

Space varying PSF estimation: the telescope tomography approach

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Outlines

- **Telescope tomography**
- **Fast propagation through telescope**
- **Proximity operators for phase retrieval**

Euclid: Mapping the geometry of the dark Universe

■ Stringent requirements on PSF knowledge

PSF shape biases the galaxy ellipticity measurement

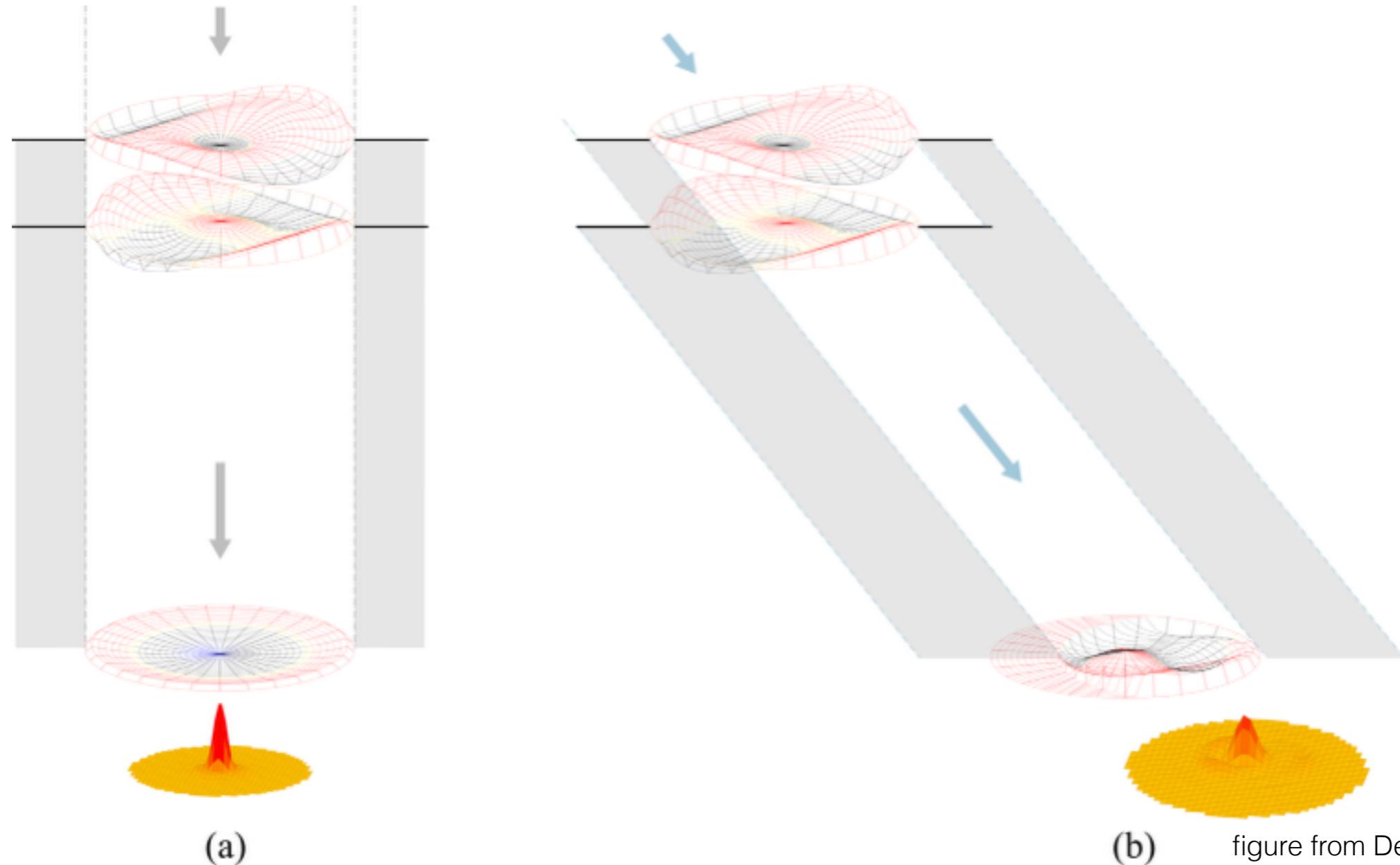
- PSF varies across the wide field of view
- Field stars give a sample the PSF estimate
- PSF have to be estimated at galaxies position.

■ Two problems

- **estimation** of the PSF (using fields stars)
- **interpolation** of the PSF on galaxies positions

Telescope tomography

- Space varying PSFs are totally described by aberrations of all mirrors



Mirror aberrations « sum up » in different manners across the field of view producing space varying PSFs.

Telescope tomography

■ Optical diffraction tomography

Reconstructing the 3D distribution of a biological sample using:

- illumination from different angles
- interferometric measurements of the wavefront in the detector plane
- light propagation modeled by BPM

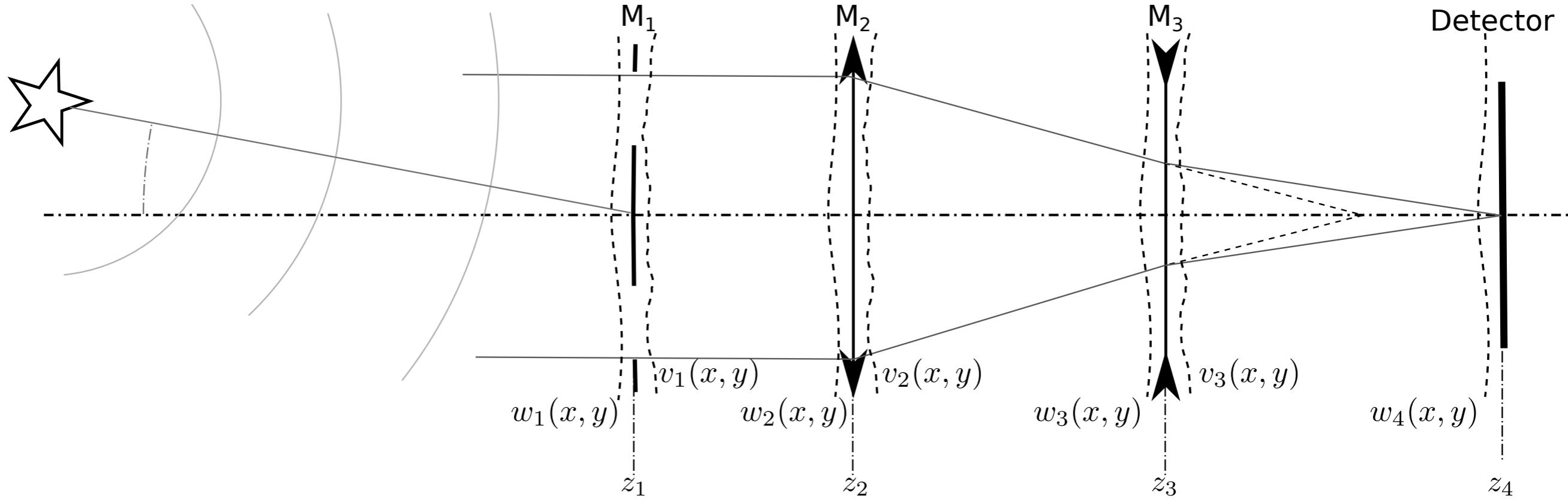
■ Euclid telescope tomography

Estimating the aberrations of all telescope mirrors using:

- field stars: plane wave sources with different incoming angles
- star image (PSF): inline hologram of the wavefront
- light propagation modeled by BPM

(Kamilov et al. *Optica* 2015)

Forward model



Wavefront w_k on optical surface k

$$w_k(\alpha) = \mathbf{H}_{k-1} \mathbf{M}_{k-1} \mathbf{A}_{k-1}(\alpha_k) w_{k-1}(\alpha)$$

■ Incoming wave

Star are infinitely far: w_1 is a tilted plane wave

■ Propagation \mathbf{H}_k

Forward model

Wavefront w_k on optical surface k

$$w_k(\alpha) = H_{k-1} M_{k-1} A_{k-1}(\alpha_k) w_{k-1}(\alpha)$$

■ Mirrors M_k

- change optical path
- cuts light outside of the pupil

■ Aberrations A_k

Introduced by polishing errors and misalignment:

Expressed on a suitable basis: $A_k(\alpha_k) = \text{diag}(\exp(i Z_k \alpha_k))$.

■ Measurements

$$d_n = |w_{K,n}|^2 + e_n$$

Aberrations estimation

Wavefront w_k on optical surface k

$$w_k(\alpha) = H_{k-1} M_{k-1} A_{k-1}(\alpha_k) w_{k-1}(\alpha)$$

$$d_n = |w_{K,n}|^2 + e_n$$

■ ‘Three times’ non-linear problem

- Aberrations expressed in phase
- Propagation through multiple mirrors
- Intensity only measurements

■ Huge size problem

- 24000 x 24000 pixels
- PSF undersampled by a factor 2
- at least 1800 stars per images

Framework

■ The reconstruction problem

$$\alpha^+ = \arg \min_{\alpha} \sum_{s=1}^S \sum_{n=1}^N \frac{1}{\sigma_n^2} \left(|w_{K,n}^s(\alpha)|^2 - d_n^s \right)^2$$

■ The constrained formulation

$$\alpha^+ = \arg \min_{\alpha} \sum_{s=1}^S \sum_{n=1}^N \frac{1}{\sigma_n^2} \left(|t_n^s|^2 - d_n^s \right)^2 \text{ subject to } w_K^s(\alpha) = t^s$$

■ Its augmented Lagrangian form

$$\mathcal{L}(\alpha, t, u) = \sum_{s=1}^S \sum_{n=1}^N \frac{1}{\sigma_n^2} \left(|t_n^s|^2 - d_n^s \right)^2 + \frac{\rho}{2} \sum_{s=1}^S \|w_K^s(\alpha) - t^s - u^s\|_2^2$$

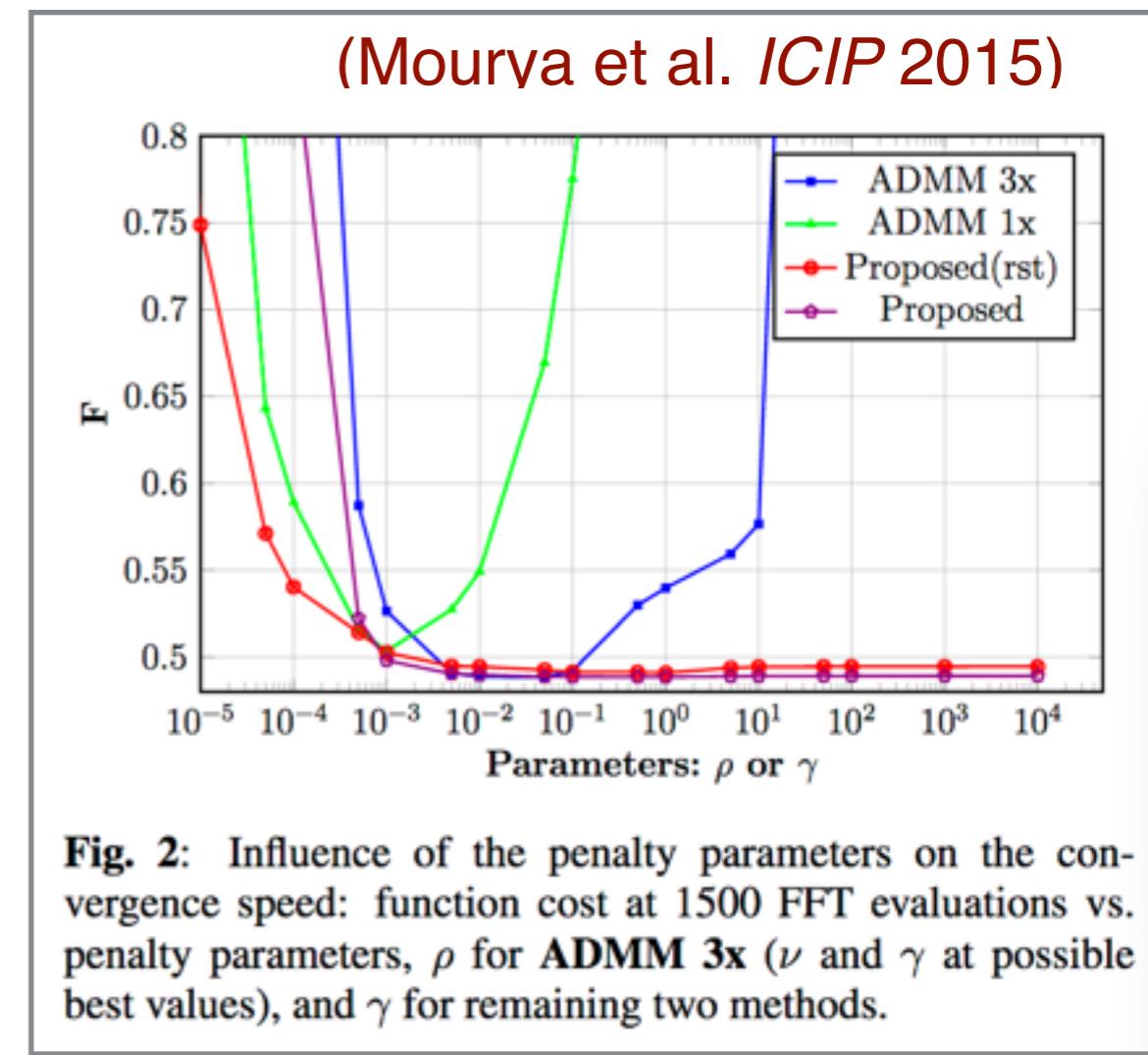
Augmented Lagrangian without alternating direction

■ A hierarchical optimization

$$\alpha = \arg \min_{\alpha} \sum_{s=1}^S \sum_{n=1}^N \left| w_{K,n}^s(\alpha) - t_n^s(\alpha) - u_n^s \right|^2$$

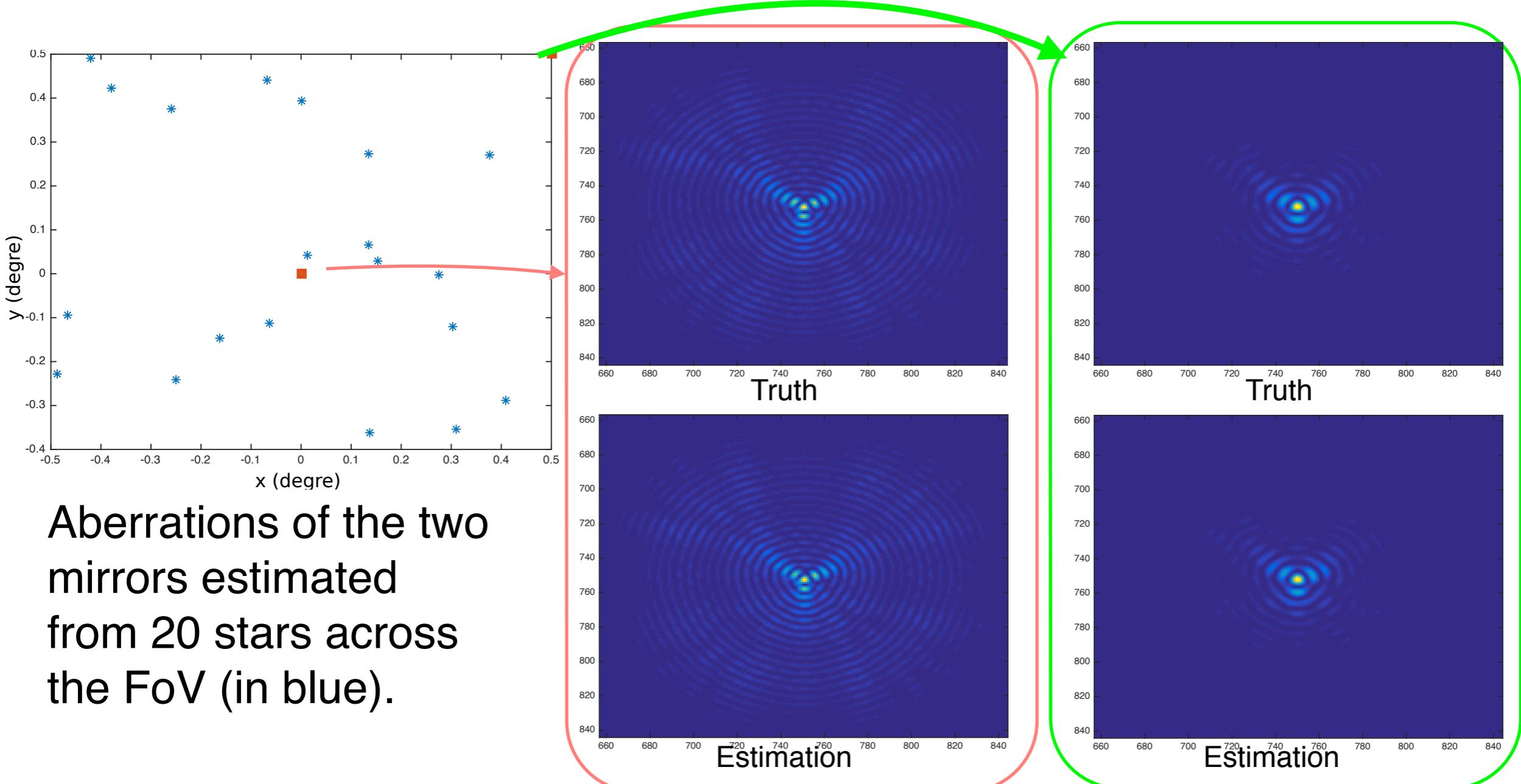
$$\text{with } t_n^s(\alpha) = \arg \min_{t \in \mathbb{C}} \frac{1}{\sigma_n^2} \left(|t|^2 - d_n^s \right)^2 + \frac{\rho}{2} \left| t - w_{K,n}^s(\alpha) + u_n^s \right|^2$$

Solved using a continuous iterative optimization method (e.g. VMLMB).



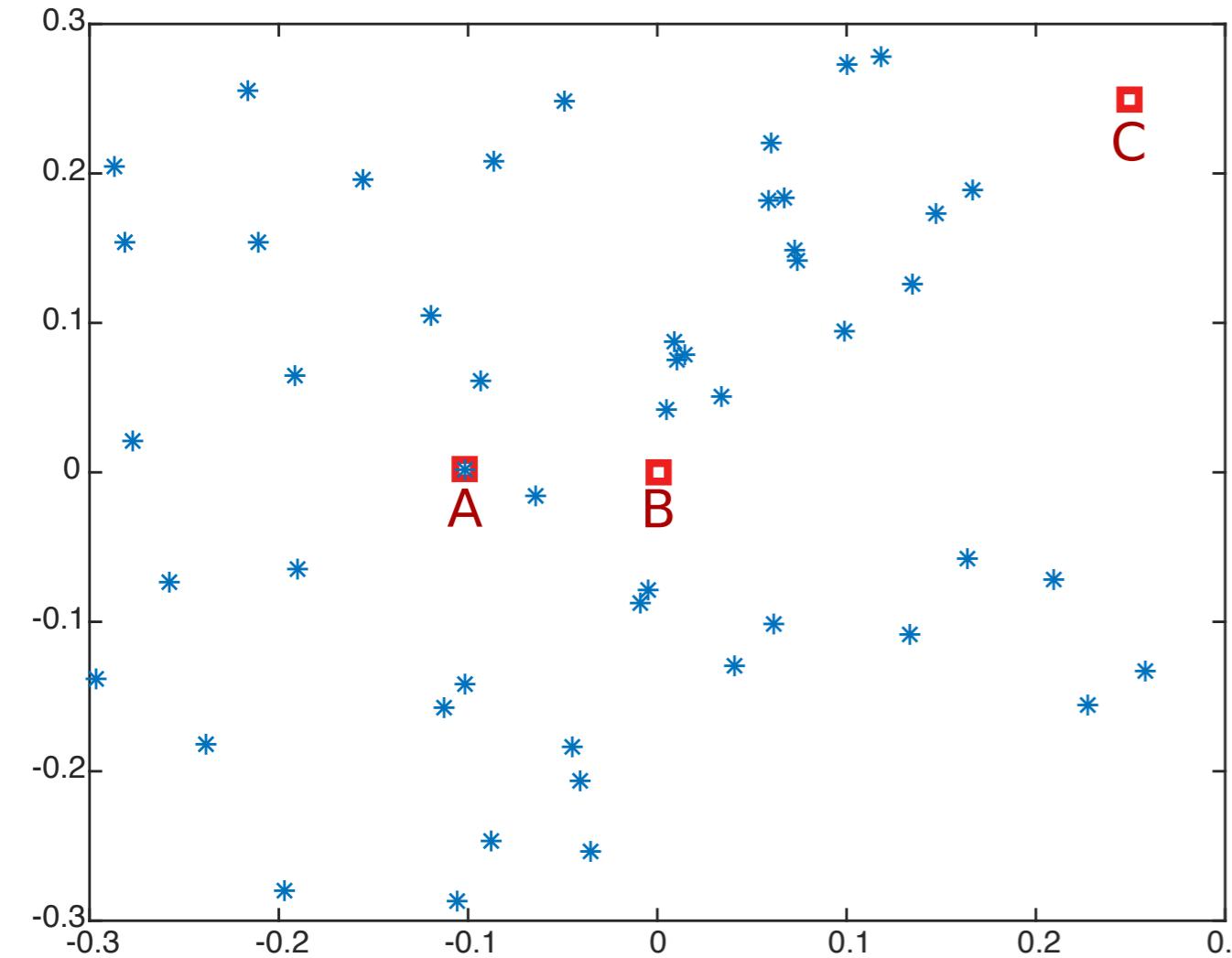
Telescope tomography

■ Preliminary results on HST-like telescope strong aberrations (narrow band, no noise)

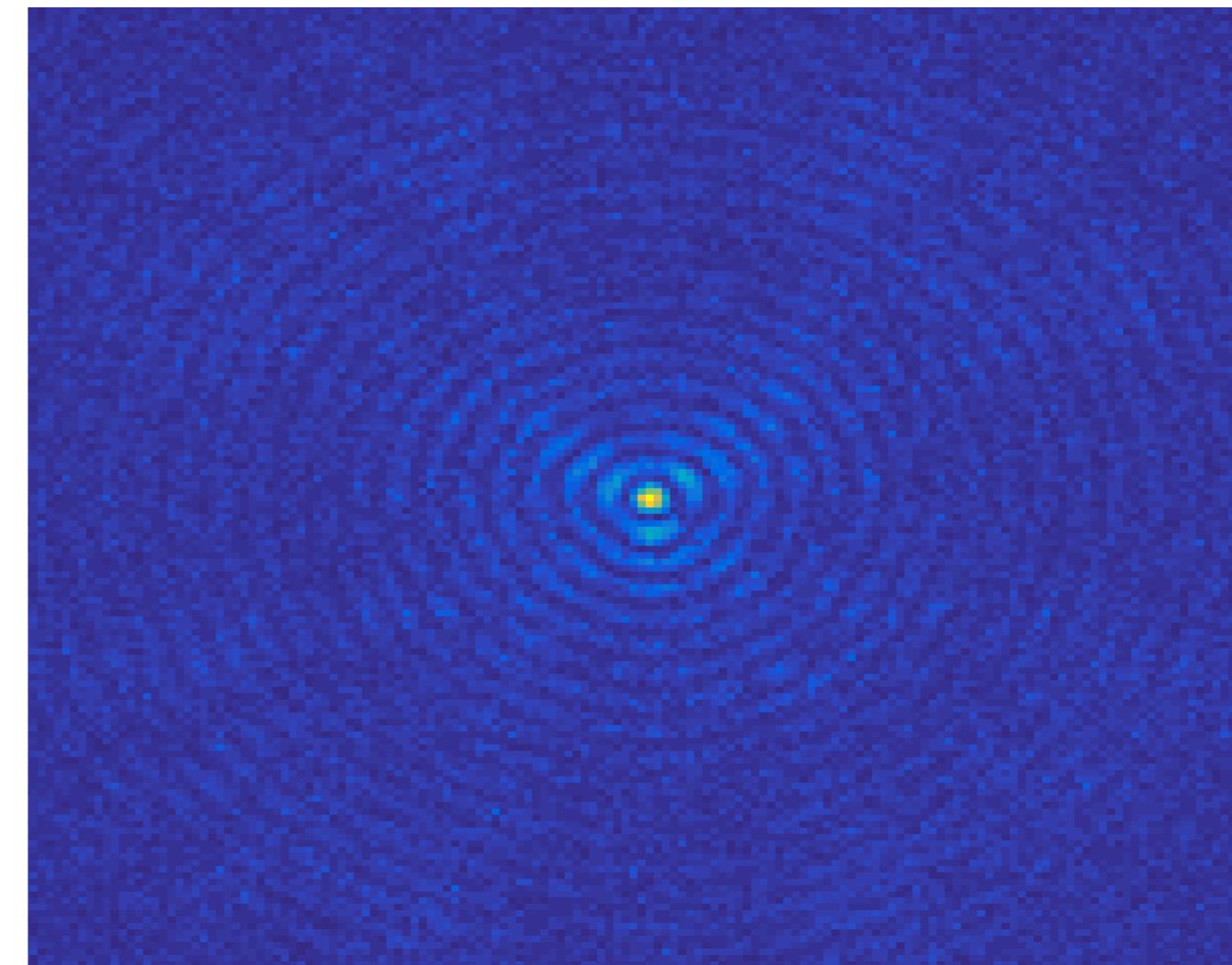


Telescope tomography

■ Results on HST-like telescope with noise

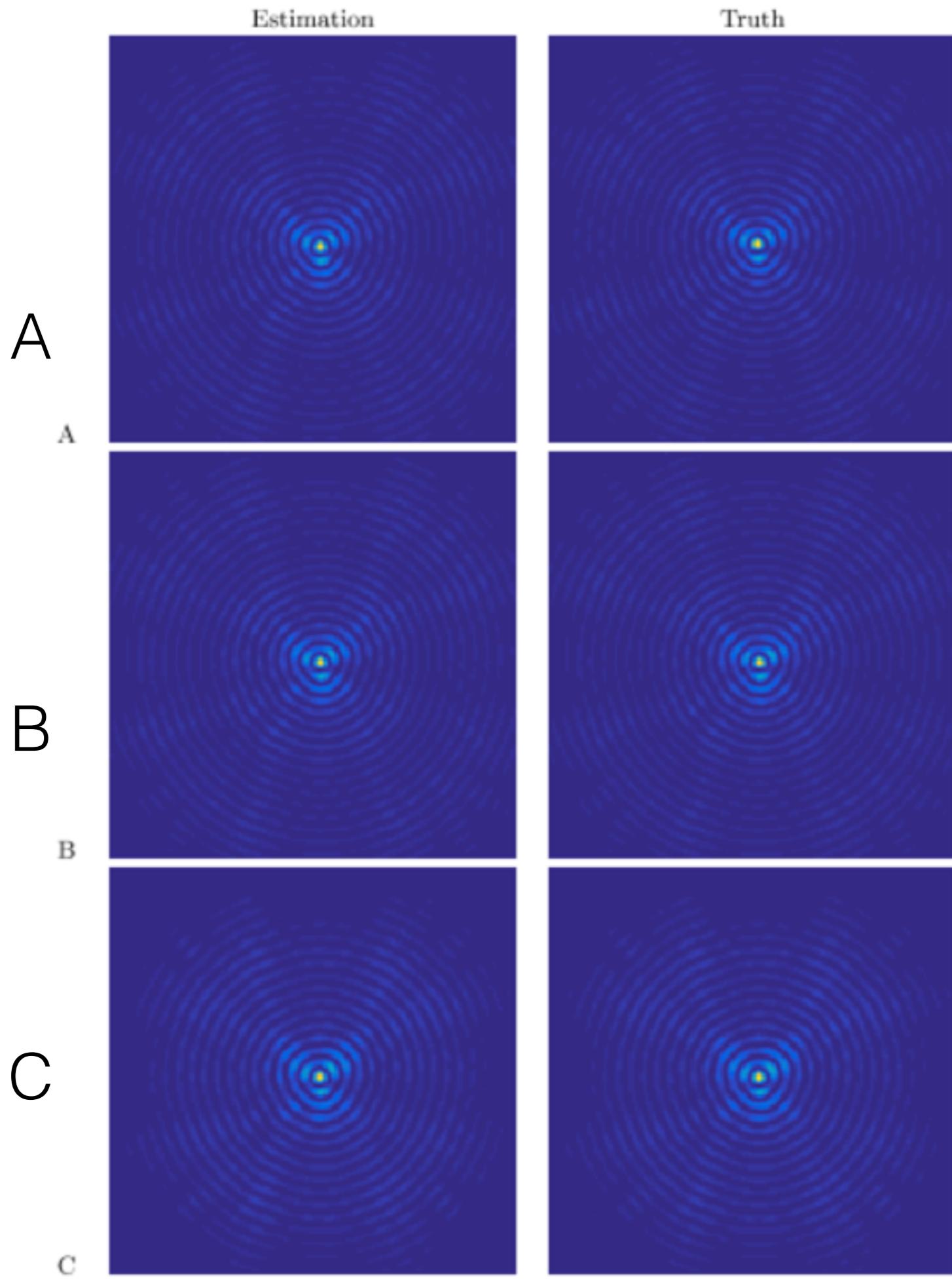


Field of view

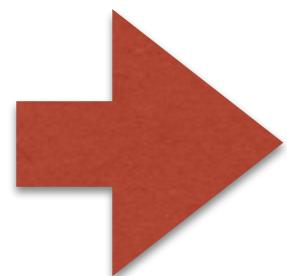


Star A

50 stars, 26 000 photons per star (256 at max), 5 e- of noise
Strong aberrations on both mirrors



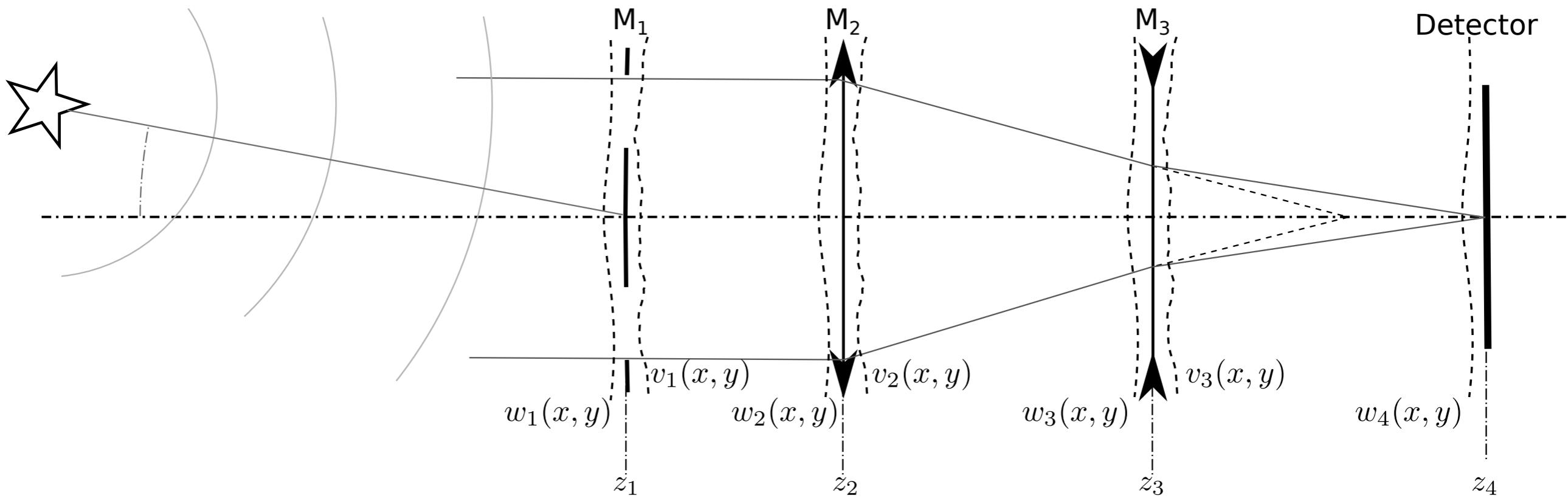
Fast light propagation through telescope



$$\boldsymbol{\alpha} = \arg \min_{\boldsymbol{\alpha}} \sum_{s=1}^S \sum_{n=1}^N |w_{K,n}^s(\boldsymbol{\alpha}) - t_n^s(\boldsymbol{\alpha}) - u_n^s|^2$$

$$\text{with } t_n^s(\boldsymbol{\alpha}) = \arg \min_{t \in \mathbb{C}} \frac{1}{\sigma_n^2} \left(|t|^2 - d_n^s \right)^2 + \frac{\rho}{2} |t - w_{K,n}^s(\boldsymbol{\alpha}) + u_n^s|^2$$

Forward model



Wavefront w_k on optical surface k

$$w_k(\alpha) = \mathbf{H}_{k-1} \mathbf{M}_{k-1} \mathbf{A}_{k-1}(\alpha_k) w_{k-1}(\alpha)$$

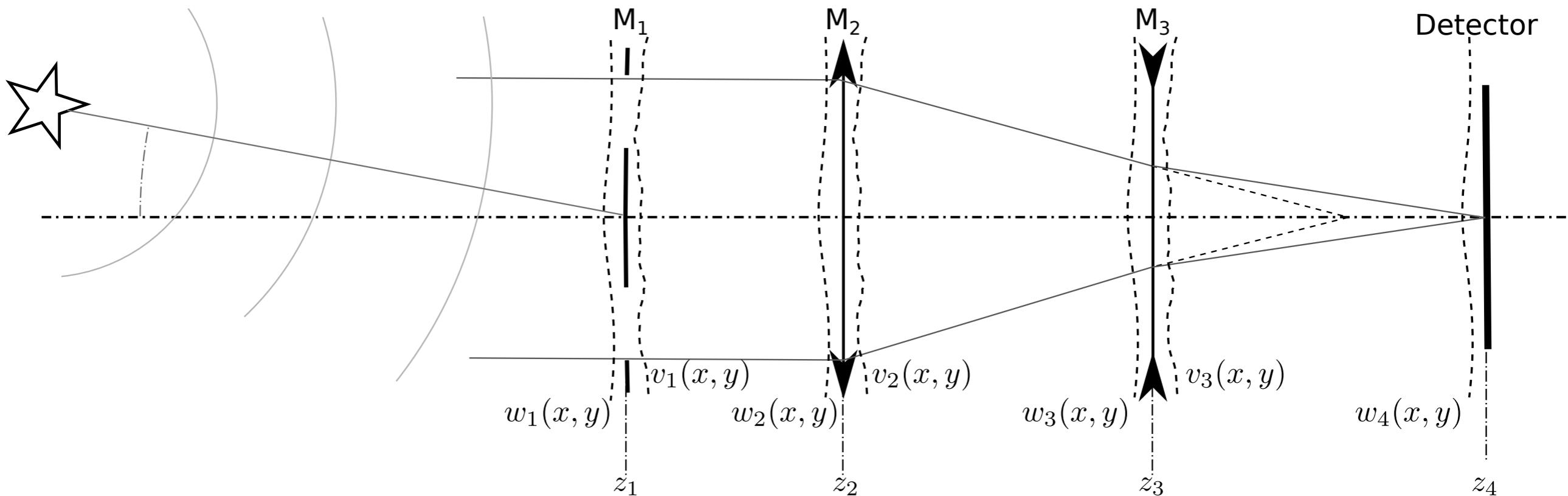
■ Incoming wave

Star are infinitely far: w_1 is a tilted plane wave

$$w_1(x, y) = \exp(i(x \sin(\theta_1)/\lambda + y \sin(\theta_2)/\lambda))$$

a **shift** in Fourier domain

Forward model



$$w_k(\alpha) = \mathbf{H}_{k-1} \mathbf{M}_{k-1} \mathbf{A}_{k-1}(\alpha_k) w_{k-1}(\alpha)$$

■ Propagation: paraxial / Fresnel approximation

$$\mathbf{H}_k = \mathbf{F}^{-1} \cdot \text{diag} \left(\hat{h}_{(z_{k+1} - z_k)} \right) \cdot \mathbf{F}$$

$$\hat{h}(\omega, z) = e^{-j z \frac{\|\omega\|^2}{2 k}}$$

Mirrors Model

■ Mirror transmittance

$$\mathbf{M}_k = \text{diag}(m_k)$$

$$m_k(\mathbf{x}) = \begin{cases} \exp\left(i 2 k \frac{\|\mathbf{x}\|^2}{R_k + \sqrt{R_k^2 - (1 - \epsilon_k^2) \|\mathbf{x}\|^2}}\right), & \text{if } \|\mathbf{x}\| \leq D_k \\ 0, & \text{otherwise.} \end{cases}$$

↑ ↗ ↑
curvature eccentricity

diameter
↓

■ Aberrations

$$\mathbf{A}_k = \text{diag}(a_k(\boldsymbol{\alpha}_k))$$

$$a_k(\boldsymbol{\alpha}_k) = \exp(i 2 \mathbf{Z}_k^T \cdot \boldsymbol{\alpha}_k)$$

↑ ↗
suitable basis (e.g. Zernike)

aberration coefficients

Sampling issues

■ Mirror transmittance

Maximum instant frequency for mirror k

$$\nu_k^{\max} = \frac{D_k}{\lambda \sqrt{D_k^2 (\epsilon_k^2 - 1) + 4R_k^2}},$$

■ Propagator

Sampling requirements

Maximum instant frequency for propagator k

$$L_k^{\max} = z \rho \lambda = \text{width of the field of view needed}$$

Periodicity issues

$$L_i = D_i + \frac{(z_{i+1} - z_i) \lambda}{2 \pi^2 D_i \vartheta}$$

 fraction of energy that spills from an aperture to another

Sampling issues

■ Numbers...

For HST (M1 to M2):

- maximum pixel size: $2.3 \mu\text{m}$
- number of pixels $\gg 10^6 \times 10^6$

■ Solution

inspired from Sziklas & Siegman, 1975

Coordinate transform: propagation in a prime domain:

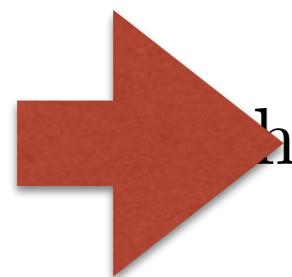
$$w_k(x) = w'_k(x') \exp \left(i \left(\beta_k \|x\|^2 + \langle \theta'_k \cdot x \rangle \right) \right)$$

Mirrors are much smoother in primed coordinate

maximum pixel size: 3.8 mm $\rightarrow 1500 \times 1500$ pixels

Proximity operator for phase retrieval

$$\boldsymbol{\alpha} = \arg \min_{\boldsymbol{\alpha}} \sum_{s=1}^S \sum_{n=1}^N |w_{K,n}^s(\boldsymbol{\alpha}) - t_n^s(\boldsymbol{\alpha}) - u_n^s|^2$$


$$t_n^s(\boldsymbol{\alpha}) = \arg \min_{t \in \mathbb{C}} \frac{1}{\sigma_n^2} \left(|t|^2 - d_n^s \right)^2 + \frac{\rho}{2} |t - w_{K,n}^s(\boldsymbol{\alpha}) + u_n^s|^2$$

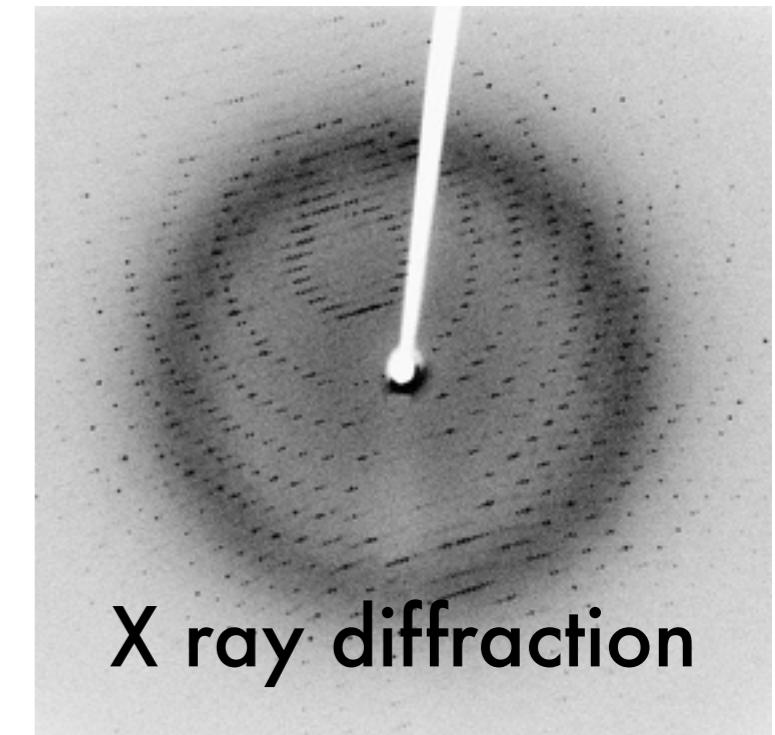
Phase retrieval in coherent imaging

■ A twofold forward model

Wave optics model

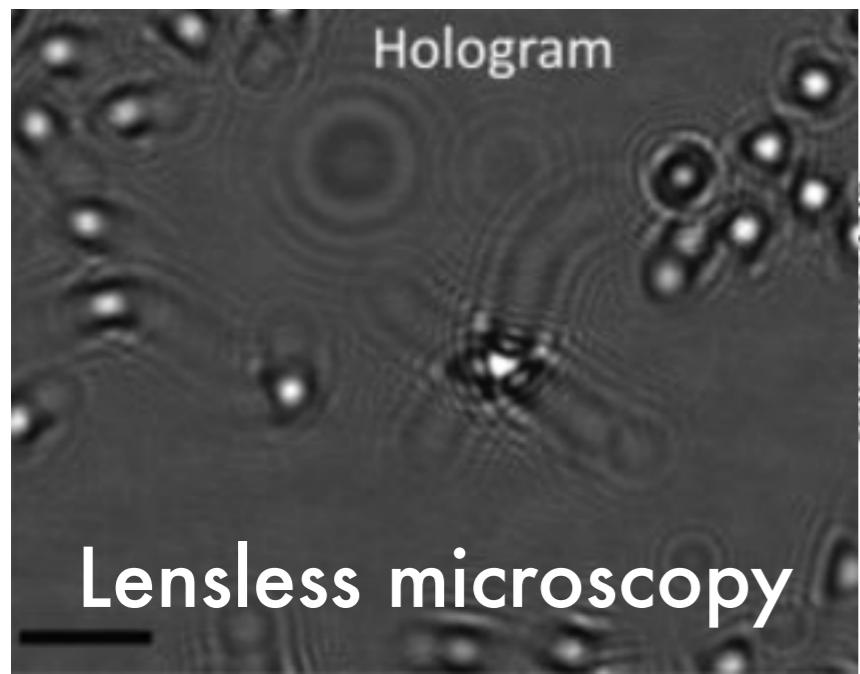
$$\mathbf{y} = \mathcal{H}(\mathbf{x})$$

Intensity measurements $d_k = |\mathbf{y}_k|^2 + n_k$



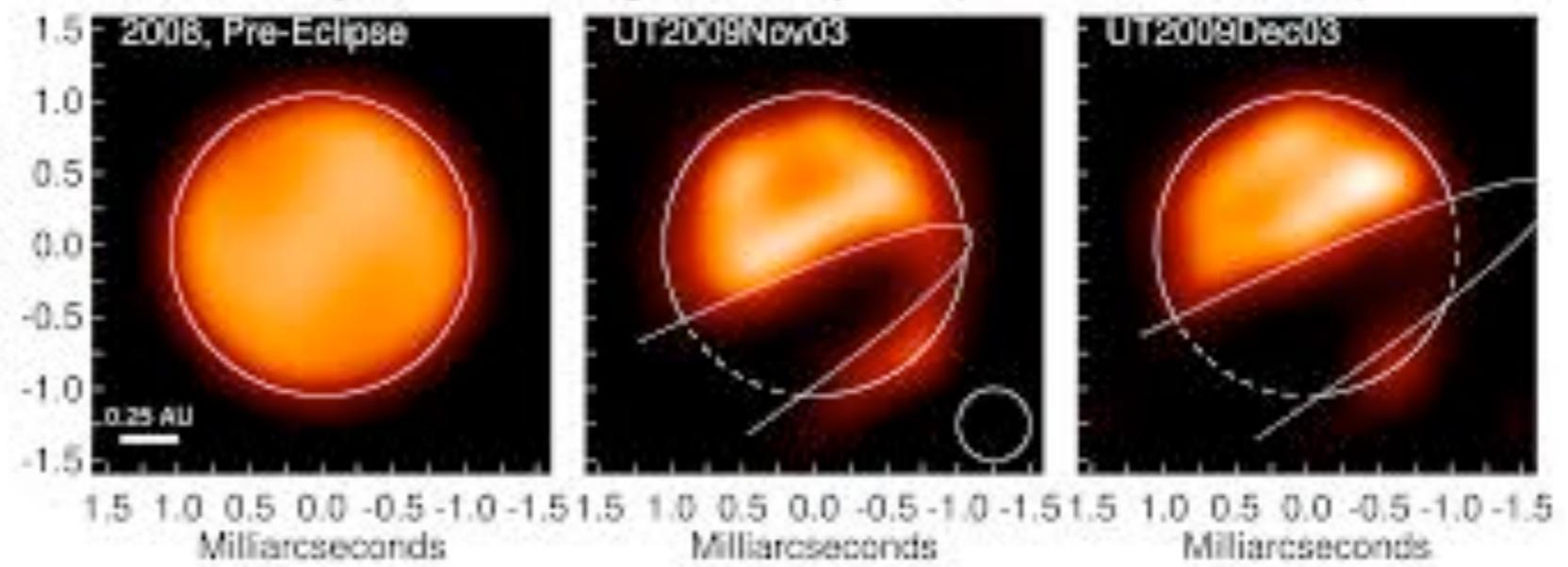
■ Arise in numerous applications

Hologram



Lensless microscopy

Epsilon Aurigae Eclipse (CHARA-MIRC)



Stellar interferometry

The phase retrieval problem

■ Estimating complex valued signal from intensity

$$d_2 = |x_2|^2 + n$$

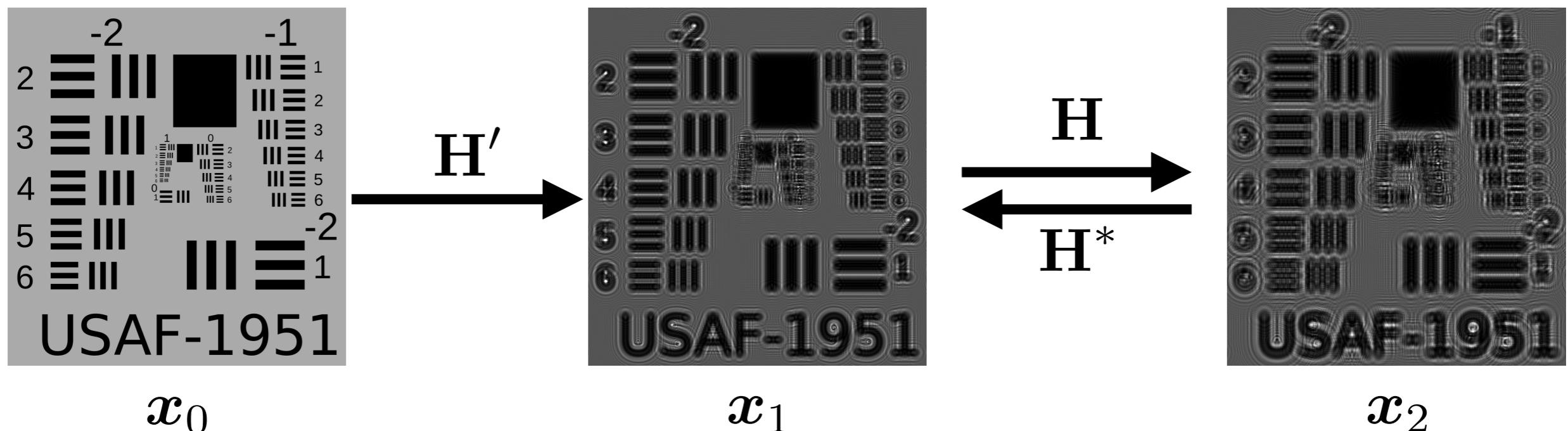
■ An abundant literature

- iterative projections [Misell 1973, Fienup 1980, Bauschke 2003,...]
- semi-definite programming (Phase lifting) [Candes 2013, Fogel 2013]
- phase marginalization in variational bayesian approach [Dremeau 2015]

Gerchberg Saxton Algorithm

■ The simplest setup

- estimate the wavefront \mathbf{x}_1 in plane z_1 given intensity measurements at two different depth z_1 and z_2



■ Propagation

- Propagation using either Fresnel or Fraunhoffer approximation:

$$\mathbf{x}_2 = \mathbf{H} \mathbf{x}_1$$

$$\mathbf{x}_1 = \mathbf{H}^* \mathbf{x}_2$$

Gerchberg-Saxton Algorithm

■ A non convex POCS algorithm

Algorithm 1. Gerchberg-Saxton algorithm

```
1: procedure GS( $d_A, d_B$ )
2:    $\mathbf{x}^{(0)} = \sqrt{d_A}$                                  $\triangleright$  Initialization
3:   for  $n = 1, 2, \dots, \text{maxiter}$  do
4:      $\mathbf{y}^{(n+1/2)} = \mathbf{H} \cdot \mathbf{x}^{(n)}$      $\triangleright$  Propagation to the  $z_B$  plane
5:      $\mathbf{y}^{(n)} = P_B(\mathbf{y}^{(n+1/2)})$            $\triangleright$  Projection
6:      $\mathbf{x}^{(n+1/2)} = \mathbf{H}^{-1} \cdot \mathbf{y}^{(n)}$    $\triangleright$  Back propagation to the  $z_A$ 
    plane
7:      $\mathbf{x}^{(n)} = P_A(\mathbf{x}^{(n+1/2)})$            $\triangleright$  Projection
8:   return  $\mathbf{x}^{(\text{maxiter})}$        $\triangleright$  The complex amplitude in the  $z_A$ 
  plane
```

■ Projection step

$$P(x_k | d_k) = \begin{cases} \frac{x_k}{|x_k|} \sqrt{d_k}, & \text{if } |x_k| > 0 \\ \sqrt{d_k}, & \text{otherwise.} \end{cases}$$

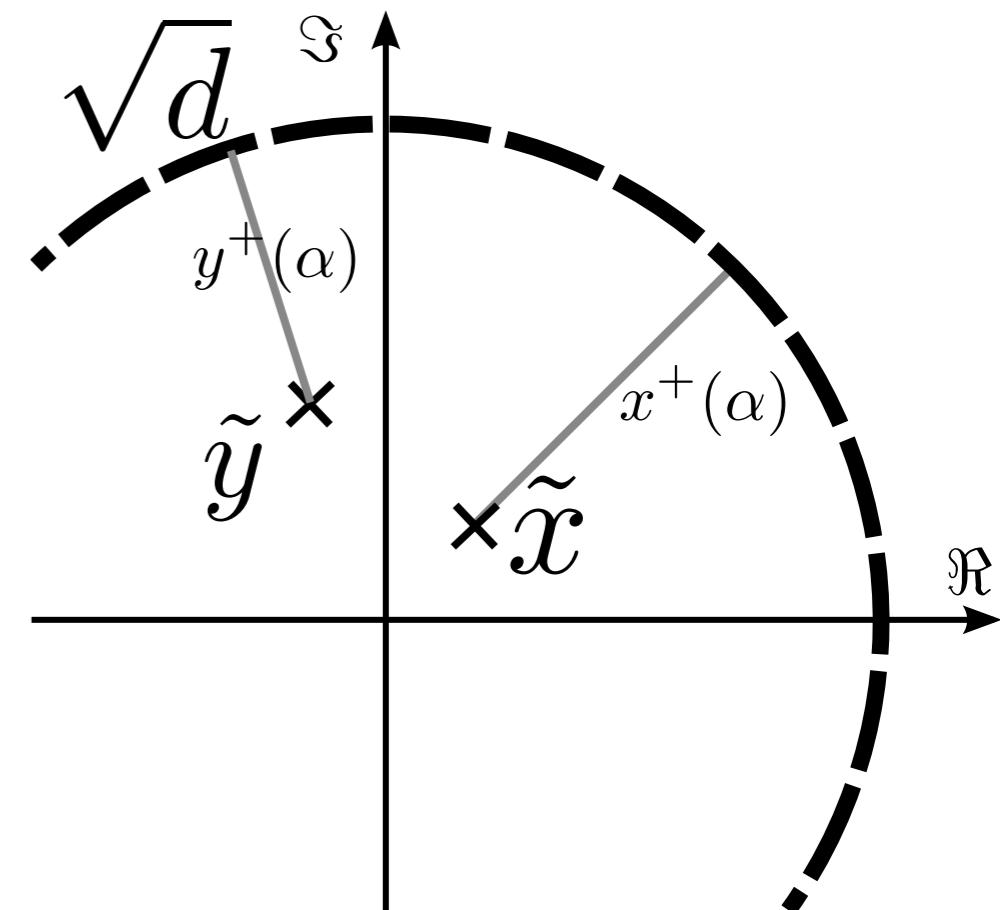
Proximity operator for intensity

■ Modulus only proximity operator

$$\text{prox}_\ell(\tilde{\rho}) = \arg \min_{\rho \geq 0} \left\{ \ell(d - \rho^2) + \frac{1}{2} (\rho - \tilde{\rho})^2 \right\}$$

■ Properties

- lower semi-continuous
- prox-bounded
- not prox-regular in $\{0\}$
- non-expansive only where $\tilde{\rho} > \sqrt{d}$



Proximity operator for intensity

■ Maximum likelihood $f_k(x) = \ell_k(d_k - |x_k|^2)$

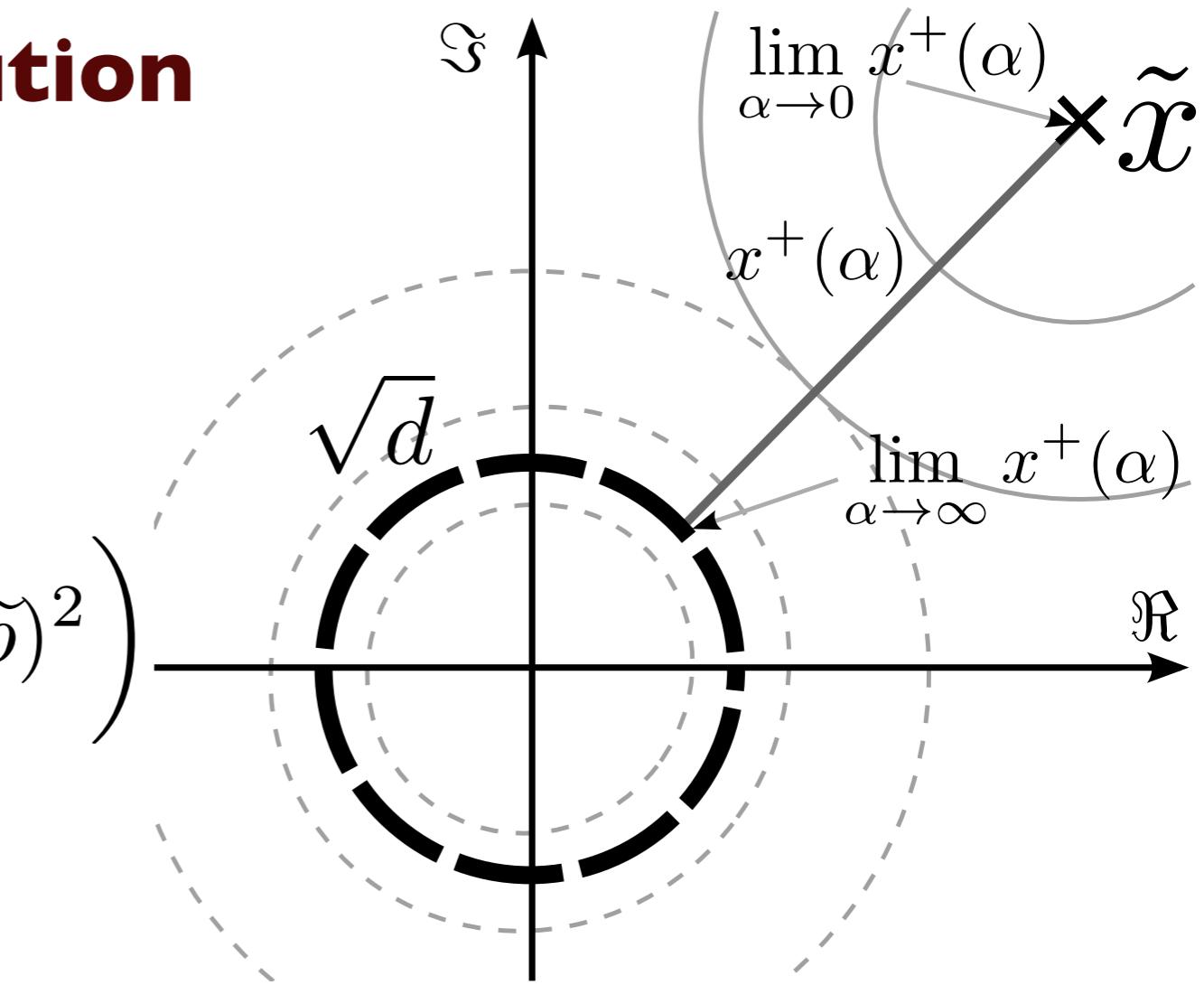
- is the neg-loglikelihood of the noise distribution

■ Phase independent solution

$$\rho^+ e^{i\phi^+} = \text{prox}_{\alpha f} (\rho e^{i\phi})$$

$$\phi^+ = \phi$$

$$\rho^+ = \arg \min_{\rho \geq 0} \left(\alpha f(\rho) + \frac{1}{2} (\rho - \tilde{\rho})^2 \right)$$



Proximity operator for intensity

■ **Gaussian likelihood** $f(x) = \frac{1}{\sigma^2} (|x|^2 - d)^2$

$$\text{prox}(\tilde{\rho}) = \arg \min_{\rho \geq 0} \left\{ \frac{1}{\sigma^2} (\rho^2 - d)^2 + \frac{1}{2} (\rho - \tilde{\rho})^2 \right\}$$

$$\begin{aligned} q_G(\rho) &= \frac{d}{d\rho} \left(\frac{1}{\sigma^2} (\rho^2 - d)^2 + \frac{1}{2} (\rho - \tilde{\rho})^2 \right) \\ &= 4 \frac{1}{\sigma^2} \rho^3 + \rho \left(1 - 4 \frac{1}{\sigma^2} d \right) - \tilde{\rho}. \end{aligned}$$

3rd order polynomial with only one root in \mathbb{R}^+ computed using Cardano Method

Proximity operator for intensity

■ **Poisson likelihood** $f(x) = |x|^2 - d \log(|x|^2 + b)$

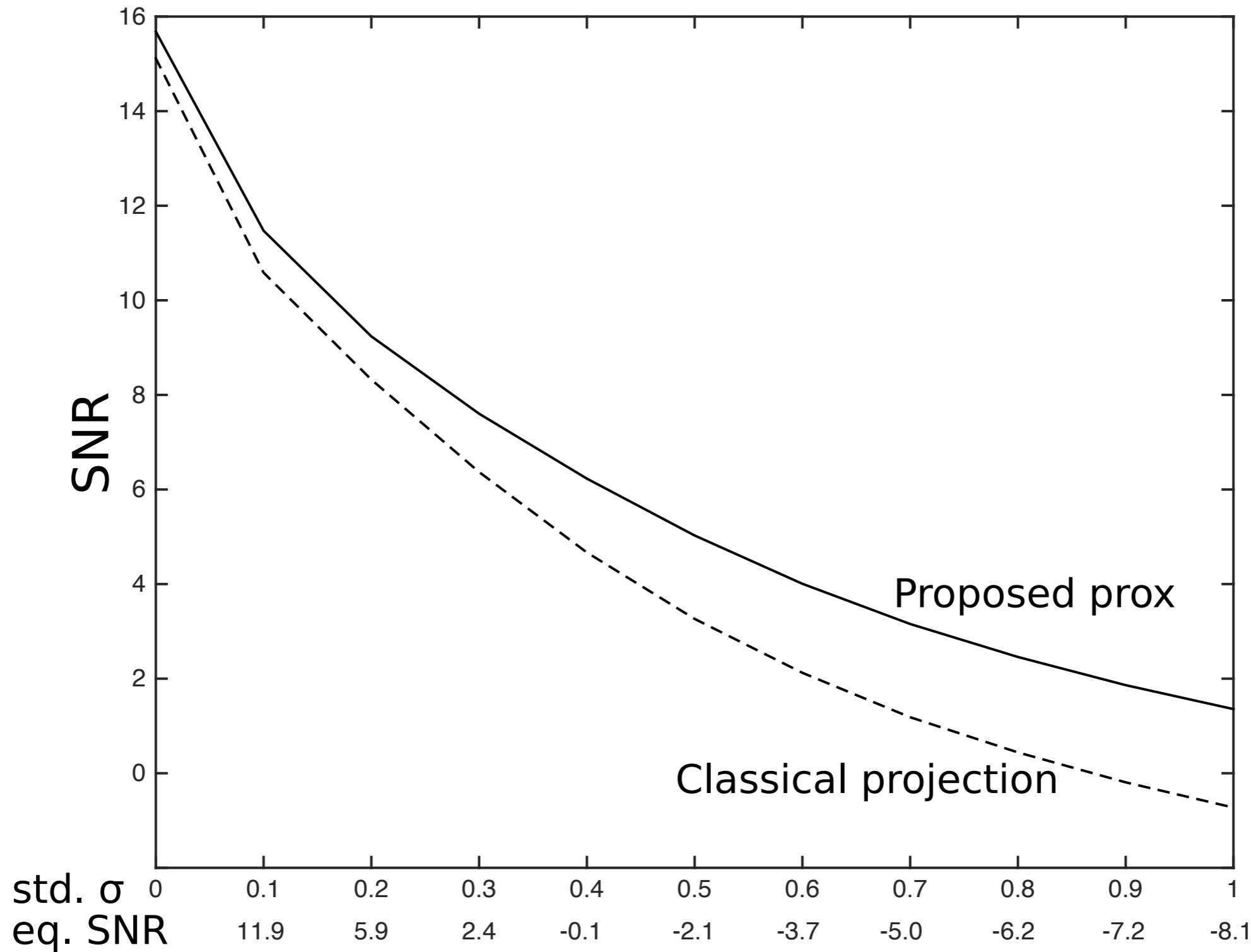
$$\begin{aligned} q_P(\rho) &= \frac{d}{d\rho} \left(\alpha f(\rho) + \frac{1}{2} (\rho - \tilde{\rho})^2 \right) \\ &= (2\alpha + 1)\rho^3 - \tilde{\rho}\rho^2 + ((2\alpha + 1)b - 2\alpha d)\rho - b\tilde{\rho} \end{aligned}$$

3rd order polynomial with only one root in \mathbb{R}^+
computed using Cardano Method

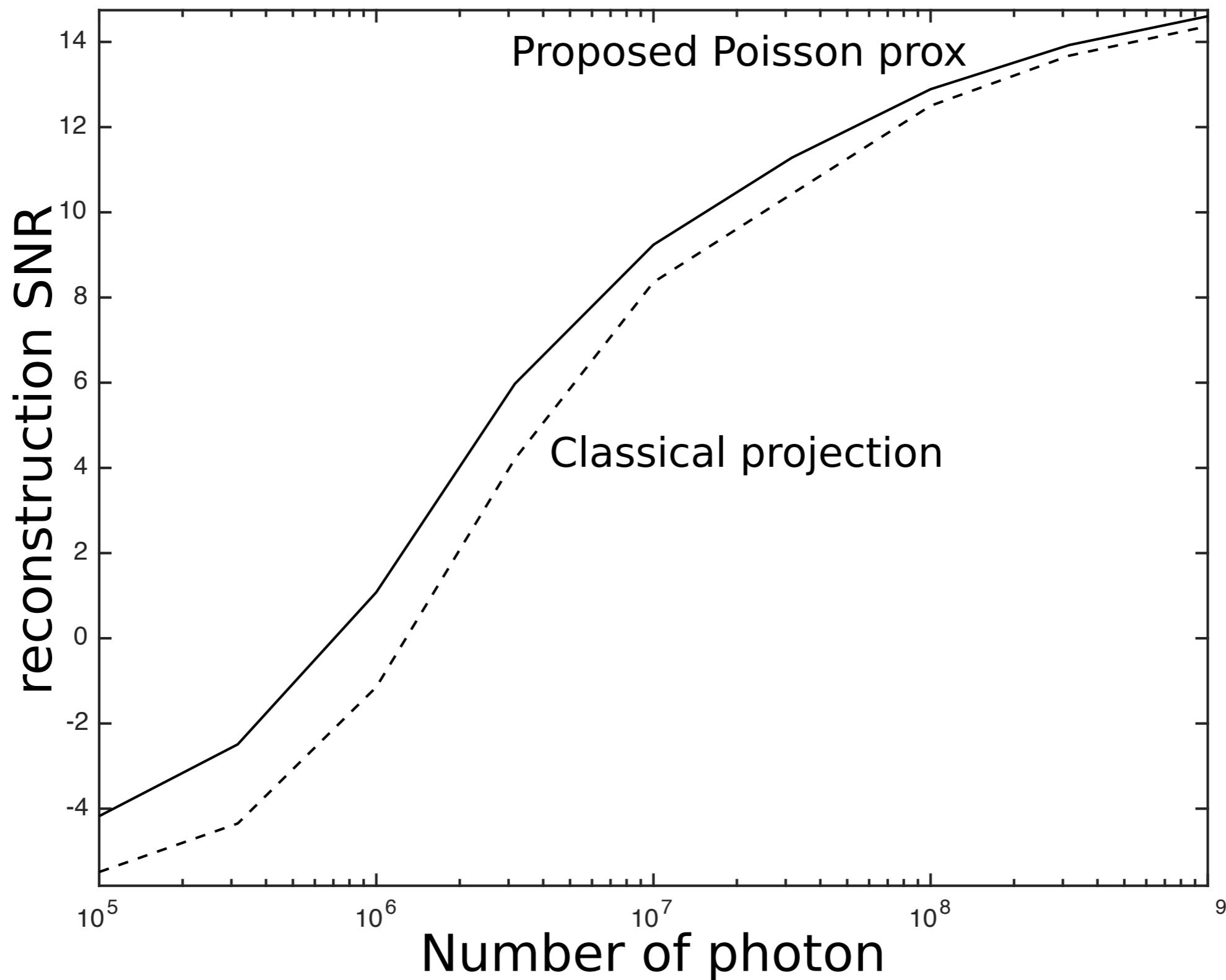
if $b=0$ then it reduces to a 2nd order polynomial

$$\rho^+ = \frac{\tilde{\rho} + \sqrt{8d\alpha(1+2\alpha) + \tilde{\rho}^2}}{2+4\alpha}.$$

Numerical results : Gaussian noise



Numerical results : Poisson noise



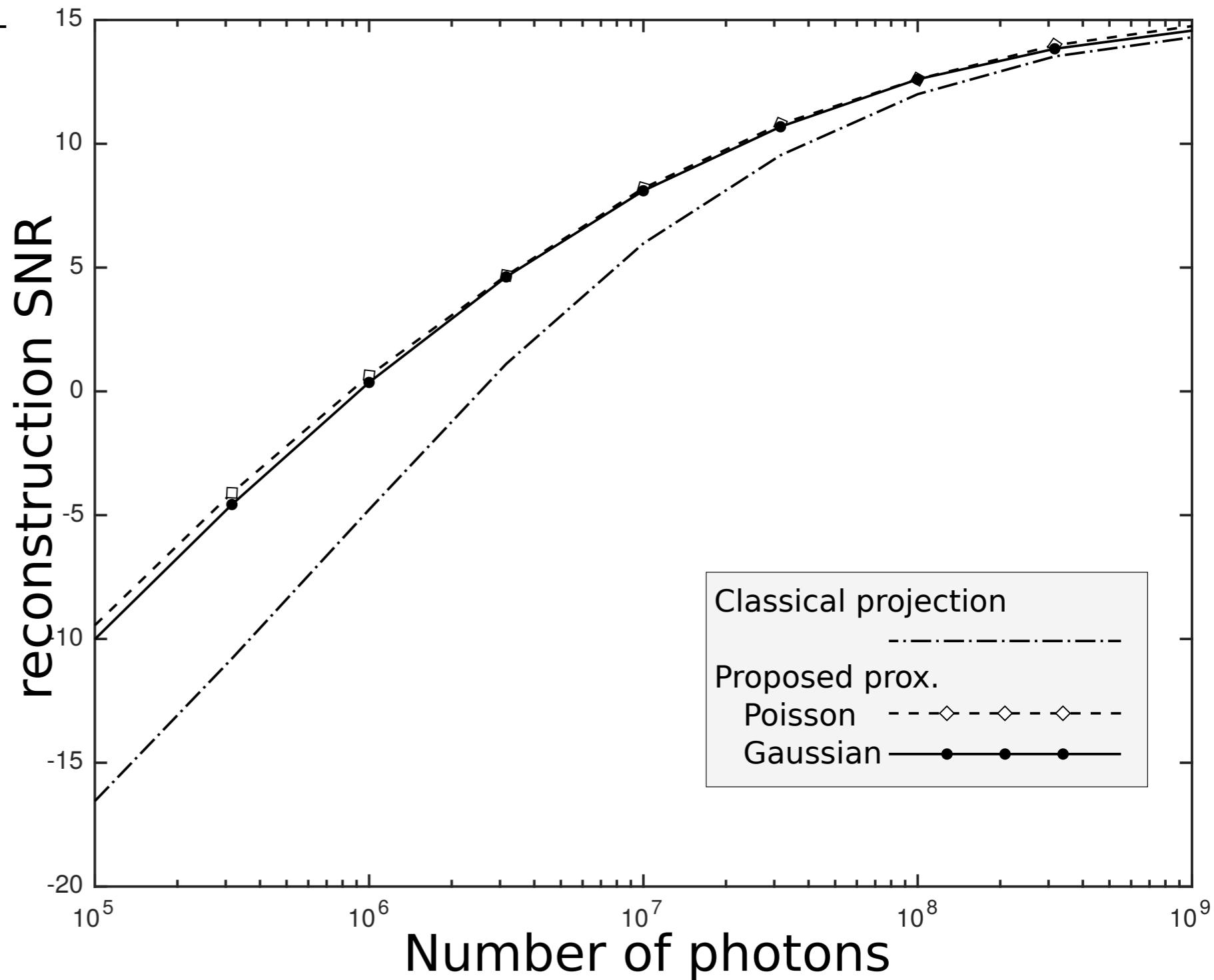
Numerical results : low light condition

■ Gaussian Vs Poisson model

Dark current $b=3e^-$

**Gaussian
approximation**

$$\sigma^2 = \max(d_k, b)$$



Proximity operator for sum of intensities

■ Sum of intensities model

- Undersampled fringes, broadband imaging,...

$$d_k = \|\mathbf{y}_k\|_2^2 + n_k ,$$

setting $\mathbf{y}_k = \eta \mathbf{u}$, with $\eta \geq 0$ and $\|\mathbf{u}\|_2 = 1$

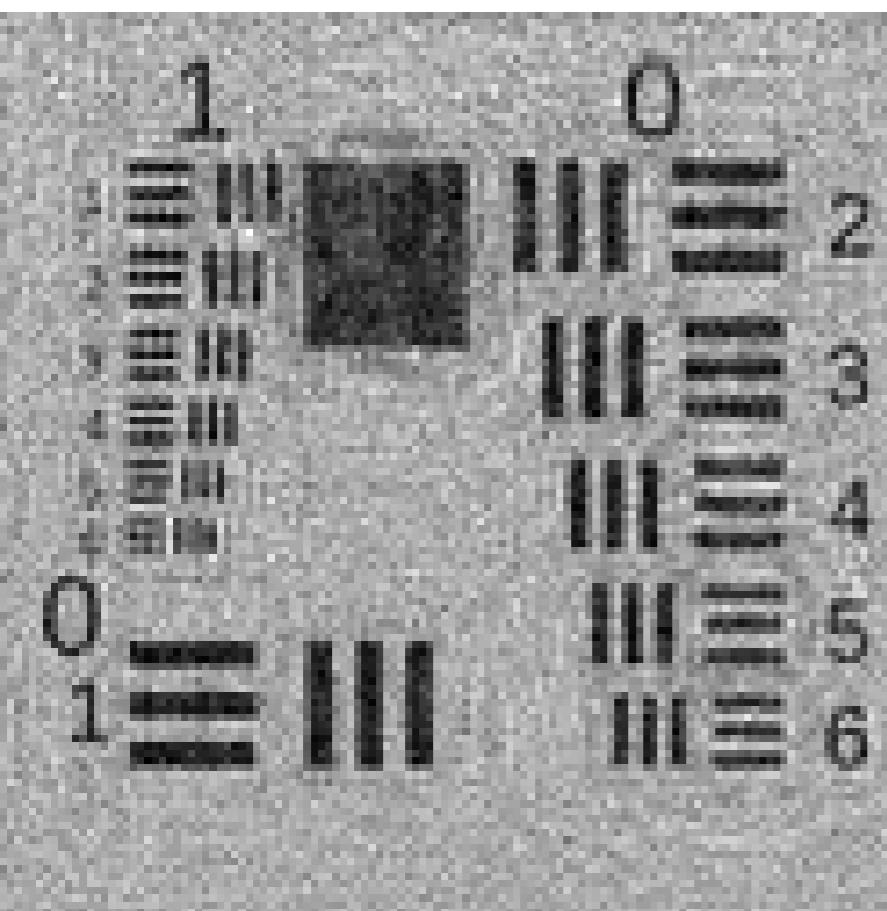
$$\text{prox}_f(\tilde{\mathbf{y}}) = \begin{cases} \eta^+, & \text{if } \|\tilde{\mathbf{y}}\|_2 = 0 \\ \eta^+ \frac{\tilde{\mathbf{y}}}{\|\tilde{\mathbf{y}}\|_2}, & \text{otherwise.} \end{cases}$$

$$\begin{aligned} \text{With } \eta^+ &= \arg \min_{\eta} \left(\min_{\mathbf{u}, \|\mathbf{u}\|=1} \left(f(\eta) + \frac{1}{2} \|\eta \mathbf{u} - \tilde{\mathbf{y}}\|_2^2 \right) \right), \\ &= \arg \min_{\eta>0} \left(f(\eta) + \frac{1}{2} (\eta - \|\tilde{\mathbf{y}}\|_2)^2 \right), \end{aligned}$$

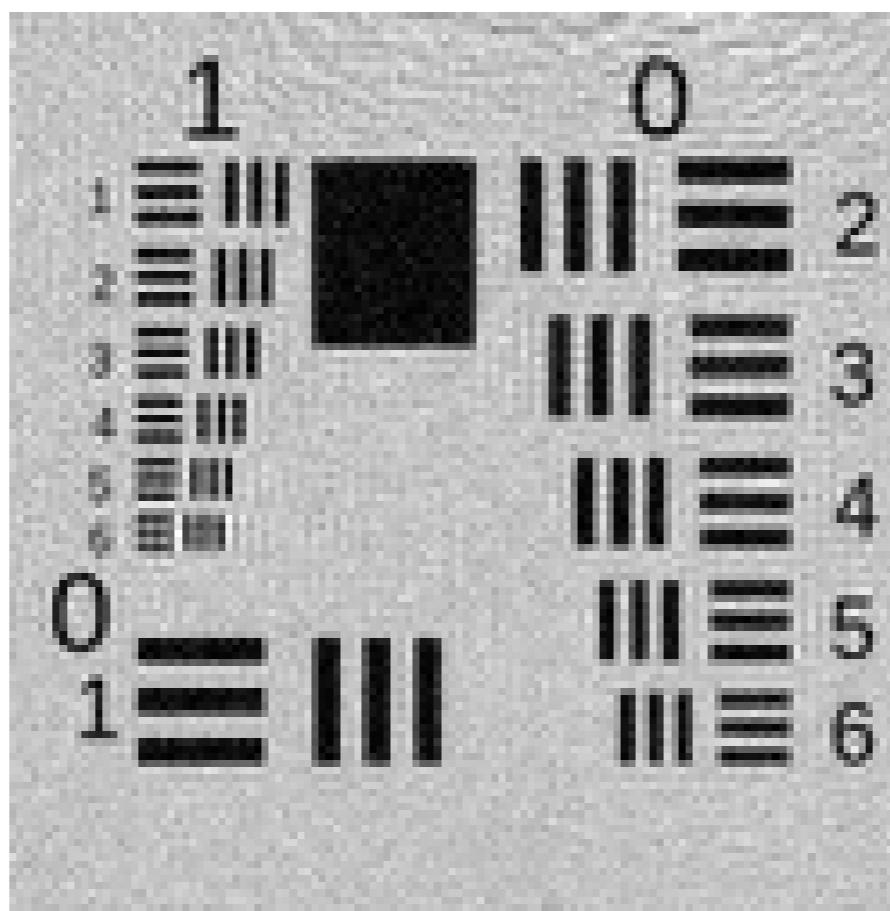
Proximity operator for sum of intensities

■ Results: trading noise for resolution

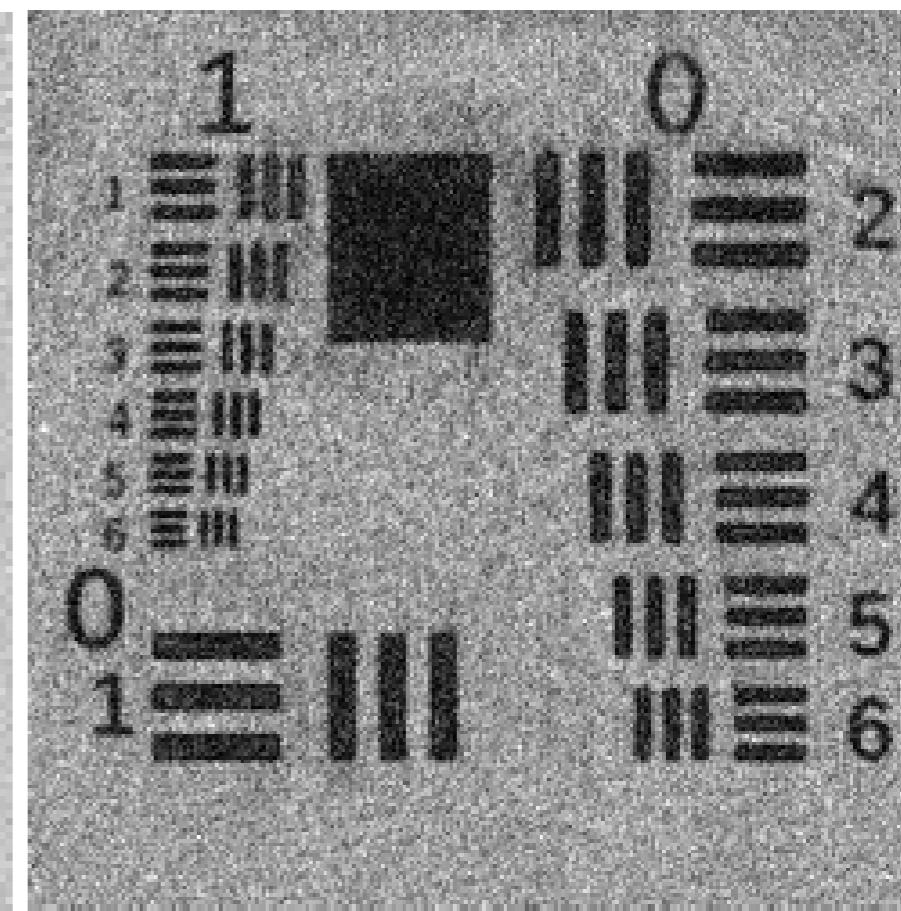
- Data down-sampled by 2x2
- Measurements at 8 different depths



Classical,
2 planes



Classical,
8 planes



Super resolution
8 planes

Image backpropagated at z_0

Telescope tomography

■ Pros

- Optically motivated mode (restriction to the plausible PSF only),
- Irreducible formulation: should be immune to binarity bias,
- Gives the PSF everywhere (no interpolation needed),
- Can easily take into account variation of aberrations with time.

■ Cons

- Only accounts for optical aberrations (no jitter, CTI,...),
- Quite computationally intensive (but can be easily parallelized),
- Propagation model needs to be accurate,
- Needs a good knowledge on Euclid optical design,
- Yet to be tested (broadband and under sampled case).