



Identification of system parameters during AO-corrected observations

C. Béchet¹, J. Kolb¹, P.-Y. Madec¹,
M. Tallon², E. Thiébaud²

¹ European Southern Observatory, Germany

² Centre de Recherche Astrophysique de Lyon, France



Outline

1. Context & constraints
 - a) towards AOF and the ELTs
 - b) examples of sensitivity to misregistrations
 - c) main goals

2. Criterion for parameters identification
 - a) AO equations
 - b) estimation method

3. First numerical results
 - a) simulations on AOF-size AO
 - b) on-going and future work



1. Context & constraints

a) Towards Adaptive Optics Facility (AOF) @ the VLT

- deformable secondary mirror (DSM)
- large : 1170 actuators / 40x40 subapertures
- no calibration source before the DSM
- **adaptive telescopes** / non-stationnary AO systems
 - atmospheric parameters, like Cn2 profile evolution
 - **system parameters, like misregistrations**

⇒ requires new calibration strategies

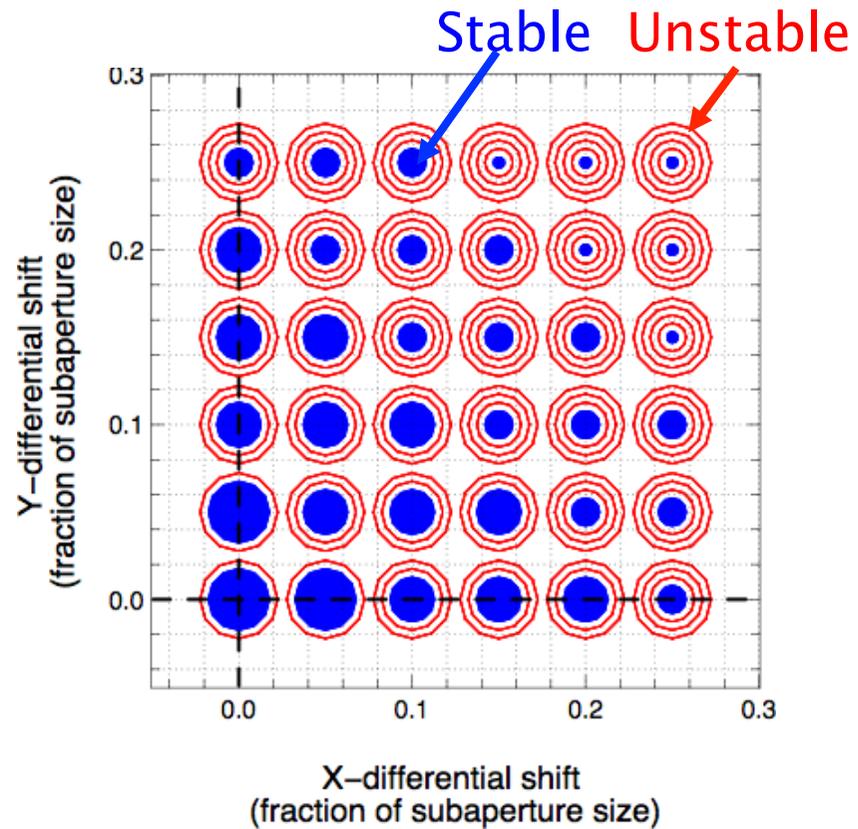
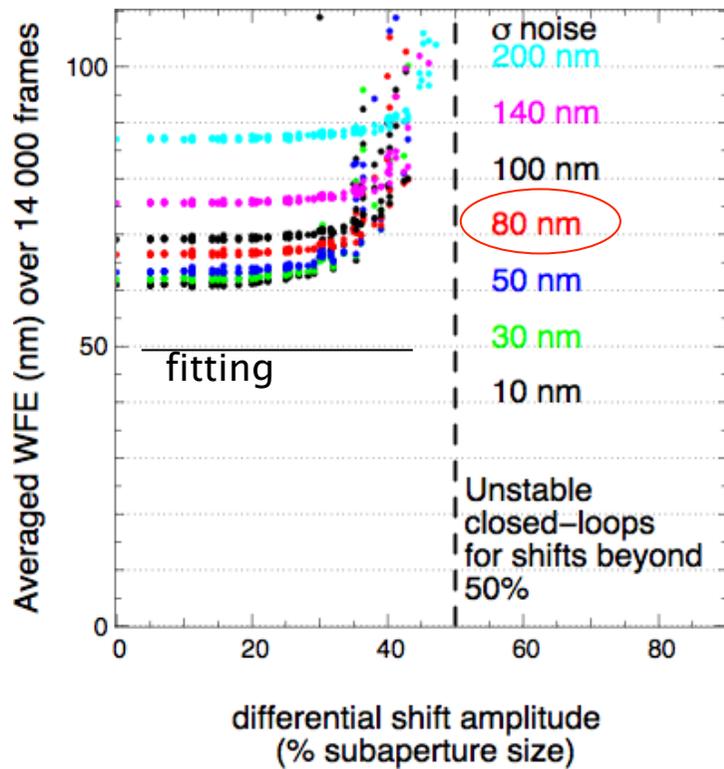
- ❖ pseudo-synthetic interaction matrix
- ❖ on-sky calibrations of long-term fixed parameters
- ❖ **on-sky identification during closed-loop AO at a minute-scale**



★ 1. Context & constraints

- b) examples of sensitivity to misregistrations
 - 2D-shifts and rotation of the DM w.r.t. the WFS alignments

AOF-size SCAO but « square aperture » and Fried geometry

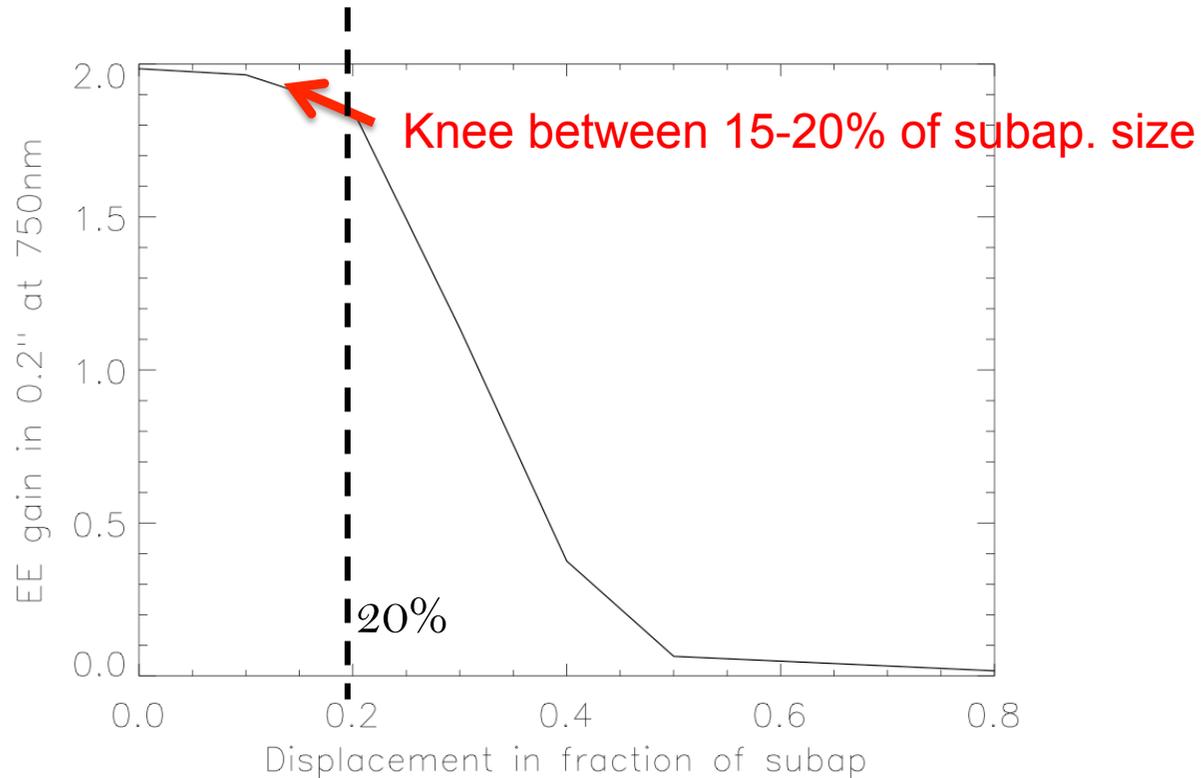




★ 1. Context & constraints

b) examples of sensitivity to misregistrations

from AO performance analysis report of GALACSI (LTAO with AOF)



Remember:

For a given system, it depends on the reconstruction and control algorithms



1. Context & constraints

c) main goals

1. estimation accuracy requirements

- AOF driver : better than 10% subaperture size
- AO performance analysis report of GALACSI (LTAO with AOF)

2. no additional disturbance introduced

- no impact on astrophysical data
- fully exploit AO data



3. numerical update of interaction matrix (IM) and control matrix (CM)

- identification ~ every minute
- not moving the DM, but updating the pseudo synthetic IM/CM
- only when misregistrations are larger than a threshold (lost frames)



2. Criterion for parameters identification

a) AO equations

- measurement equation

$$\mathbf{d}_k = \mathbf{S} \cdot \mathbf{w}_k - \mathbf{G}(p) \cdot \mathbf{a}_k + \mathbf{e}_k$$

Diagram illustrating the measurement equation with annotations:

- \mathbf{d}_k : AO closed-loop data
- \mathbf{S} : propag. & sensing model
- \mathbf{w}_k : turbulent atmosphere
- $\mathbf{G}(p)$: interaction matrix (circled in red)
- \mathbf{a}_k : DMs commands
- \mathbf{e}_k : measurement noise

- parametric model for the interaction matrix
- Note: time index different from usual writing -> clearer for identification

- control law in AOF real-time computer (tau = AO correction delay)

$$\mathbf{a}_{k+\tau} = \alpha \mathbf{a}_{k+\tau-1} + \beta \mathbf{a}_{k+\tau-2} + \gamma \mathbf{C} \cdot \mathbf{d}_k$$



★ 2. Criterion for parameters identification

- a) AO equations
- measurement equation

$$d_k = \mathbf{S} \cdot \mathbf{w}_k - \mathbf{G}(p) \cdot \mathbf{a}_k + \mathbf{e}_k$$

$$d_k = -\mathbf{G}(p) \cdot \mathbf{a}_k + \mathbf{z}_k$$

➤ parametric model for the interaction matrix

with $\mathbf{z}_k = \underbrace{\mathbf{S} \cdot \mathbf{w}_k}_{\text{contribution of turbulent atmosphere}} + \underbrace{\mathbf{e}_k}_{\text{measurement noise}}$

- measures the system parameters *through* the applied commands
- « noise » includes: turbulence and AO usual measurement noise



2. Criterion for parameters identification

b) criterion for parameters estimation

$$\mathbf{d}_k = -\mathbf{G}(\mathbf{p}) \cdot \mathbf{a}_k + \mathbf{z}_k$$

⇒ non-linear fitting method

$$f(\mathbf{p}) = \frac{1}{2} \sum_{k=1}^N (\mathbf{d}_k + \mathbf{G}(\mathbf{p}) \cdot \mathbf{a}_k)^T \cdot \mathbf{C}_z^{-1} \cdot (\mathbf{d}_k + \mathbf{G}(\mathbf{p}) \cdot \mathbf{a}_k) \quad ???$$



★ 2. Criterion for parameters identification

b) criterion for parameters estimation

$$\mathbf{d}_k = -\mathbf{G}(\mathbf{p}) \cdot \mathbf{a}_k + \mathbf{z}_k$$

⇒ non-linear fitting method

~~$$f(\mathbf{p}) = \frac{1}{2} \sum_{k=1}^N (\mathbf{d}_k + \mathbf{G}(\mathbf{p}) \cdot \mathbf{a}_k)^T \cdot \mathbf{C}_z^{-1} \cdot (\mathbf{d}_k + \mathbf{G}(\mathbf{p}) \cdot \mathbf{a}_k)$$~~ ???

main issues:

- o on-sky **calibration** approaches do not apply, since we do not choose the applied commands
- o applied commands are almost the inverse of the turbulence (which is now a « noise » source) ⇒ strongly correlated



★ 2. Criterion for parameters identification

b) Criterion for parameters estimation

$$\mathbf{d}_k = -\mathbf{G} \cdot \mathbf{a}_k + \mathbf{z}_k$$



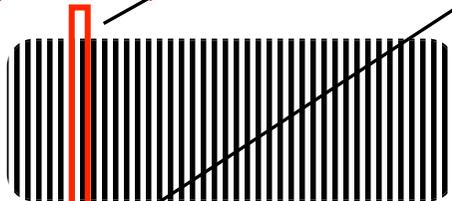
$$\begin{aligned} \delta \mathbf{d}_k &= \mathbf{d}_{k+1} - \mathbf{d}_k \\ &= -\mathbf{G} \cdot \delta \mathbf{a}_k + \delta \mathbf{z}_k \end{aligned}$$

⇒ **non-linear fitting method**

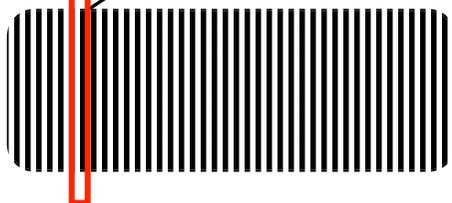
⇒ aim is to have decorrelated signal and noise (commands increments w.r.t. turbulence increments)

$$f_1(\mathbf{p}) = \frac{1}{2} (\delta \mathbf{d}_k + \mathbf{G}(\mathbf{p}) \cdot \delta \mathbf{a}_k)^T \cdot \mathbf{C}_{\delta \mathbf{z}}^{-1} \cdot (\delta \mathbf{d}_k + \mathbf{G}(\mathbf{p}) \cdot \delta \mathbf{a}_k)$$

{k+1, k}



recorded AO data history



recorded AO commands history



★ 2. Criterion for parameters identification

b) Criterion for parameters estimation

$$\mathbf{d}_k = -\mathbf{G} \cdot \mathbf{a}_k + \mathbf{z}_k \quad \longrightarrow$$

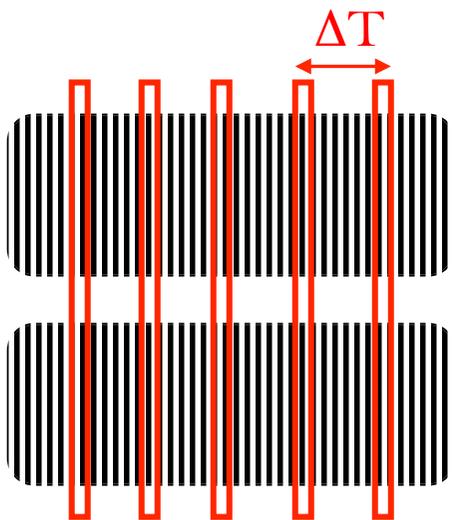
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⇒ **non-linear fitting method**

⇒ aim is to have decorrelated signal and noise (commands increments w.r.t. turbulence increments)

$$f_{N,\Delta T}(\mathbf{p}) = \frac{1}{2} \sum_{k=1}^N (\delta \mathbf{d}_{k\Delta T} + \mathbf{G}(\mathbf{p}) \cdot \delta \mathbf{a}_{k\Delta T})^T \cdot \mathbf{C}_{\delta \mathbf{z}}^{-1} \cdot (\delta \mathbf{d}_{k\Delta T} + \mathbf{G}(\mathbf{p}) \cdot \delta \mathbf{a}_{k\Delta T})$$

$$\mathbf{p}^* = \text{Arg min}_p f_{N,\Delta T}(\mathbf{p})$$



recorded AO data history

recorded AO commands history



★ 2. Criterion for parameters identification

b) Criterion for parameters estimation

$$\mathbf{d}_k = -\mathbf{G} \cdot \mathbf{a}_k + \mathbf{z}_k$$



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main difficulties :

- o on-sky **calibration** approaches do not apply, since we do not choose the applied commands
- o very good AO correction mainly means poor signal-to-noise ratio for identification



★ 2. Criterion for parameters identification

b) Criterion for parameters estimation

$$\mathbf{d}_k = -\mathbf{G} \cdot \mathbf{a}_k + \mathbf{z}_k$$



$$\begin{aligned} \delta \mathbf{d}_k &= \mathbf{d}_{k+1} - \mathbf{d}_k \\ &= -\mathbf{G} \cdot \delta \mathbf{a}_k + \delta \mathbf{z}_k \end{aligned}$$

⇒ **non-linear fitting method**

$$f_{N,\Delta T}(\mathbf{p}) = \frac{1}{2} \sum_{k=1}^N (\delta \mathbf{d}_{k\Delta T} + \mathbf{G}(\mathbf{p}) \cdot \delta \mathbf{a}_{k\Delta T})^T \cdot \mathbf{C}_{\delta \mathbf{z}}^{-1} \cdot (\delta \mathbf{d}_{k\Delta T} + \mathbf{G}(\mathbf{p}) \cdot \delta \mathbf{a}_{k\Delta T})$$

advantages:

- weighting $\mathbf{C}_{\delta \mathbf{z}}^{-1}$ is much closer to a diagonal matrix (even if 2 times the measurement noise level)
- $\delta \mathbf{z}_{k\Delta T}$ and $\delta \mathbf{a}_{k\Delta T}$ not much correlated (because AO correction delay)
- But is $\mathbf{G}(\mathbf{p}) \cdot \delta \mathbf{a}_{k\Delta T}$ informative enough?



3. Numerical results

a) Simulations

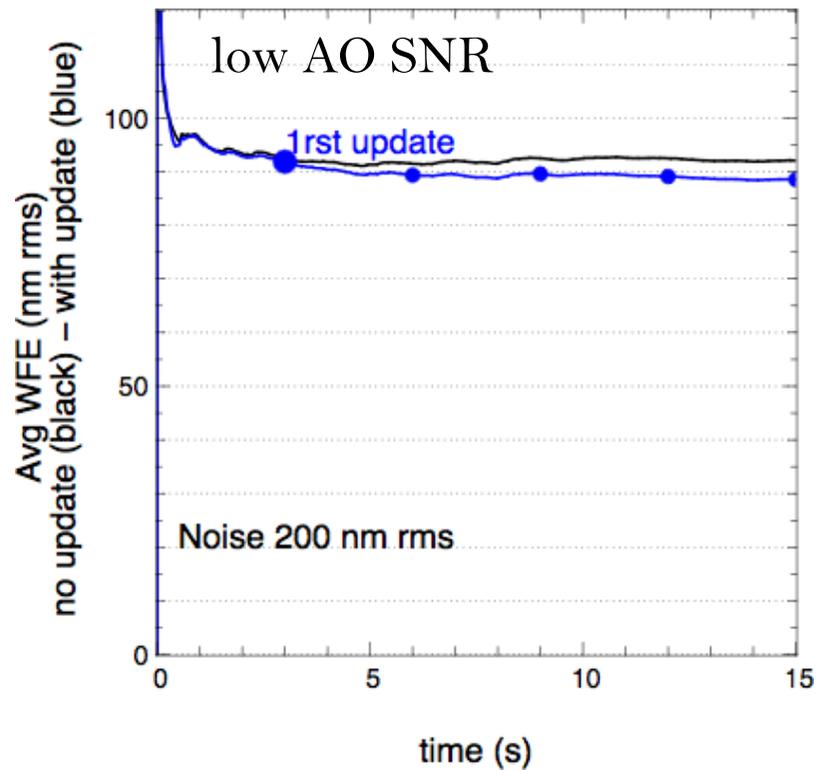
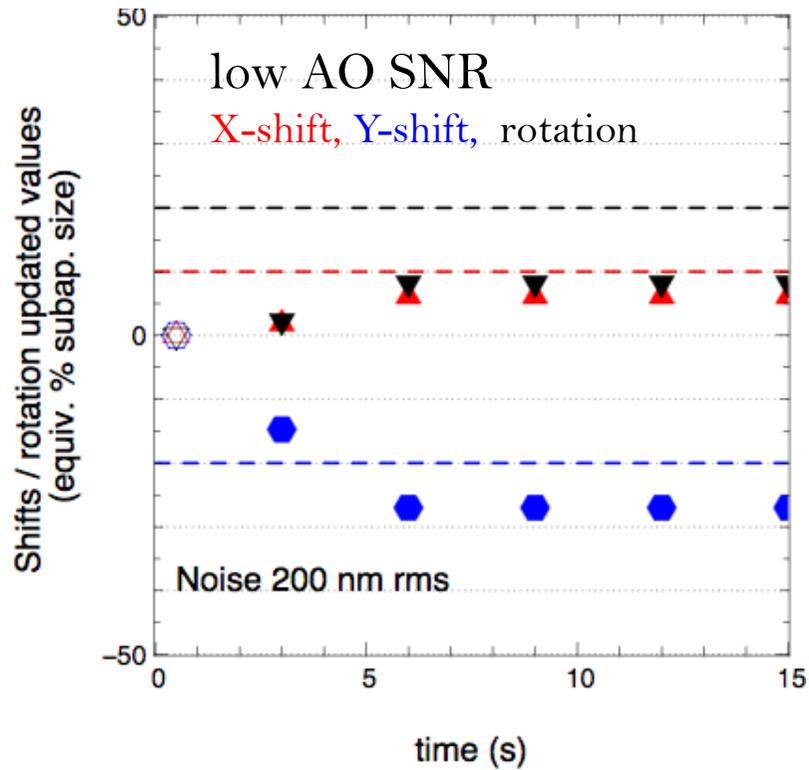
- o first checks with simulations of a 40x40 « square aperture » SCAO
 - parameters = 2D-shifts and rotation angle
 - $\Delta T = 23$ frames
 - $N = 130$
 - residuals do not include information from subapertures on the edges
 - Levenberg-Marquardt algorithm modified with « Trust region » approach
 - 2 iterations
 - large # residuals but computations are easy to highly parallelize

$$f_{N,\Delta T}(\mathbf{p}) = \frac{1}{2} \sum_{k=1}^N (\delta \mathbf{d}_{k\Delta T} + \mathbf{G}(\mathbf{p}) \cdot \delta \mathbf{a}_{k\Delta T})^T \cdot \mathbf{C}_{\delta z}^{-1} \cdot (\delta \mathbf{d}_{k\Delta T} + \mathbf{G}(\mathbf{p}) \cdot \delta \mathbf{a}_{k\Delta T})$$



3. Numerical results

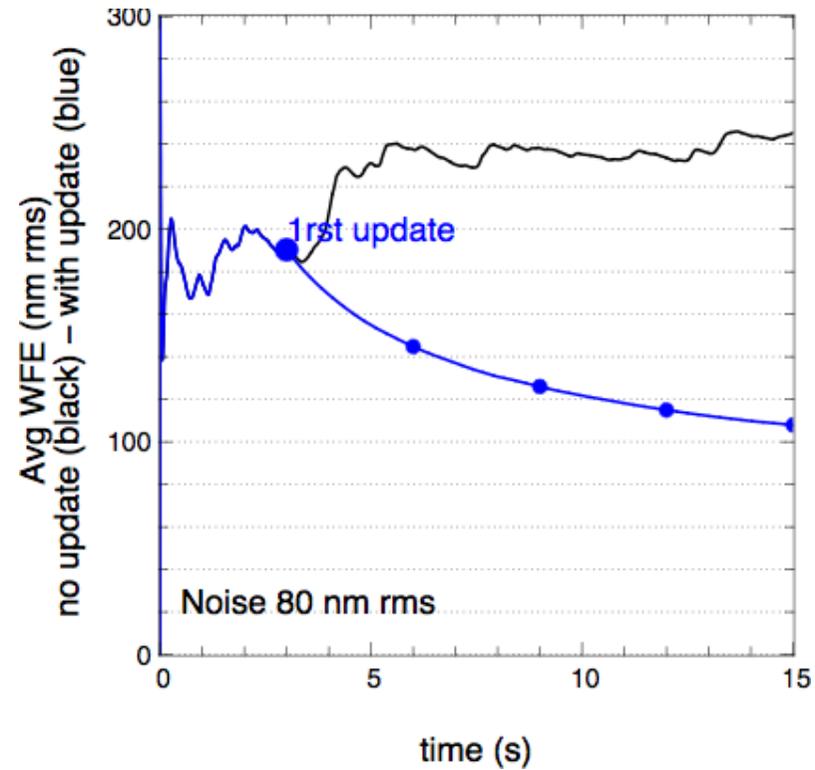
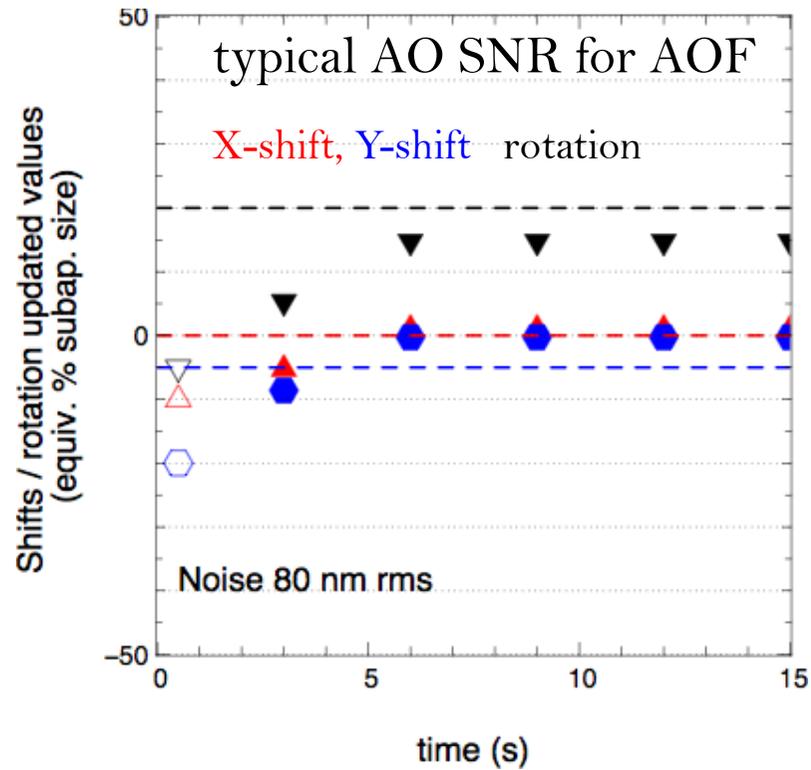
- ✓ medium initial misregistrations
- ✓ convergence of the method
- ✓ improved parameters knowledge by identification
- ✓ remaining error related to the threshold chosen for effective update of the CM
- ✓ small benefit observed in terms of WFE (measurement noise dominates)





3. Numerical results

- ✓ very large initial misregistrations
- ✓ requires 0.1 integrator gain to close the loop
- ✓ convergence of the method
- ✓ stabilization of the AO correction
- ✓ significant benefit in terms of WFE

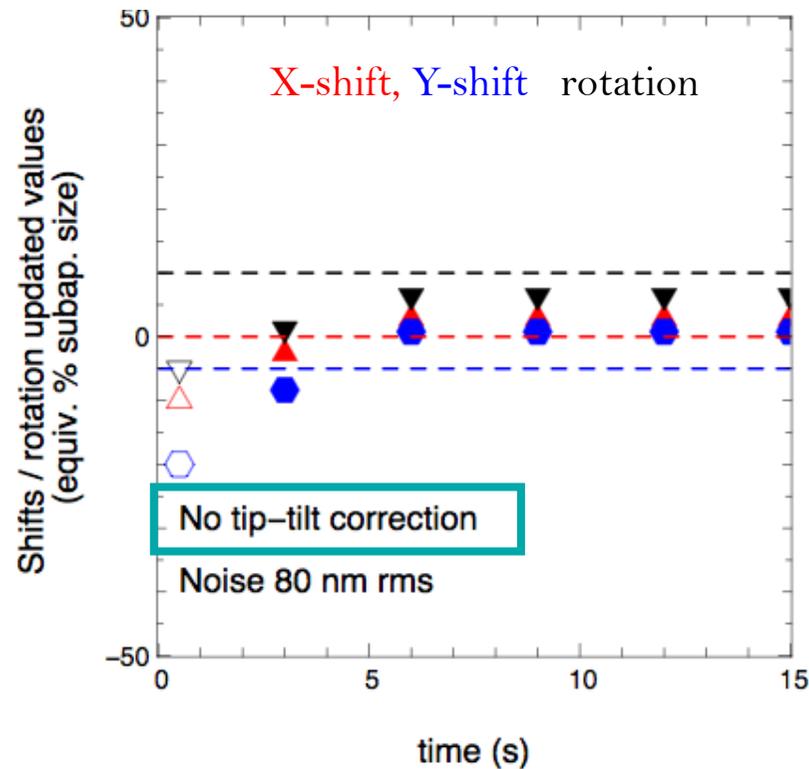




3. Numerical results

a) Simulations

- o first checks with simulations of a 40x40 « square aperture » SCAO
 - estimation accuracy => ongoing theoretical analysis
 - closed-loop CM update after parameters identification
 - LGS-related issue: not considering the tip-tilt correction

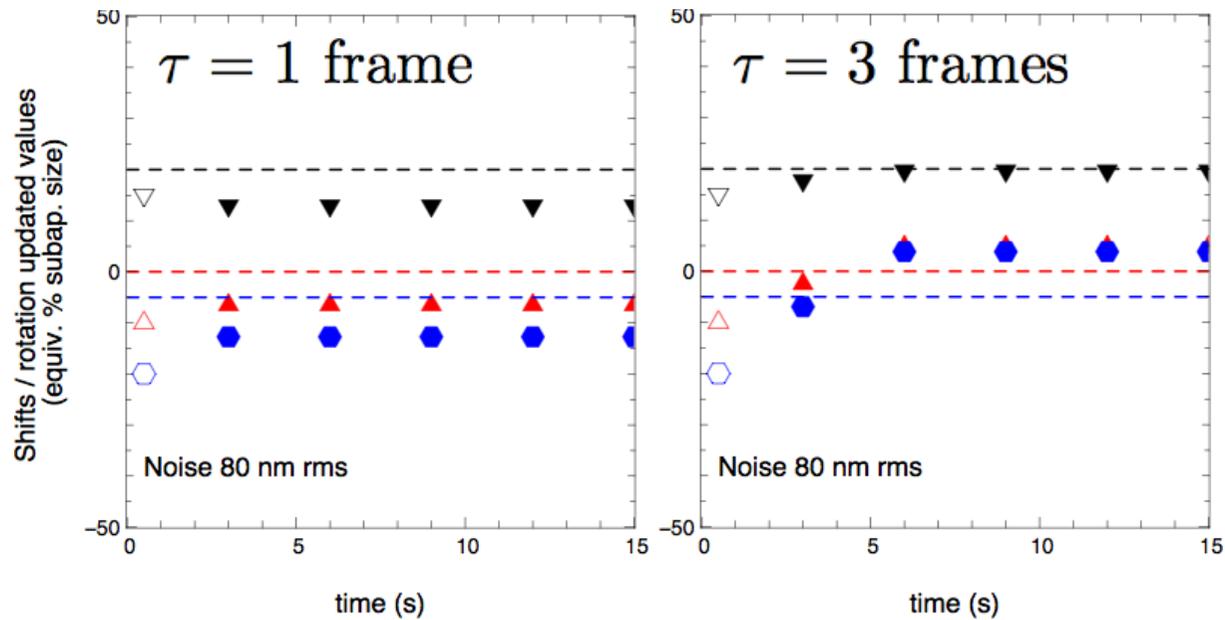




3. Numerical results

a) Simulations

- o first checks with simulations of a 40x40 « square aperture » SCAO
 - estimation accuracy => ongoing theoretical analysis
 - closed-loop CM update after parameters identification
 - LGS-related issue: not considering the tip-tilt correction
 - robustness to delay assumption error



estimation changes.
Which diagnostic?

=>the criterion is
minimum for the true
delay value



3. Numerical results

b) on-going and future work

- simulations of AOF exact system (influence functions of its DM, no Fried geometry)
- confirm parallel theoretical analysis on estimation accuracy, w.r.t
 - atmospheric conditions,
 - noise level,
 - control parameters,
 - fitting parameters (ΔT , N , # of iterations, ...)
- multi-LGS identification (LTAO & GLAO of the AOF)



Conclusion

- ✓ global estimation approach to identify system misregistrations during closed-loop AO-corrected observations
- ✓ goals driven by AOF requirements
 - ✓ estimation accuracy ($\sim 10\%$ subaperture size for shifts)
 - ✓ numerical update of the CM when large misregistrations appear
- ✓ originality & assets
 - ✓ no additional disturbance \Rightarrow fully exploits AO data
 - ✓ computationally reasonable
 - ✓ can handle undetermined TT (low-orders) from LGS
 - ✓ robustness to uncertainty on AO-correction delay
- ✓ results
 - ✓ accuracy theoretical analysis on-going
 - ✓ first promising simulations results (& some NAOS data analysis)