



Tomography for Multiconjugate Adaptive Optics Systems using Laser Guide Stars

Donald Gavel
Lick Observatory
Center for Adaptive Optics
UC Santa Cruz

CfAO Summer School
on Adaptive Optics
August, 2004

Outline:

- Why tomography?
- MCAO system concepts
- An MCAO reconstruction algorithm that works for laser guide star beams
- Simulation examples
- Conclusions



Why tomography on ELTs*?

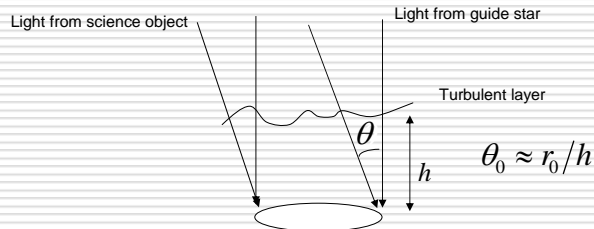
- AO correction on a wide field
multiplex the science
- Reduce laser guide star cone effect
more severe on larger telescopes

*ELT = Extremely Large Telescopes (i.e. $D > 10$ meters)



Anisoplanatism error

- If the guide star is not the science object...

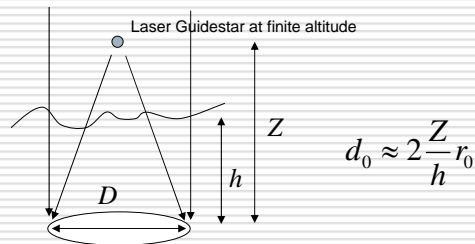


- Typical: $r_0 = 20$ cm, $\theta_0 = 4$ arcsec at $\lambda = 0.5 \mu$
- Mean height of turbulence: $h_0 = r_0/\theta_0 = 10$ km

Residual wavefront error: $\sigma_A^2 = (\theta/\theta_0)^{5/3}$



Laser guidestar specific error: Cone effect

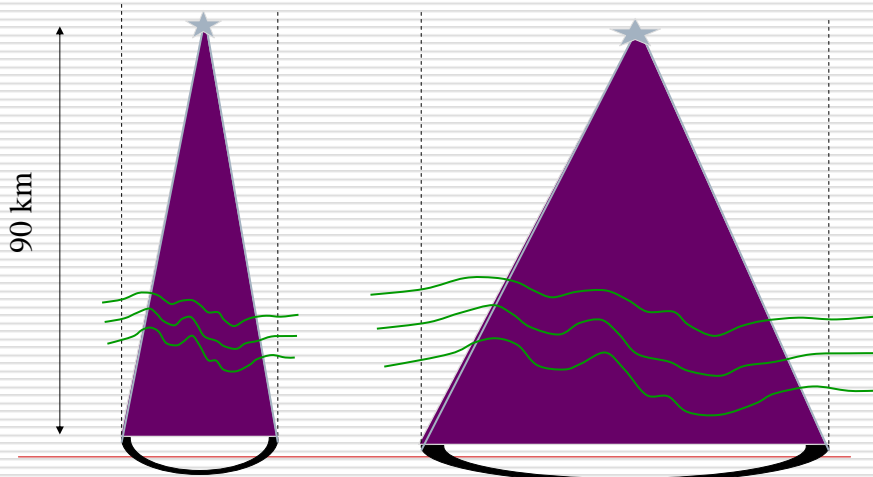


- Example: $h=8$ km, $r_0=20$ cm at $\lambda = 0.5 \mu$
- $d_0=4.5$ m

Residual wavefront error: $\sigma_{cone}^2 = (D/d_0)^{5/3}$



Cone effect gets worse as telescope size increases



CfAO Adaptive Optics Summer School, August, 2004

5



Mitigating cone effect with multiple laser guidestars



CfAO Adaptive Optics Summer School, August, 2004

6

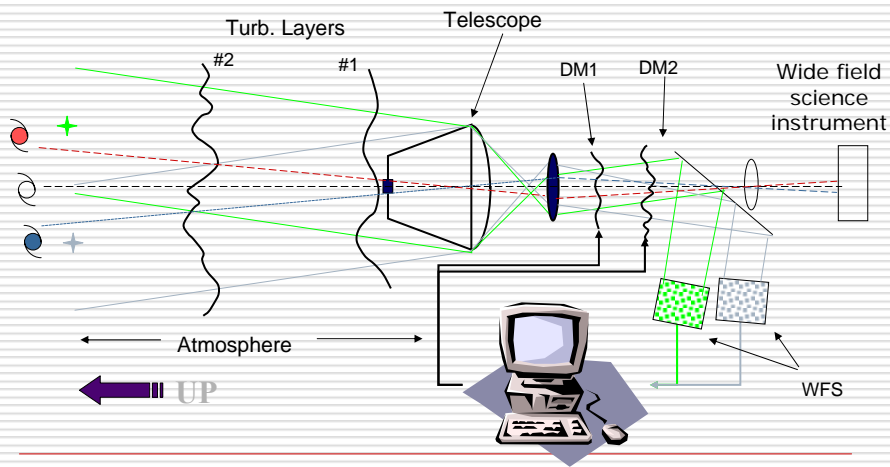


AO system concepts for the near IR bands

- MCAO – Multi-conjugate adaptive optics
 - Diffraction-limited imaging
 - Correction over a “wide” field (~30 arcsec)
 - Multiple DMs, Multiple LGS
- MOAO – Multi-object adaptive optics
 - Diffraction-limited spectroscopy (IFUs)
 - Correction over small fields (~5 arcsec) at many focal plane positions
 - Each unit has a single DM, corrects for the science direction
 - Multiple LGS measure the volume of turbulence

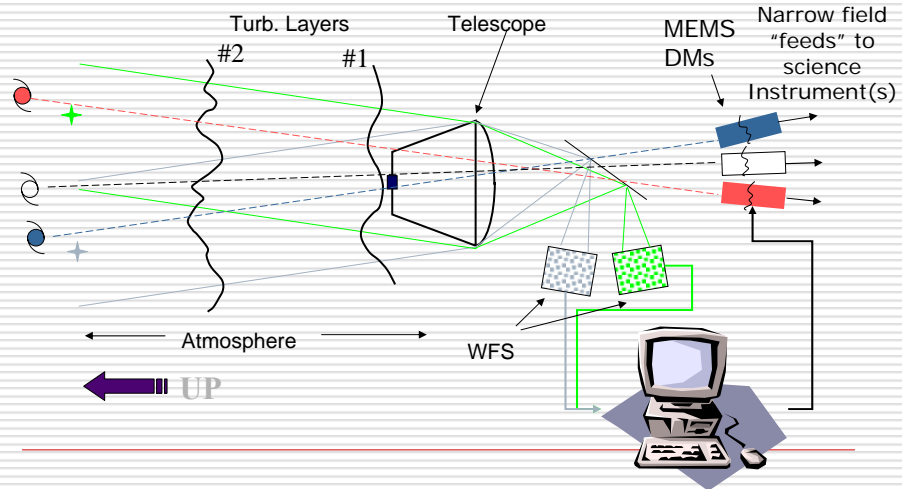


Multiconjugate AO Set-up





Multi-Object AO Setup



CfAO Adaptive Optics Summer School, August, 2004

9



MCAO performance example

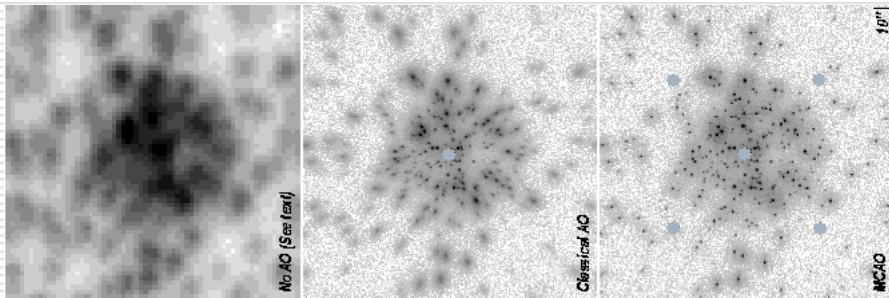
No AO

Classical AO

MCAO

1 DM / 1 NGS

2 DMs / 5 NGS



320 stars / K band / 0.7'' seeing

165''

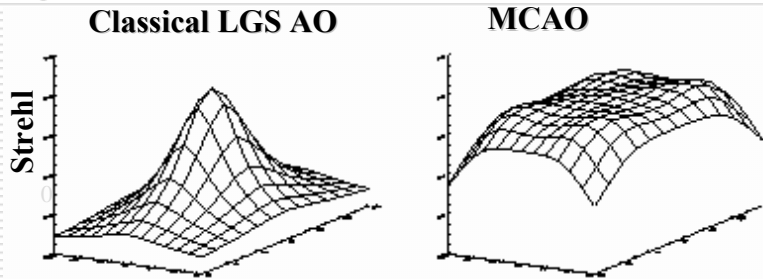
Stars magnified for clarity

CfAO Adaptive Optics Summer School, August, 2004

10



MCAO Performance



Surface plots of Strehl ratio over a 1.5 arc min FoV.
13x13 actuator system, K band, CP turbulence.



MCAO Pros and Cons

PROS:

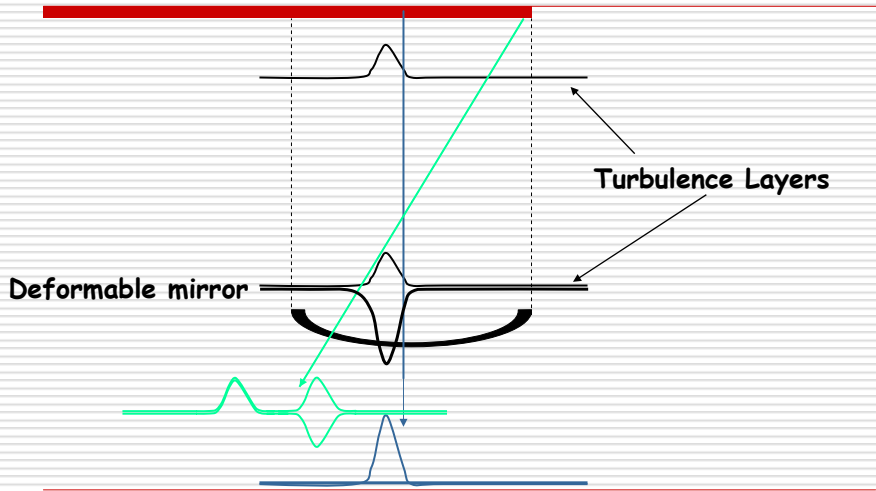
- ❑ **Enlarged Field of View**
 - PSF variability problem drastically reduced
- ❑ **Cone-effect solved**
- ❑ Gain in SNR (less sensitive to noise, predictive algorithms)
- ❑ Marginally enlarged Sky Coverage (LGS systems)

CONS

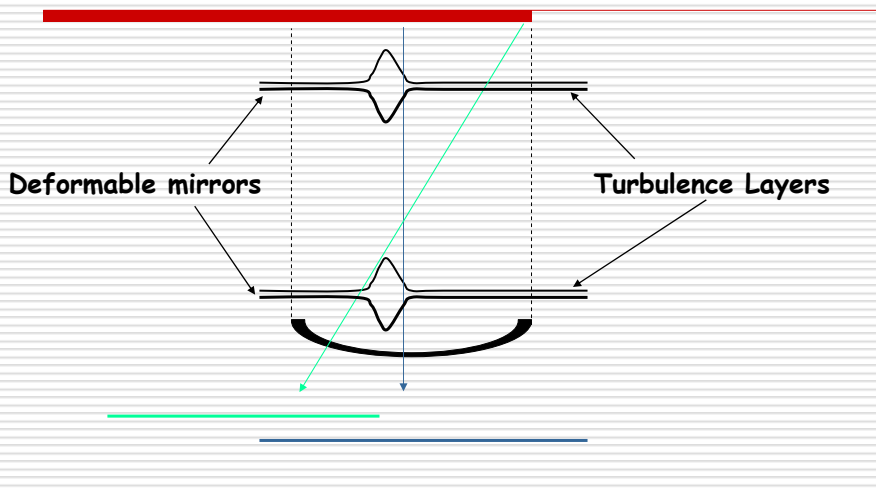
- ❑ **Complexity:** Multiple Guide stars and DMs
- ❑ **Other limitations:** Generalized Fitting, anisoplanatism, aliasing



Why multiconjugate?



Why multiconjugate?

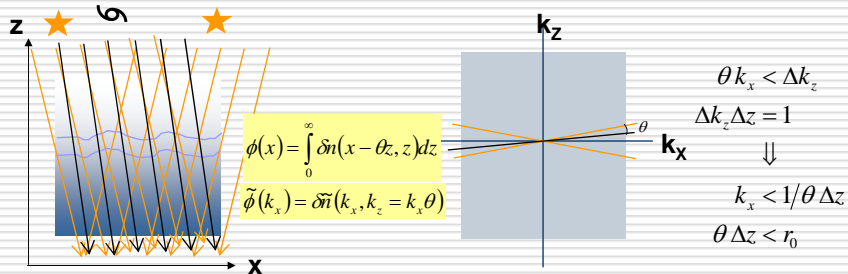




What is tomography?



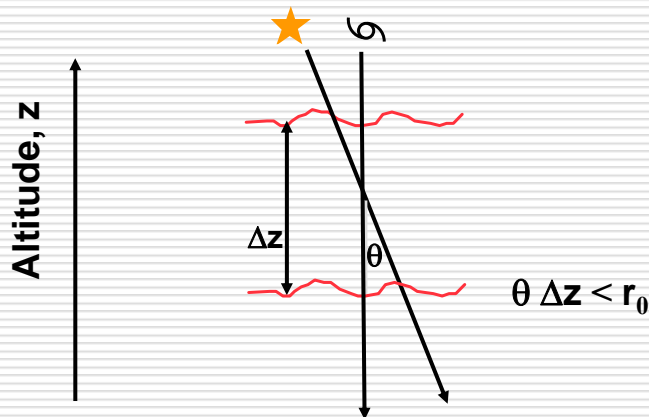
Fourier slice theorem (Kak, 1988)



- Each wavefront sensor measures the integral of index variation along the ray lines
- The line integral along z determines the $k_z=0$ Fourier spatial frequency component
- Projections at several angles sample the k_x, k_y, k_z volume
- Each line in k -space has a thickness ~ thickness of layers in atmosphere
- Fill desired k_x space over bow-tie of field angles
- Angle between guidestars must be less than the ratio of r_0 to layer thickness



Spatial interpretation of the $k_x < 1/[\Delta z(\theta - \theta_{gs})]$ requirement





The M*AO wavefront reconstruction problem



Measurements from guide stars:

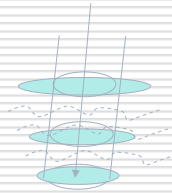
$$s_k(x) = M(\phi(x, \theta_k)) + v_k(x) \quad k = 1, \dots, n_{gs}$$

s = wfs slopes

ϕ = atmospheric phase integrated along direction θ

M = wfs operator

v = wfs noise



Problem(s):

- MCAO: Find commands for multi-conjugate DMs that best correct the wavefront in any direction over a given field of view
- MOAO: Find the best estimate of the wavefront in a particular science direction



Spatial domain reconstructors¹



- Obvious extension of present single-conjugate AO reconstructor methods

Control on each DM Reconstructor matrix Measurements from each WFS,...

$$\psi(\mathbf{x}, z) = \mathbf{G}\mathbf{s}[\delta n(\mathbf{x}, z)]$$

...a function of index of refraction variations along the path

- Minimum-variance solution:

$$\mathbf{G} = \Sigma_{\delta n} \mathbf{H}^T (\mathbf{H} \Sigma_{\delta n} \mathbf{H}^T + \Sigma_{\nu})^{-1} \quad \text{where} \quad \mathbf{H} = \frac{\partial \mathbf{s}}{\partial \psi}$$

Covariance of refractive index variations Covariance of measurement noise ("push" matrix)

¹Ellerbroek, B., *Efficient computation of minimum-variance wave-front reconstructors with sparse matrix techniques*, JOSA-A, 19, 9, Sept. 2002, pp 1803-1816.



An aside:

Minimum variance solutions

- The normal equation:

$$\boldsymbol{\psi} = \mathbf{G}\mathbf{s}$$

$$\min_{\mathbf{G}} J = \langle |\delta\mathbf{n} - \boldsymbol{\psi}|^2 \rangle$$

$$\frac{\partial J}{\partial \mathbf{G}} = 0 \Rightarrow \langle (\delta\mathbf{n} - \boldsymbol{\psi})\mathbf{s}^T \rangle = 0$$

$$\langle \delta\mathbf{n}\mathbf{s}^T \rangle - \mathbf{G}\langle \mathbf{s}\mathbf{s}^T \rangle = 0$$

$$\Rightarrow \mathbf{G} = \langle \delta\mathbf{n}\mathbf{s}^T \rangle \langle \mathbf{s}\mathbf{s}^T \rangle^{-1}$$

(For the sake of simplifying the explanation we're imagining here that atmospheric turbulent layers exist only at DM conjugate locations. We'll treat the more general case later on.)

- All minimum variance solutions have this form

The best linear reconstructor is the cross-correlation of the unknown with the data, times the inverse of the auto-correlation of the data



Minimum variance solution

- Assuming a linear model for WFS response to index variations:

$$\mathbf{s} = \mathbf{H}\delta\mathbf{n} + \mathbf{v}$$

H represents the line integral along z through the turbulence, followed by the x,y gradient operations done by the Hartmann (slope) wavefront sensor.

- We use the normal equation to determine the reconstructor matrix

$$\mathbf{G} = \langle \delta\mathbf{n}\mathbf{s}^T \rangle \langle \mathbf{s}\mathbf{s}^T \rangle^{-1} = \boldsymbol{\Sigma}_{\delta\mathbf{n}}\mathbf{H}^T (\mathbf{H}\boldsymbol{\Sigma}_{\delta\mathbf{n}}\mathbf{H}^T + \boldsymbol{\Sigma}_{\mathbf{v}})^{-1}$$



Relationship of minimum variance to least-squares

$$\begin{aligned} \mathbf{G} &= \langle \delta n \mathbf{s}^T \rangle \langle \mathbf{s} \mathbf{s}^T \rangle^{-1} \\ &= \boldsymbol{\Sigma}_{\delta n} \mathbf{H}^T (\mathbf{H} \boldsymbol{\Sigma}_{\delta n} \mathbf{H}^T + \boldsymbol{\Sigma}_v)^{-1} \\ &= (\mathbf{H} \boldsymbol{\Sigma}_v^{-1} \mathbf{H}^T + \boldsymbol{\Sigma}_{\delta n}^{-1})^{-1} \boldsymbol{\Sigma}_v^{-1} \mathbf{H}^T \quad \text{Matrix inversion theorem} \end{aligned}$$

In the limit of high signal-to-noise: $\boldsymbol{\Sigma}_v \rightarrow I\sigma_v^2$ small

$$= (\mathbf{H} \mathbf{H}^T)^{-1} \mathbf{H}^T \quad \begin{array}{l} \text{Least-squares solution} \\ \text{(best fit to measurement):} \\ \text{Moore-Penrose pseudo-inverse} \end{array}$$



Fourier domain approach

- A function is represented as a sum of sine waves

f is spatial frequency from now on. Sorry for the change of notation but I need *k* to enumerate guide stars.

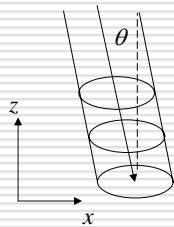
$$\phi(\mathbf{x}) = \iint \tilde{\phi}(\mathbf{f}) e^{i2\pi\mathbf{f}\cdot\mathbf{x}} df_x df_y$$

We're taking the Fourier transforms only in the transverse (x,y) plane

- Shifts Δx in the spatial domain \rightarrow sine waves add $2\pi f \Delta x$ phase, or multiply Fourier coefficient by $e^{2\pi i f \Delta x}$



Phase at the pupil from a source at field position θ



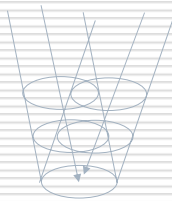
$$\Delta x = \theta z$$

$$\phi(x) = \int_0^{\infty} \delta n(x - \theta z) dz$$

$$\tilde{\phi}(f) = \int_0^{\infty} \delta \tilde{n}(f, z) e^{i2\pi f \cdot \theta z} dz$$



A minimum variance solution in the Fourier domain²



Measurements from guide stars:

$$\tilde{s}_k(f) = M(f) \tilde{\phi}(f, \theta_k) + \tilde{v}_k(f) \quad k = 1, \dots, n_{gs}$$

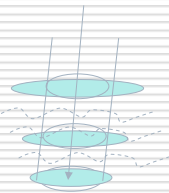
s = wfs slopes

ϕ = atmospheric phase integrated along direction θ

M = wfs operator

v = wfs noise

We'll assume $M=1$, for simplicity for now, but with a justification later



Problem: MOAO: Find a linear combination of guide star data that best predicts the wavefront in a given science direction, θ

$$\tilde{\psi}(f, \theta) = \sum_{k=1}^N \tilde{g}_k(f, \theta) \tilde{s}_k(f) = \tilde{\mathbf{g}}^T \tilde{\mathbf{s}}$$

We'll deal with the MCAO case later

²Tokovinin, A., Viard, E., "Limiting precision tomographic phase estimation," *JOSA-A*, 18, 4, Apr. 2001, pp873-882.



Fourier domain reconstructor

- The normal equation in the Fourier domain is:

$$\langle \langle \tilde{\phi} - \tilde{\psi} \rangle \tilde{s}_k \rangle = 0$$

- so the minimum variance solution is

$$\tilde{g}_k(f) = S_{\phi\phi_k}(f) [S_{\phi\phi_k}(f) + S_v(f)]^{-1}$$

← One equation per frequency

where

$$S_\phi(f) = \langle \tilde{\phi}(f) \tilde{\phi}_k^*(f) \rangle$$

$$S_v(f) = \langle \tilde{v}_k(f) \tilde{v}_k^*(f) \rangle$$

$$S_s(f) = \langle \tilde{s}_k(f) \tilde{s}_k^*(f) \rangle = S_\phi(f) + S_v(f)$$

Matrix is
 $n_{gs} \times n_{gs}$



Minimum variance solution in the Fourier domain

$$\tilde{\mathbf{g}}^T(\mathbf{f}) = \tilde{\mathbf{c}}^{T*}(\mathbf{f}) [\mathbf{A}^*(\mathbf{f}) + \nu'(\mathbf{f}) \mathbf{I}]^{-1}$$

where

$$\tilde{c}_k(\mathbf{f}) = \int C_n^2(h) e^{-2\pi i \mathbf{f}(\theta_k - \theta)h} dh / c_0 \quad = \langle \tilde{\phi} \tilde{\phi}_k^* \rangle / W_0 c_0$$

$$\tilde{a}_{kk'}(\mathbf{f}) = \int C_n^2(h) e^{-2\pi i \mathbf{f}(\theta_k - \theta_{k'})h} dh / c_0 \quad = \langle \tilde{s}_k \tilde{s}_{k'}^* \rangle / W_0 c_0$$

$$c_0 = \int C_n^2(h) dh$$

$$\nu'(\mathbf{f}) = W_\nu(\mathbf{f}) / (W_0(\mathbf{f}) c_0)$$

$$W_0(\mathbf{f}) = 9.69 \times 10^{-3} (2\pi/\lambda)^2 f^{-11/3}$$

$$c_0 = 2.36 (\lambda/2\pi) r_0^{-5/3}$$

Noise-to-'signal' ratio

Kolmogorov power spectrum of turbulence



An interpretation³...

$$\tilde{\psi}(f, \theta) = \int e^{2\pi i \theta h f} C_n^2(h) \sum_{k=1}^N e^{-2\pi i \theta_k h f} \tilde{s}'_k(f) dh / c_0$$

$\tilde{s}'(f) = [\mathbf{A}^*(f) + \mathbf{I}v^*(f)]^{-1} \tilde{s}(f)$
 "Consistency-filtered"
 wavefront measurement

$$\psi(x, \theta) = \int C_n^2(h) \sum_{k=1}^N s'_k(x - h(\theta - \theta_k)) dh / c_0$$

$$\psi(x, \theta) = \int \delta n_{est}(x - h\theta, h) dh$$

$$\delta n_{est}(x, h) = \sum_{k=1}^N C_n^2(h) s'_k(x + h\theta_k) / c_0$$

³Gavel, *Tomography for multiconjugate adaptive optics systems using laser guide stars*, SPIE 5490, June, 2004.



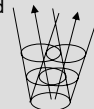
The minimum variance solution = back projection tomography

Measure

$$s_k(x, \theta) = \int \delta n(x - h\theta_k, h) dh + v_k$$

Reconstruct

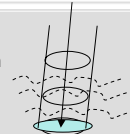
- 1 Back-Project filtered measurements through the volume toward the guide stars



$$\delta n_{est}(x, h) = \sum_{k=1}^N C_n^2(h) s'_k(x + h\theta_k) / c_0$$

- 2

Propagate light from Science target



$$\psi(x, \theta) = \int \delta n_{est}(x - h\theta, h) dh$$



Why the Hartmann wavefront sensor model can be "ignored" (M=1) without loss of generality

$$\mathbf{s}_k(x) = \nabla \phi_k(x) + \mathbf{v} \Rightarrow \tilde{\mathbf{s}}_k(f) = -2\pi i \mathbf{f} \tilde{\phi}_k(f) + \tilde{\mathbf{v}}$$

Perform an invertible transformation on the WFS data

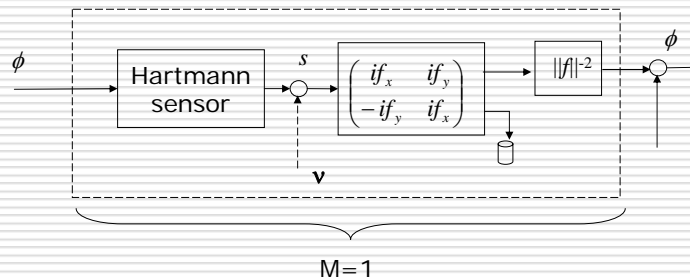
$$\begin{pmatrix} \tilde{s}_0(f) \\ \tilde{s}_1(f) \end{pmatrix} = \begin{pmatrix} if_x & if_y \\ -if_y & if_x \end{pmatrix} \begin{pmatrix} \tilde{s}_x(f) \\ \tilde{s}_y(f) \end{pmatrix} = 2\pi \begin{pmatrix} f_x^2 + f_y^2 \\ 0 \end{pmatrix} \phi + \begin{pmatrix} if_x v_x + if_y v_y \\ -if_y v_x + if_x v_y \end{pmatrix}$$

- Since the 2nd equation doesn't depend on the data, $M = \|f\|^2$ is an appropriate scalar measurement model (the 1st equation above).
- Without loss of generality, we'll assume the WFS has a post-filter $1/2\pi\|f\|^2$. This allows us to set $M=1$ (up to some frequency cut-off associated with the WFS spatial sampling). The noise spectrum is modified accordingly $\langle v(f)^2 \rangle \sim 1/f^2$

i.e. We can assume the wavefront sensors measure phase directly



The "direct phase" Hartmann sensor

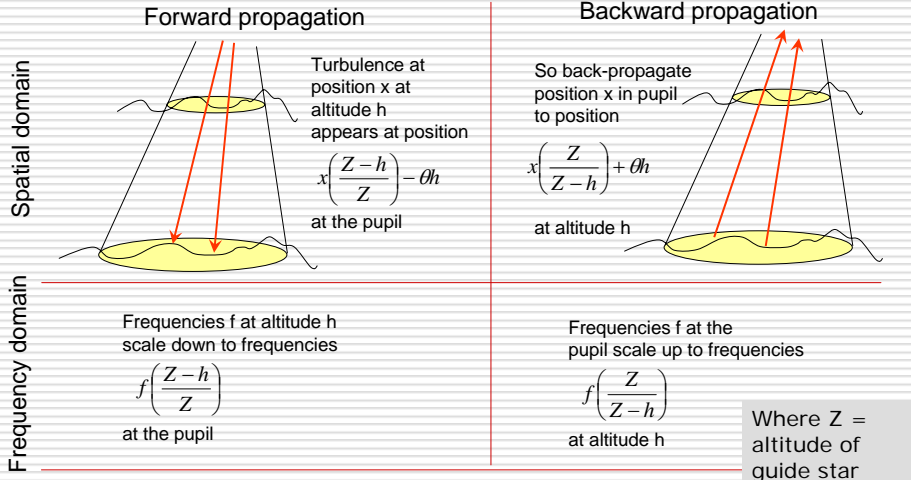


Important: you must consider the boundary conditions when using the Fourier transform on aperture-masked slope data. See:

⁴Poyneer, L., Gavel, D., and Brase, J., *Fast wave-front reconstruction in large adaptive optics systems with use of the Fourier transform*, **JOSA-A**, 19, 10, October, 2002, pp2100-2111.



We can extend the back projection method to **spherical waves**, resulting in a tomography solution for **guide stars at finite altitude, Z**



Where $Z =$ altitude of guide star



Guide Stars at Finite Altitude: The Math³

Minimum error variance spherical wave tomography formulae



Back-project

$$\delta n_{est}(\mathbf{x}, h) = \sum_{k=1}^N C_n^2(h) s_k' \left(\mathbf{x} \frac{Z}{Z-h} + h \boldsymbol{\theta}_k \right) / c_0$$

Forward propagate

$$\psi(x, \theta) = \int \delta n_{est}(x - h\theta, h) dh$$

Consistency filters

$$\tilde{\mathbf{s}}'(f) = [\mathbf{A}^{ss}(f) + \mathbf{I} \mathbf{V}^{*s}(f)]^{-1} \tilde{\mathbf{s}}(f)$$

$$c_k^2(\mathbf{f}, h) = (Z/(Z-h))^{-1/2} C_n^2(h) \exp(-2\pi i \mathbf{f} \cdot \boldsymbol{\theta}_k h) / c_0$$

$$a_{kk}^s(\mathbf{f}) = \int_0^Z (Z/(Z-h'))^{-5/2} C_n^2(h') \exp(2\pi i \mathbf{f} \cdot (\boldsymbol{\theta}_k - \boldsymbol{\theta}_k) h') dh' / c_0$$

$$v_{kk}^s(\mathbf{f}) = v_k^2(\mathbf{f}) \delta_{kk} / (9.69 \times 10^{-3} f^{-1/3} c_0)$$

Covariance & PSF calculation

$$S_c(f) = c_0 W_0(f) \left[1 - \int_0^Z \int_0^Z \sum_k \tilde{g}_k(\mathbf{f}, h) \times \left(\frac{Z}{Z-h'} \right)^2 \left(\frac{Z-h}{Z-h'} \right)^{-1/2} C_n^2(h') e^{-2\pi i \mathbf{f} \cdot (\boldsymbol{\theta}_k - \boldsymbol{\theta}_k) \frac{Z-h}{Z-h'} - 2\pi i \mathbf{f} \cdot \boldsymbol{\theta}_k h'} dh dh' / c_0 \right. \\ \left. + \int_0^Z \sum_k [\tilde{g}_k(\mathbf{f}, h)]^2 v_k^s \left(\mathbf{f} \frac{Z-h}{Z} \right) dh \right]$$

$$D_c(x) = FT^{-1}\{S_c(f)\}$$

$$MTF(x) = MTF_{telescope}(x) \times \exp\left[-\frac{1}{2} D_c(x)\right]$$

$$PSF = |FT\{r(x)\}|^2$$

³Gavel, *Tomography for multiconjugate adaptive optics systems using laser guide stars*, SPIE 5490, June, 2004.

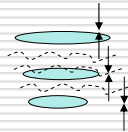


For MCAO one can optimally project the volume estimates onto a finite number of DMs to minimize generalized anisoplanatism⁵



$$\tilde{a}_m(f) = \int \tilde{g}_m^{DM}(f, h) \tilde{\alpha}_{est}(f, h) dh \quad m = 1, \dots, n_{DM}$$

DM commands
Combining gains
Index estimates



$$\tilde{g}^{DM}(f, h) = \mathbf{A}^{DM^{-1}}(f) \tilde{\mathbf{b}}^{DM}(f, h)$$

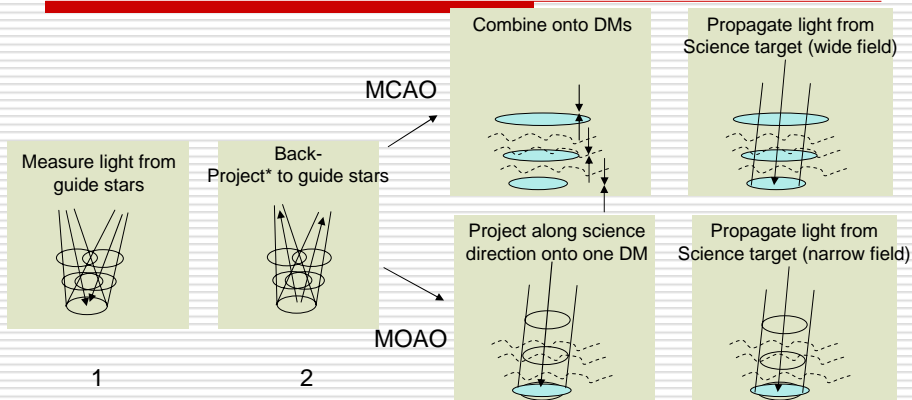
$$\tilde{b}_{m'}^{DM} = J_0 [2\pi f \Theta(H_m - h)]$$

$$\tilde{a}_{mm'}^{DM} = J_0 [2\pi f \Theta(H_m - H_{m'})]$$

⁵Tokovinin, A., Le Louarn, M., Sarazin, M., "Isoplanatism in a multiconjugate adaptive optics system," *JOSA-A*, 17, 10, Oct. 2000, pp1819-1827.



Tomographic Wavefront Reconstruction a quick summary



*after the all-important filtering step, which makes the back projections consistent with all the data



Tomography solution domains: a comparison



Spatial Domain

- Obvious extension of present AO reconstructor methods
- Big matrix calculations; requires large number of computations $O(n^2)$

$$\Psi = \mathbf{GM}[\phi(\mathbf{x})]$$

$$\mathbf{G} = \sum_{\partial n_i} \mathbf{H}^T (\mathbf{H} \Sigma_{\partial n_i} \mathbf{H}^T + \Sigma_v)^{-1}$$

- Requires reformulating for every parameter change
- Not obviously “tomography”

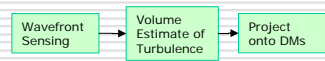
Fourier Domain

- The natural “Karhunen-Loeve” space (uncorrelated modes)
- Smaller calculational problem $O(n \log n)$

$$\tilde{\psi}(f) = \tilde{\mathbf{g}}^T(f) \tilde{\mathbf{s}}(f)$$

$$\tilde{\mathbf{g}}(f) = S_\phi(f) (S_\phi(f) + S_v(f))^{-1}$$

- It is tomography!
- Insight –breaks into parts



- Finite aperture causes difficulties



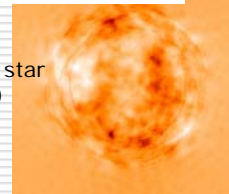
Problem from finite apertured data: Artifacts due to the sharp boundary



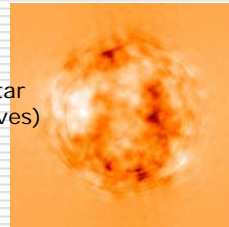
- Let's **mimic** the infinite aperture algorithm, i.e. back-prop filtered measurements, but only from a finite aperture
- Artifacts from the **hard edges** of the aperture **degrade** the reconstruction
- Affects both NGS and LGS, but LGS more severely
- An “**extension**” scheme fixes the problem. **Extension is the minimum-variance estimate of the unknown data given the known data**

Volume estimate at 15 km

Natural guide star (plane waves)



Laser guide star (spherical waves)

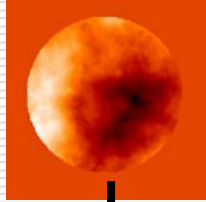




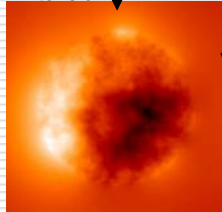
Minimum Variance *Extension* of wavefront measurements to outside the aperture



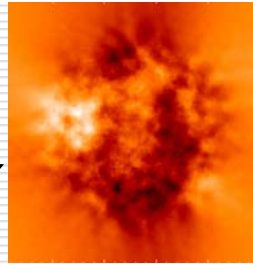
Wavefront measurement



Extension



Volume estimate at 15 km
(spherical waves)



Simulation results

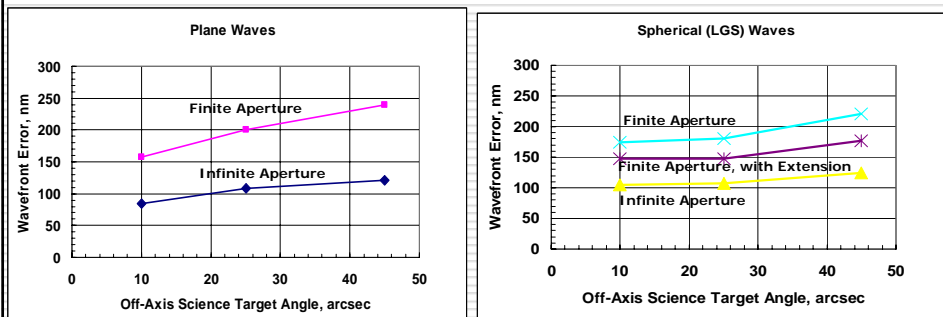


- Telescope Diameter $D = 30$ meters
- $r_0 = 20$ cm, CP C_n^2 profile (7 layer)
- $Z_{LGS} = 90$ km
- Tomography only (no anisoplanatism or other error budget terms)

- Compare NGS (plane wave) and LGS (spherical wave)
- Compare Fourier and Spatial Domain solutions



Tomographic Error --- Generalized Cone Effect

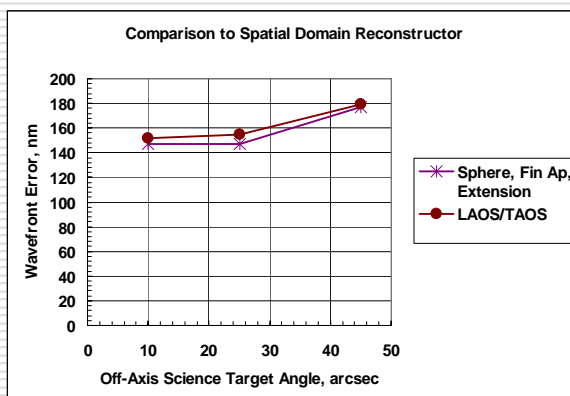


Two types of “cone” effects:

- Expanding of spherical waves
- Finite aperture - by far the biggest effect, partially offset by extension



Results compared to a spatial domain reconstructor*

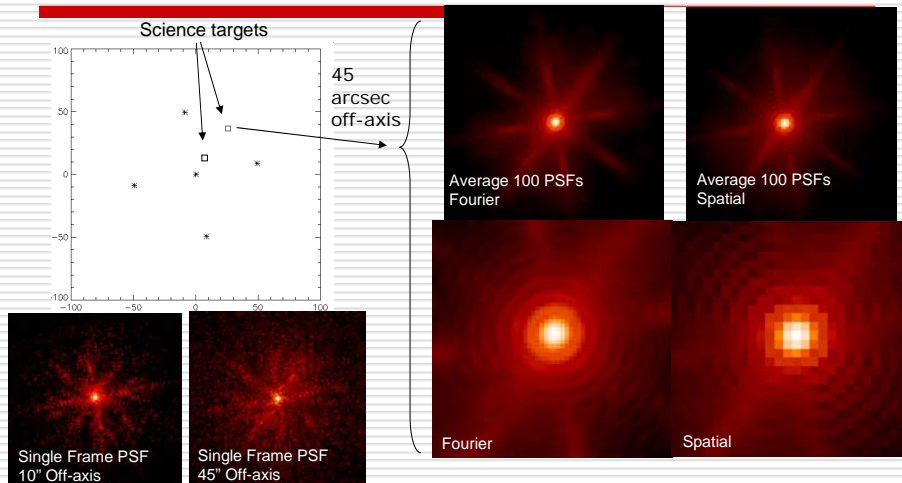


*Linear Adaptive Optics Simulator, Tools for Adaptive Optics Simulation – Brent Ellerbroek, NOAO



Point Spread Function

Finite Aperture, 90km LGS, with Extension, $\lambda = 2$ microns



Conclusions



- Multiple guide-star AO provides
 - a wide field of correction
 - a means of defeating laser guide star cone-effect error
- Two M*AO system architectures:
 - MCAO – one DM per turbulent layer
 - MOAO – one DM per science target
- Minimum variance M*AO reconstructors estimate the index variations in the atmospheric volume above the telescope
- One minimum variance reconstructor (Fourier domain method) is intuitively interpreted as back-projection tomography



Conclusions

- Fourier domain methods are
 - computationally fast
 - provide problem insight (the solution breaks into a series of independent operations, allowing easier parametric design and analysis)
 - can be applied to NGS (plane waves) and LGS (spherical waves) beacons
 - and there is very close simulation agreement between spatial-domain and Fourier domain minimum-variance solutions

Grateful acknowledgement is given to the NSF Center for Adaptive Optics for supporting this work under the Analysis, Modeling, and Simulation of AO for ELTs project
