

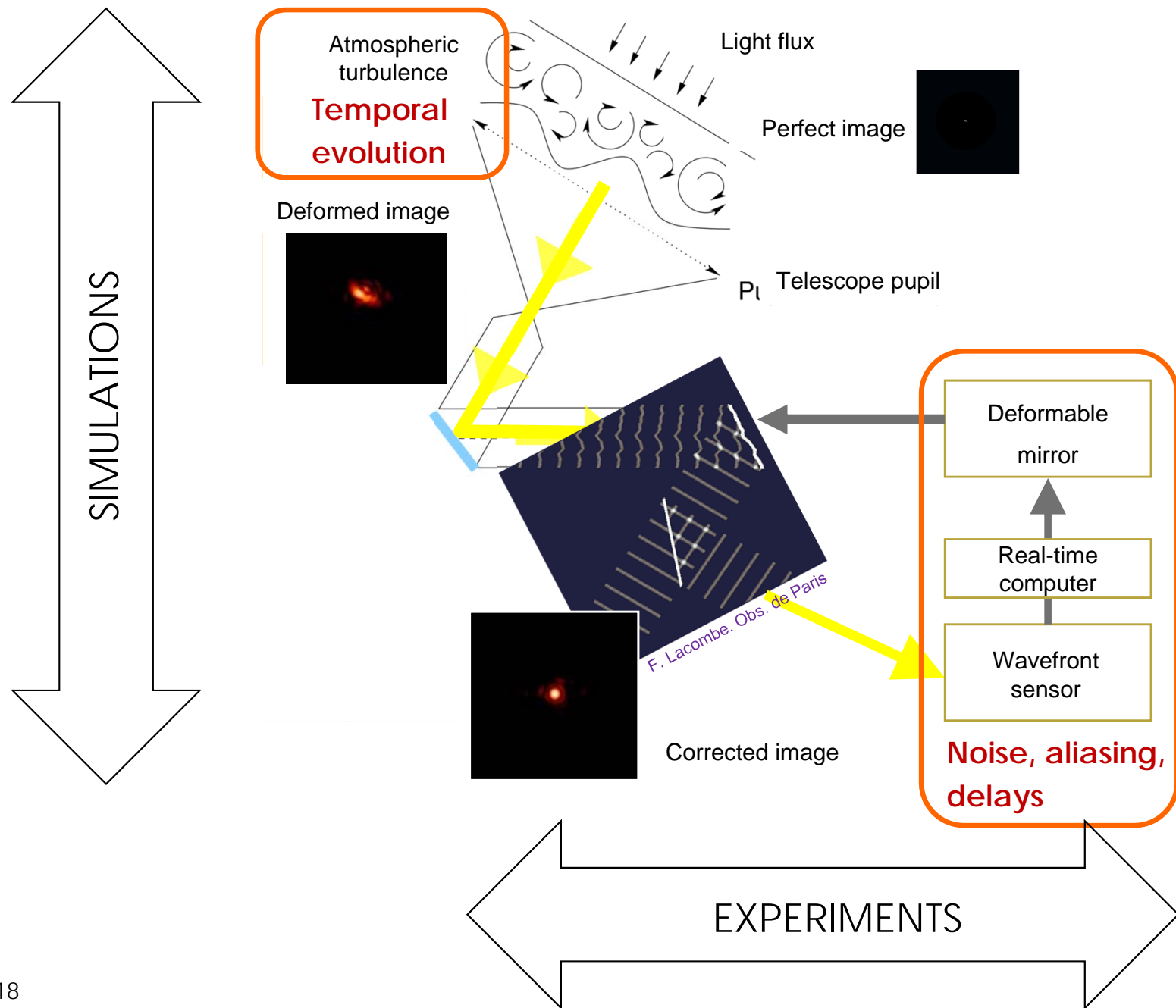
A few words about AO control performance evaluation

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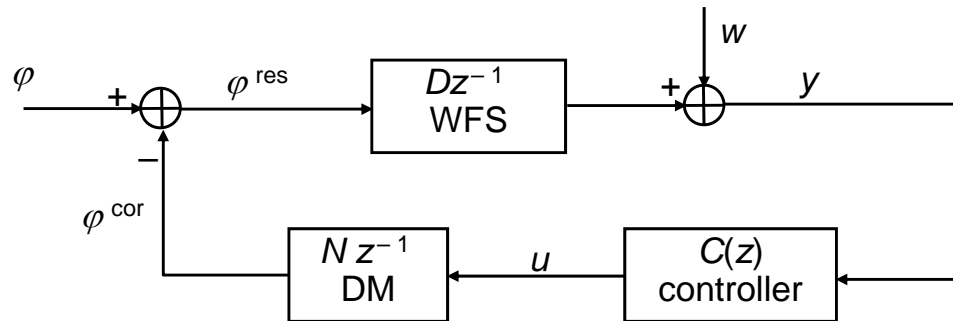
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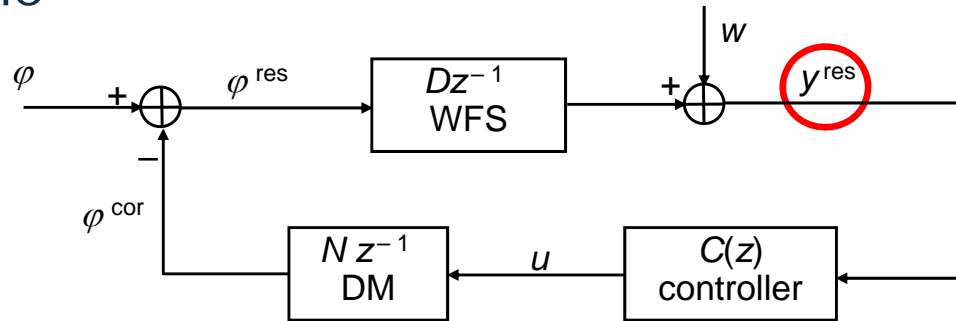
- Real system or replay from WFS data
 - equivalent block-diagram representations
- The scalar case: noise propagation and performance
- Performance evaluation in the multivariable case
- Why it fails, what we can do

- The control engineer point of view



- WFS: Wave Front Sensor (Shack-Hartmann, pyramid...)
- DM: Deformable Mirror
- Controller: u applied with a zero-order-hold
 \Rightarrow constant over one frame (WFS integration time)

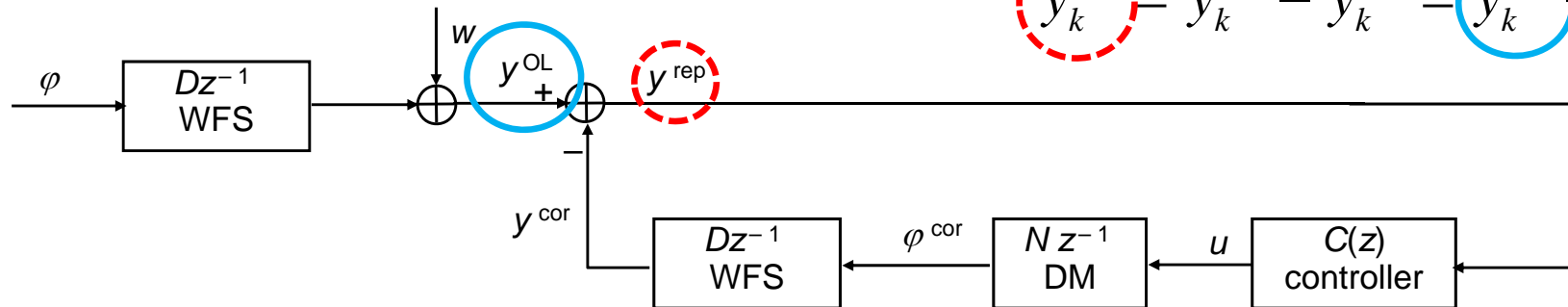
- On-sky scheme



$$y_k^{\text{res}} = D\varphi_{k-1}^{\text{res}} + w_k = y_k^{\text{OL}} - D\varphi_{k-1}^{\text{cor}}$$

- Evaluate control performance in replay

$$y_k^{\text{rep}} = y_k^{\text{OL}} - y_k^{\text{cor}} = y_k^{\text{OL}} - D\varphi_{k-1}^{\text{cor}}$$



- For a scalar loop: $\sigma_{\text{res}}^2 = \frac{\text{Var}(y^{\text{rep}}) - \sigma_w^2}{D^2}$

- Noise propagation

$$\varphi_k^{\text{res}} = -\varphi_k^{\text{cor}} = -Nu_{k-1}$$

$$\sigma_{\text{noise prop}}^2 = N^2 \text{Var}(u)$$

- Example: integrator

$$u_k = u_{k-1} + gy_k^{\text{res}}$$

$$= u_{k-1} - gDNu_{k-2} + gw_k$$

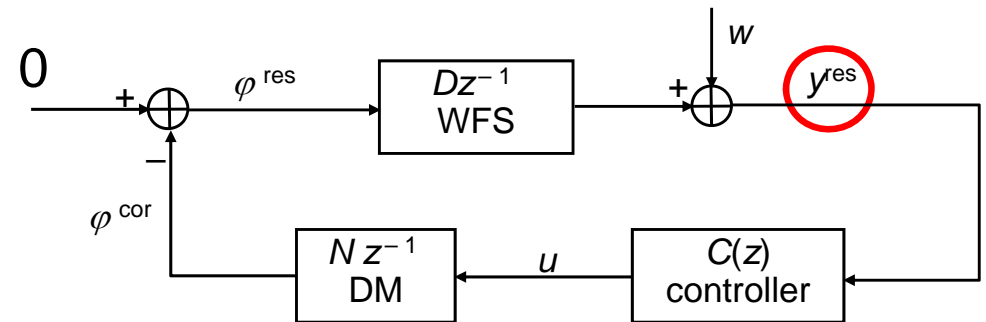
$$X_{k+1} = AX_k + Bw_k$$

$$X_k = \begin{pmatrix} u_{k-1} \\ u_{k-2} \end{pmatrix}, \quad A = \begin{pmatrix} 1 & -gDN \\ 1 & 0 \end{pmatrix}, \quad B = \begin{pmatrix} g \\ 0 \end{pmatrix}$$

Solution to the Lyapunov equation

$$\text{Var}(X) = A \text{Var}(X) A^T + g^2 \sigma_w^2 \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$$

$$\sigma_{\text{noise prop}}^2 = N^2 \text{Var}(X_{1,1})$$



- Noise propagation

- Controller in state-space form

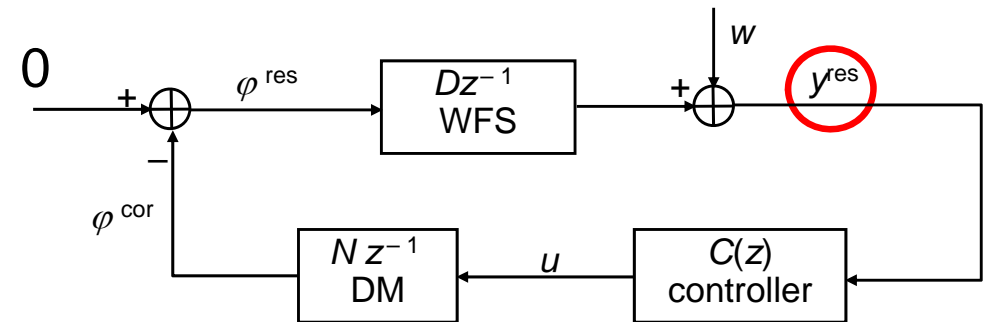
$$X_k^c = A^c X_{k-1}^c + B^c (-DNu_{k-2} + w_k)$$

- Closed-loop formulation

$$X_k = \begin{pmatrix} X_k^c \\ u_{k-2} \end{pmatrix} \Rightarrow X_{k+1} = AX_k + Bw_k$$

$$u_k = PX_k$$

- Variance evaluation (Lyapunov) $\text{Var}(X) = A\text{Var}(X)A^T + B\Sigma_w B^T$



Scalar case

$$\sigma_{\text{noise prop}}^2 = N^2 \text{Var}(X_{1,1})$$

Noise propagation variance

$$\sigma_{\text{noise prop}}^2 = \text{trace}(NP \text{Var}(X) P^T N^T)$$

213 actuators
Noise: 0.2 rad²
 $r_0 = 10 \text{ cm} @ 0.5 \mu\text{m}$
1 turbulent layer
Performance: rad² @ 1.65 μm

- Noise propagated variance
 - Lyapunov: $\sigma_{\text{noise prop}}^2 = 0.097 \text{ rad}^2$
 - Simulations: $\sigma_{\text{noise prop}}^2 = 0.097 \pm 0.001 \text{ rad}^2$

Noise propagation error is easy to estimate
for any linear controller

- Residual phase variance

- Scalar case
$$\sigma_{\text{res}}^2 = \frac{\text{Var}(y^{\text{rep}}) - \sigma_w^2}{D^2}$$

- Reconstructed phase
$$\hat{\phi} = R^{\text{MAP}} y^{\text{OL}} \quad (y^{\text{OL}} = D\phi + w)$$

$$\hat{\phi} = R^{\text{MAP}} D\phi + R^{\text{MAP}} w$$

- Closed-loop formulation

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

$$\hat{\sigma}_{\text{res}}^2 = \text{Var}(\hat{\phi}^{\text{res}}) - R^{\text{MAP}} \Sigma_w (R^{\text{MAP}})^T$$

- Residual phase variance with LQG regulator on 495 Zernike modes

- Simulation of the real system: $\sigma_{\text{res}}^2 = 0.54 \text{ rad}^2$

- Phase reconstruction with 209 Zernike modes: $\hat{\sigma}_{\text{res}}^2 = 0.48 \text{ rad}^2$

- Phase reconstruction with 495 Zernike modes: $\hat{\sigma}_{\text{res}}^2 = 0.29 \text{ rad}^2$

213 actuators (15x15)

Noise: 0.2 rad^2

$r_0 = 10 \text{ cm} @ 0.5 \mu\text{m}$

1 turbulent layer

Performance:

$\text{rad}^2 @ 1.65 \mu\text{m}$

MAP phase estimation does not allow for
error budget evaluation

- Residual phase variance with LQG regulator on 495 Zernike modes

- Simulation of the real system: $\sigma_{\text{res}}^2 = 0.54 \text{ rad}^2$

Variance
increase of 37 %

- Phase reconstructed with 209 Zernike modes: $\hat{\sigma}_{\text{res}}^2 = 0.48 \text{ rad}^2$

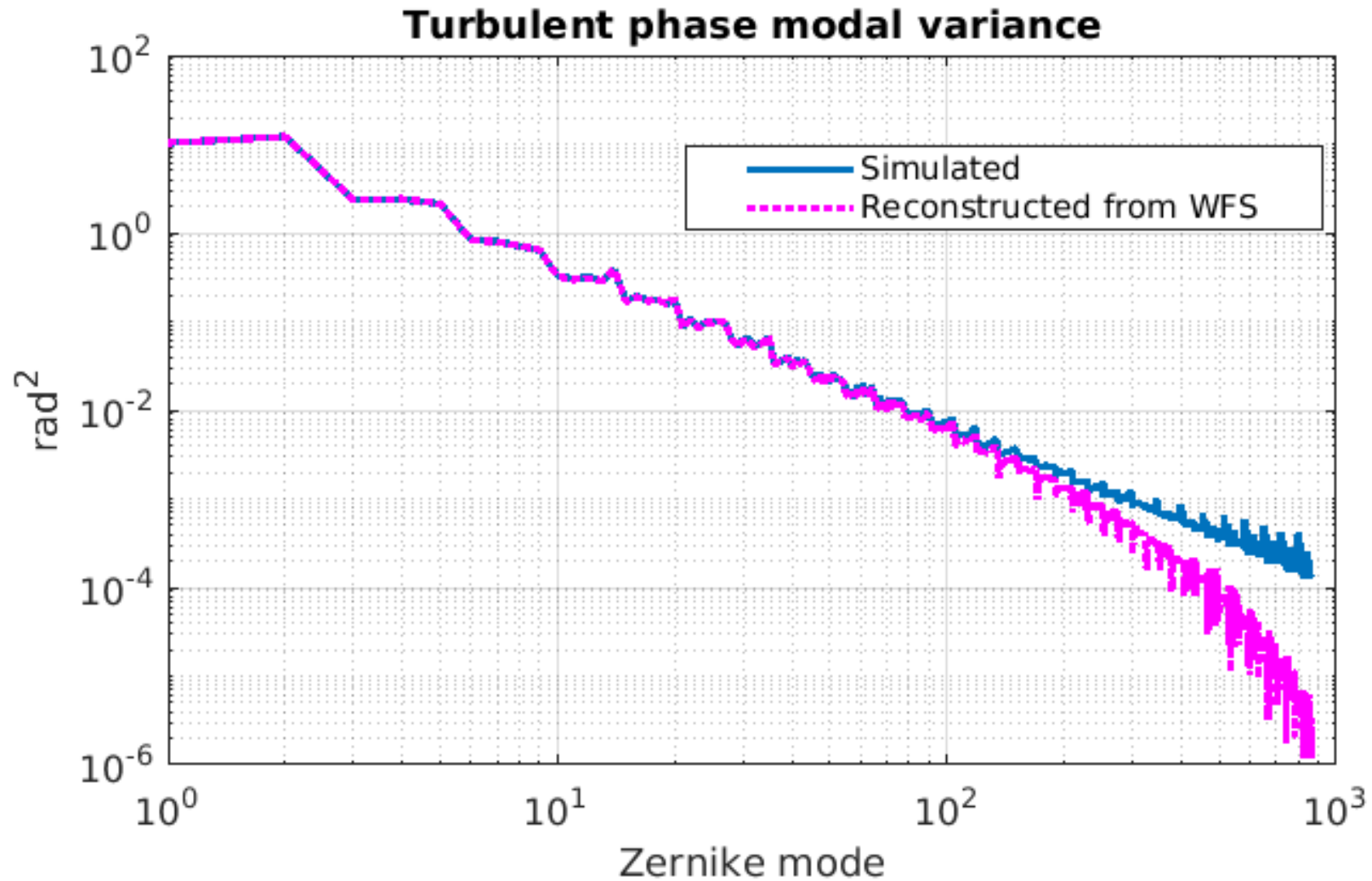
- Residual phase variance with integral action regulator

- Simulation of the real system: $\sigma_{\text{res}}^2 = 0.74 \text{ rad}^2$

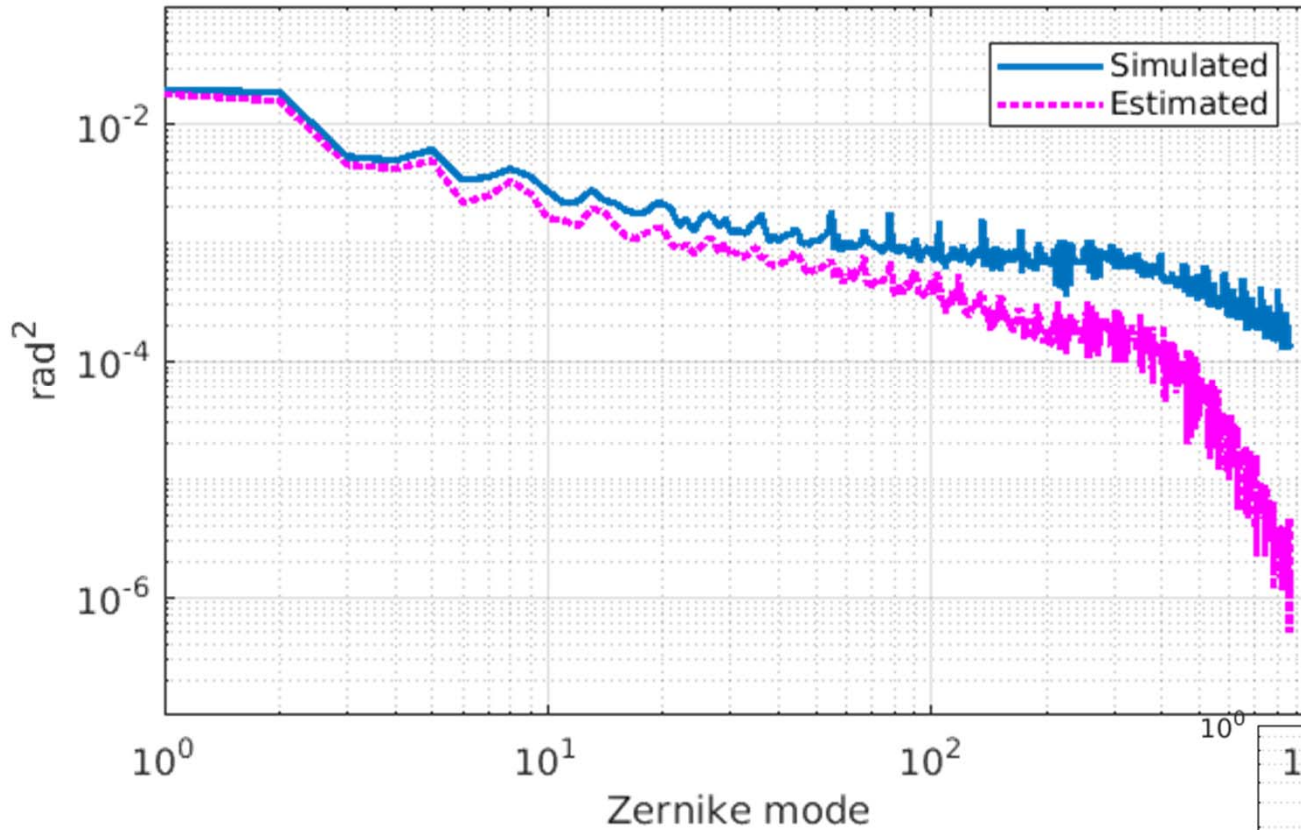
Variance
increase of 17 %

- Phase reconstructed with 209 Zernike modes: $\hat{\sigma}_{\text{res}}^2 = 0.56 \text{ rad}^2$

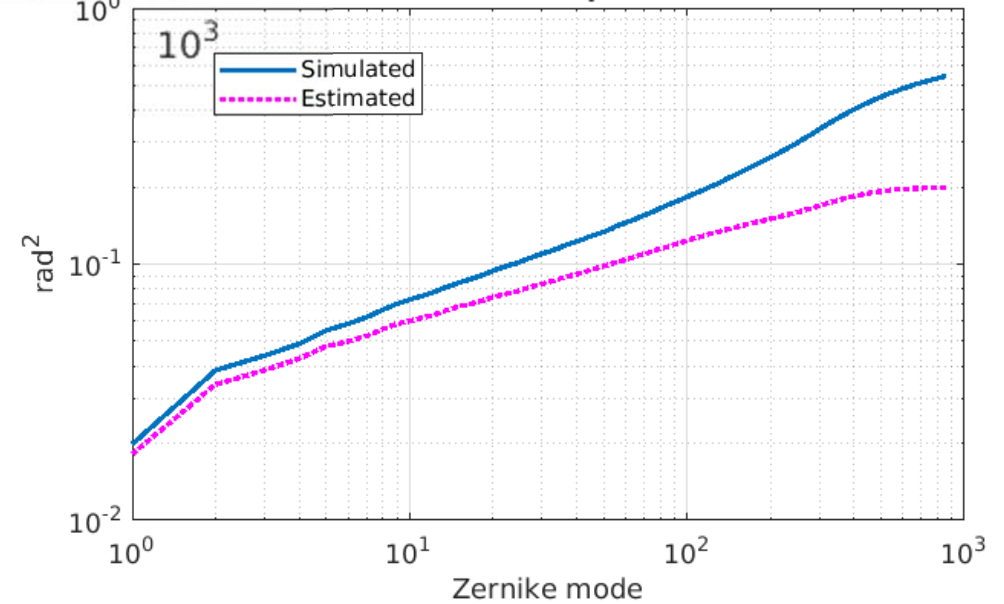
The gap in performance between controllers is not preserved



Residual phase modal variance



Cumulated residual phase variance



- Improve residual phase variance estimation
 - Spatio-temporal priors for high-order modes
 - Taylor+Kolmogorov

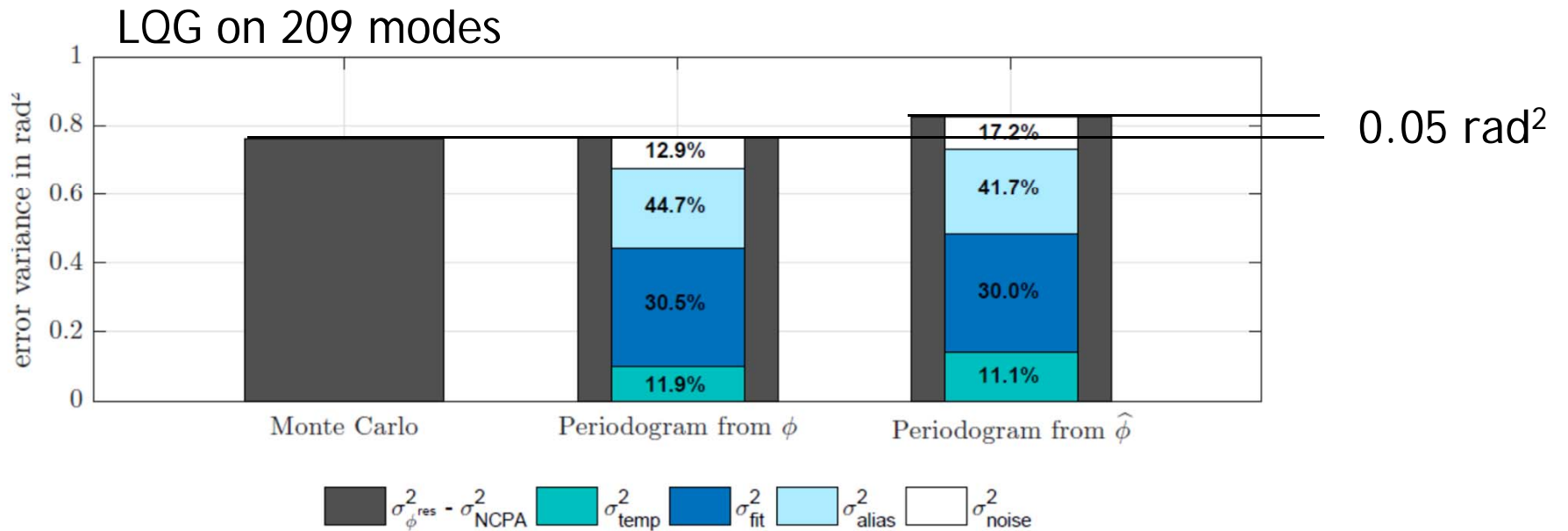
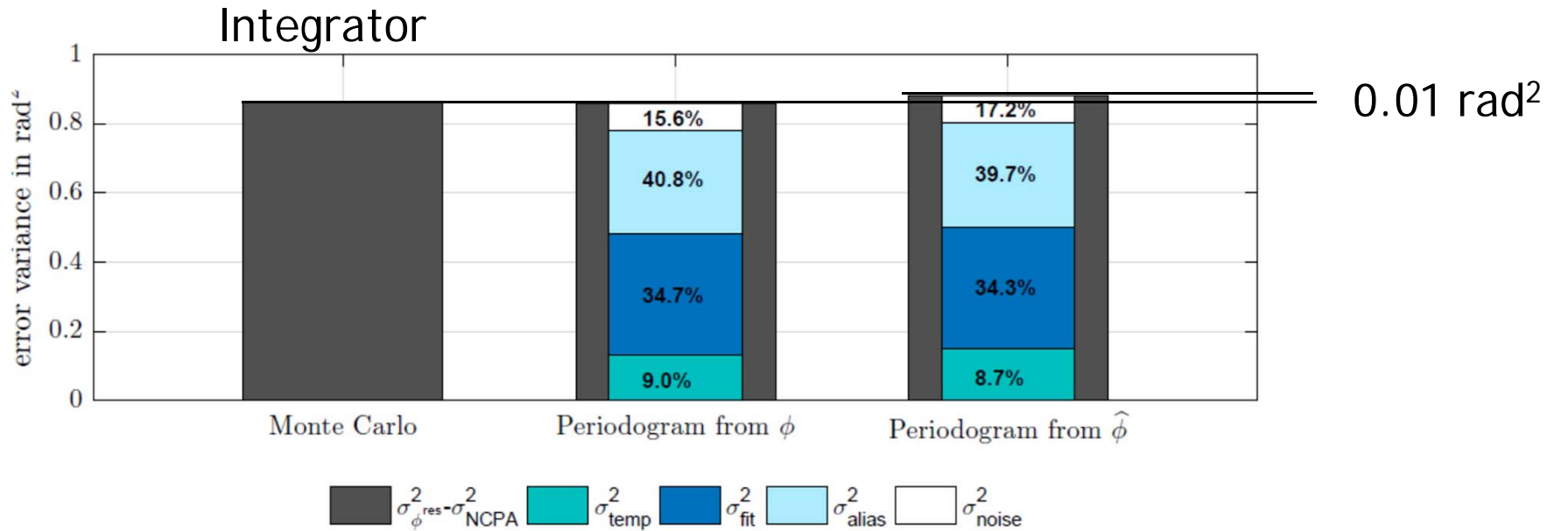
Performance criterion evaluation using Parseval

$$\sigma_{\text{res}}^2 = \text{trace} \frac{1}{\pi} \int_0^\pi T^{\text{rej}} S_\varphi T^{\text{rej}*} d\omega + \underbrace{\text{trace} \frac{1}{\pi} \int_0^\pi T^{\text{noise}} \Sigma_w T^{\text{noise}*} d\omega}_{\sigma_{\text{noise prop}}^2}$$

Power spectral density estimation with extrapolation

$$\hat{S}_\varphi = \left(\begin{array}{c|cc} S_{\hat{\varphi}} & & 0 \\ \hline 0 & \hat{S}_{\bar{n}+1} & 0 \\ & 0 & \ddots \end{array} \right)$$

R. Juvénal & al. Linear controllers error budget assessment for SCAO systems. JOSA A 2018.



- Noise propagation evaluation using state-space formulation is reliable
- Direct performance evaluation from WFS data fails
 - Residual phase variance is badly estimated from residual WFS measurements
- Spatial and temporal priors need to be used for phase PSD estimation
- Phase PSD estimation from WFS data needs priors on high-order modes
- Allows for “control dependent” error terms evaluation in error budget breakdown
- The method is able to cope with linear regulators in SCAO mode

Thank you

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