

Optimal modal Fourier transform wave-front Control

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We want fast and smart algorithms

- § Large systems and high frame rates require fast algorithms:
 - § *Matrix-based methods, which apply the control matrix more quickly*
 - § **Local control, sparse matrix methods, conjugate gradient & multigrid**
 - § *Fourier transform reconstruction (FTR), which filters the slopes*
- § Our controller should use knowledge about the phase aberration and noise to its advantage:
 - § *Can use a priori model of atmosphere and noise*
 - § **MVU, weighted least-squares**
 - § *Can incorporate real-time information about the phase aberration and noise level as determined from telemetry of AO system*

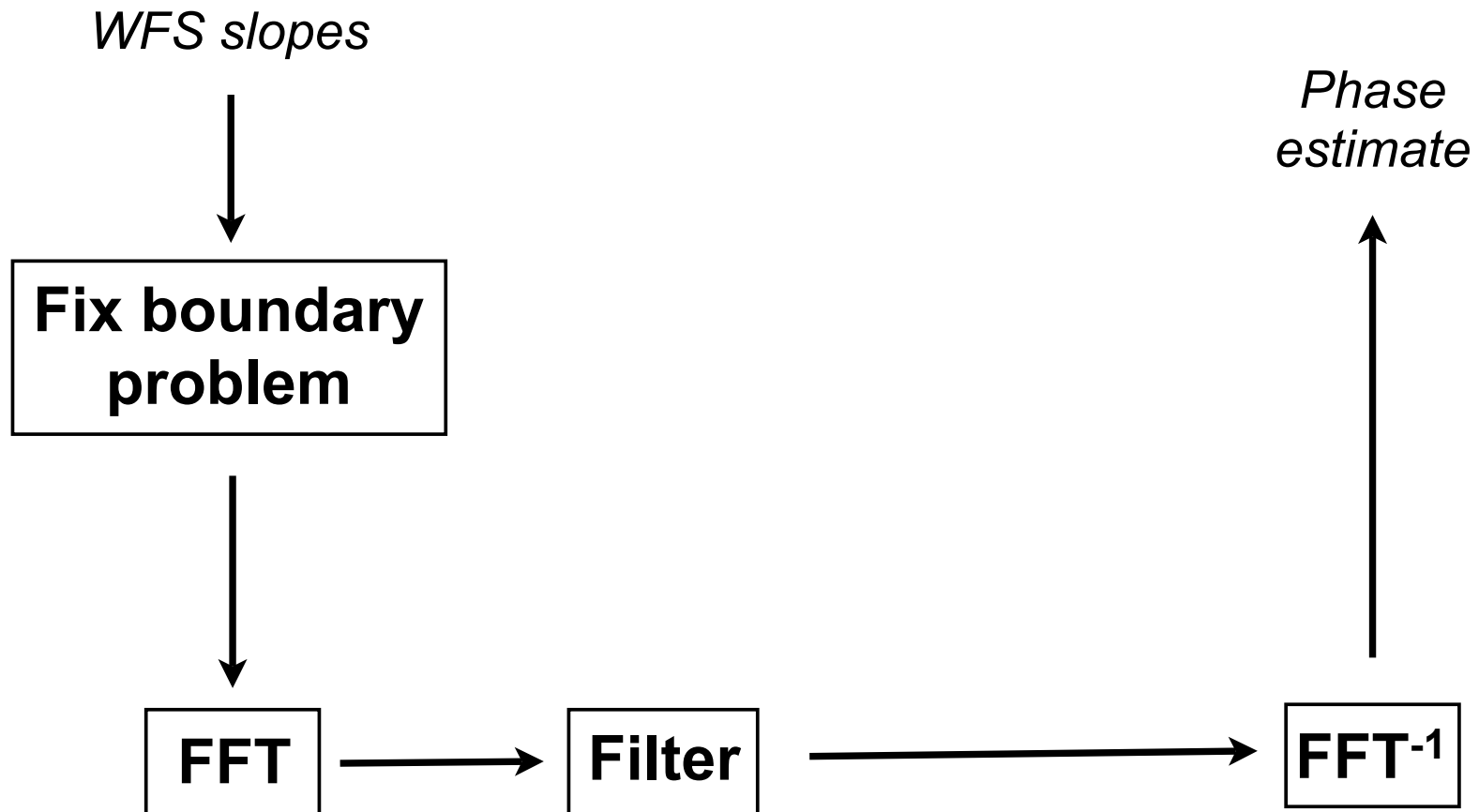
What is modal control?

- § For n actuators, create an orthonormal basis of n functions
 - § *these basis functions are normally chosen to concentrate signal power in few modes*
- § Instead of controlling the phase value at each actuator, we control the amount of each mode present
 - § *this is expressed as the modal coefficient*
- § The AO control problem simplifies to finding the optimal control law for each mode, independently of all the others
- § Explicit modal control with optimization is used in Altair and ESO's NAOS

Optimal Fourier Control is our solution

- § Our modal set is the Fourier basis. This works even on an **arbitrary aperture**.
- § Reconstruction at each time step is with FTR.
- § Closed-loop modal coefficients are used to estimate optimal gains for control law for each mode. Gains are implemented as a filter.
- § Computationally feasible for 64x64 ExAO right now.
- § Extra benefits include
 - § *Modal coefficients are available for free, unlike matrix-based modal control, which requires extra computation.*
 - § *There is a natural relationship between filter structure and PSF structure.*

FTR works by filtering the slopes



The Fourier basis is the modal set

§ FTR uses the DFT on $N \times N$ real signals. This leads to a real cosine and sine ONB with $N/2$ modes:

§ For $[k,l]$ either $[0,0]$, $[0,N/2]$, $[N/2,0]$ or $[N/2,N/2]$, there is only a cosine mode

$$C_{k,l}[m, n] = \frac{1}{N} \cos \left(\frac{2\pi}{N} [km + ln] \right)$$

§ All other frequencies have a sine and cosine

$$C_{k,l}[m, n] = \frac{\sqrt{2}}{N} \cos \left(\frac{2\pi}{N} [km + ln] \right)$$

$$S_{k,l}[m, n] = \frac{\sqrt{2}}{N} \sin \left(\frac{2\pi}{N} [km + ln] \right)$$

Use the DFT to get modal coefficients

§ FTR uses the DFT

$$X[k, l] = \frac{1}{N} \sum_{m=0}^{N-1} \sum_{n=0}^{N-1} x[m, n] \exp\left(\frac{-j2\pi(km + ln)}{N}\right)$$

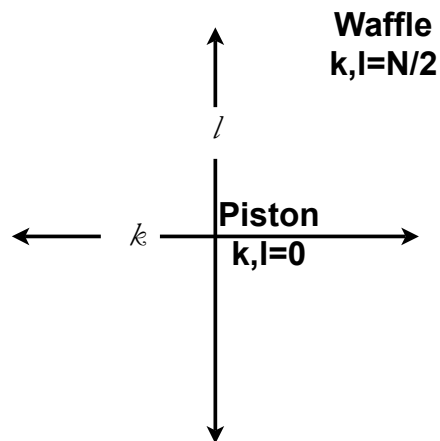
§ Modal coefficients are obtainable directly from the DFT

$$\langle x[m, n], \mathcal{C}_{k,l}[m, n] \rangle = \frac{1}{D_{k,l}} \operatorname{Re} \{ X[k, l] \}$$

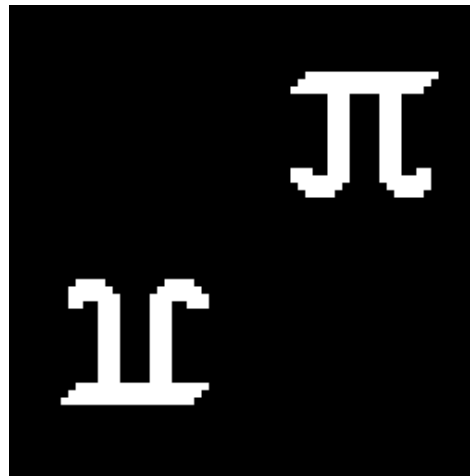
$$\langle x[m, n], \mathcal{S}_{k,l}[m, n] \rangle = \frac{-1}{D_{k,l}} \operatorname{Im} \{ X[k, l] \}$$

Modes correspond to PSF locations

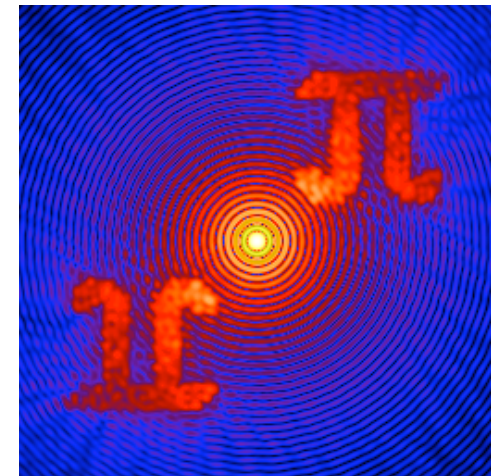
- § Each Fourier mode lives at a specific spatial frequency pair $[k,l]$
- § Because the PSF is approximately the PSD of the residual phase (to second order), each Fourier mode appears at a specific location in the PSF



Frequency domain



PSD/power at each mode



PSF

Modes are eigenfunctions

§ Fourier modes are eigenfunctions of linear, shift-invariant (LSI) systems

§ *The modes for the slopes (on a square aperture) are the same as the modes for the phase*

§ *A cosine of phase at frequency $[k,l]$ produces x- and y-slopes only at the cosine and sine of that frequency $[k,l]$*

§ *Where $M_x[k,l]$ describes the filter which measures the x-slopes from actuator commands*

phase	x-slope
$C_{k,l}[m, n]$	$AC_{k,l}[m, n] + BS_{k,l}[m, n]$
$S_{k,l}[m, n]$	$-BC_{k,l}[m, n] + AS_{k,l}[m, n]$

$$A = \text{Re}\{M_x[k, l]\}, B = -\text{Im}\{M_x[k, l]\}$$

Filter comes from modal responses

§ We simply pseudo-invert the measurement matrix

$$\begin{array}{l} \text{x-slope cosine and sine} \\ \text{y-slope cosine and sine} \end{array} = \begin{bmatrix} A & -B \\ B & A \\ C & -D \\ D & C \end{bmatrix} \begin{array}{l} \text{phase cosine and sine} \end{array}$$

§ And obtain the reconstruction matrix for the phase modes as:

$$\frac{1}{A^2 + B^2 + C^2 + D^2} \begin{bmatrix} A & B & C & D \\ -B & A & -D & C \end{bmatrix}$$

§ This produces an extremely sparse modal control matrix

Filter inverts the measurement process

§ Filter derived from modal coefficients...

$$\hat{P}[k, l] = \frac{(A + jB)X[k, l] + (C + jD)Y[k, l]}{|(A + jB)|^2 + |(C + jD)|^2}$$

§ ... is exactly the same as if we knew the measurement filters

$$\hat{P}[k, l] = \frac{M_x^*[k, l]X[k, l] + M_y^*[k, l]Y[k, l]}{|M_x[k, l]|^2 + |M_y[k, l]|^2}$$

Many filtering options now available

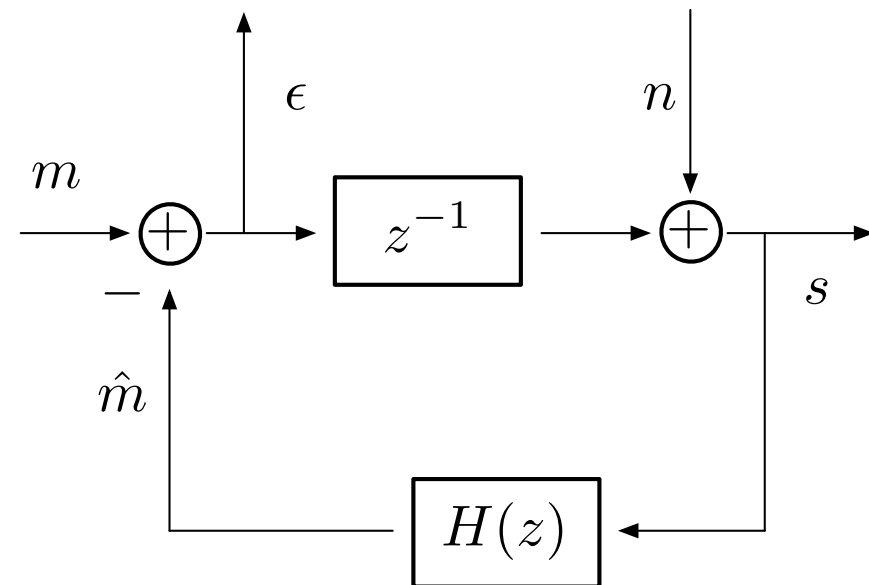
- § Best point-based model filter is Modified-Hudgin
- § Assuming ideal continuous models for WFS and DM, can derive the Ideal filter
 - § *Ideal filter and Mod-Hud very similar*
- § Given any LSI AO system or simulation, we can **measure** the coefficients that describe the modal responses and produce a Custom filter
 - § *captures influence function response of DM*
- § Each filter **exactly reconstructs** given the assumed model, except for invisible modes of piston and sometimes waffle.

We can do all this with an aperture

- § Fourier basis in an arbitrary aperture is a tight Frame that allows analysis and synthesis like an ONB.
- § If we window the data, we can use a fast DFT to get modal coefficients.
- § New method of slope management called edge correction ensures high-quality coefficient estimation by making outside region of phase flat.
- § Result - we get the modal coefficients for free at each time step in the FTR process.

Optimal modal control scheme

- § We follow Altair's implementation and assume an approximate model of control system (exact in simulation case) for each of the independent modes.
- § We control a mode with feedback in the presence of noise.



Block diagram of control loop for a modal coefficient

Optimize the squared-residual error

- § Since the noise at any step is independent of past errors, if we minimize on the measurement s , we minimize on the residual error.
- § If we had perfect knowledge we would minimize

$$\mathcal{J} = \int \left| \frac{1}{1 + \exp(-j\omega)H(\omega)} \right|^2 [M(\omega) + N(\omega)] d\omega$$

- § But we don't... so we have to estimate the open-loop PSD from the closed-loop measurements using

$$\hat{M}(\omega) + \hat{N}(\omega) = |1 + \exp(-j\omega)H_0(\omega)|^2 \hat{S}(\omega)$$

Gain estimation for FTR (1)

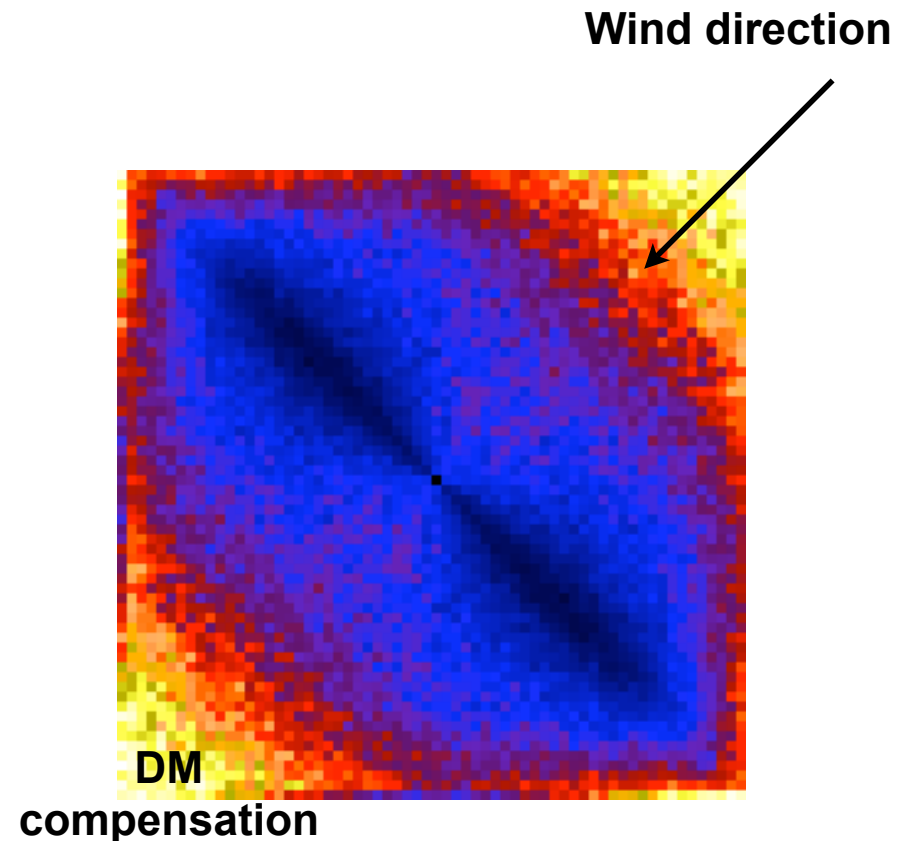
- § From closed-loop telemetry, we estimate the closed-loop measurement PSDs
- § Convert these to open-loop PSD estimates
- § Find the control law which minimizes the error for the sine and cosine modes together

$$\operatorname{argmin} H(z) \left\{ \int \left| \frac{1}{1 + \exp(-j\omega)H(\omega)} \right|^2 |1 + \exp(-j\omega)H_0(\omega)|^2 [\hat{S}_S(\omega) + \hat{S}_C(\omega)] d\omega \right\}$$

§ Where our control law is simple: $H(z) = \frac{g}{1 - cz^{-1}}$

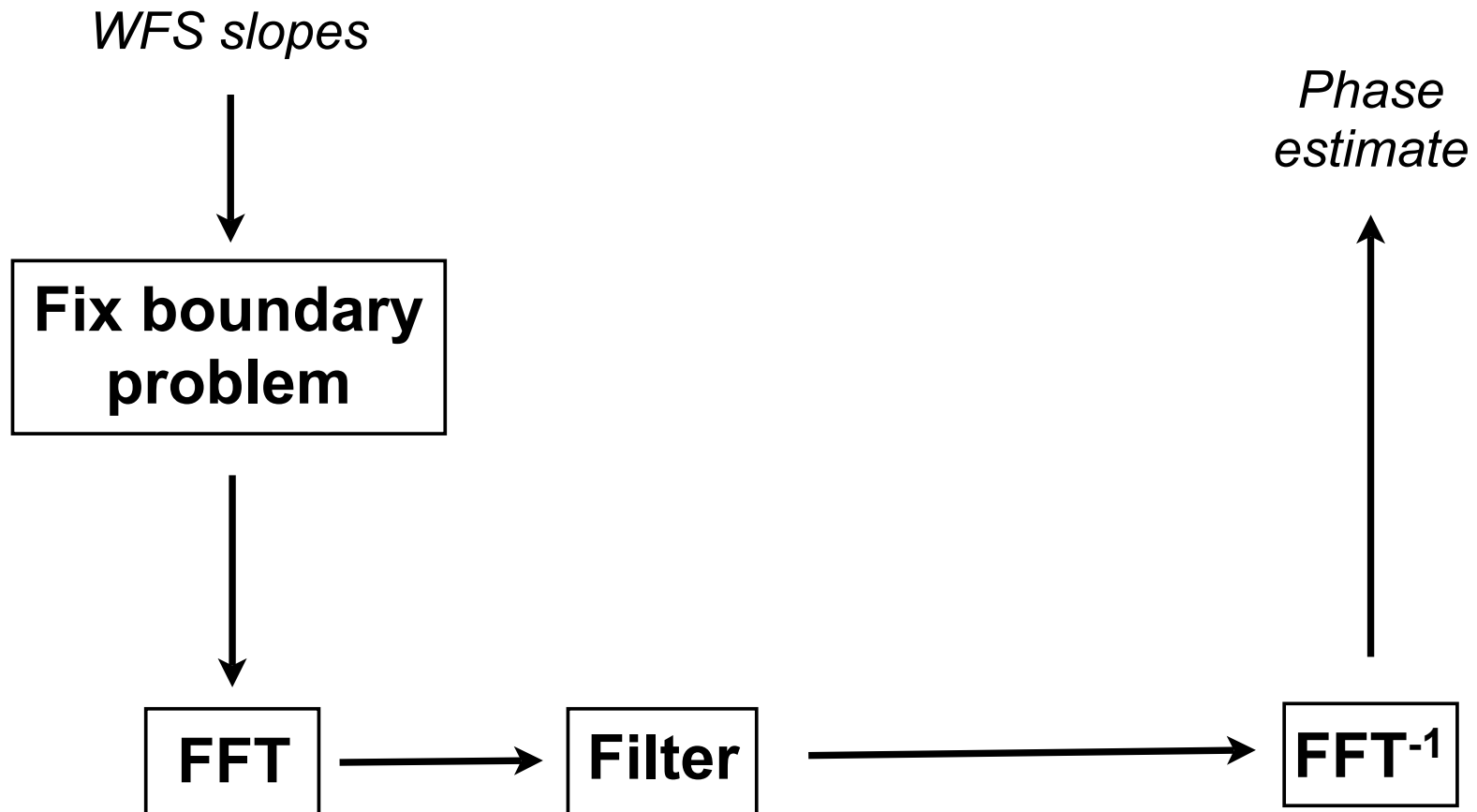
Gain estimation for FTR (2)

- § For a single variable (gain g) we can solve the optimization problem efficiently.
- § At each frequency $[k,l]$ we have a gain - we construct the filter of these gains using Hermitian symmetry. This filter is then incorporated into the reconstruction filter.

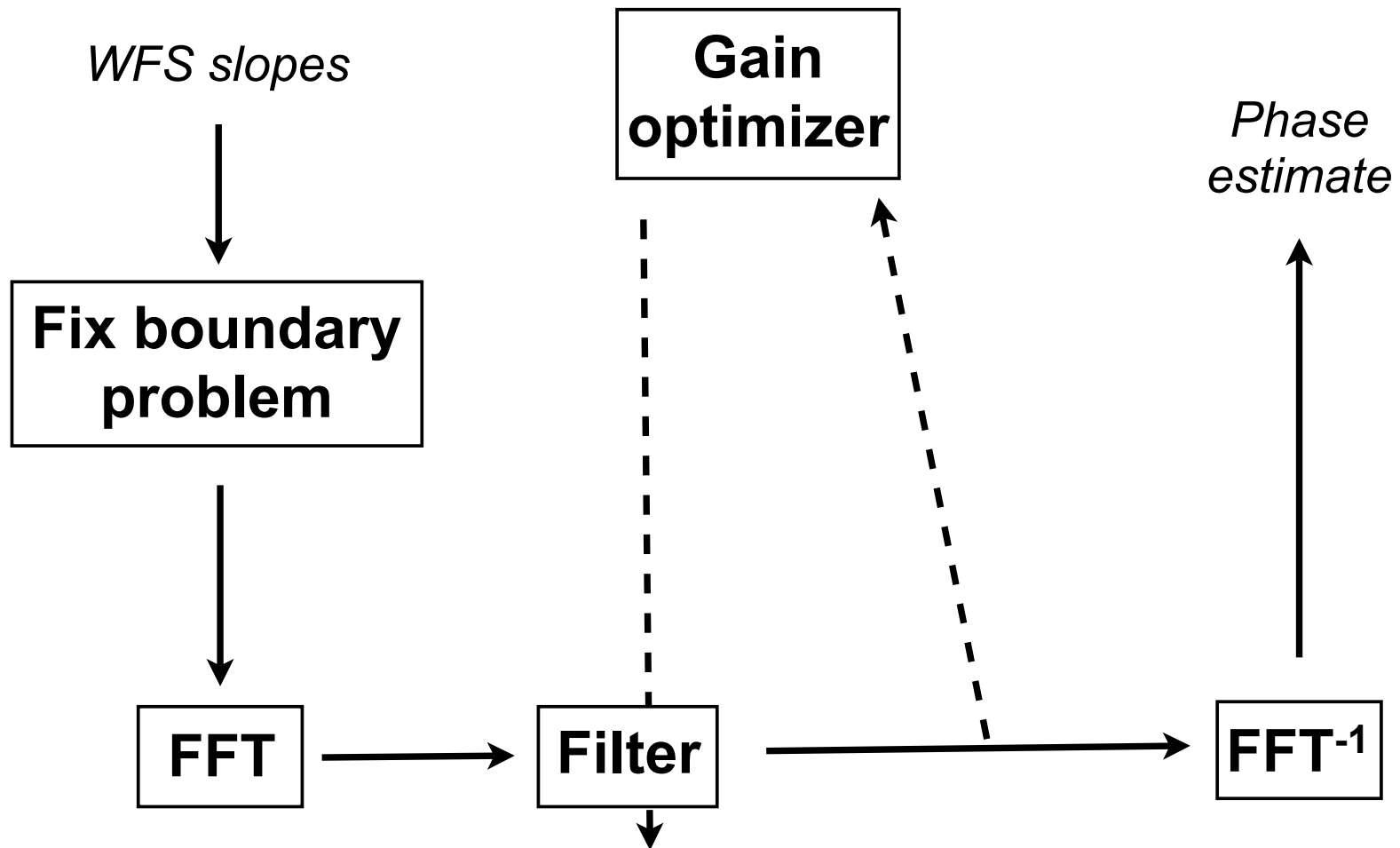


Example filter, $N=64$

Gains are incorporated into filter



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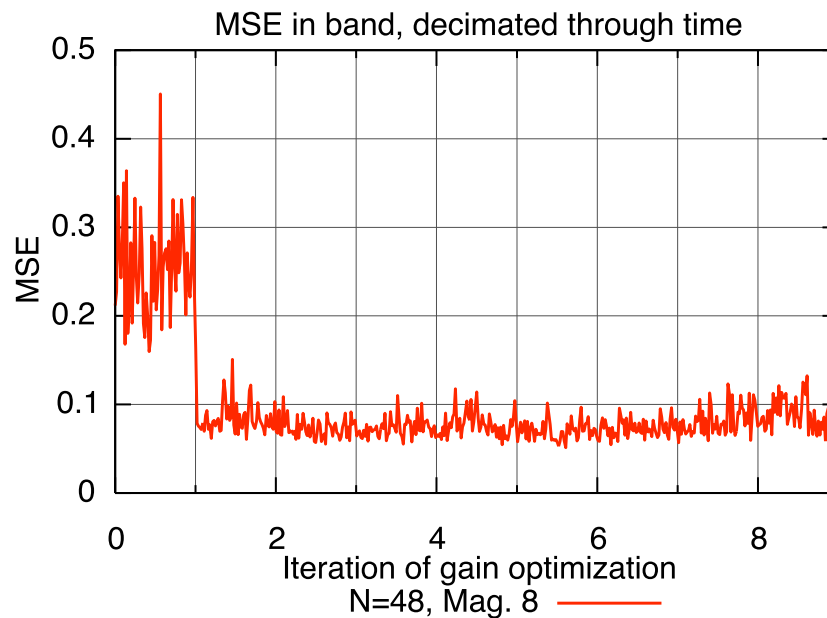


Details of end-to-end ExAOC simulation

§ Features of ExAOC simulation include:

- § *Fourier Optics for Spatially-filtered WFS onto CCD, quadcell config*
- § *Altair-based DM model using influence functions*
- § *Input phase aberration is a very long screen shifted at 10m/s*
- § *FTR reconstruction*
- § *Modal coefficients obtained in reconstruction stage*
- § *Gain optimization as describe above*
- § *Full diagnostics including long-exposure PSDs and PSFs from the residual wavefront and instantaneous residual error in different spatial frequency bands*
- § *Run either single long case to watch optimization or many short cases with a specific filter to analyze general case performance*

Significant reduction in residual error



**N=48, NGS Mag 8 example for 8 iterations
of gain optimization**

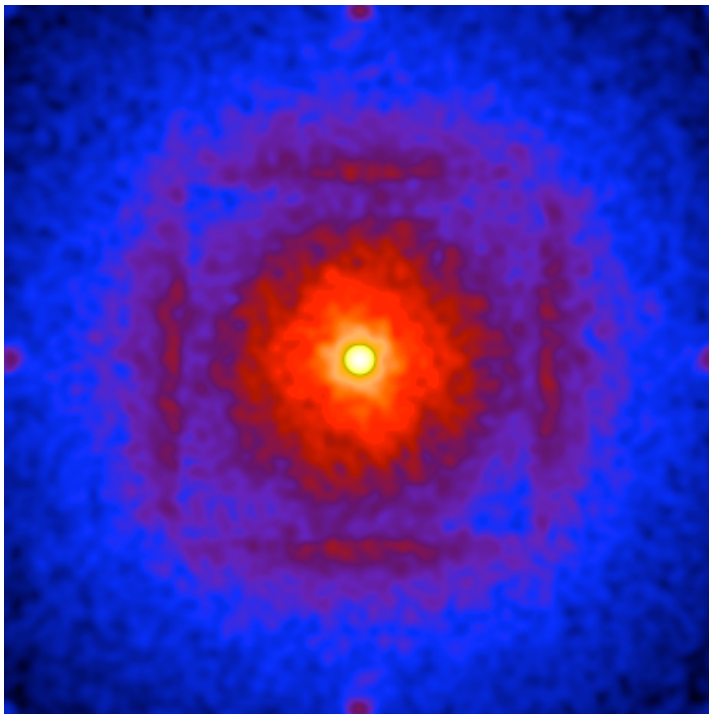
§ Use of optimal gains
improves performance

§ *significant reduction in
residual MSE at each
timestep*

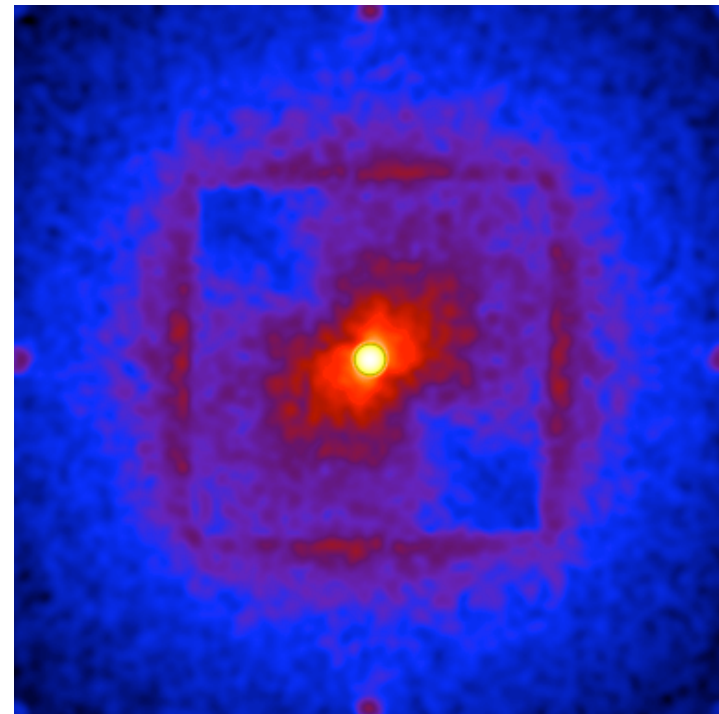
§ *less variation in MSE at
each timestep*

Comparison shows improvement

- § N=48 case with WFS SNR of 2.16
- § Strehl increased from 0.75 to 0.87 (+12%)
- § MSE in band reduced from 0.224 to 0.074 (3 times less)

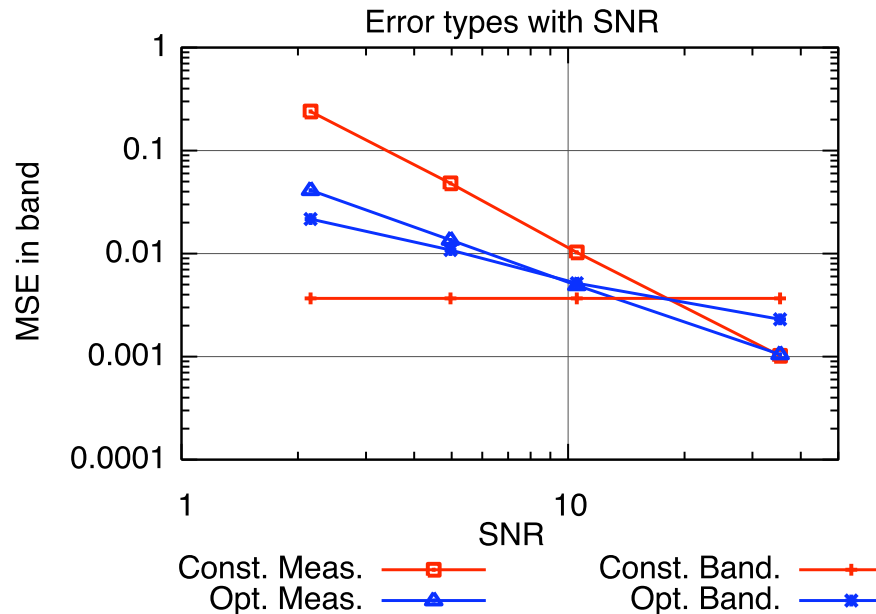


Before



After

Trade bandwidth and sensor errors



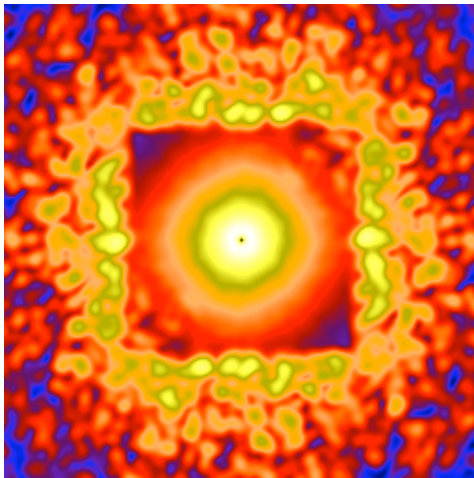
**Data for N=48, median over a set
of 25 random phase screens**

§ At high SNRs, optimal gains produce equivalent or more measurement error but less temporal error than before

§ At low SNRs, optimal gains produce less measurement error but more temporal error than before

We can control modes independently

- § Given an optimal gain profile, we compare three filters
 - § (1) constant gain of 0.6 for all modes
 - § (2) optimized gains
 - § (3) constant gains with optimized for a smaller region of filter
- § PSD of case (3) is almost exactly the combination of parts of the other two responses in the right places



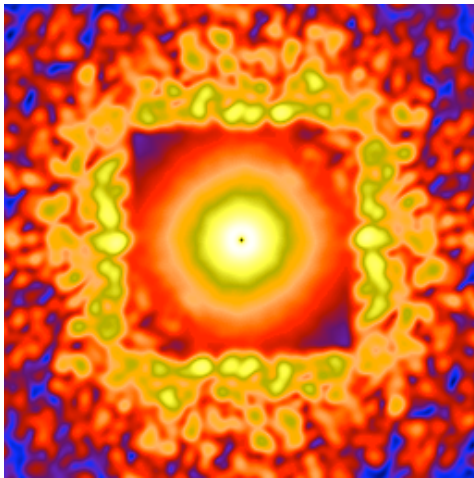
Constant

Mix

Optimal

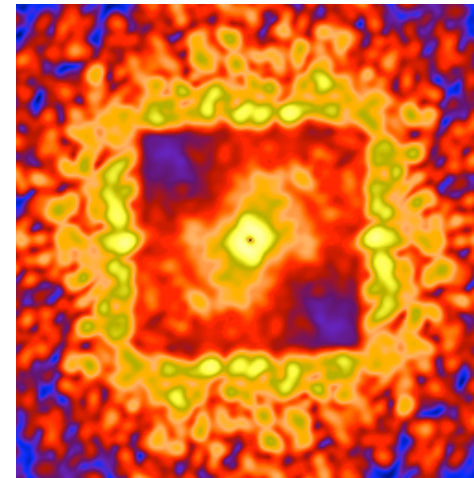
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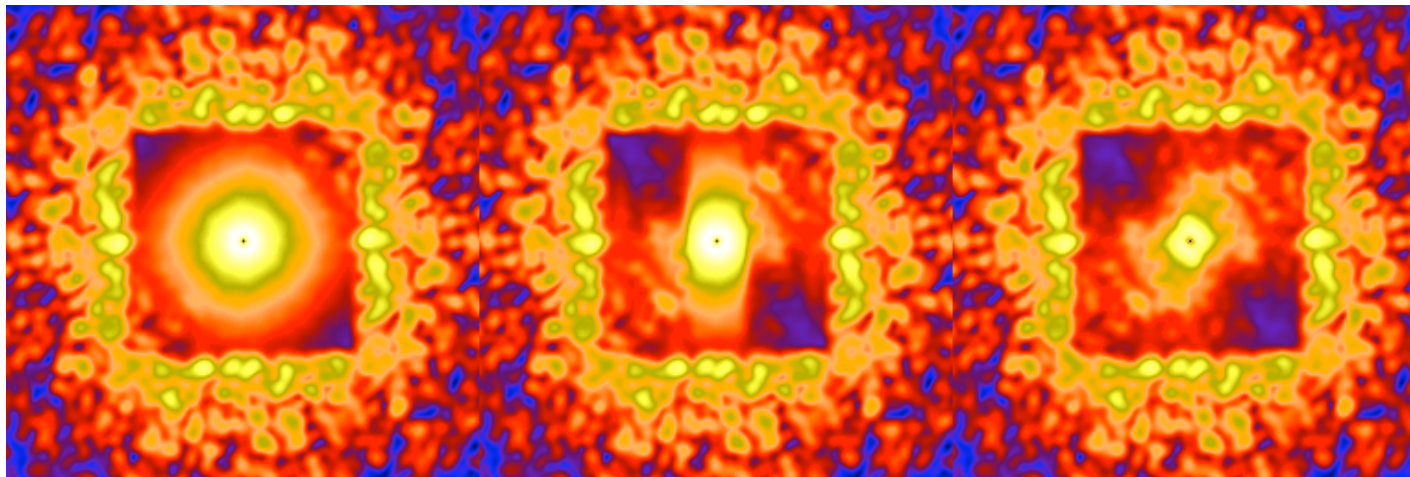
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Optimal

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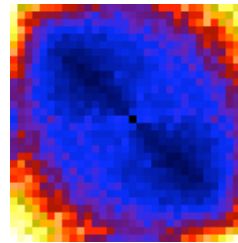
Constant

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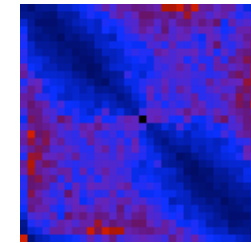
Optimal

Optimal gains compensate for DM

§ Test: run Mod-Hud and Custom filter on same aberration and determine steady-state optimal gains

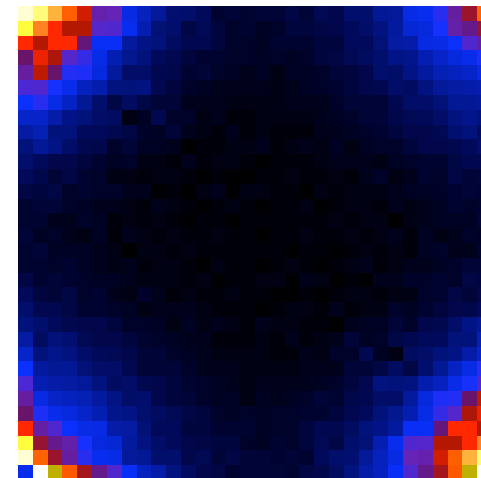


No DM knowledge



DM knowledge

§ Ratio of these gains is almost exactly the inversion of the DM response, as determined empirically from AO system models



Inversion of DM influence function

Computational load is satisfiable today

§ FTR each timestep: $15N^2 \lg N + 20N^2$

§ Estimating periodograms for t steps of telemetry:

$$N^2(5 + 2.5 \lg t)$$

§ Averaging the periodograms and finding the optimal gain (k is for evaluations in root-finding):

$$N^2(1 + k) + 4k$$

§ Assuming $k = 10$ (using fast method), a 64x64 system at 2.5k kHz has a maximum load of 1.43 GFLOPs/sec.

Plans for Optimal Fourier Control

- § Short term continuation of theory:
 - § *Explore complex gain filters and higher-order control laws*
 - § *Contribute to ExAOC system design with performance predictions*
 - § *Verify system measurement procedure (custom filters) at LAO ExAO testbed*
- § Long term experimental verification of performance:
 - § *Implement at LAO testbed in ExAOC control system*
- § Paper preprint (PDF) available