{\mathbf{r}_2 \cdot (s_q + \Delta s_{q2})}{\lambda}} \\ \vdots & \vdots & \ddots & \vdots \\ e^{-j2\pi \frac{\mathbf{r}_p \cdot (s_1 + \Delta s_{1p})}{\lambda}} & e^{-j2\pi \frac{\mathbf{r}_p \cdot (s_2 + \Delta s_{2p})}{\lambda}} & \dots & e^{-j2\pi \frac{\mathbf{r}_p \cdot (s_q + \Delta s_{qp})}{\lambda}} \end{bmatrix}$$

$$\mathbf{A}' = \begin{bmatrix} e^{-j2\pi \frac{\mathbf{r}_1 \cdot \Delta s}{\lambda}} & 0 & \dots & 0 \\ 0 & e^{-j2\pi \frac{\mathbf{r}_2 \cdot \Delta s}{\lambda}} & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & e^{-j2\pi \frac{\mathbf{r}_p \cdot \Delta s}{\lambda}} \end{bmatrix} \mathbf{A}$$

$$\mathbf{R}(\theta) = \mathbf{G} \mathbf{A} \Sigma_s \mathbf{A}^H \mathbf{G}^H + \Sigma_n$$

A solution: using the deterministic constrains

$$\mathbf{R}(\boldsymbol{\theta}) = \mathbf{G} \mathbf{A} \boldsymbol{\Sigma}_s \mathbf{A}^H \mathbf{G}^H + \boldsymbol{\Sigma}_n$$

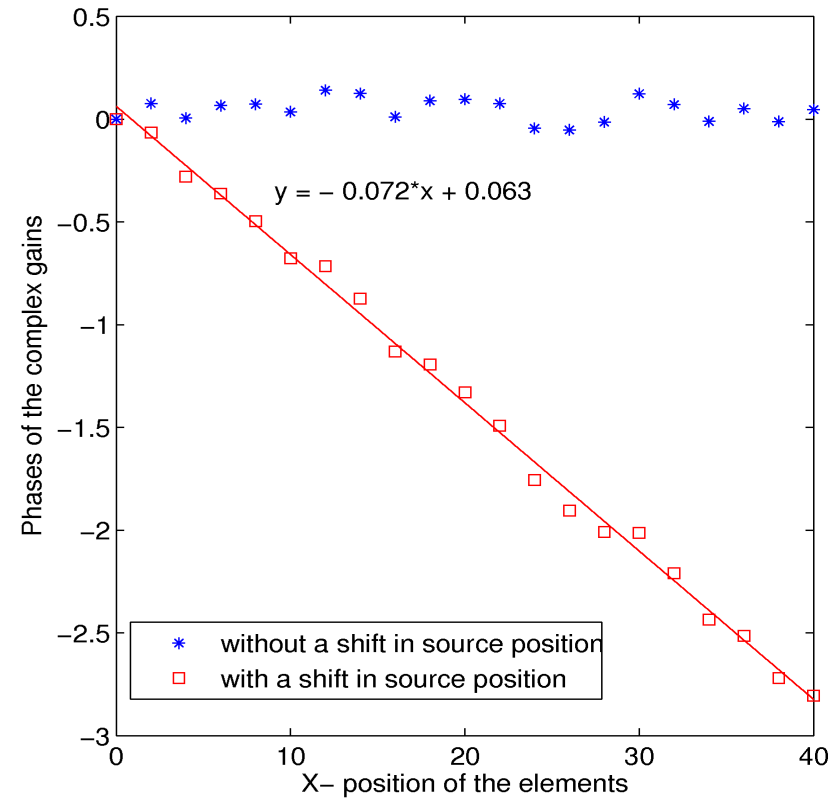
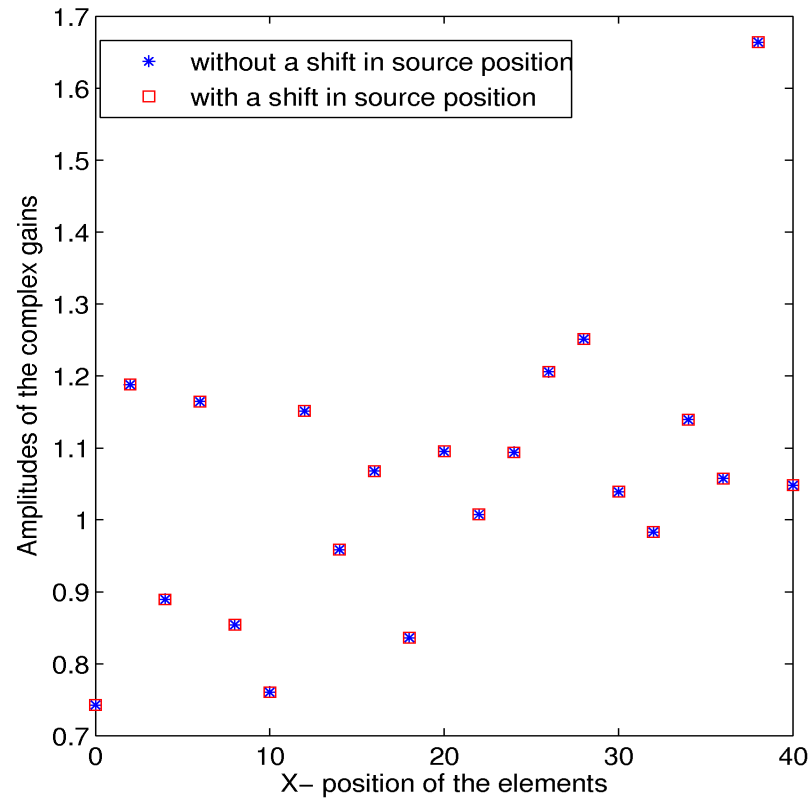
$$\mathbf{G} = \begin{bmatrix} \gamma_1 e^{j(\phi_1 - 2\pi \frac{\mathbf{r}_1 \cdot \Delta \mathbf{s}}{\lambda})} & 0 & \dots & 0 \\ 0 & \gamma_2 e^{j(\phi_2 - 2\pi \frac{\mathbf{r}_2 \cdot \Delta \mathbf{s}}{\lambda})} & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \gamma_p e^{j(\phi_p - 2\pi \frac{\mathbf{r}_p \cdot \Delta \mathbf{s}}{\lambda})} \end{bmatrix}$$

$$\mathbf{g}' = [\gamma_1 e^{j(\phi_1 - 2\pi \frac{\Delta l x_1 + \Delta m y_1}{\lambda})}, \gamma_2 e^{j(\phi_2 - 2\pi \frac{\Delta l x_2 + \Delta m y_2}{\lambda})}, \dots, \gamma_p e^{j(\phi_p - 2\pi \frac{\Delta l x_p + \Delta m y_p}{\lambda})}]^T$$

$$\mathbf{f}(\boldsymbol{\theta}) = \begin{bmatrix} \sum_{i=1}^p \phi_i x_i \\ \sum_{i=1}^p \phi_i y_i \end{bmatrix} = \begin{bmatrix} \boldsymbol{\phi} \mathbf{x}^T \\ \boldsymbol{\phi} \mathbf{y}^T \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \mathbf{0}_{2 \times 1}$$

We always calibrate against a single source to reduce the parameters space (Kazemi et al. 2011). In this way, we do not have the unitary ambiguity either.

A simple example: self-cal results



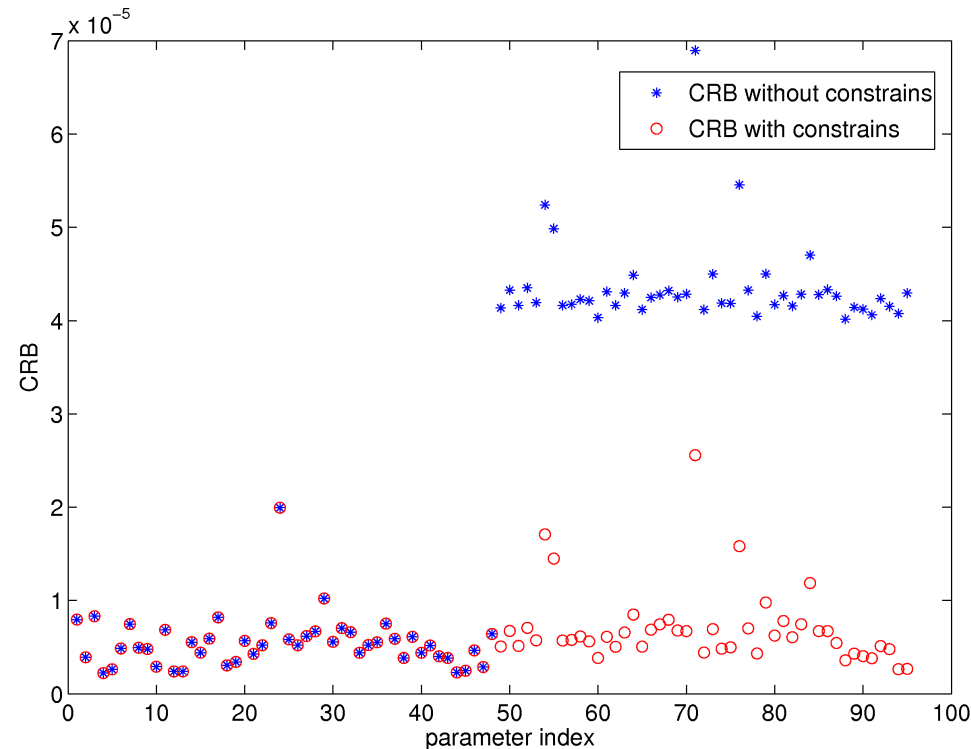
$$\Delta\phi = 0.2, \text{ where } \Delta\phi < 2\pi$$

$$\Delta l = 0.02$$

$$f = 170\text{MHz}, \text{ thus } \lambda = 1.7635\text{m}$$

$$-2\pi \frac{\Delta l}{\lambda} = -0.0713$$

CRB analysis: the constrains indeed optimize the self-cal performance

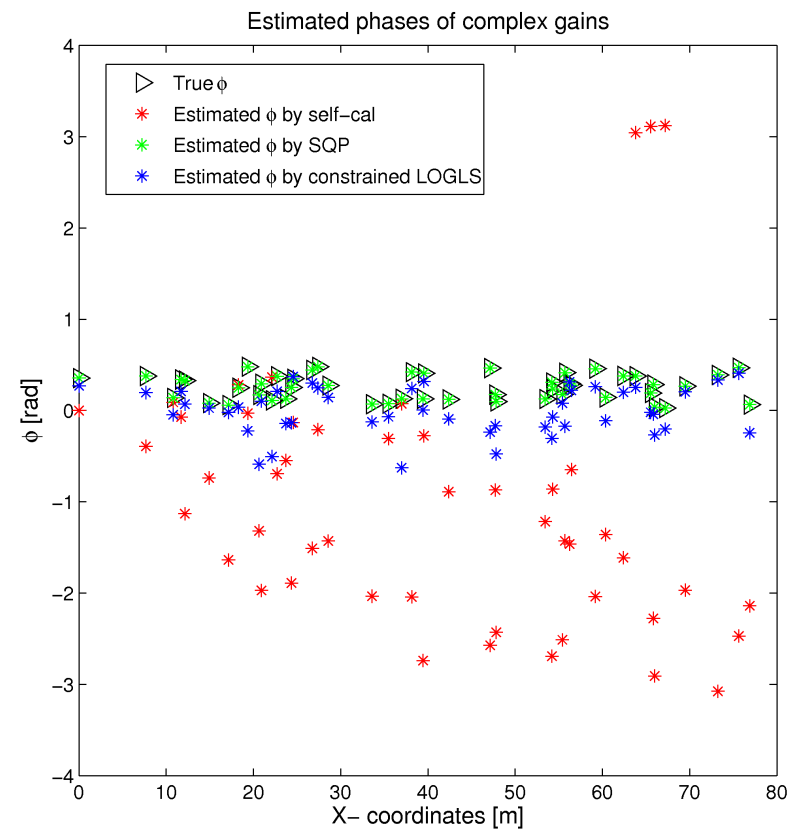
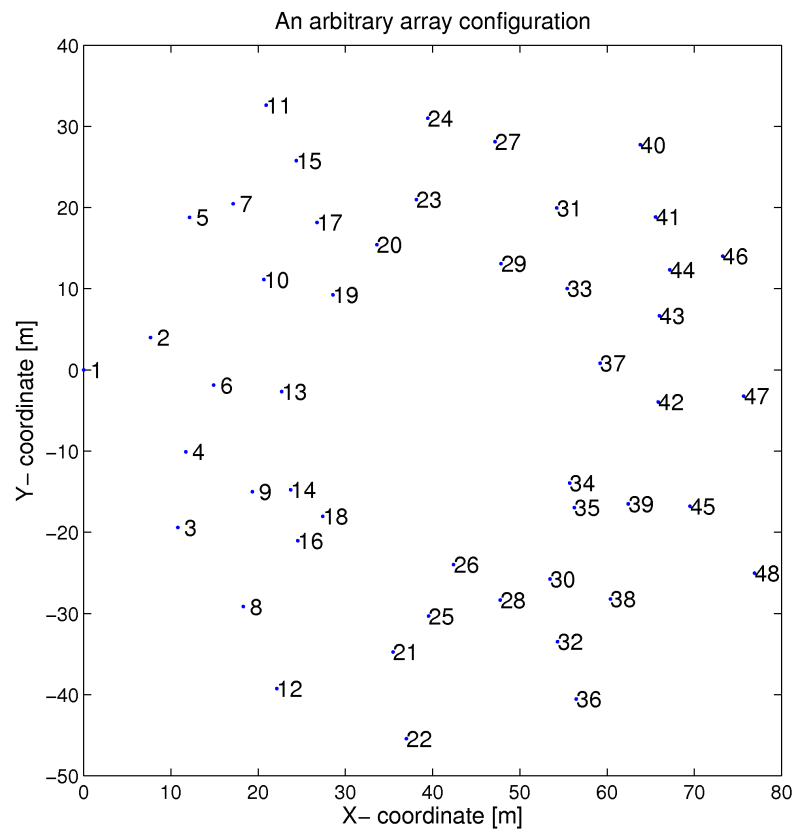


$$\mathbf{f}(\boldsymbol{\theta}) = \begin{bmatrix} \sum_{i=1}^p \phi_i x_i \\ \sum_{i=1}^p \phi_i y_i \end{bmatrix} = \begin{bmatrix} \boldsymbol{\phi} \mathbf{x}^T \\ \boldsymbol{\phi} \mathbf{y}^T \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \mathbf{0}_{2 \times 1} \quad \mathbf{C} = E \left((\hat{\boldsymbol{\theta}} - \boldsymbol{\theta})(\hat{\boldsymbol{\theta}} - \boldsymbol{\theta})^T \right) \geq \frac{1}{N} \mathbf{U}(\mathbf{U}^T \mathbf{J} \mathbf{U})^{-1} \mathbf{U}$$

Based on Wieringa (1991), Boonstra & van der Veen (2003), van der Tol et al. (2007),
Wijnholds & van der Veen (2009), and Stoica & Ng (1998)

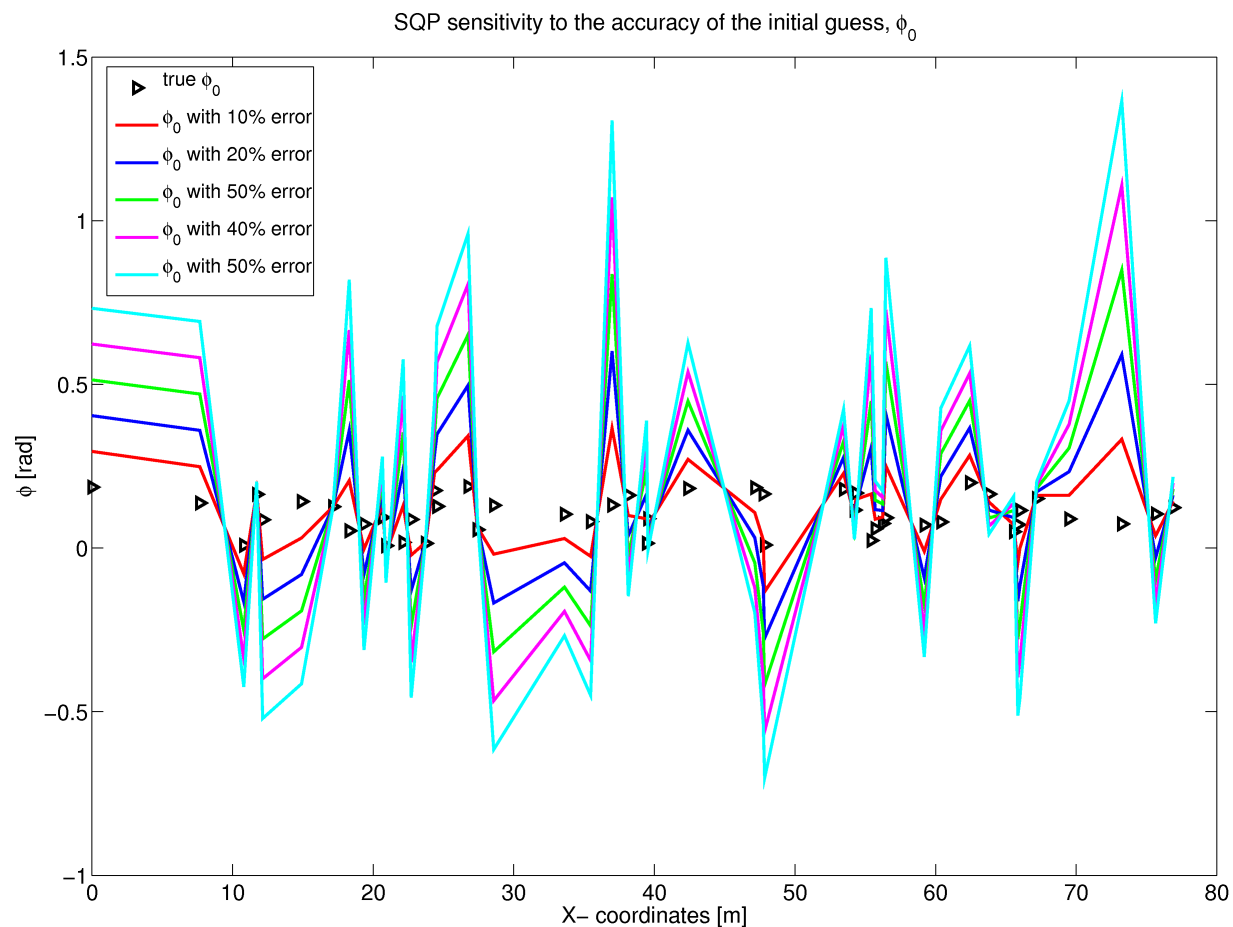
Solving a constrained non-linear LS

- NOTE: To see the slope over the array and remove it, element numbering is important.



Based on Boonstra & van der Veen (2003), Schittkowski (1985) and many more.

SQP and initial guess accuracy



- These results are after only one iteration.
- We hope to have an initial guess within 10% of accuracy.

Summary & conclusion

Summary & conclusion:

- We suggested a solution for a more accurate self-cal (for an array with arbitrary geometry).
- Using a priori information e.g. constraining some of the parameters (to reduce the parameters degrees of freedom), instead of estimating so many parameters in the calibration process.
- To see the slope and correct for it, element number is important.
- This is a one step.
- **And I have one more un-discussed slide.**

One more slide in favor of redundancy

- There are variants of redundancy calibration that we have not exploited in radio astronomy.
- They seem to be efficient and computationally cheaper than our version of redundancy (Li & Er, 2006)
- All variants of redundancy determine the complex gains up to a scalar. Why don't we consider normalizing the gain of our array elements in their design (especially for the sake of constraining the amplitudes as well)?