



UNIVERSITÉ
DE GENÈVE



Sayan Saha, Louis Legrand and Julien
Carron

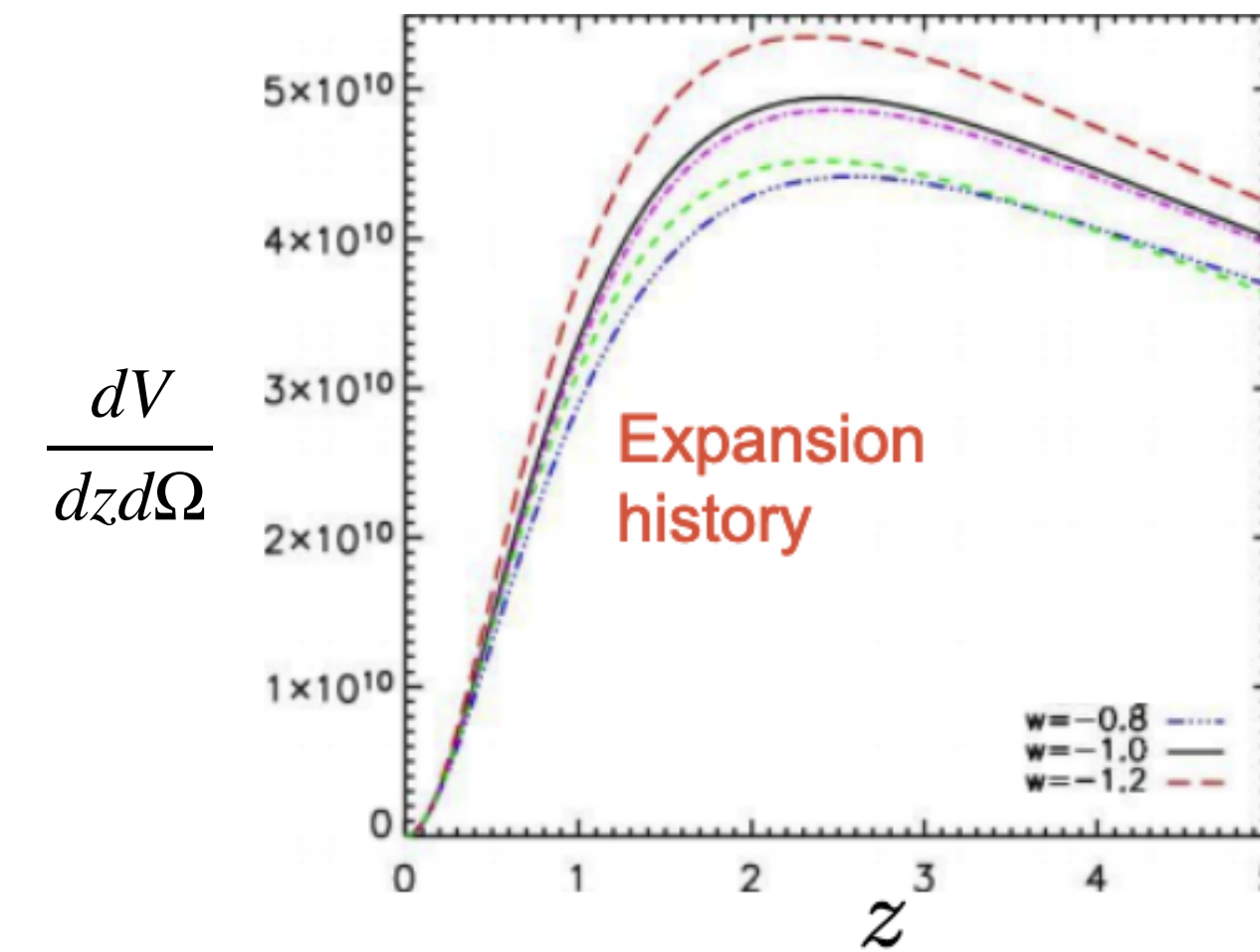
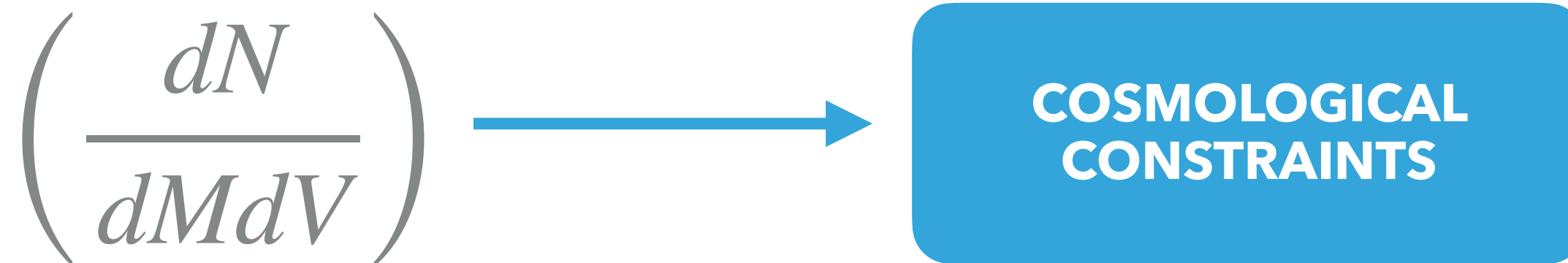
EXTRACTING CLUSTER INFORMATION FROM SMALL-SCALE CMB

OUTLINE

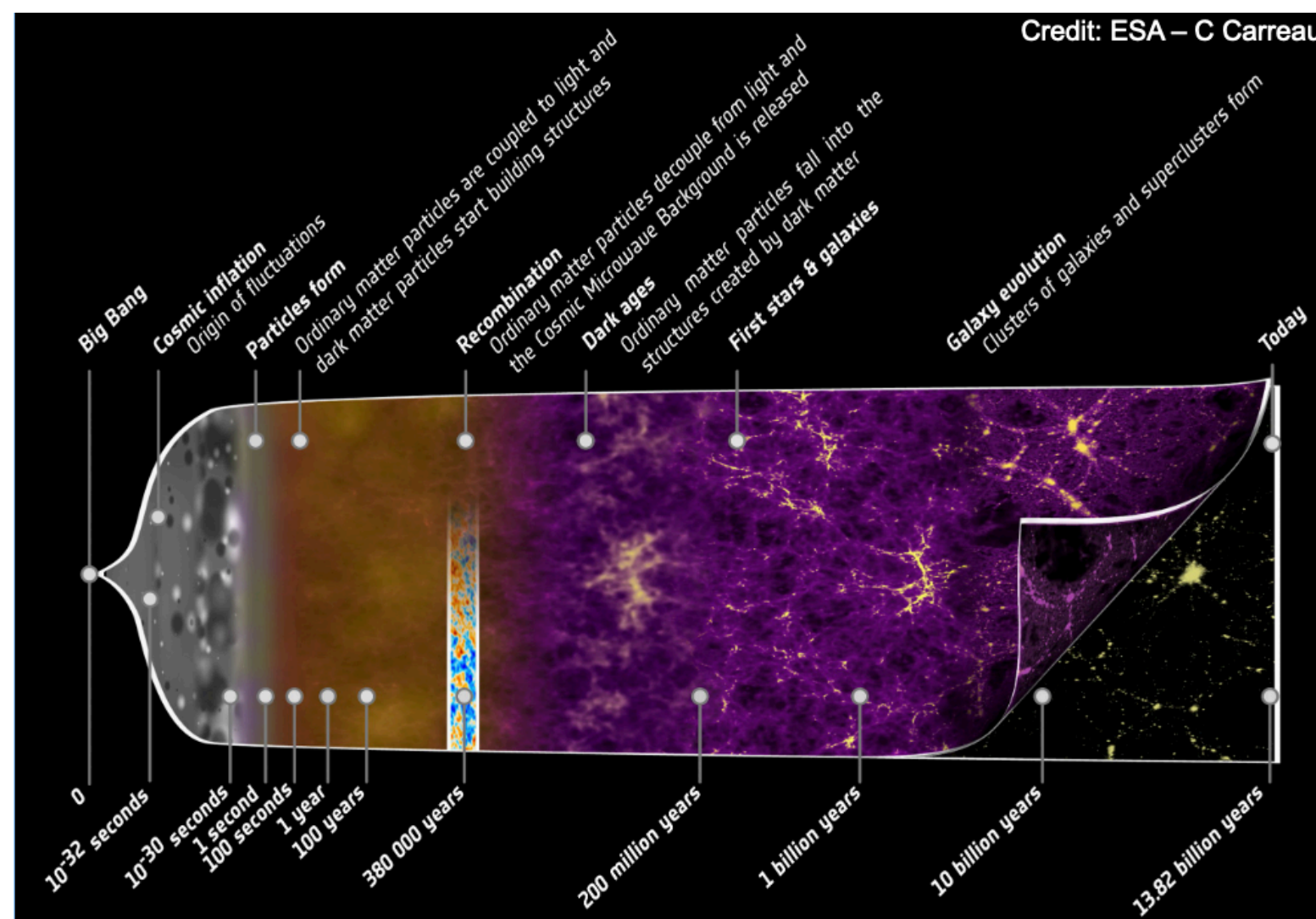
- ▶ **Motivation**
- ▶ Theoretical Model
- ▶ Analysis
- ▶ Conclusion

MOTIVATION BEHIND COSMOLOGY WITH CLUSTERS

- ▶ Cluster abundances Cosmology



Credit: Koyama & Maartens, 2006



- ▶ This history sets bounds on how small and how large a collapsed object can be.
- ▶ Uncertainties in cluster mass measurements affects our understanding of the cosmic expansion history

HOW MASS OF CLUSTERS COMES TO THE PICTURE?

- ▶ The gravitational lensing signature is directly sensitive to the mass of clusters.



galaxy cluster SMACS 0723
Credits: NASA, ESA, CSA, and STScI

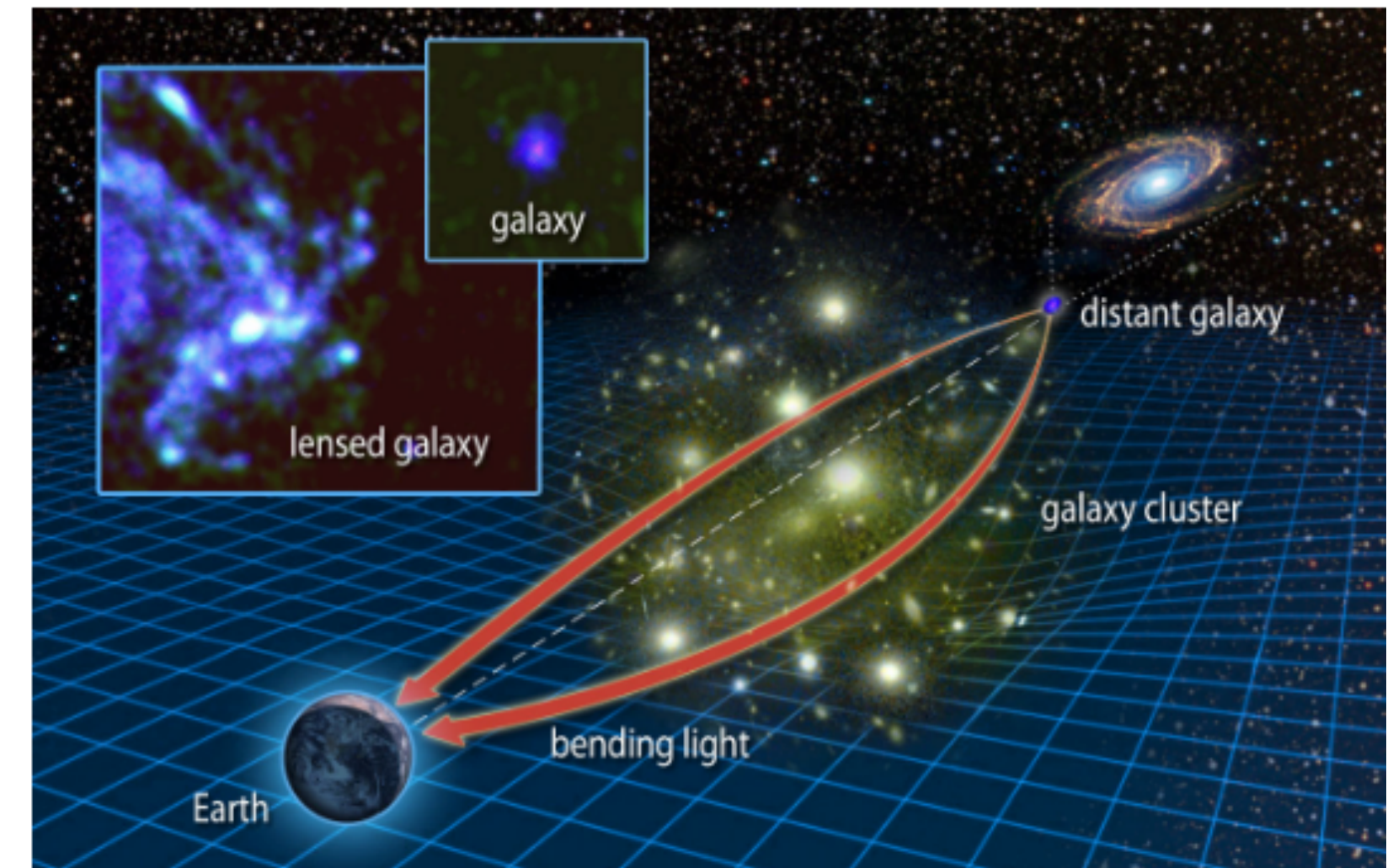
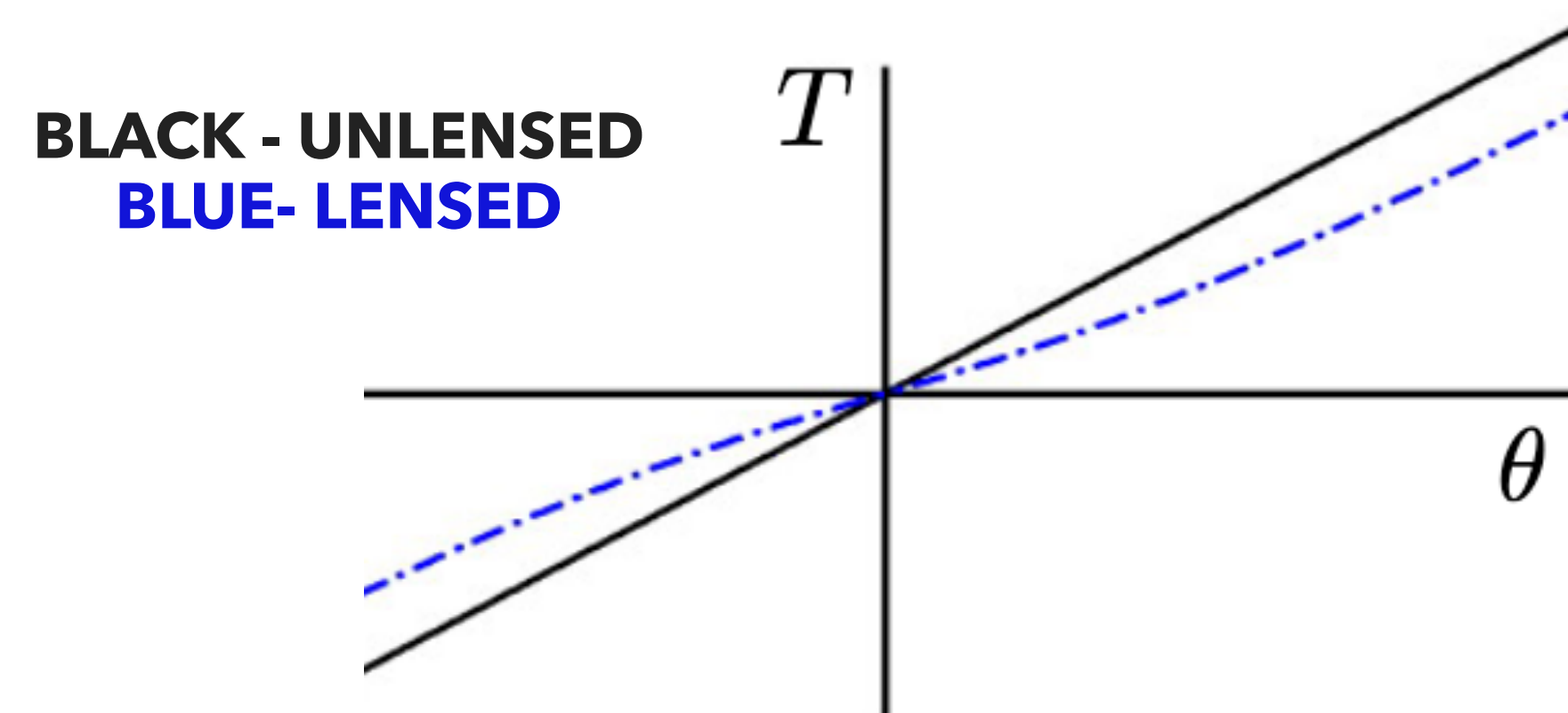


Image credit: Karen Teramura

- ▶ The mass profile of the clusters can be studied through:
 1. Strong Lensing distortions of Galaxies
 2. Weak Lensing distortions of Galaxies
 3. CMB Lensing by the galaxy clusters

CMB LENSING BY GALAXY CLUSTERS

- ▶ The lensing of CMB conserves the surface brightness.
So if CMB is uniform, the lensing cannot be detected.
Lensing can only be detected if there is anisotropies
- ▶ In case of a 1D CMB gradient that is lensed by a cluster



Seljak & Zaldarriaga 2001

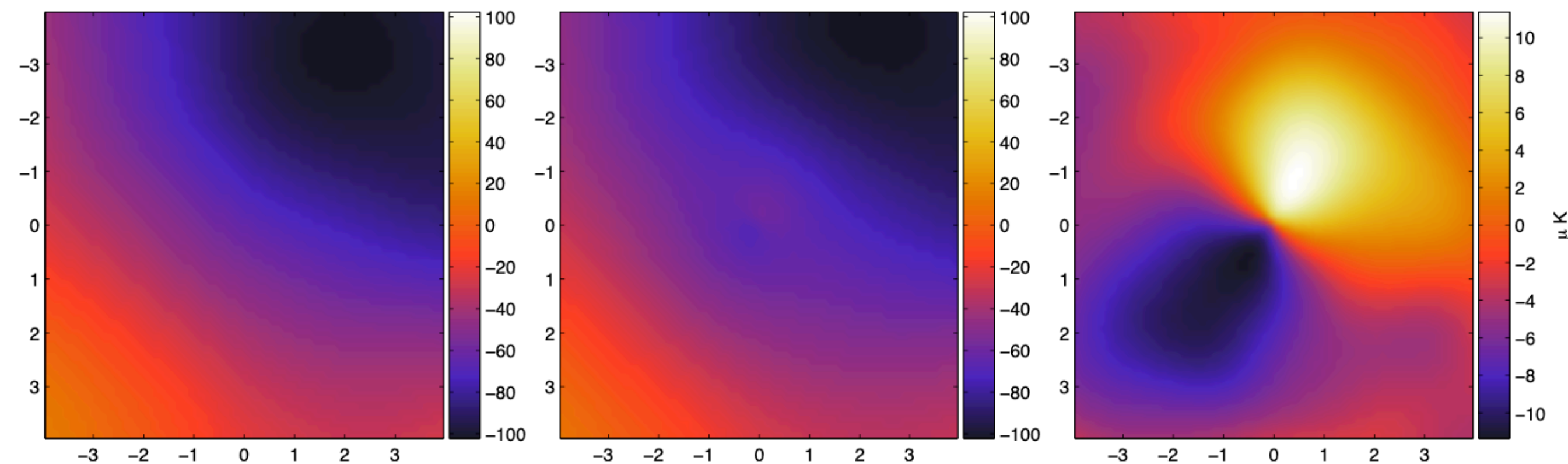
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Lewis and Challinor, 2006

WHY IS CLUSTER-LENSING OF CMB EXCITING

► Advantages

1. We can probe clusters in high redshifts
2. We know the exact redshift for our background source, i.e. CMB
3. Systematics from CMB lensing measurements are well studied

► Challenges

1. Small effects, μK fluctuations in arc-min scales.
2. tSZ and kSZ effects, radio galaxies and dusty galaxies.
3. systemic biases like differences between true and assumed cluster profiles, miss-centering

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CLUSTER MODEL (NFW PROFILE)

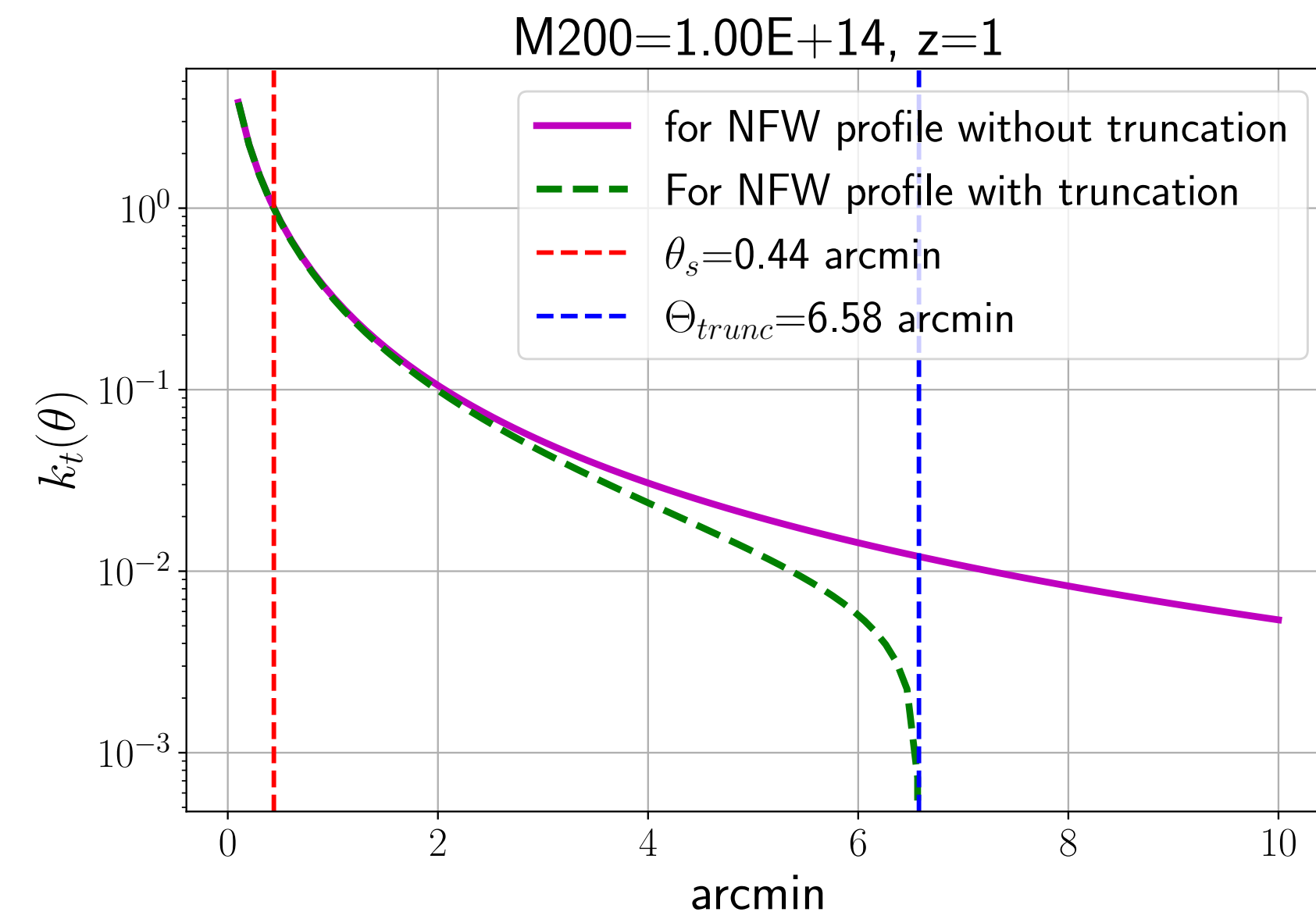
- ▶ The halo density profile

$$\rho(r) = \begin{cases} \frac{\rho_0}{\left(\frac{r}{r_s}\right)\left(1 + \frac{r}{r_s}\right)^2} & \text{if } r < R_{\text{trunc}} \\ 0 & \text{if } r > R_{\text{trunc}} \end{cases},$$

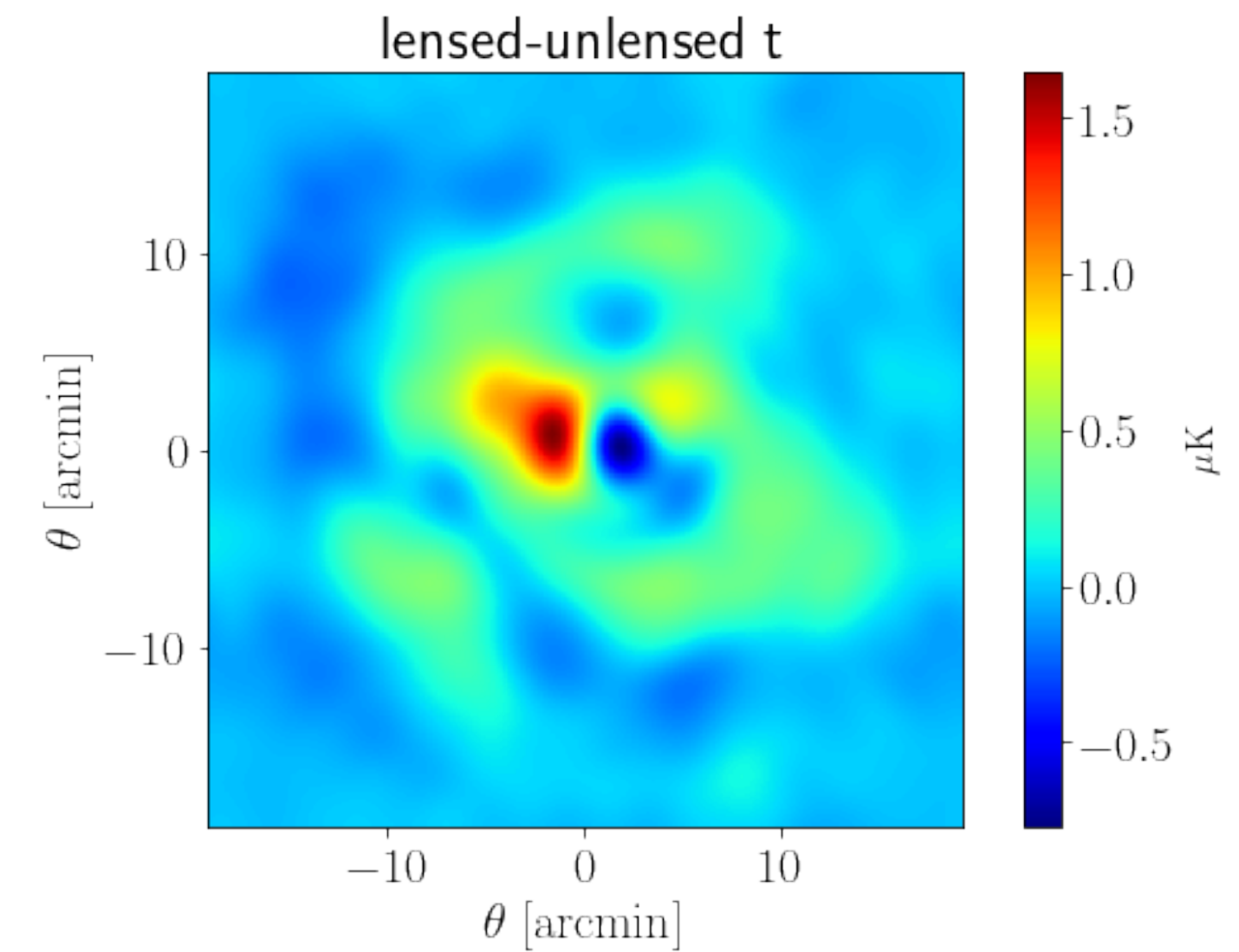
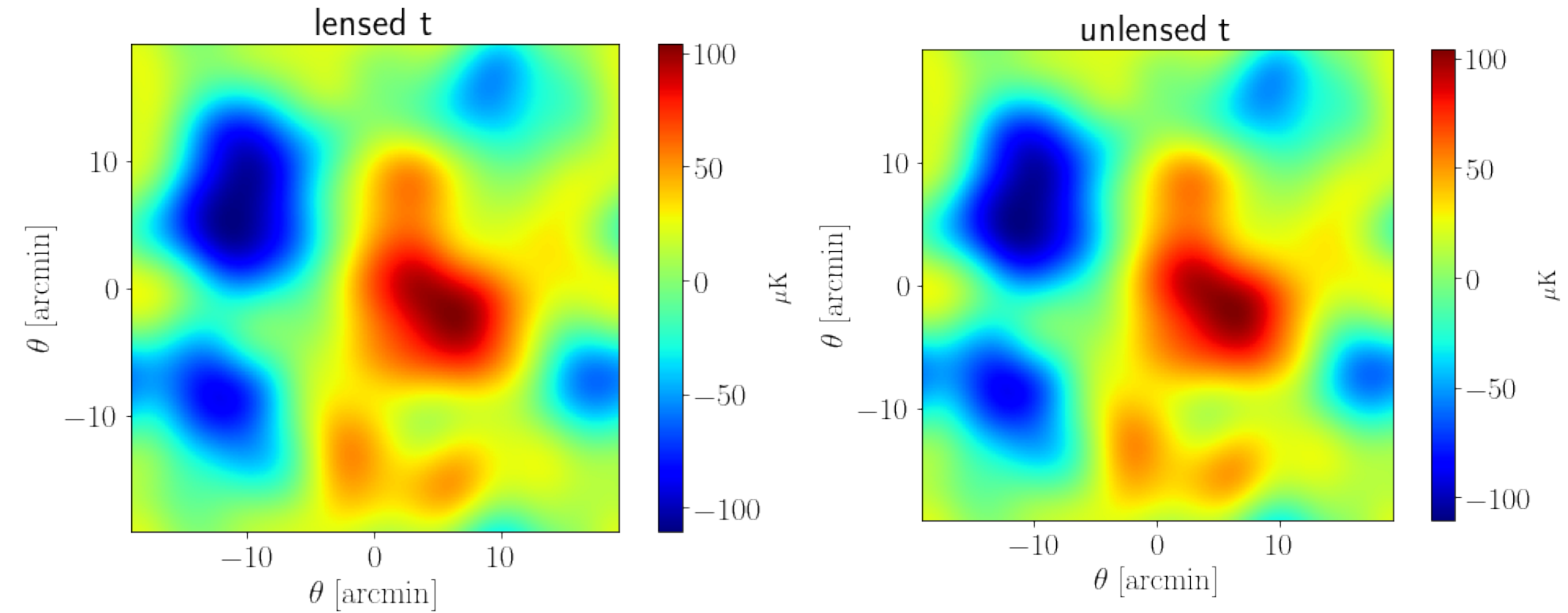
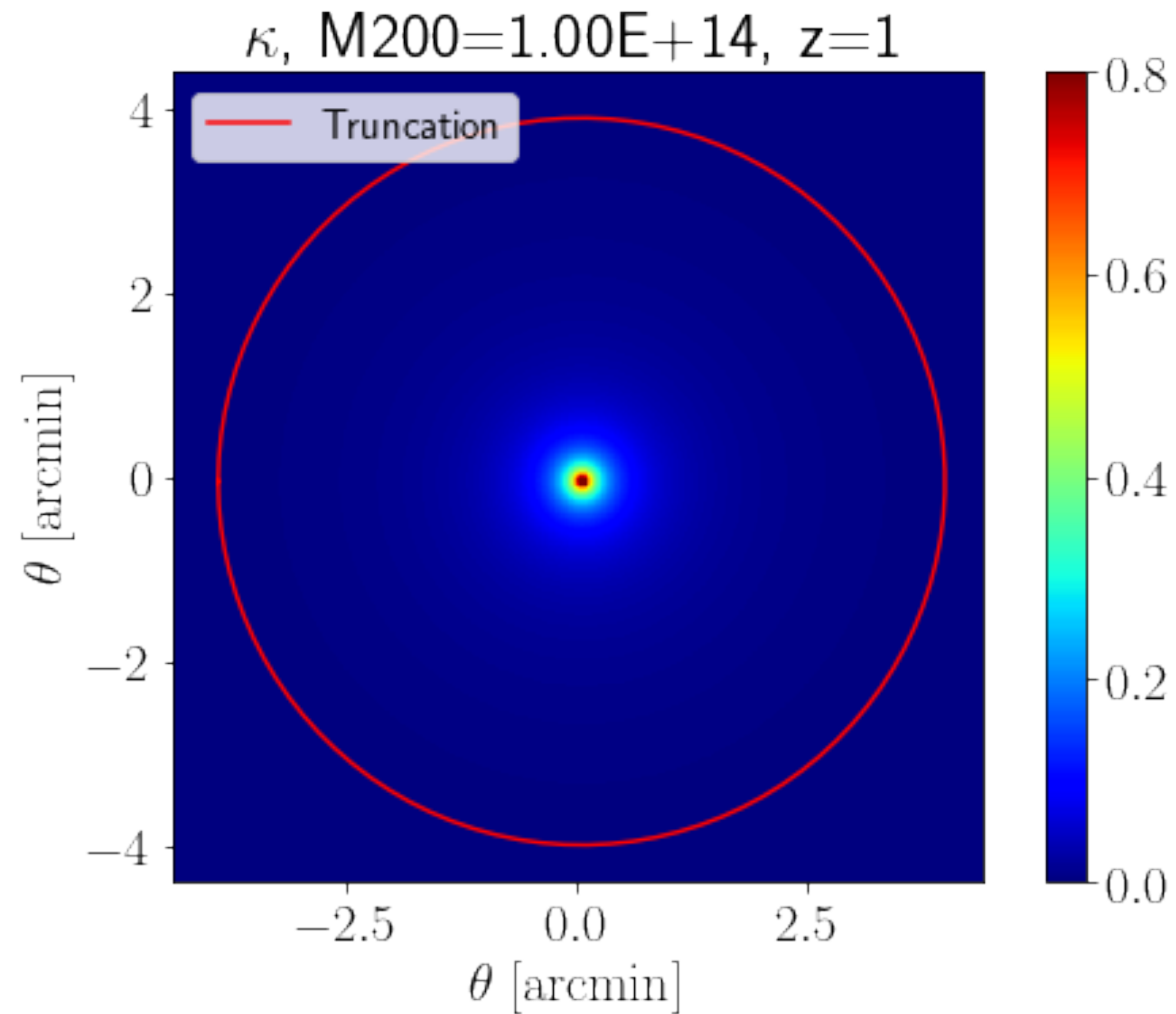
$$\kappa_{cl} = \frac{2\rho_s r_s}{\Sigma_{crit}(z)} g(x), \text{ where } x = \frac{r}{r_s} = \frac{\theta}{\theta_s}$$

- ▶ The convergence profile is

$$\kappa_{cl}(r) = \frac{\Sigma_{cl}(r)}{\Sigma_{crit}(z)}$$



CMB LENSING BY NFW PROFILE



THE TEMPLATE FUNCTION

- ▶ $\kappa_{cl}(\theta) = \kappa_0 \kappa_t(\theta, \theta_s)$
- ▶ $\kappa_t(\theta = \theta_s) = 1$ and $\kappa_{cl}(\theta = \theta_s) = \kappa_0$.

We need an estimator for κ_0

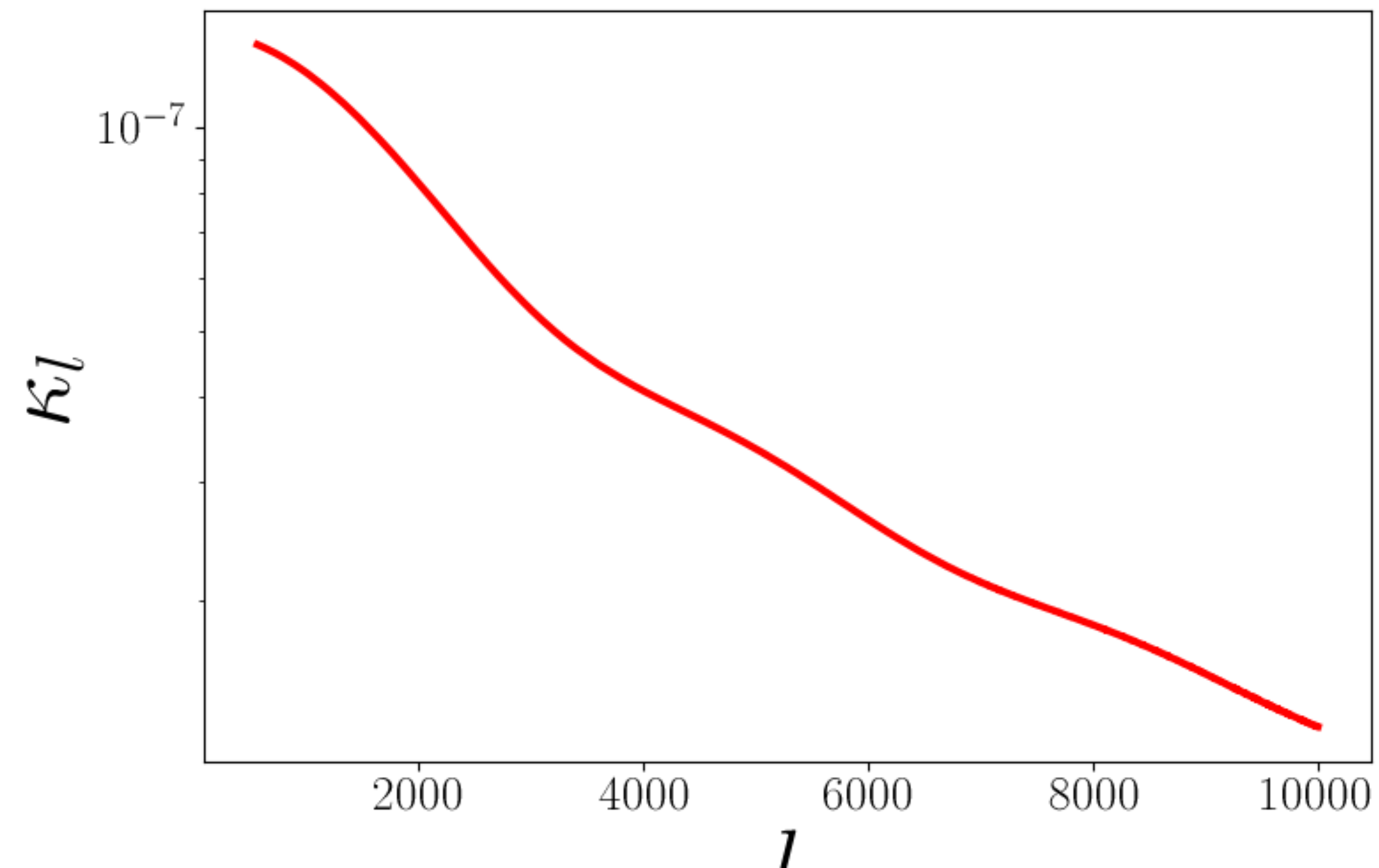
$$\kappa_0 \theta_s^2 \propto \frac{M_{200}}{\sum_{crit}(z) d_A(z)}$$

$$\frac{\sigma_M}{M} = \frac{\sigma_{\kappa_0}}{\kappa_0}$$

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CLUSTER CONVERGENCE IN FOURIER SPACE



$$\hat{\kappa}(l) = \kappa_0 \kappa_l(l)$$

LEAVE ME ALONE

κ_0 should be calculable from every mode, l

We construct an average sum from all the κ_0 values from different modes, such that it has minimum variance possible

MINIMUM VARIANCE ESTIMATOR OF κ_0

$$\hat{\kappa}_0 = \frac{\int d^2\vec{l} \frac{\kappa^t(\vec{l})\hat{\kappa}(\vec{l})}{N_{\vec{l}}}}{\int d^2\vec{l} \frac{|\kappa^t(\vec{l})|^2}{N_{\vec{l}}}}$$

With the inverse variance,

$$\frac{1}{\sigma^2} = \int d^2l \frac{|\kappa_l^t|^2}{N_l}$$

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$\hat{\kappa}_l$ = convergence estimated from data
 N_l = Noise of the estimation
 $= C_l^{\kappa\kappa} + N_0^\kappa + N_1^\kappa$

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We employ The Maximum a Posterior (MAP) Estimator by Carron et al 2017

We maximize the log posterior:

$$\ln p(\phi | X^{dat}) = \ln p(X^{dat} | \phi) - \frac{1}{2} \sum_L \frac{\phi_L^2}{C_L^{\phi\phi}}$$

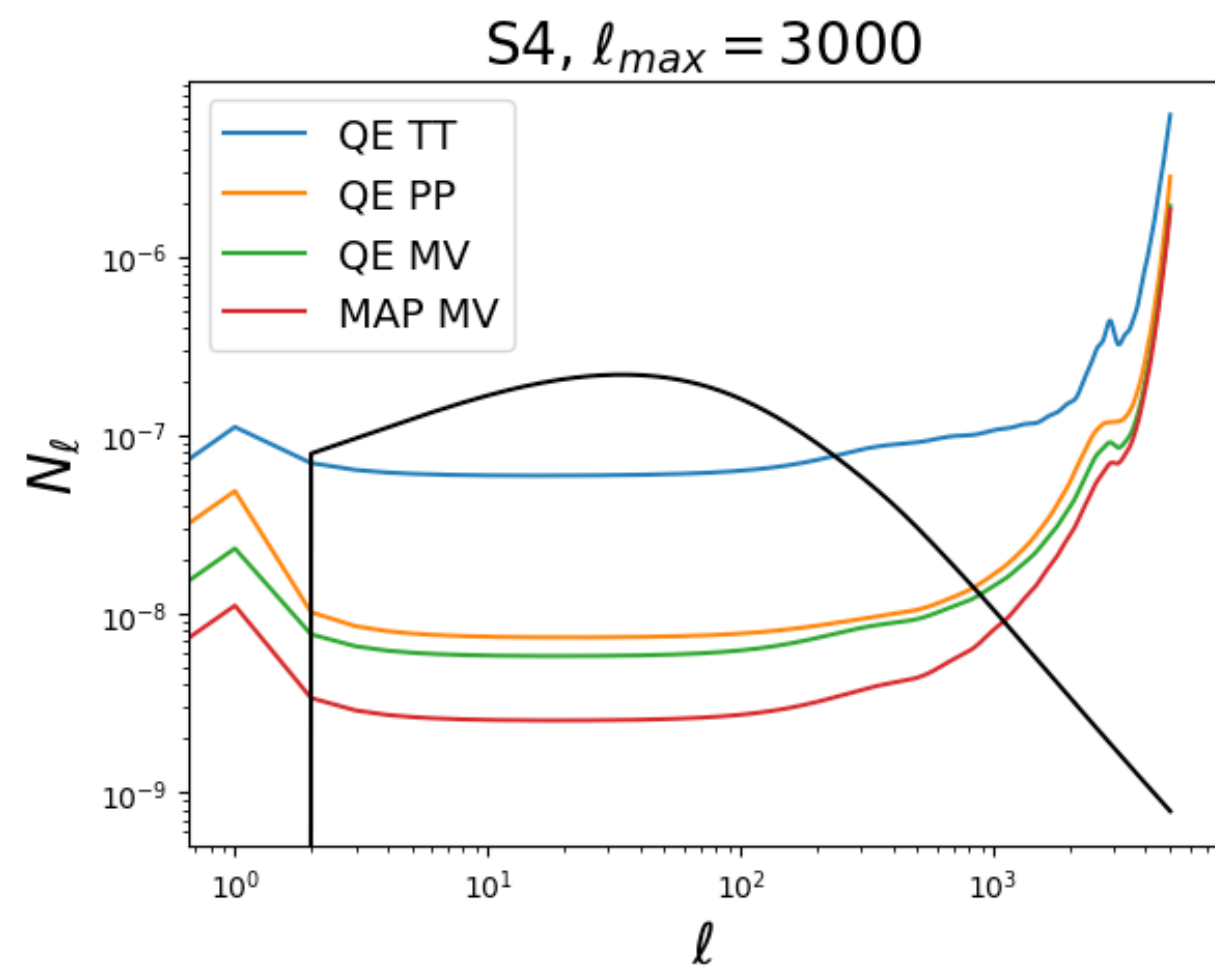
Using Gradients:

$$g_\phi = \frac{\delta \ln p(X^{dat} | \phi)}{\delta \phi} = g^{QD} - g^{MF} + g^{PR}$$

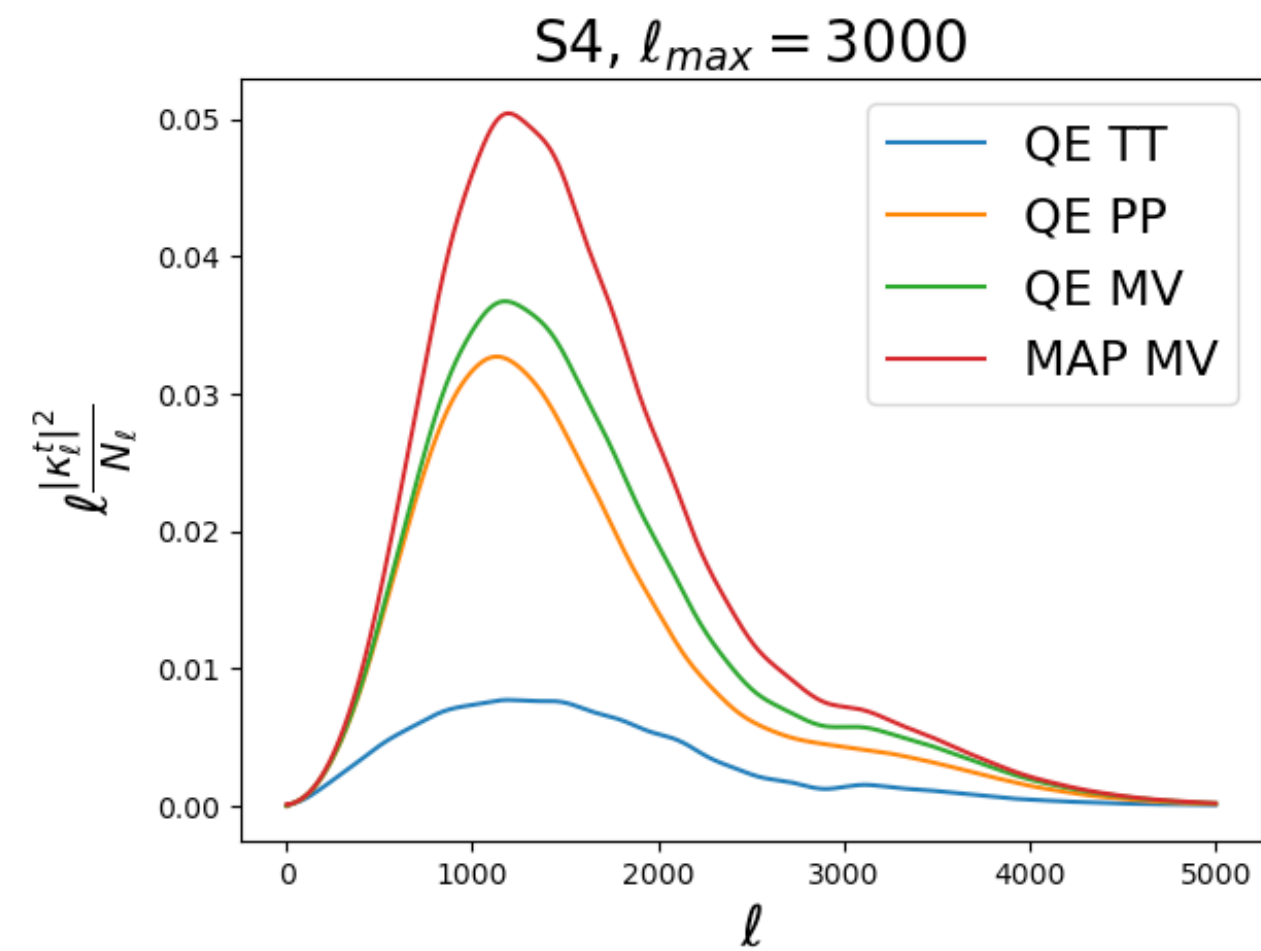
We use these g_ϕ 's iteratively to reach the maximum

IMPROVEMENT OVER QUADRATIC ESTIMATOR

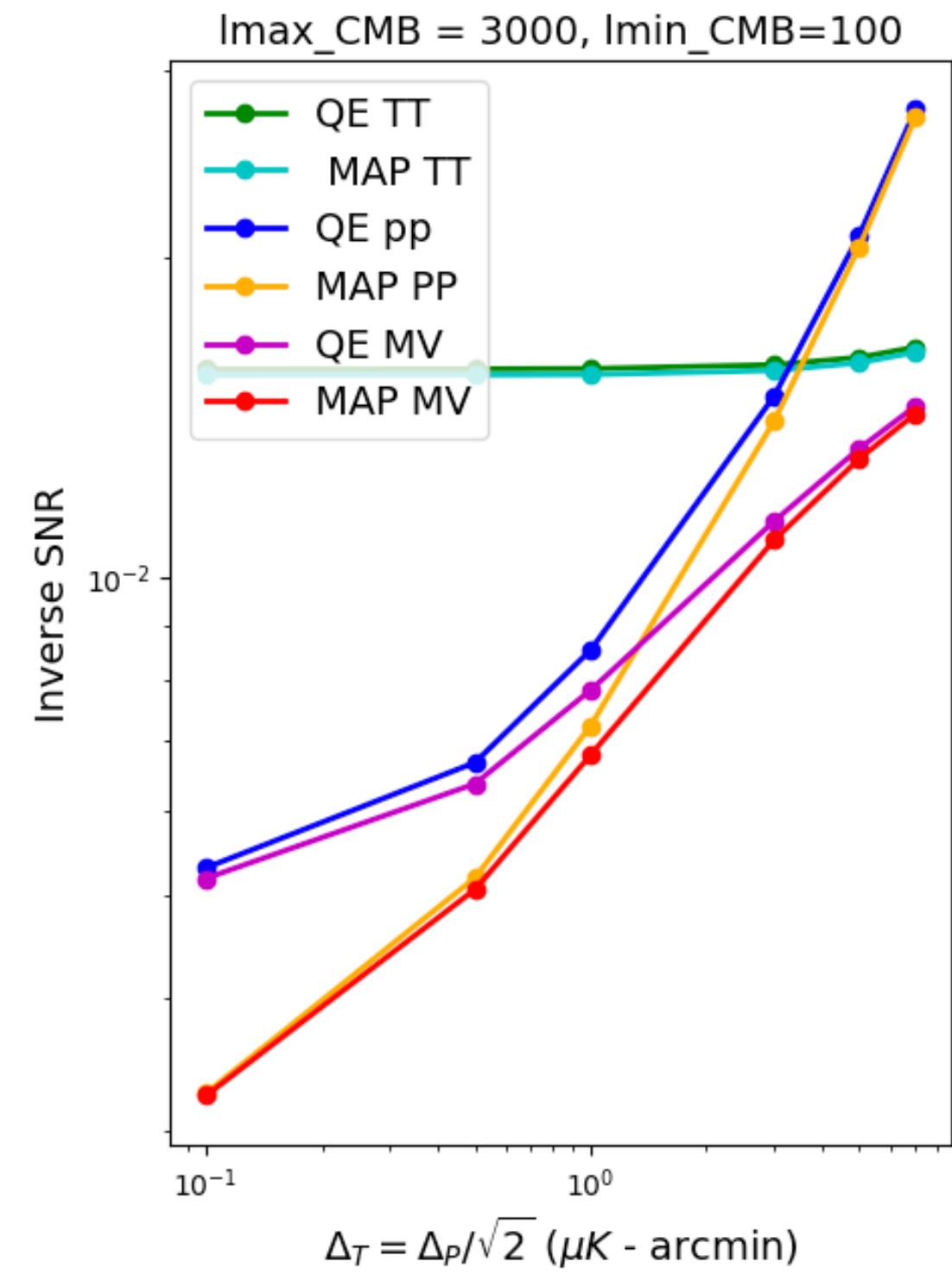
The noise of the estimator



The contribution in $\frac{1}{\sigma^2} = \int d^2l \frac{|\kappa_l^t|^2}{N_l}$ comes from certain scales



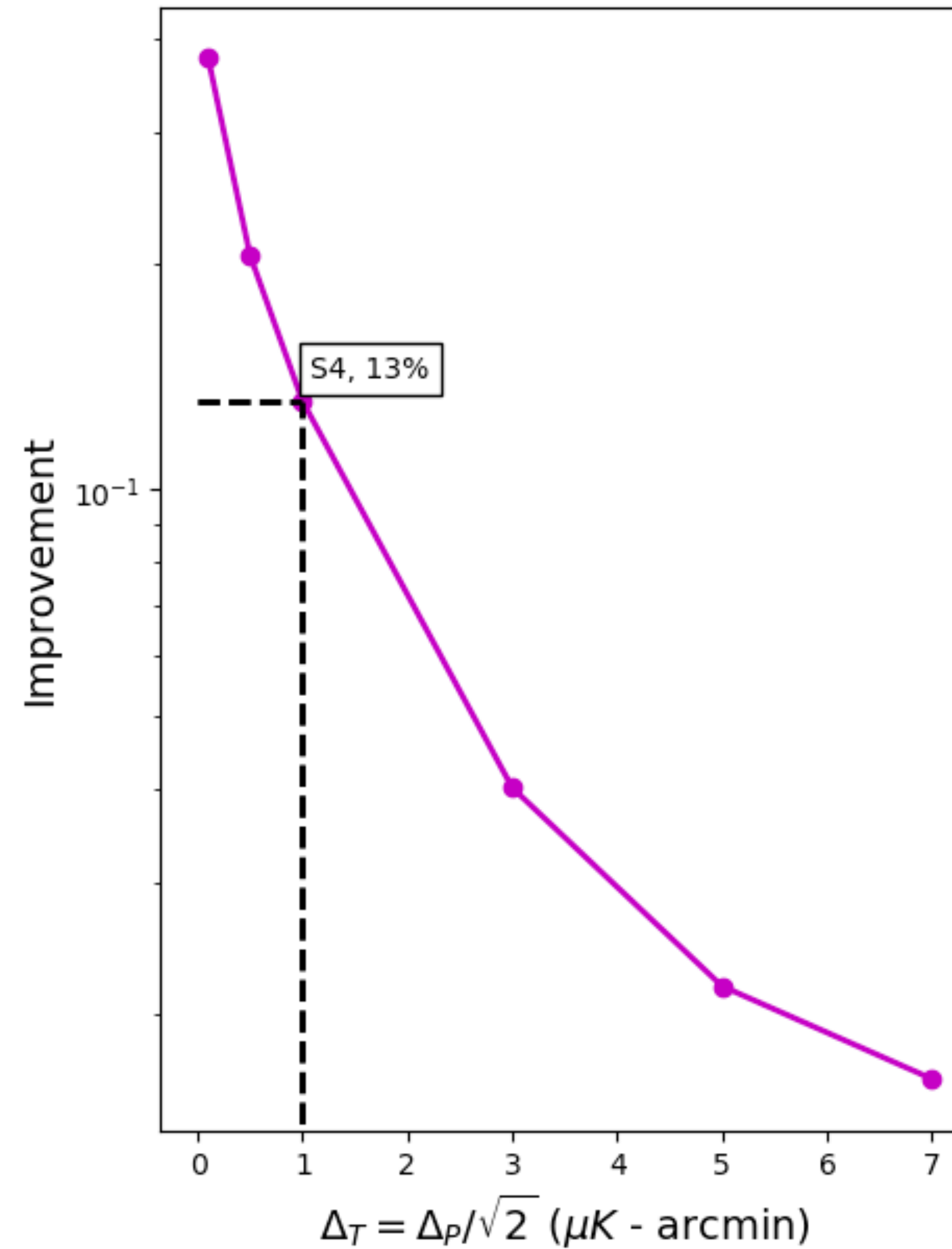
Inverse SNR for 10^5 clusters



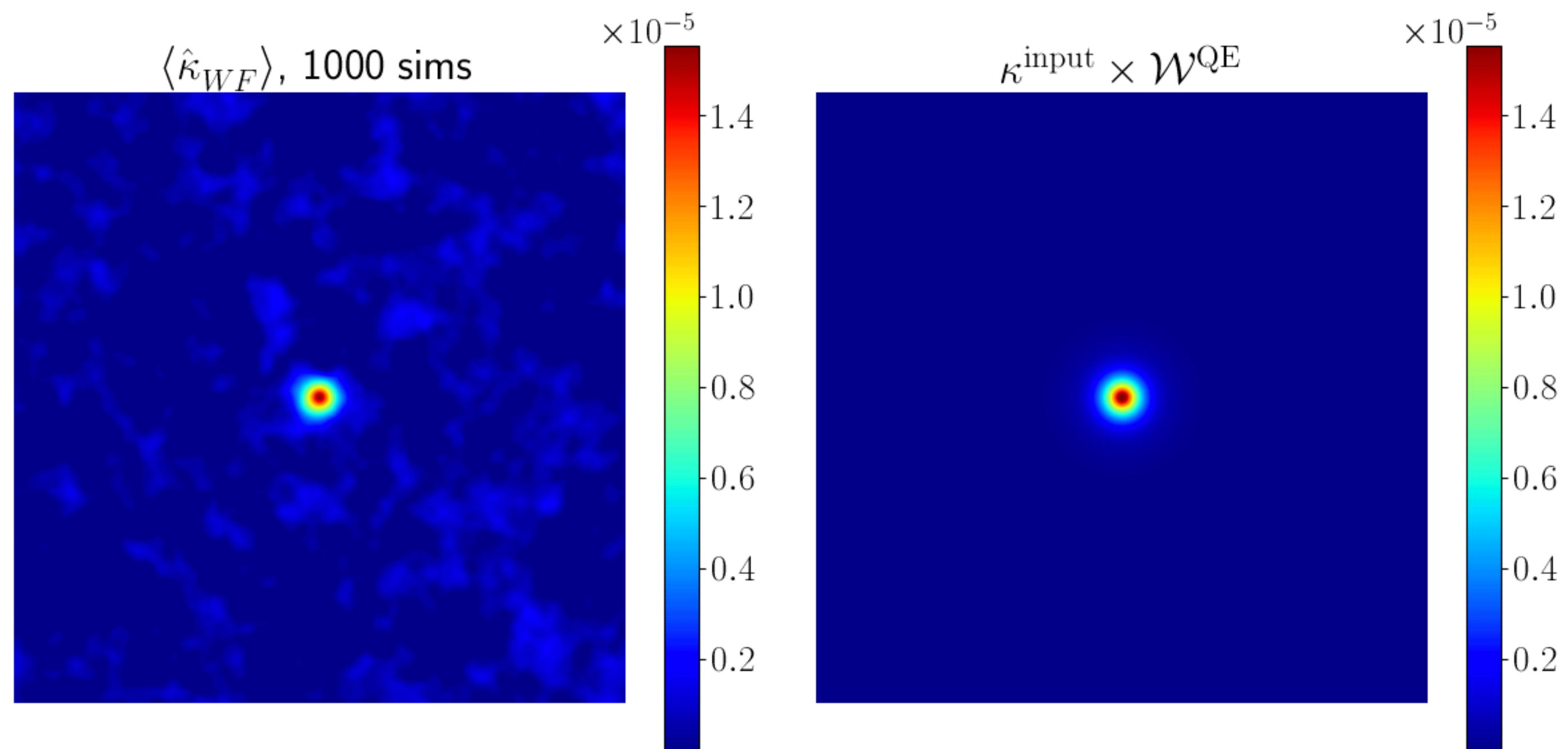
$$\Delta_T = \Delta_p / \sqrt{2} = 1 \mu K\text{-arcmin}$$

Beam = 1 arcmin

IMPROVEMENT OVER QUADRATIC ESTIMATOR



APPLICATION ON SIMULATIONS



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THANK YOU

- ▶ We study the lensing signature of galaxy-clusters in small scale CMB.
- ▶ We worked on an estimator, if you show a patch of CMB which is lensed by a galaxy cluster, it will estimate its mass (κ_0).
- ▶ In the estimator κ_0 , we use iterative estimate (MAP estimator) of $\hat{\kappa}$, instead of a quadratic estimator.
- ▶ We forecast improvement using our estimator, for CMB-S4 like experiment.
- ▶ We test our estimator on Mock data, which is lensed by galaxy-clusters.
- ▶ Furthermore, we shall test our estimator on simulations with foregrounds

[Travel to this workshop is supported under IISER Pune - Infosys Foundation Endowment Fund]

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