

MOAO real-time LQG implementation on CANARY

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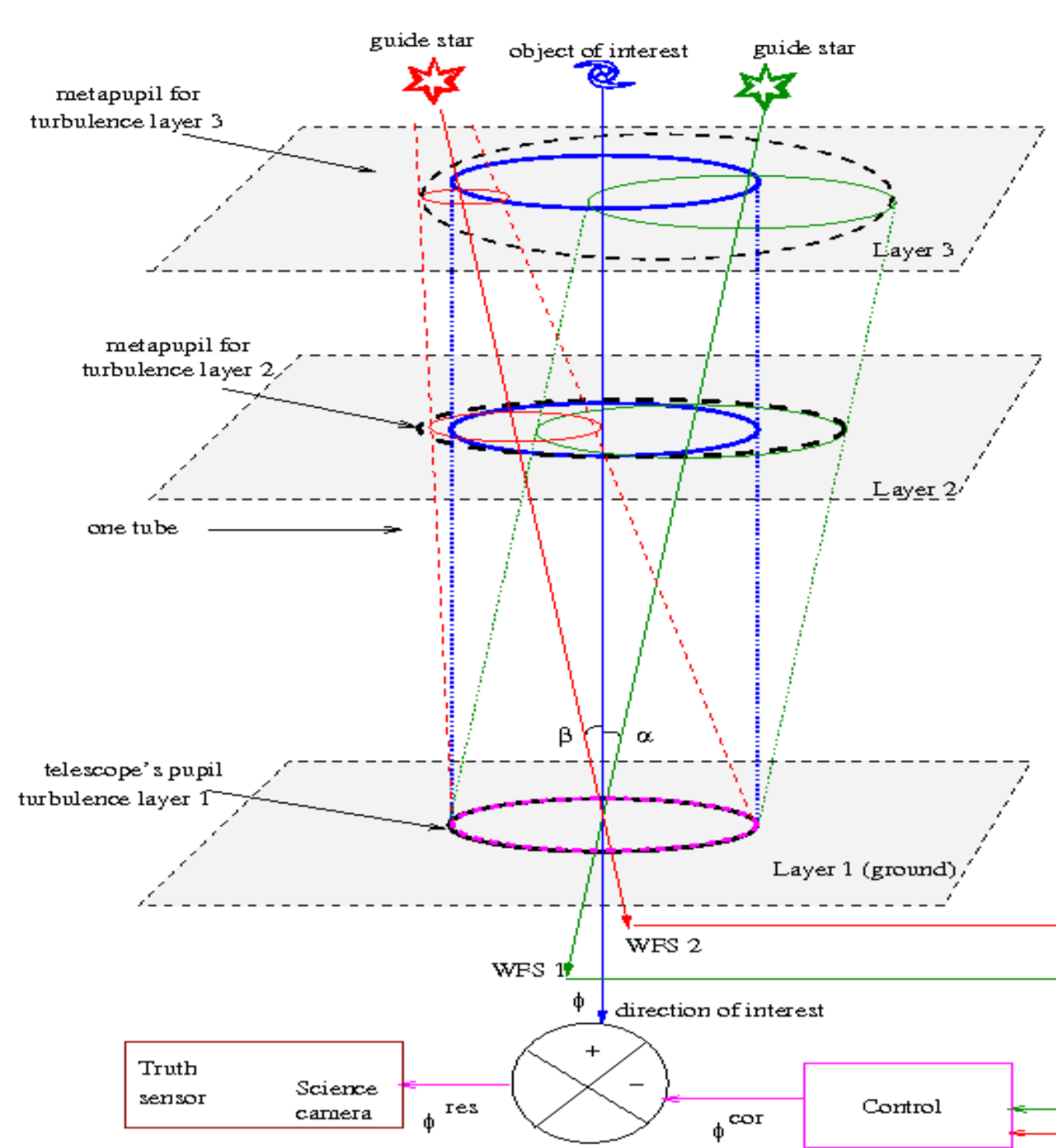
CONTEXT

Advances on LQG AO control

- LQG control accounts for **spatial and temporal priors** (expressed through explicit models) [1]
 - Better performance
 - Better stability and sensitivity to noise
- Flexible control**
 - SCAO and WFAO, open and closed loop config [2]
 - Turbulence + Vibrations rejection [3]
- Woofers-tweeters** [4]
- Large degrees of freedom** [5], [6]
- Experimental demonstrations**
 - Vibration filtering [7]
 - WFAO validation on WFAO testbed HOMER [8]
- Canary phase B constitute a perfect framework for the first LQG control on-sky demonstration**



Tomographic AO description



$$y^{ngs,i} = D^{ngs,i} M^{ngs,i} \phi + w^{ngs,i}$$

NGS WFS matrix

$$y^{lgs,i} = D^{lgs,i} M^{lgs,i} \phi + w^{lgs,i}$$

LGS WFS matrix (deals with Tip/Tilt indetermination)

$$\phi = M^{obj} \phi$$

Resulting turbulent phase
On-axis cylindrical projection

$$\phi^{cor} = N u$$

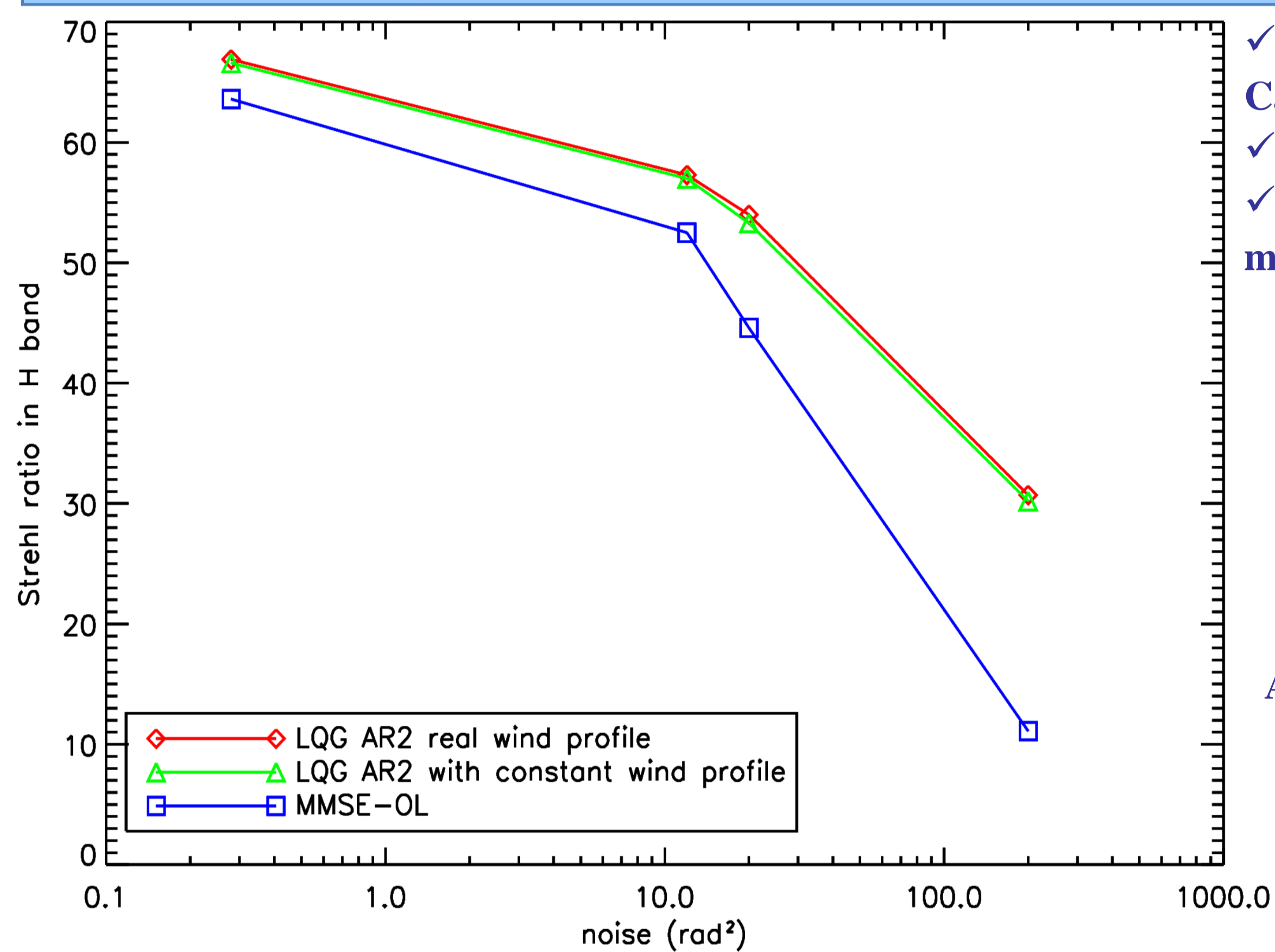
Correction phase
Influence matrix

$$\phi^{res} = \phi - \phi^{cor}$$

Residual phase

LQG MODELING AND SIMULATION RESULTS

Canary LQG MOAO Performance: Simulation results



- Perfect prior on System Calibration & Cn2 profile
- No vibrations
- Strehl Ratios for different modelisations:
 - LQG AR2 derived from true wind profile, Wind = [7.5, 12.5, 15] m/s
 - LQG AR2 derived from mean wind profile, Wind = [11.2, 11.2, 11.2] m/s
 - MMSE-OL
- As a function of: WFS Noise

LQG Control & Disturbance model

- Optimality criterion : minimum variance of the residual phase
- Solution obtained through the Stochastic Separation Principle

$$u_k = P \hat{\phi}_{k+1|k} = (N^T N)^{-1} N^T \hat{\phi}_{k+1|k}$$

$$\hat{\phi}_{k+1|k} = E[\phi_{k+1} | \mathcal{Y}_k] \text{ where } \mathcal{Y}_k = \{ y_k, y_{k-1}, \dots, y_0 \}$$

conditional expectation → Kalman filtering with explicit dynamical models

- Disturbance = Turbulence + Vibrations $\phi_k = M^{obj} \phi_k^{tur} + \phi_k^{vib}$
- Vector-valued auto-regressive (AR) processes (of order 2 here)
- We can write this model for a given turbulent layer as:

$$\phi_{k+1}^{tur,i} = A_1^{tur,i} \phi_k^{tur,i} + A_2^{tur,i} \phi_{k-1}^{tur,i} + v_k^{tur,i}$$

- Same model used for the vibrations with resonance peak [4]

$$\phi_{k+1}^{vib,i} = A_1^{vib,i} \phi_k^{vib,i} + A_2^{vib,i} \phi_{k-1}^{vib,i} + v_k^{vib,i}$$

Parameters identified off-line
Separation between Tip/Tilt & High Order in the modeling

LQG control robust to wind profile uncertainties
Good performance in low SNR thanks to spatial & temporal priors
Significant improvement compared to static MMSE reconstructor

LQG INTERFACE STRUCTURE ON CANARY

$$x_k = \begin{pmatrix} \phi_k^{tur} \\ \phi_{k-1}^{tur} \\ \phi_k^{vib} \\ \phi_{k-1}^{vib} \end{pmatrix} \begin{cases} x_{k+1} = A x_k + \Gamma v_k & \text{Evolution model} \\ y_k = C x_k + w_k & \text{Measurement model} \\ \phi_k = C_\phi x_k & \text{Performance output} \end{cases}$$

$$\begin{cases} \hat{x}_{k|k} = \hat{x}_{k|k-1} + H_\infty (y_k - C \hat{x}_{k|k-1}) & \text{Update equation} \\ \hat{x}_{k+1|k} = A \hat{x}_{k|k} & \text{Predictive equation} \end{cases}$$

STYC	
System and Turbulence identification	Learn of Learn & Apply [9]
Dynamical parameters identification	Turbulence + Vibrations
Model construction	Kalman gain computation
Computation of M_1, M_2, M_3, M_4	Consistency real-time indicators

Vector	Format	Size
x_k	n_x	$2(410 + 20) = 860$
y_k	n_y	$8 \times 2 \times 72 = 1152$
u_k	n_u	$52 + 4 = 56$

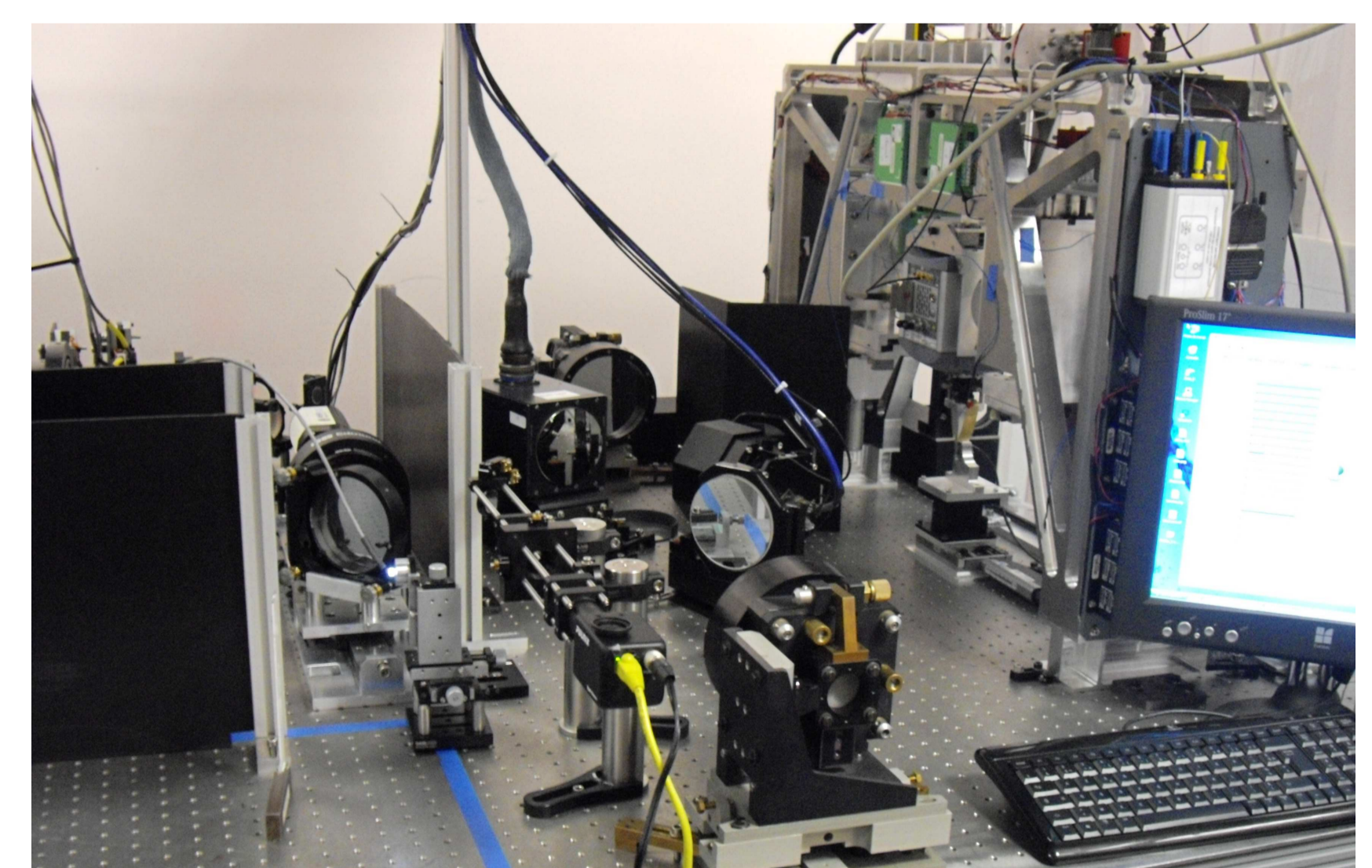
Upload Matrices

 M_1, M_2, M_3, M_4

DARC

$$\begin{cases} x_{k+1} = M_1 x_k + M_2 (y_k + M_3 u_{k-2}) \\ u_{k+1} = M_4 x_{k+1} \end{cases}$$

$$\begin{cases} M_2 = A H_\infty \\ M_1 = A - M_2 C \\ M_3 = 0 \\ M_4 = (N^T N)^{-1} N^T C_\phi \end{cases}$$



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PERSPECTIVES

- ◆ LQG interface is under development
- ◆ On-lab tests planned during Winter 2011 & Spring 2012.
- ◆ On-sky validation expected during Summer 2012

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