

# Using Small-scale Lensing Anisotropies to Constrain Dark Matter

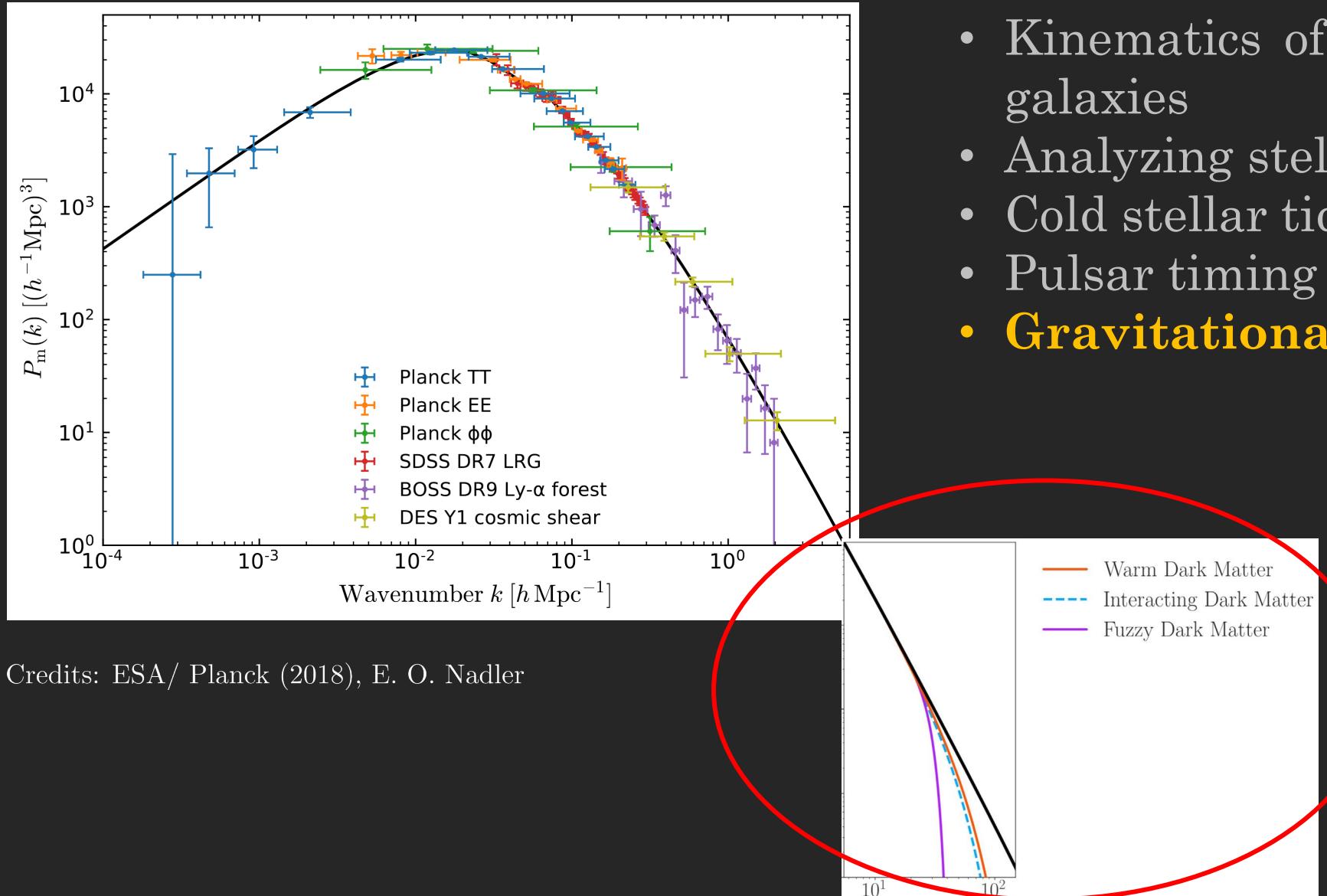
Birendra Dhanasingham

University of New Mexico, USA

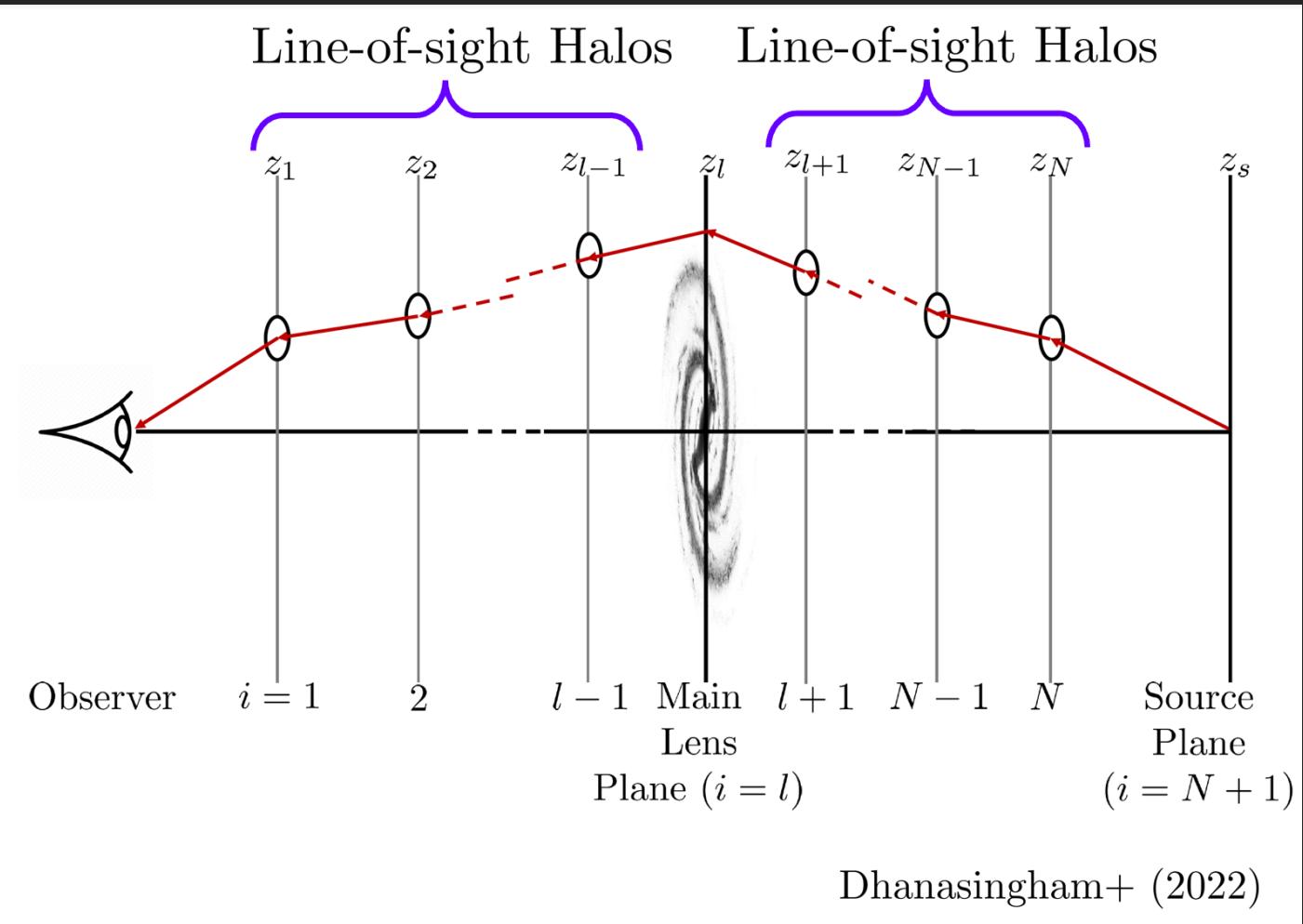
With

Francis-Yan Cyr-Racine (UNM), Annika Peter (OSU), Andrew Benson (Carnegie Observatories), and Daniel Gilman (U. of Toronto)

# Probes of Dark Matter



# Multiplane Gravitational Lensing



$$\mathbf{u} = \mathbf{x}_1 - \sum_{i=1}^N \boldsymbol{\alpha}_i(\mathbf{x}_i)$$

Recursive multiplane lens equation:

$$\mathbf{x}_j = \mathbf{x}_1 - \sum_{i=1}^{j-1} \beta_{ij} \boldsymbol{\alpha}_i(\mathbf{x}_i)$$

where

$$\beta_{ij} = \frac{D_{ij} D_s}{D_j D_{is}}$$

# A Simple Way: “Effective Multiplane Lensing”

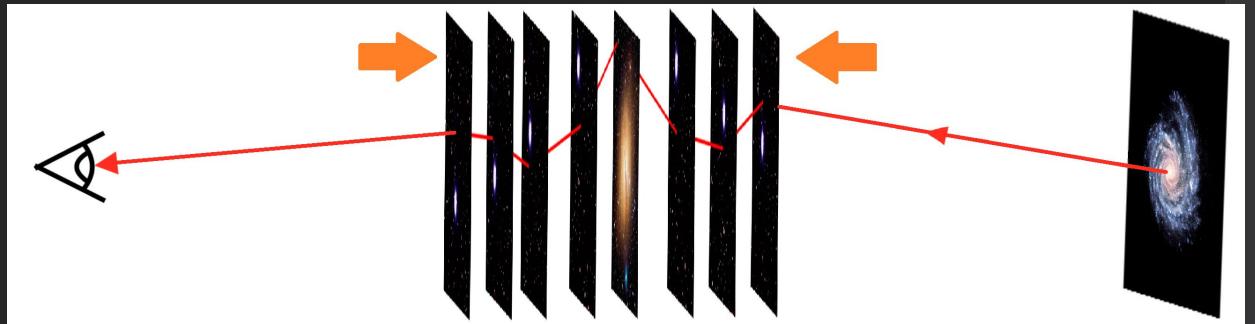
$$\boldsymbol{\alpha}_{\text{eff}}(\mathbf{x}) = \nabla \phi_{\text{eff}}(\mathbf{x}) + \nabla \times \mathbf{A}_{\text{eff}}(\mathbf{x})$$

$$\mathbf{u} = \mathbf{x} - \boldsymbol{\alpha}_{\text{eff}}(\mathbf{x})$$

where

$\phi_{\text{eff}}$  - effective scalar potential

$\mathbf{A}_{\text{eff}}$  - effective vector potential



$\mathbf{A}_{\text{eff}}$  points in line-of-sight direction. Therefore, introduce  $\zeta_{\text{eff}}(\mathbf{x})$  such that  $\mathbf{A}_{\text{eff}}(\mathbf{x}) = \zeta_{\text{eff}}(\mathbf{x})\hat{\mathbf{z}}$

Divergence Component:

$$\kappa_{\text{div}} \equiv \frac{1}{2} \nabla \cdot \boldsymbol{\alpha}_{\text{eff}} - \kappa_0 = \frac{1}{2} \nabla^2 \phi_{\text{eff}} - \kappa_0$$

Curl Component:

$$\kappa_{\text{curl}} \equiv \frac{1}{2} \nabla \times \boldsymbol{\alpha}_{\text{eff}} \cdot \hat{\mathbf{z}} = -\frac{1}{2} \nabla^2 \zeta_{\text{eff}}$$

Gilman+ (2019),  
Çağan Sengül+ (2020)

# A Simple Way: “Effective Multiplane Lensing”

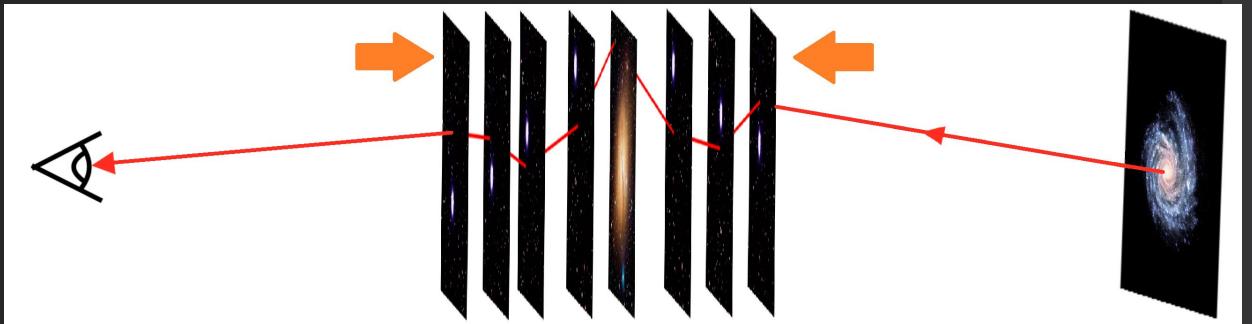
$$\boldsymbol{\alpha}_{\text{eff}}(\mathbf{x}) = \nabla \phi_{\text{eff}}(\mathbf{x}) + \nabla \times \mathbf{A}_{\text{eff}}(\mathbf{x})$$

$$\mathbf{u} = \mathbf{x} - \boldsymbol{\alpha}_{\text{eff}}(\mathbf{x})$$

where

$\phi_{\text{eff}}$  - effective scalar potential

$\mathbf{A}_{\text{eff}}$  - effective vector potential



$\mathbf{A}_{\text{eff}}$  points in line-of-sight direction. Therefore, introduce  $\zeta_{\text{eff}}(\mathbf{x})$  such that  $\mathbf{A}_{\text{eff}}(\mathbf{x}) = \zeta_{\text{eff}}(\mathbf{x})\hat{\mathbf{z}}$

Divergence Component:

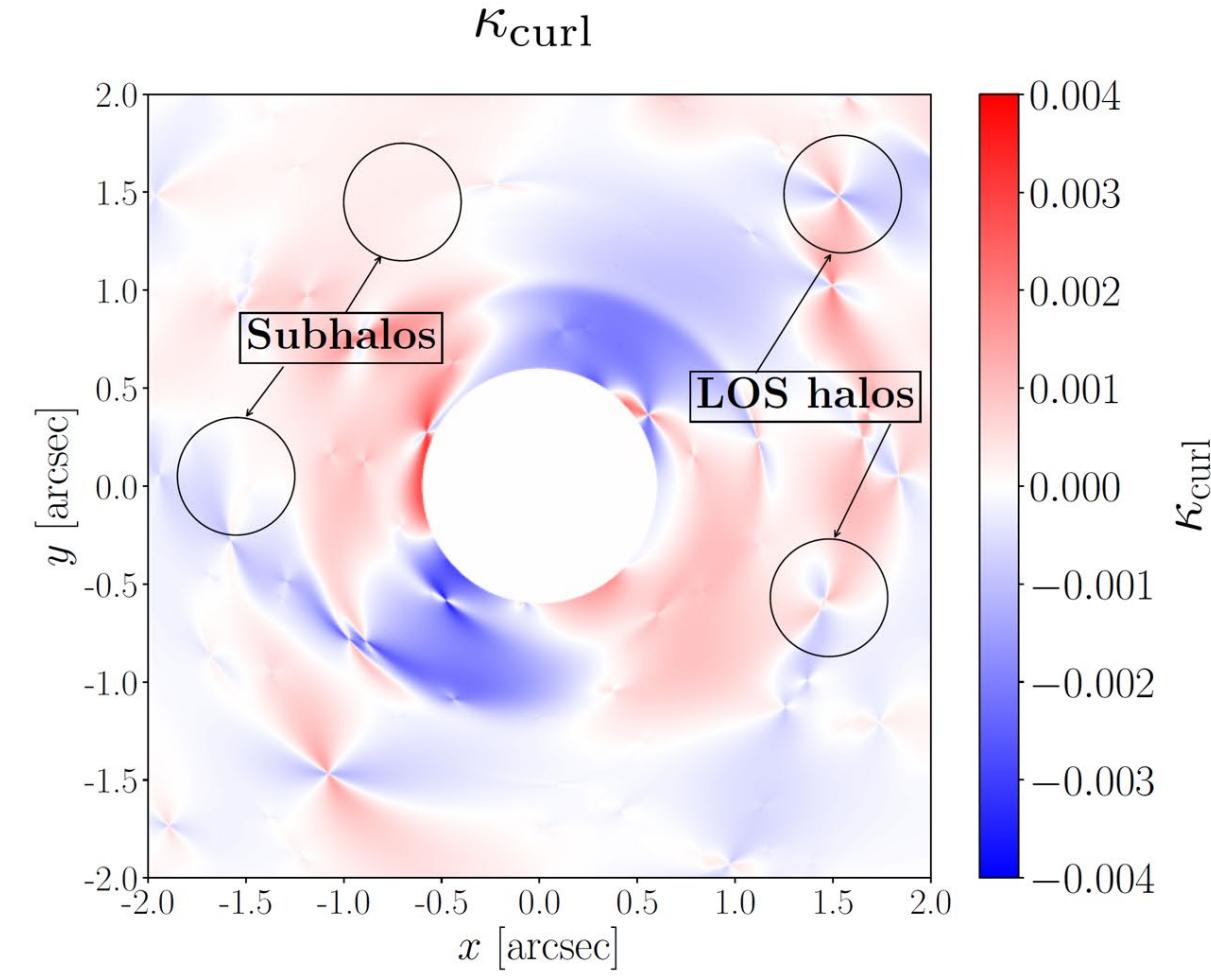
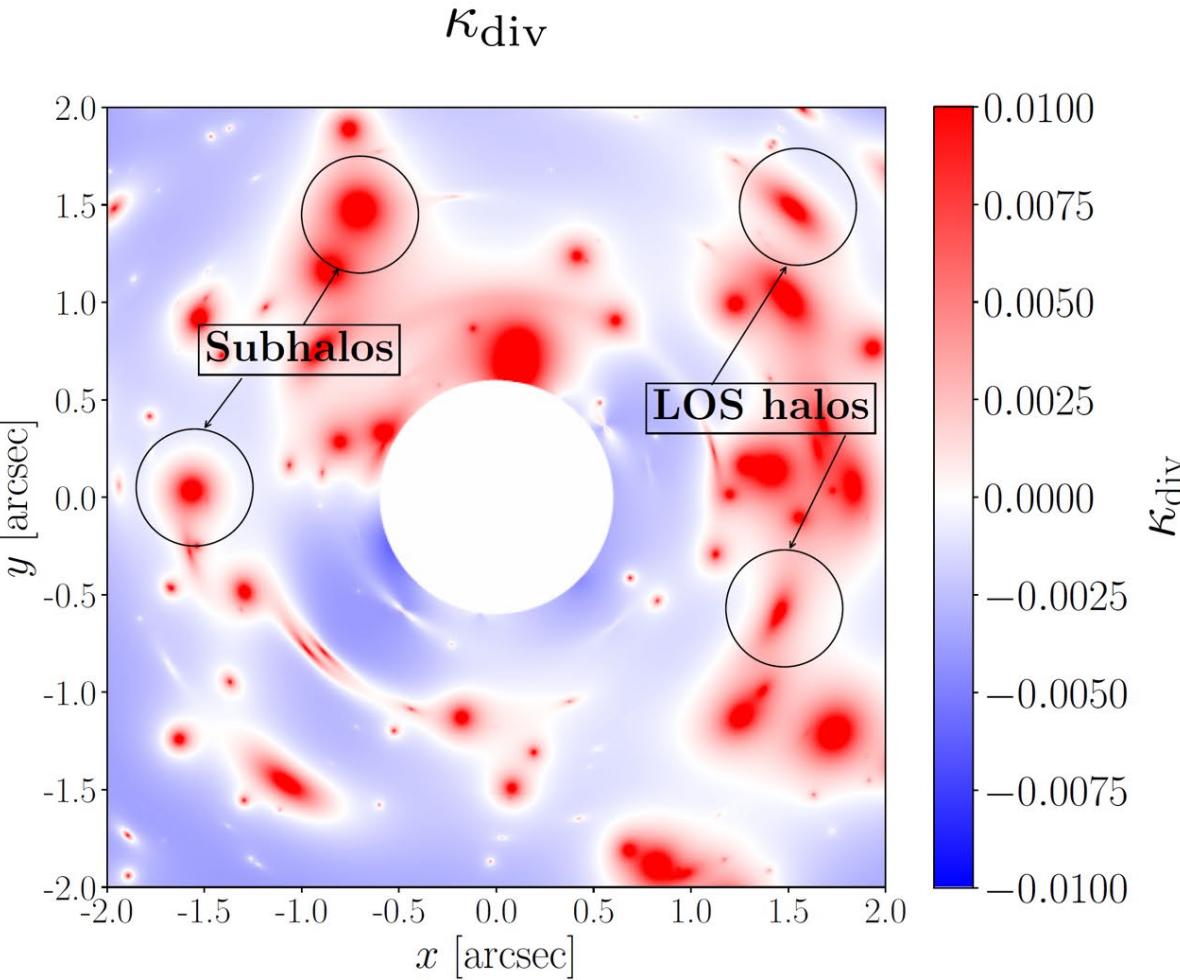
$$\kappa_{\text{div}} \equiv \frac{1}{2} \nabla \cdot \boldsymbol{\alpha}_{\text{eff}} - \kappa_0 = \frac{1}{2} \nabla^2 \phi_{\text{eff}} - \kappa_0$$

vanish for  
single-plane  
lensing

Curl Component:

$$\kappa_{\text{curl}} \equiv \frac{1}{2} \nabla \times \boldsymbol{\alpha}_{\text{eff}} \cdot \hat{\mathbf{z}} = -\frac{1}{2} \nabla^2 \zeta_{\text{eff}}$$

# Divergence and curl of the effective deflection field



(Gilman+ (2021), Çağan Sengül+ (2020))

# The Two-point Correlation Function

**The correlation function**

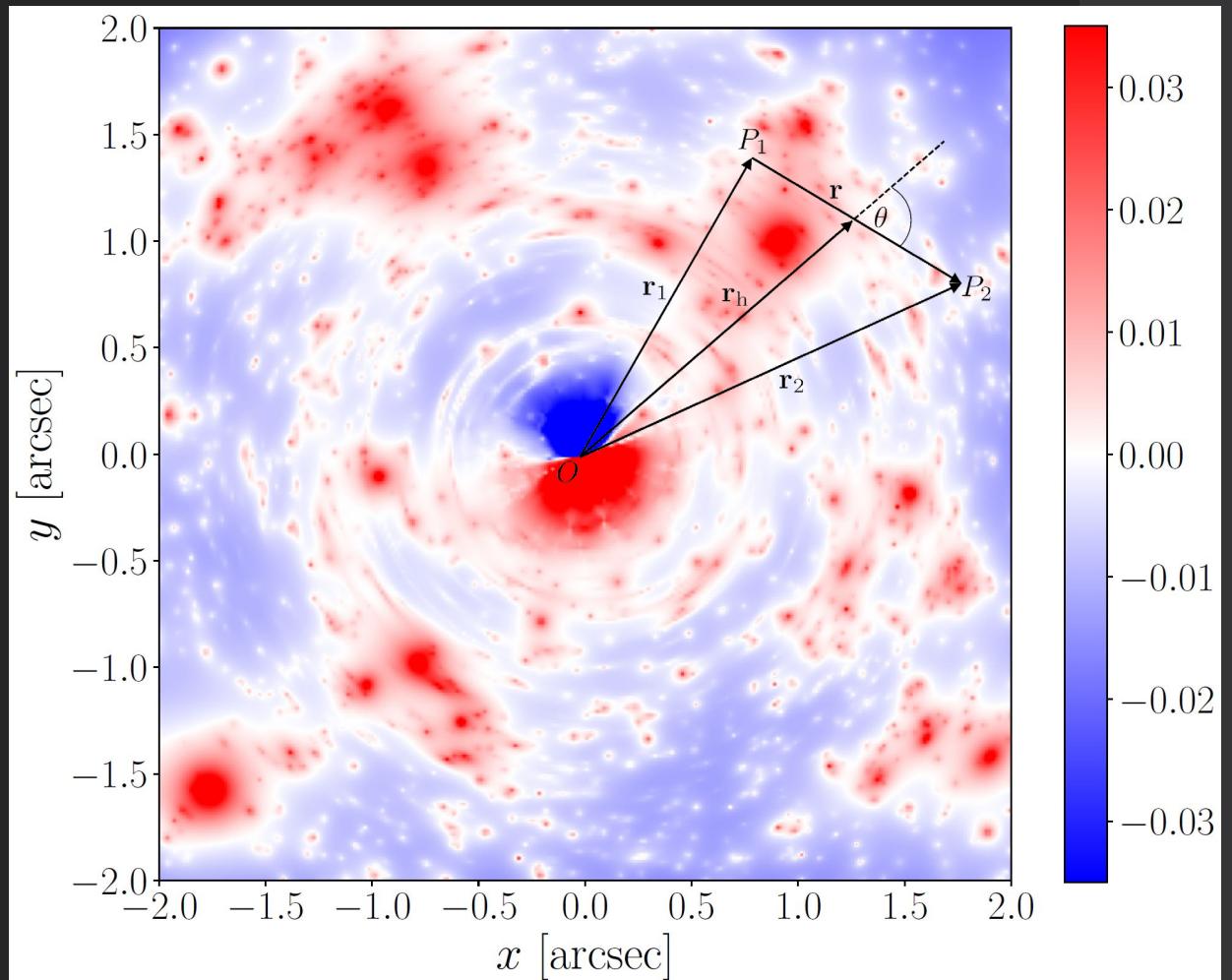
$$\xi(\mathbf{r}) = \xi(\mathbf{r}_2 - \mathbf{r}_1) = \frac{1}{A} \int_A d^2\mathbf{r}_1 \kappa_{\text{div}}(\mathbf{r}_1) \kappa_{\text{div}}(\mathbf{r} + \mathbf{r}_1)$$

$$\xi(\mathbf{r}) = \xi(r, \mu_{\mathbf{r}}) = \sum_{\ell=0}^{\infty} \xi_{\ell}(r) T_{\ell}(\mu_{\mathbf{r}}), \text{ where } \mu_{\mathbf{r}} = \cos \theta$$

**The Correlation Multipoles**

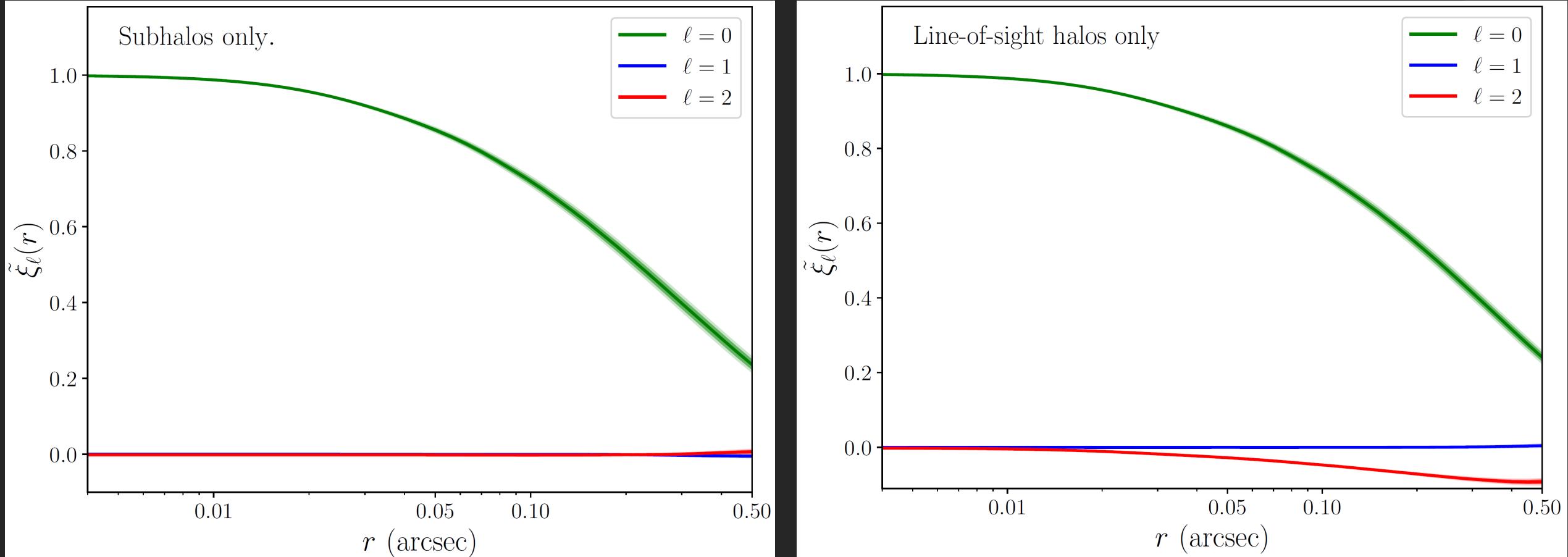
$$\xi_{\ell}(r) = \frac{2-\delta_{\ell 0}}{\pi} \int_{-1}^1 d\mu_{\mathbf{r}} \frac{\xi(r, \mu_{\mathbf{r}}) T_{\ell}(\mu_{\mathbf{r}})}{\sqrt{1-\mu_{\mathbf{r}}^2}},$$

(E.g., Hezaveh+ (2016b), Brennan+ (2019),  
Díaz Rivero+ (2018a, 2018b),  
Cyr-Racine+ (2019), Bayer+ (2018),  
Çağan Şengül+ (2020))



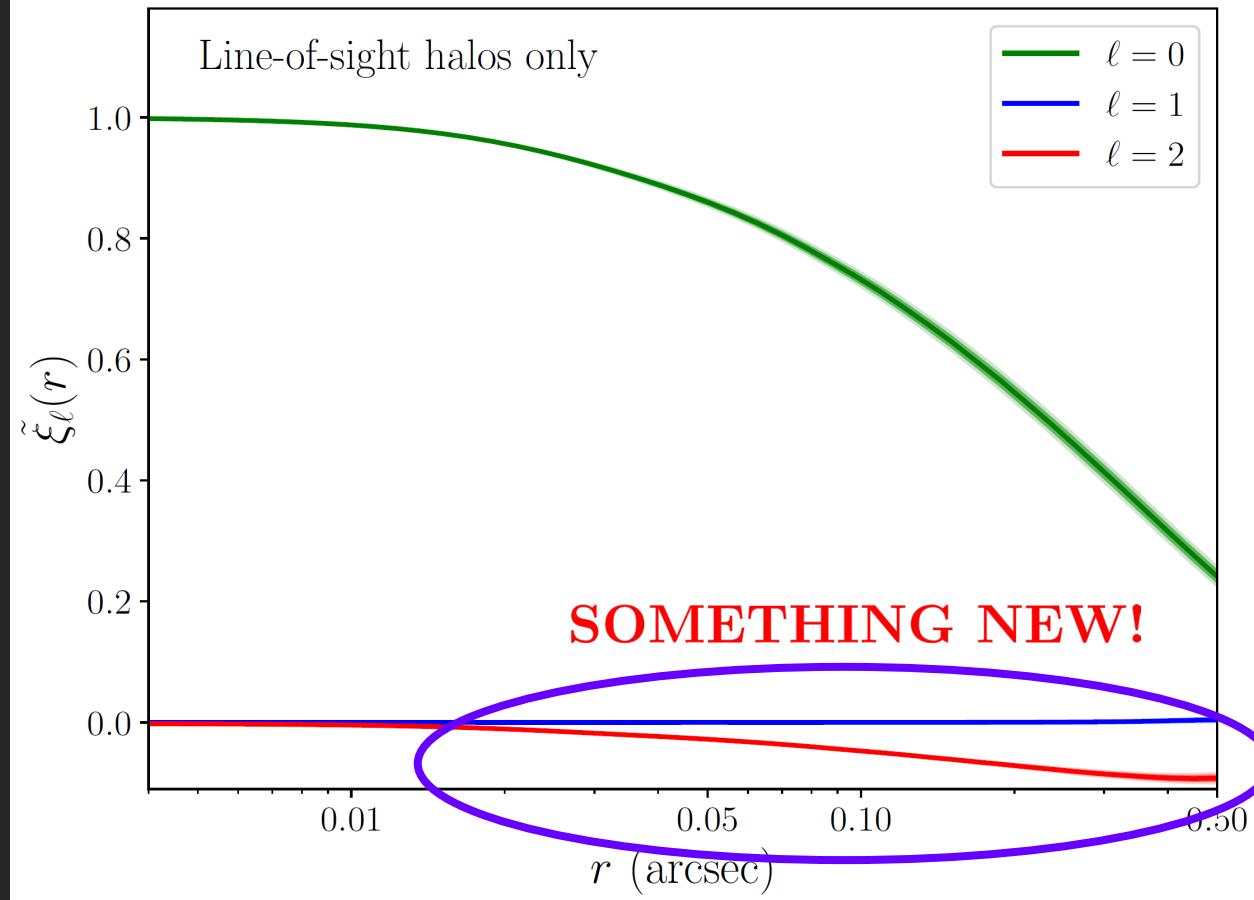
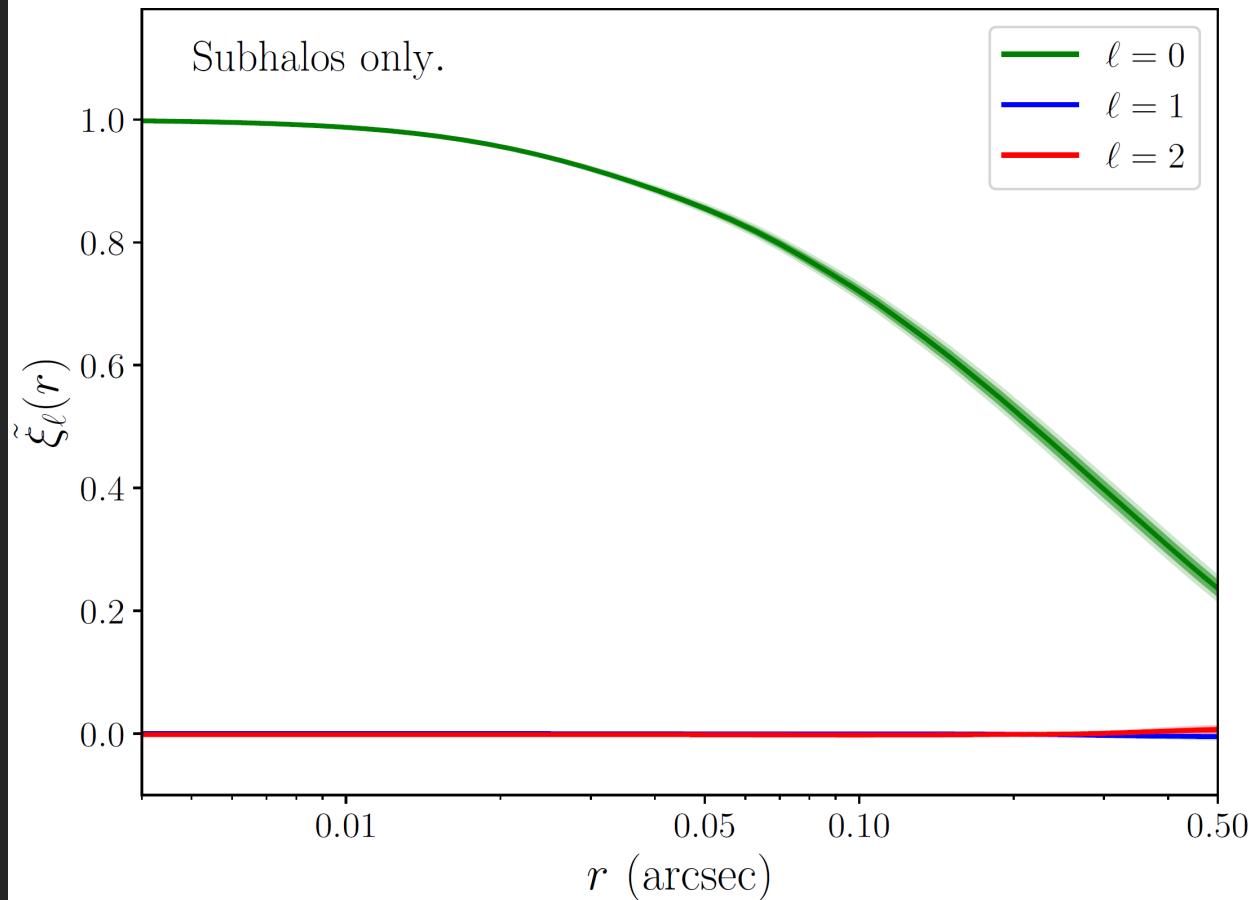
Dhanasingham+ (2022)

# Masked-normalized Correlation Function Multipoles ( $\xi_\ell$ ) of $\kappa_{\text{div}}$ Fields



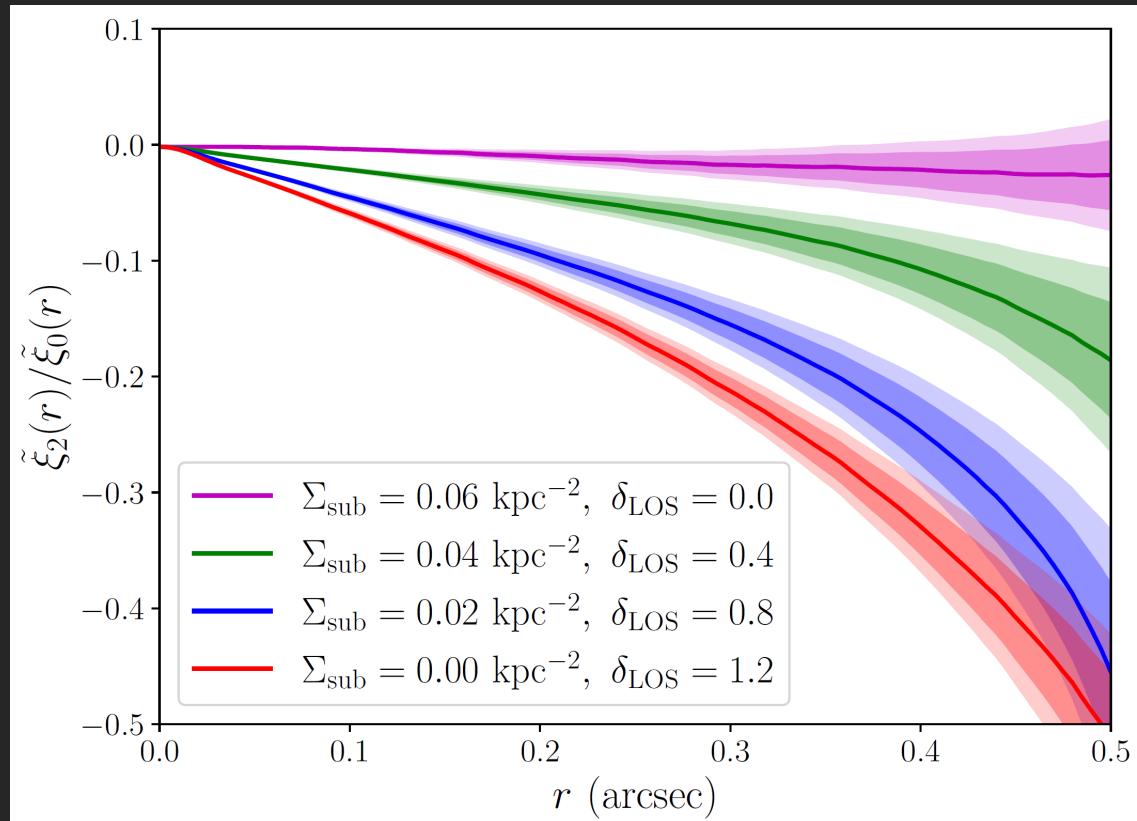
Dhanasingham+ (2022)

# Masked-normalized Correlation Function Multipoles ( $\xi_\ell$ ) of $\kappa_{\text{div}}$ Fields

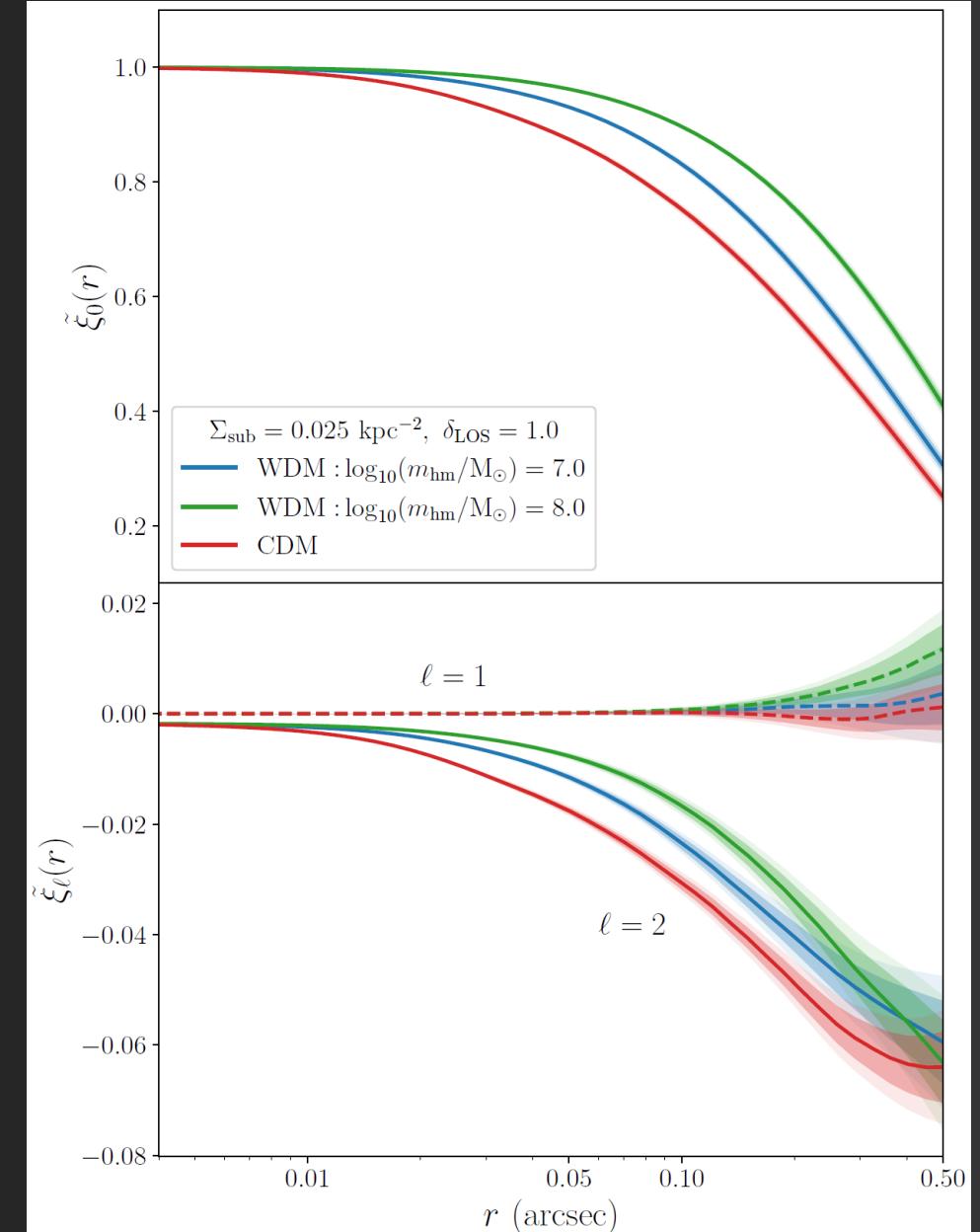


Dhanasingham+ (2022)

# Multipoles probe dark matter microphysics



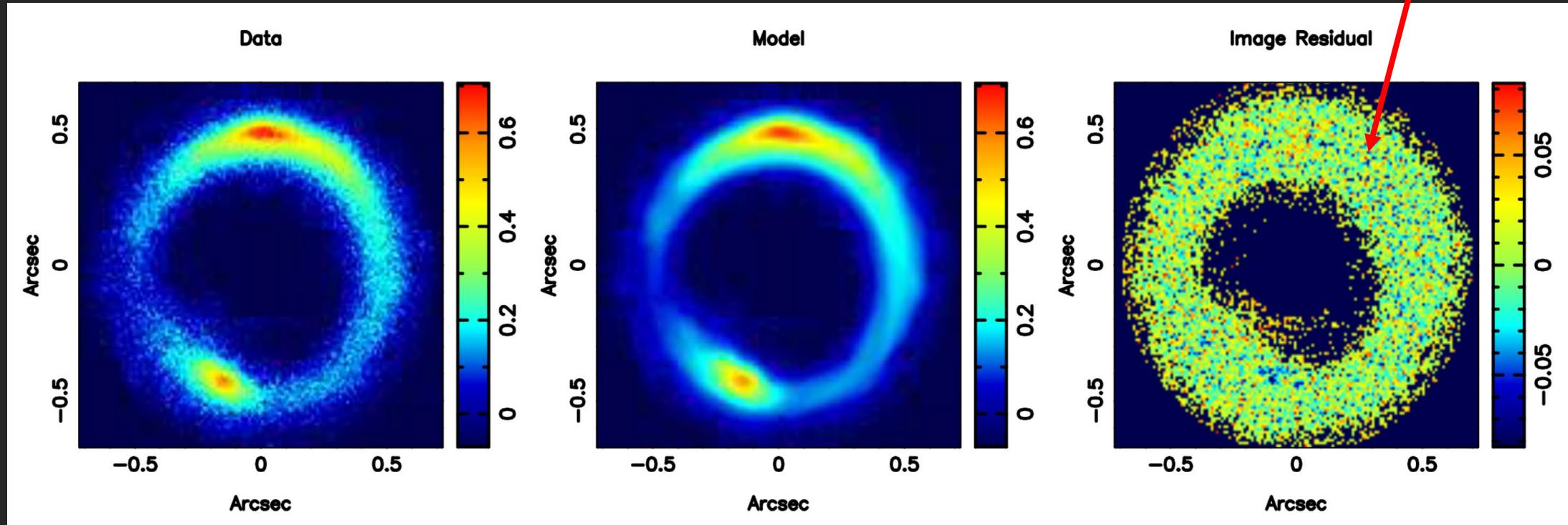
Dhanasingham+ (2022)



Dhanasingham & Cyr-Racine (in prep.)

# Lens Modeling & Residuals

Noise + (Sub+LOS) Signal +  
Source & Macrolens Mismodelling



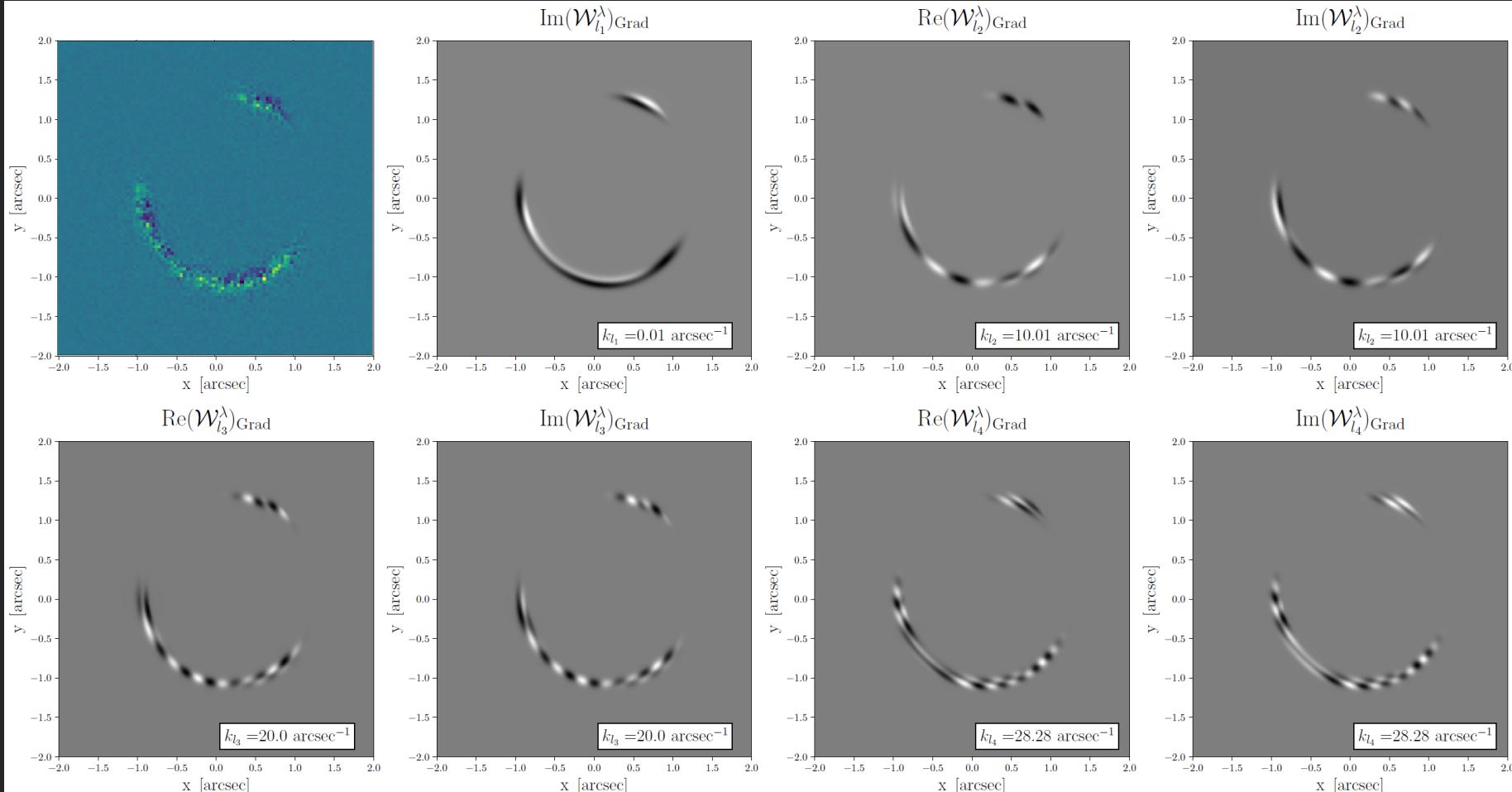
Vegetti+ (2012)

$$\delta O_{\text{obs}} \approx -\nabla S_{\text{rec}}|_{\mathbf{x} - \boldsymbol{\alpha}_{\text{bf}}(\mathbf{x})} \cdot \delta \boldsymbol{\alpha}(\mathbf{x}) + \delta S[\mathbf{x} - \boldsymbol{\alpha}_{\text{bf}}(\mathbf{x})] + N(\mathbf{x})$$

# Residuals And Decomposition

$$\delta O_{\text{obs}}(\mathbf{x}) = \delta O_{\text{div}}(\mathbf{x}) + \delta O_{\text{curl}}(\mathbf{x}) + N(\mathbf{x}) \approx \underbrace{\sum_{m=1}^{N_{\text{modes}}} \mathcal{B}_m \mathcal{W}_m}_{\delta O_{\text{div}}} + N(\mathbf{x})$$

where  $\mathcal{W}_m(\mathbf{x}) = -\nabla_{\mathbf{u}} S(\mathbf{u})|_{\mathbf{u}=\mathbf{x}-\nabla \phi_o(\mathbf{x})} \cdot \nabla \phi_m(\mathbf{x}).$



(Cyr-Racine+ (2019), Hezaveh+ (2016b))

# Test Sensitivity Using Mock Observations

- **Two instruments**

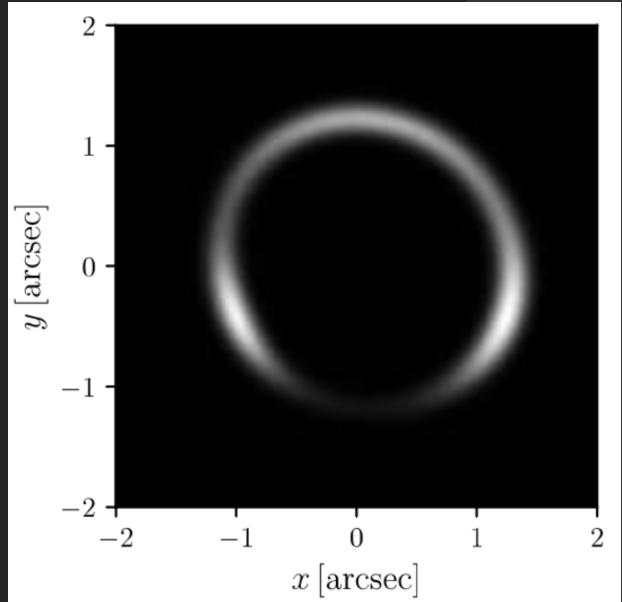
	HST-like	ELT-like
Resolution (arcsec/pixel)	0.05	0.01
FWHM <sub>PSF</sub> (arcsec)	0.07	0.15
$T_{\text{exp}}$ (hours)	1.0	1.0
$\sigma_{\text{bkg}}$ (counts/s/pixel)	0.005	0.02
$N_{\text{obs}}$	70	4

- **Likelihood**

$\mathcal{L} \propto \frac{e^{-\frac{1}{2}\delta\mathbf{O}_{\text{obs}}^T \mathbf{C}^{-1} \delta\mathbf{O}_{\text{obs}}}}{\sqrt{|\mathbf{C}|}}$  where the covariance matrix is given by  $\mathbf{C} \equiv \mathbf{C}_S + \mathbf{C}_N$

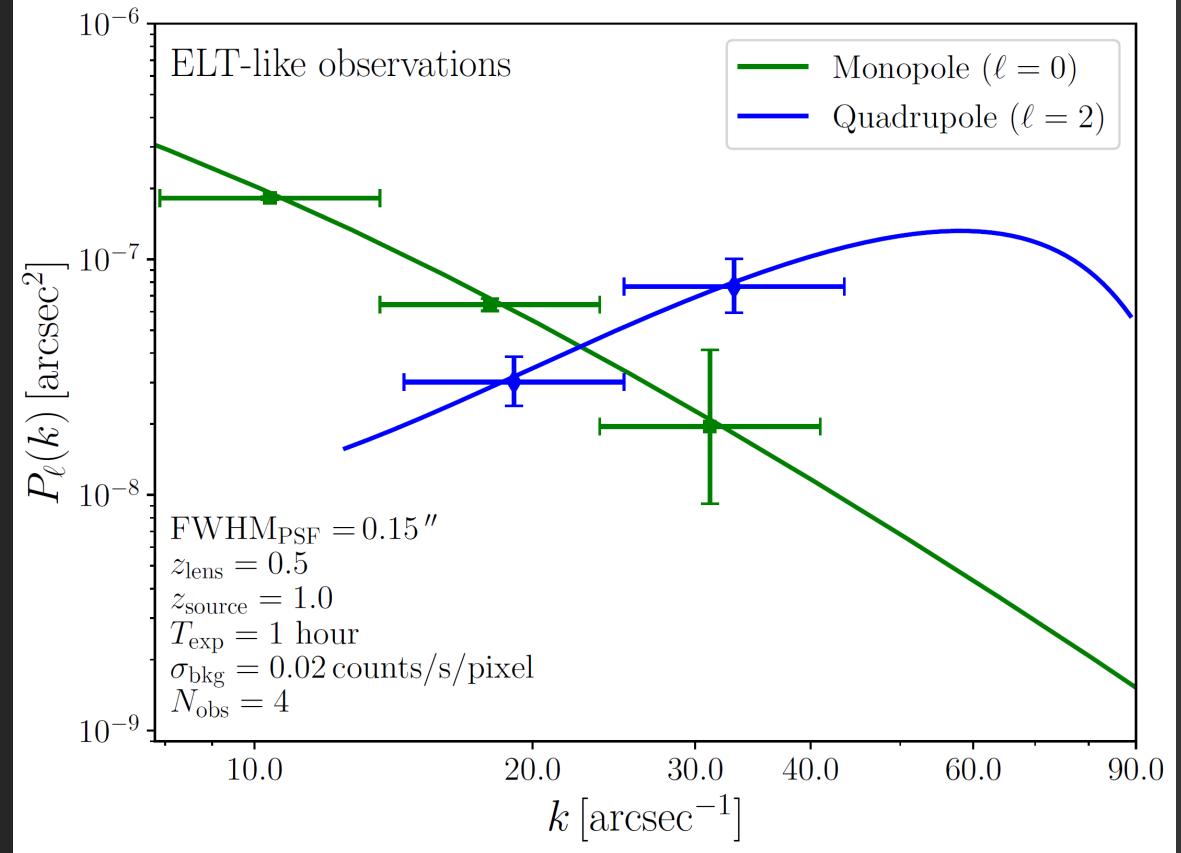
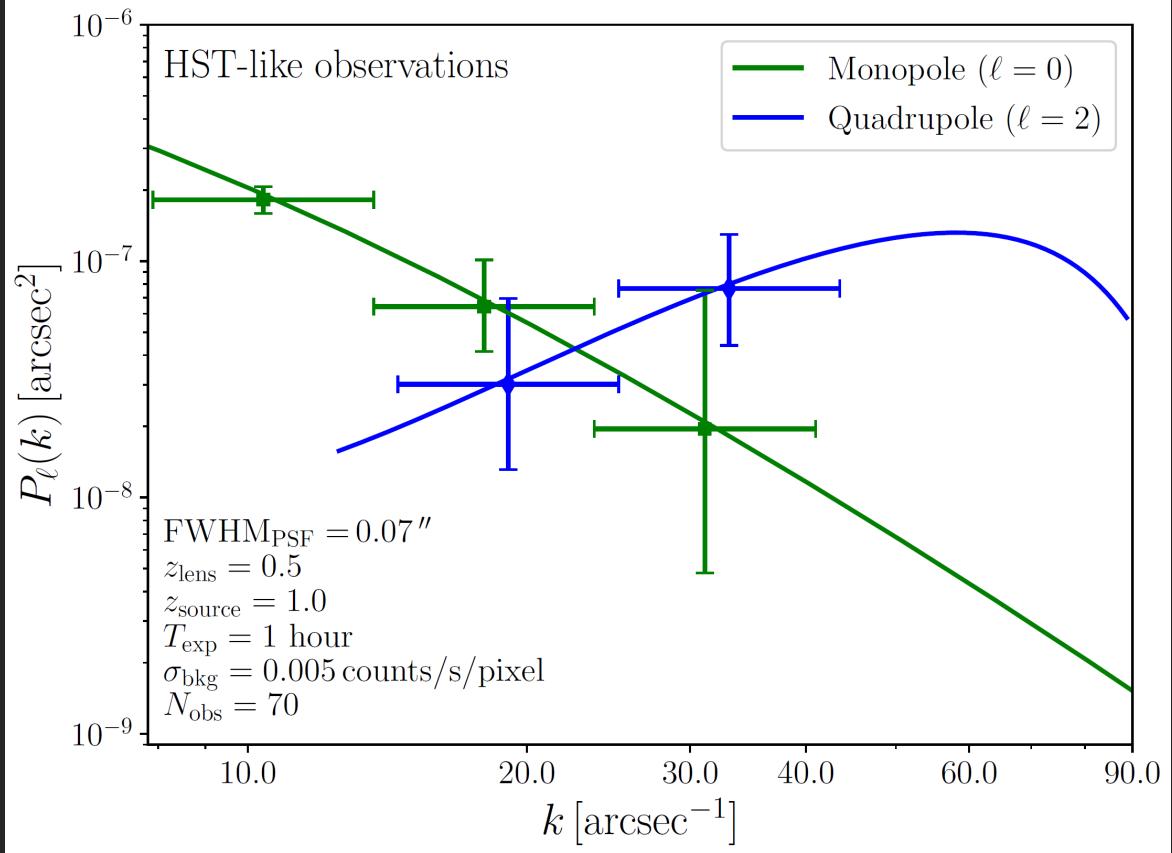
- **Fisher Matrix**

$$F_{ij} \equiv - \left\langle \frac{\partial^2 \ln \mathcal{L}}{\partial \ln P_{\ell_1, i} \partial \ln P_{\ell_2, j}} \right\rangle$$



Dhanasingham+ (2022)

# Fisher Forecast Results



Dhanasingham+ (2022)

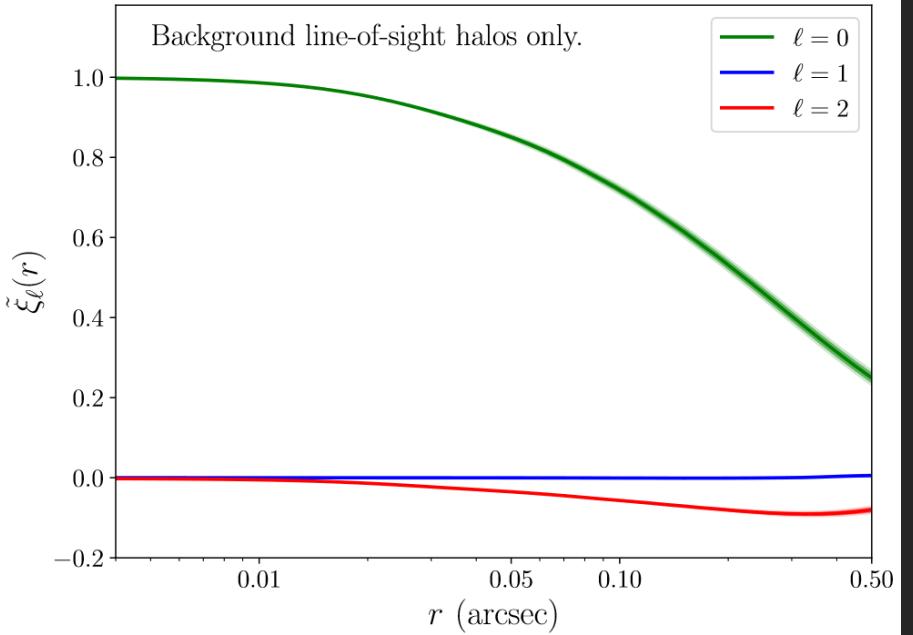
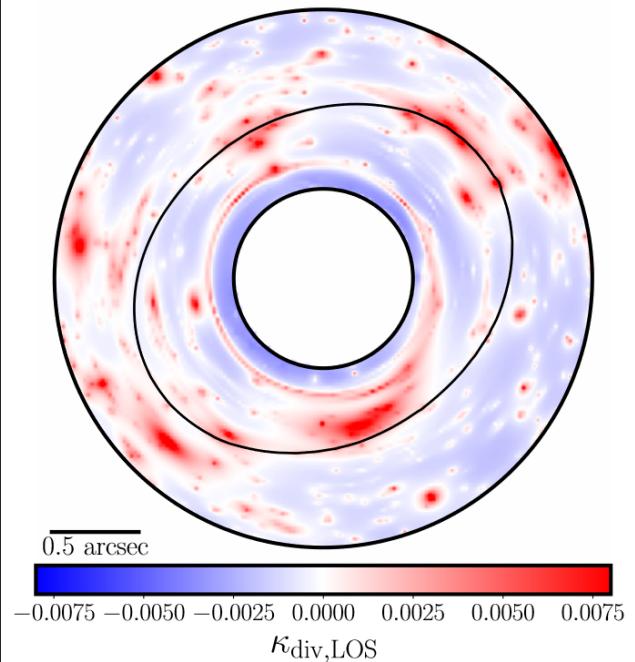
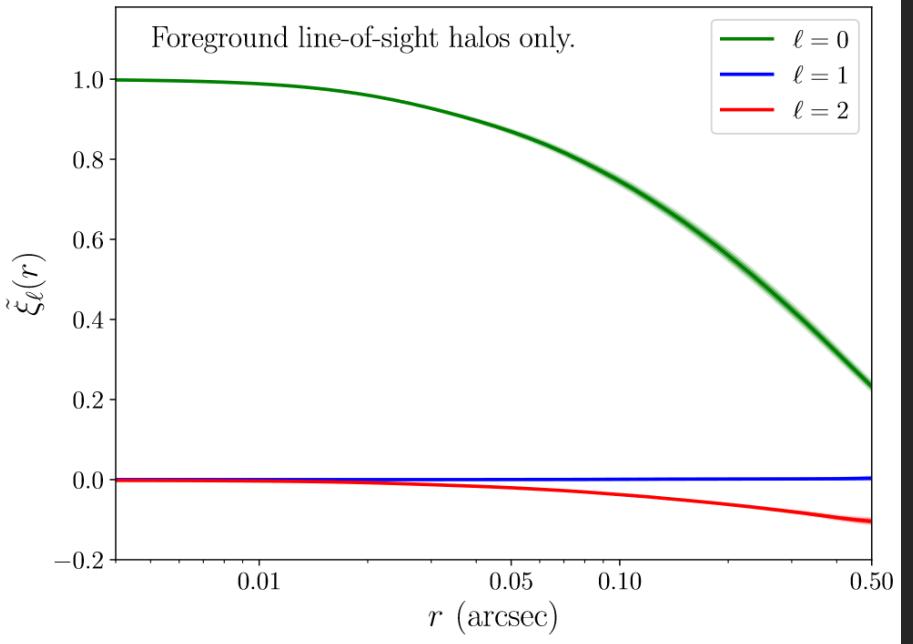
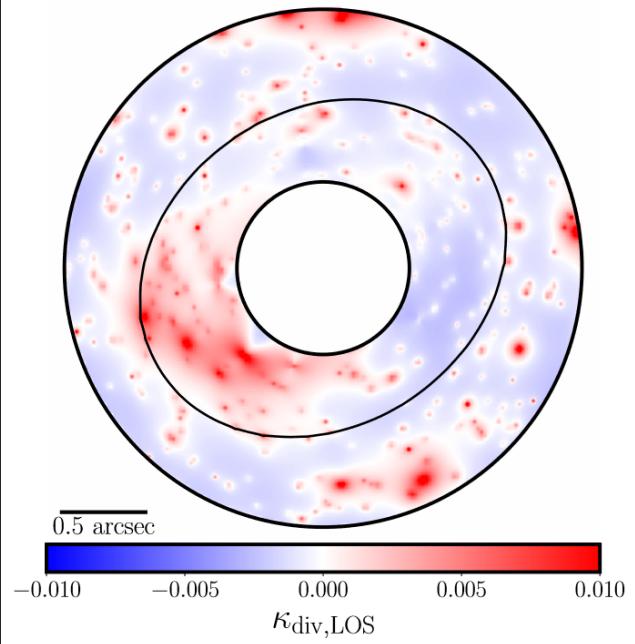
# Takeaway Message

The line-of-sight halos in a strong lens system create anisotropic signatures in  $\kappa_{\text{div}}$  maps and add non-zero quadrupole moment to the 2-point correlation function multipoles.

This detectable non-zero quadrupole can be used to study dark matter microphysics.

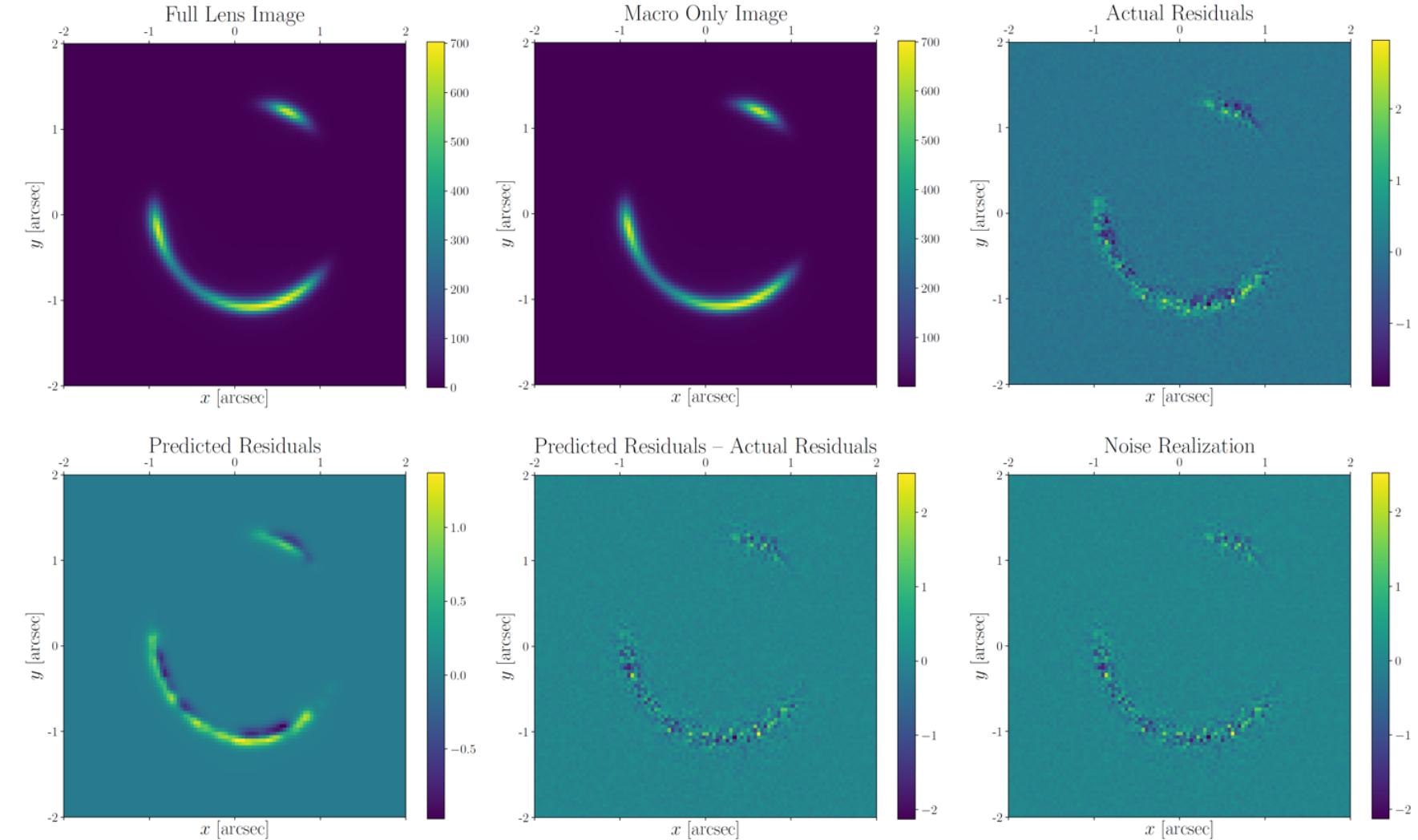
# Thank you!

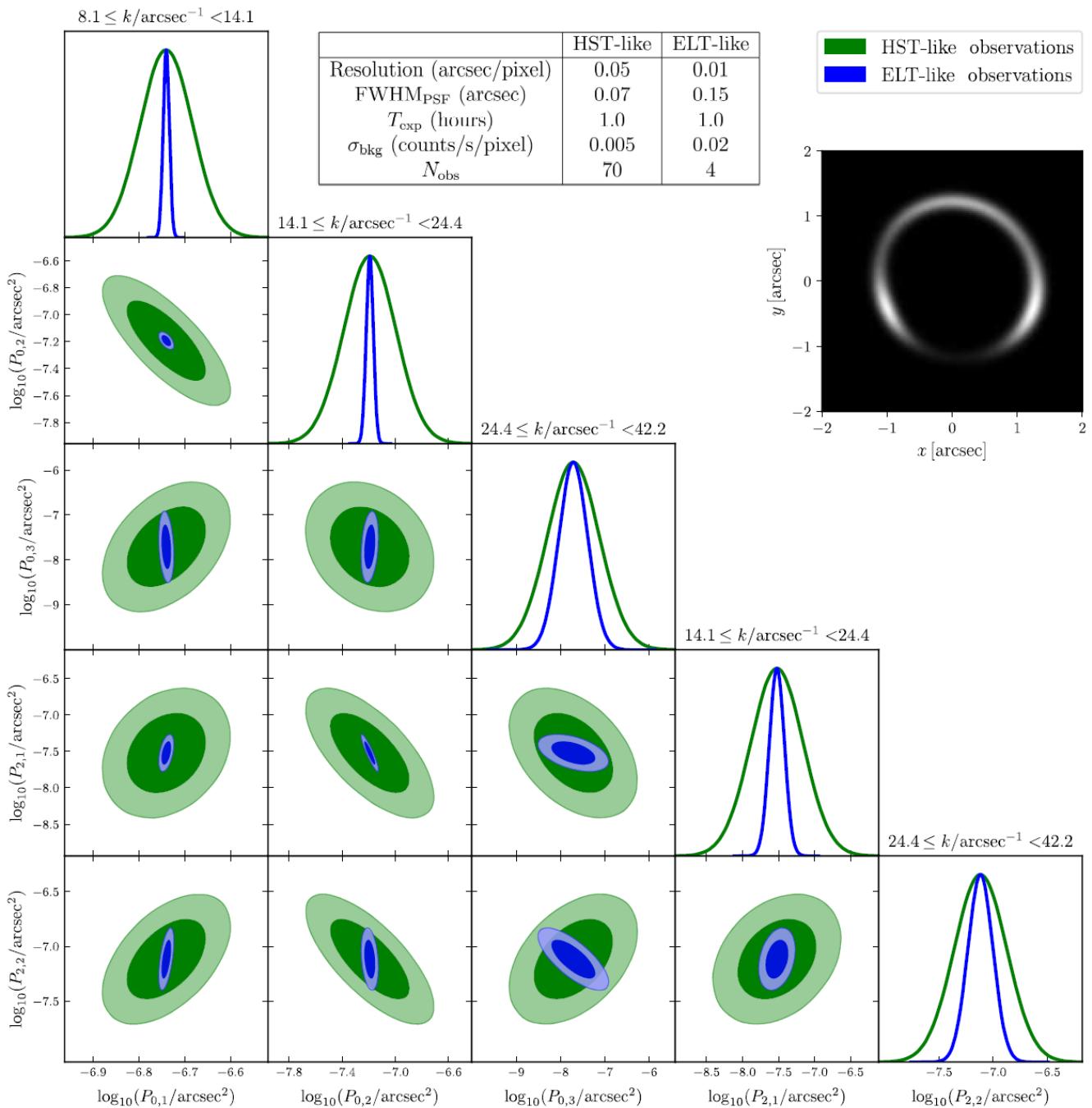
# Back-up Slides



# UNDERSTANDING RESIDUALS

$$\delta O_{\text{obs}} \approx -\nabla S_{\text{rec}}|_{\mathbf{x} - \boldsymbol{\alpha}_{\text{bf}}(\mathbf{x})} \cdot \delta \boldsymbol{\alpha}(\mathbf{x}) + \delta S[\mathbf{x} - \boldsymbol{\alpha}_{\text{bf}}(\mathbf{x})] + N(\mathbf{x})$$





# UNDERSTANDING RESIDUALS

- The true source

$$S_{\text{true}} = S_{\text{rec}} + \delta S \quad \text{where } S_{\text{rec}} - \text{reconstructed source}$$

- The true deflection field

$$\boldsymbol{\alpha}_{\text{true}}(\mathbf{x}) = \boldsymbol{\alpha}_{\text{bf}}(\mathbf{x}) + \delta \boldsymbol{\alpha}(\mathbf{x}) \quad \text{where } \boldsymbol{\alpha}_{\text{bf}} - \text{best-fit deflection field}$$

- Image residuals between the best-fit image and the true lensed image

$$\delta O_{\text{obs}} = S_{\text{true}}[\mathbf{x} - \boldsymbol{\alpha}_{\text{true}}(\mathbf{x})] - S_{\text{rec}}[\mathbf{x} - \boldsymbol{\alpha}_{\text{bf}}(\mathbf{x})] + N(\mathbf{x})$$

$$\delta O_{\text{obs}} \approx -\nabla S_{\text{rec}}|_{\mathbf{x}-\boldsymbol{\alpha}_{\text{bf}}(\mathbf{x})} \cdot \delta \boldsymbol{\alpha}(\mathbf{x}) + \delta S[\mathbf{x} - \boldsymbol{\alpha}_{\text{bf}}(\mathbf{x})] + N(\mathbf{x})$$

where

$$\delta \boldsymbol{\alpha} = \boldsymbol{\alpha}_{\text{true}} - \boldsymbol{\alpha}_{\text{bf}} = \nabla \phi_{\text{eff}} + \nabla \times \mathbf{A}_{\text{eff}} - \nabla \phi_{\text{eff,bf}} = \nabla \delta \phi_{\text{eff}} + \nabla \times \mathbf{A}_{\text{eff}}$$

# RESIDUALS AND DECOMPOSITION

$$\delta O_{\text{obs}} \approx -\nabla S_{\text{true}}|_{\mathbf{x}=\boldsymbol{\alpha}_o(\mathbf{x})} \cdot \delta \boldsymbol{\alpha}(\mathbf{x}) + N(\mathbf{x}) = \delta O_{\text{div}} + \delta O_{\text{curl}} + N(\mathbf{x})$$

where  $\delta \boldsymbol{\alpha} = \nabla \phi_{\text{eff,sub+LOS}} + \nabla \times \mathbf{A}_{\text{eff,LOS}}$

- Divergence part of the deflection field (Fourier Bases)

$$\nabla \phi_{\text{div}}(\mathbf{x}) = \nabla \phi_{\text{eff,sub+LOS}}(\mathbf{x}) = \sum_{m=1}^{N_{\text{modes}}} \mathcal{B}_m \nabla \phi_m(\mathbf{x})$$

where

$$\mathcal{B}_m = \frac{1}{A} \int_A d^2 \mathbf{x} \quad \nabla \varphi_m^*(\mathbf{x}) \cdot \nabla \phi_{\text{div}}(\mathbf{x}). \quad A - \text{Area of the image}$$

- Residuals

$$\delta O_{\text{obs}}(\mathbf{x}) = \delta O_{\text{div}}(\mathbf{x}) + \delta O_{\text{curl}}(\mathbf{x}) + N(\mathbf{x}) \approx \underbrace{\sum_{m=1}^{N_{\text{modes}}} \mathcal{B}_m \mathcal{W}_m}_{\text{Divergence Contribution}} + \underbrace{N(\mathbf{x})}_{\text{Noise}}$$

where

$$\mathcal{W}_m(\mathbf{x}) = -\nabla_{\mathbf{u}} S(\mathbf{u})|_{\mathbf{u}=\mathbf{x}-\nabla \phi_o(\mathbf{x})} \cdot \nabla \phi_m(\mathbf{x}).$$

- Residuals

$$\delta O_{\text{obs}}(\mathbf{x}) = \delta O_{\text{div}}(\mathbf{x}) + \delta O_{\text{curl}}(\mathbf{x}) + N(\mathbf{x}) \approx \sum_{m=1}^{N_{\text{modes}}} \mathcal{B}_m \mathcal{W}_m + N(\mathbf{x})$$

where

$$\mathcal{W}_m(\mathbf{x}) = -\nabla_{\mathbf{u}} S(\mathbf{u})|_{\mathbf{u}=\mathbf{x}-\nabla \phi_o(\mathbf{x})} \cdot \nabla \varphi_m(\mathbf{x}).$$

- Likelihood

$$\mathcal{L} \propto \frac{e^{-\frac{1}{2} \delta \mathbf{O}_{\text{obs}}^T \mathbf{C}^{-1} \delta \mathbf{O}_{\text{obs}}}}{\sqrt{|\mathbf{C}|}} \text{ where the covariance matrix is given by } \mathbf{C} \equiv \mathbf{C}_S + \mathbf{C}_N$$

$$\mathbf{C}_{S,ij} \equiv \langle \delta \mathbf{O}_{\text{div},i} \delta \mathbf{O}_{\text{div},j}^\dagger \rangle$$

- Fisher Matrix

$$F_{ij} \equiv - \left\langle \frac{\partial^2 \ln \mathcal{L}}{\partial \ln P_{\ell_1,i} \partial \ln P_{\ell_2,j}} \right\rangle$$