

Cosmic Shear Challenges

Joe Zuntz
University College London

Finding the ellipticity of lots and lots
of very small, very blurry, very
noisy, objects that are hopefully
galaxies with essentially zero bias

Joe Zuntz
University College London

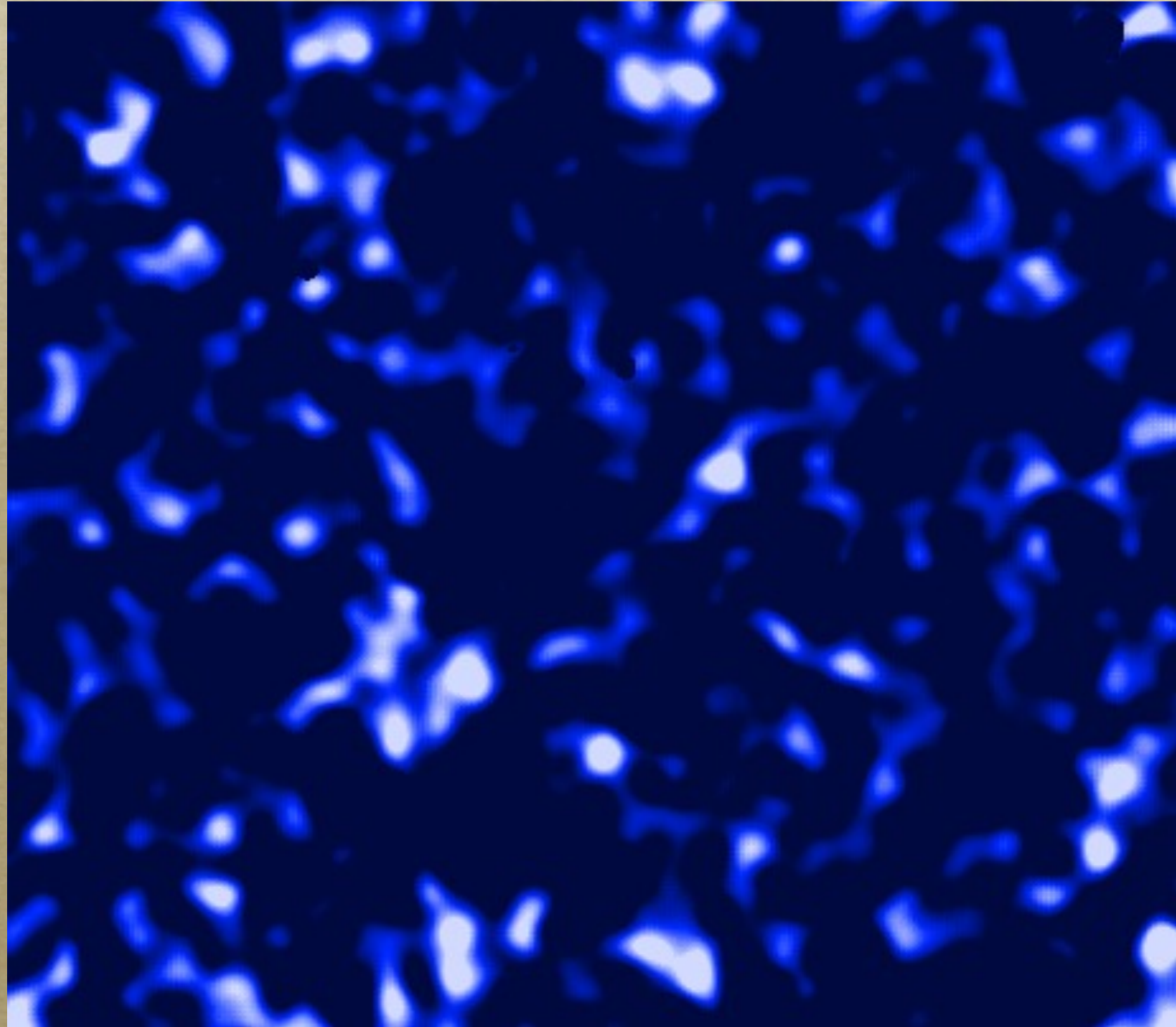
Overview

- The physics and cosmology of cosmic shear
- Why measuring shear is very hard
- Our model-fitting approach
- Challenges to model-fitters

The science goal

- *Percent-level measurements of $w(z)$ and modified-gravity measurements*
- *From structure growth and geometric distance*

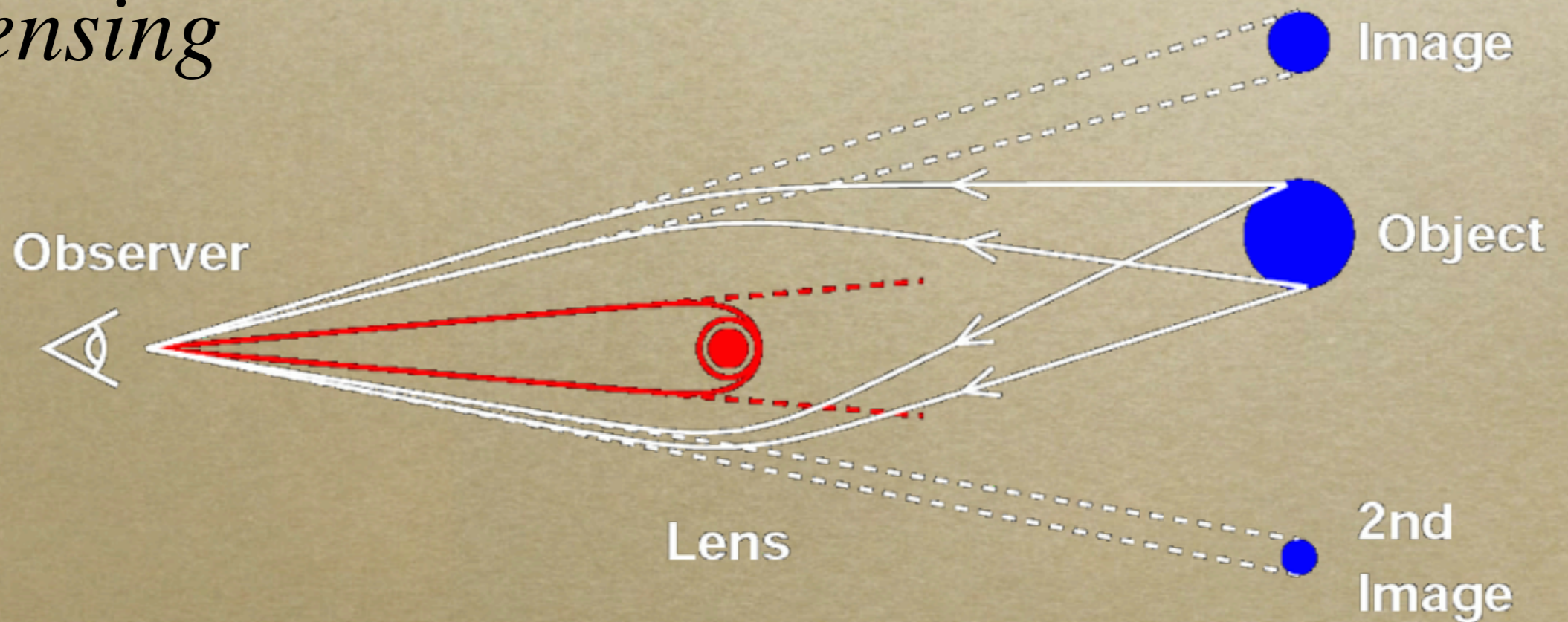
A Map of the Dark Universe



*CFHTLenS:
Heymans et al*

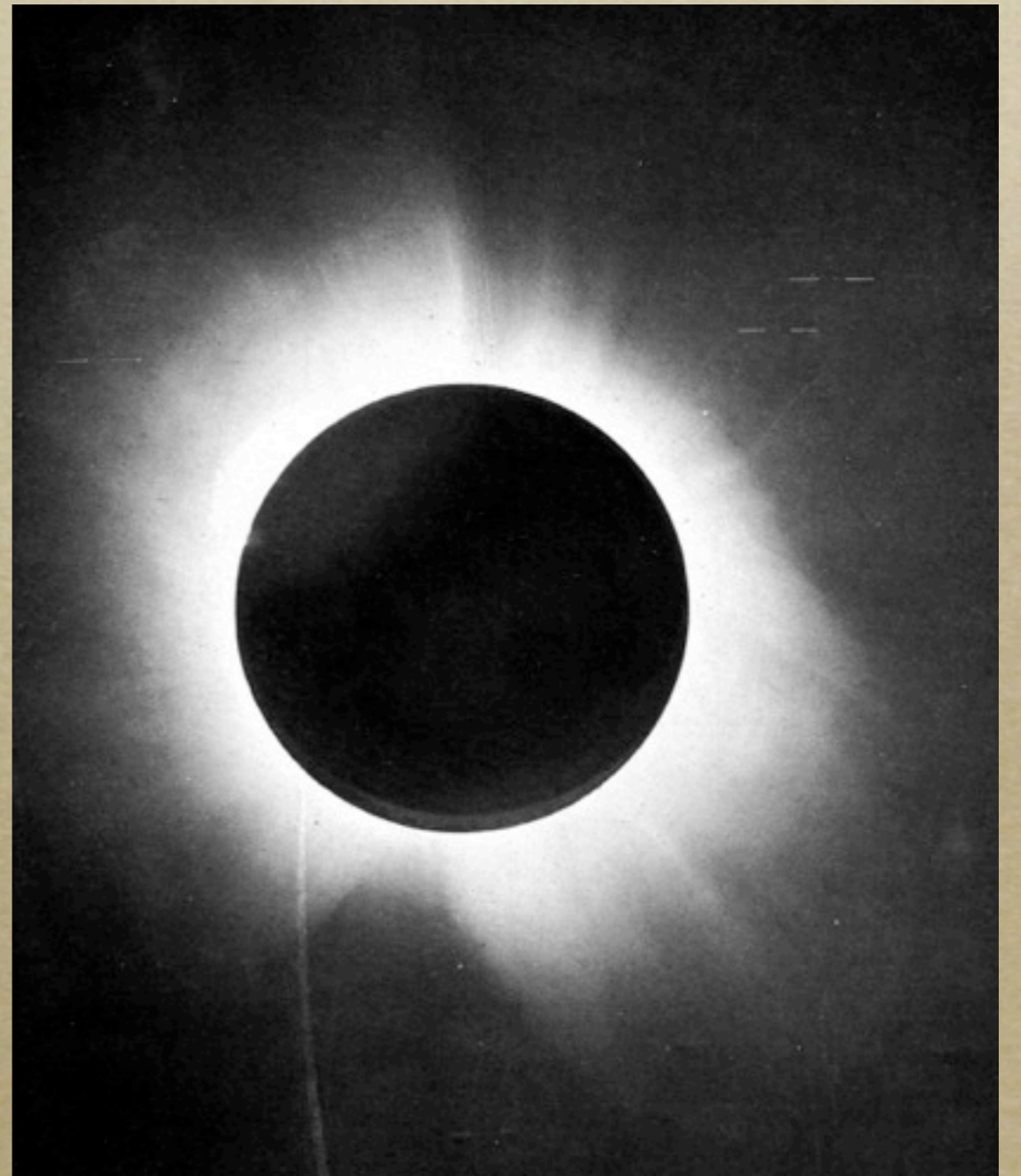
The physics of lensing

- *Everyone familiar with single object lensing*



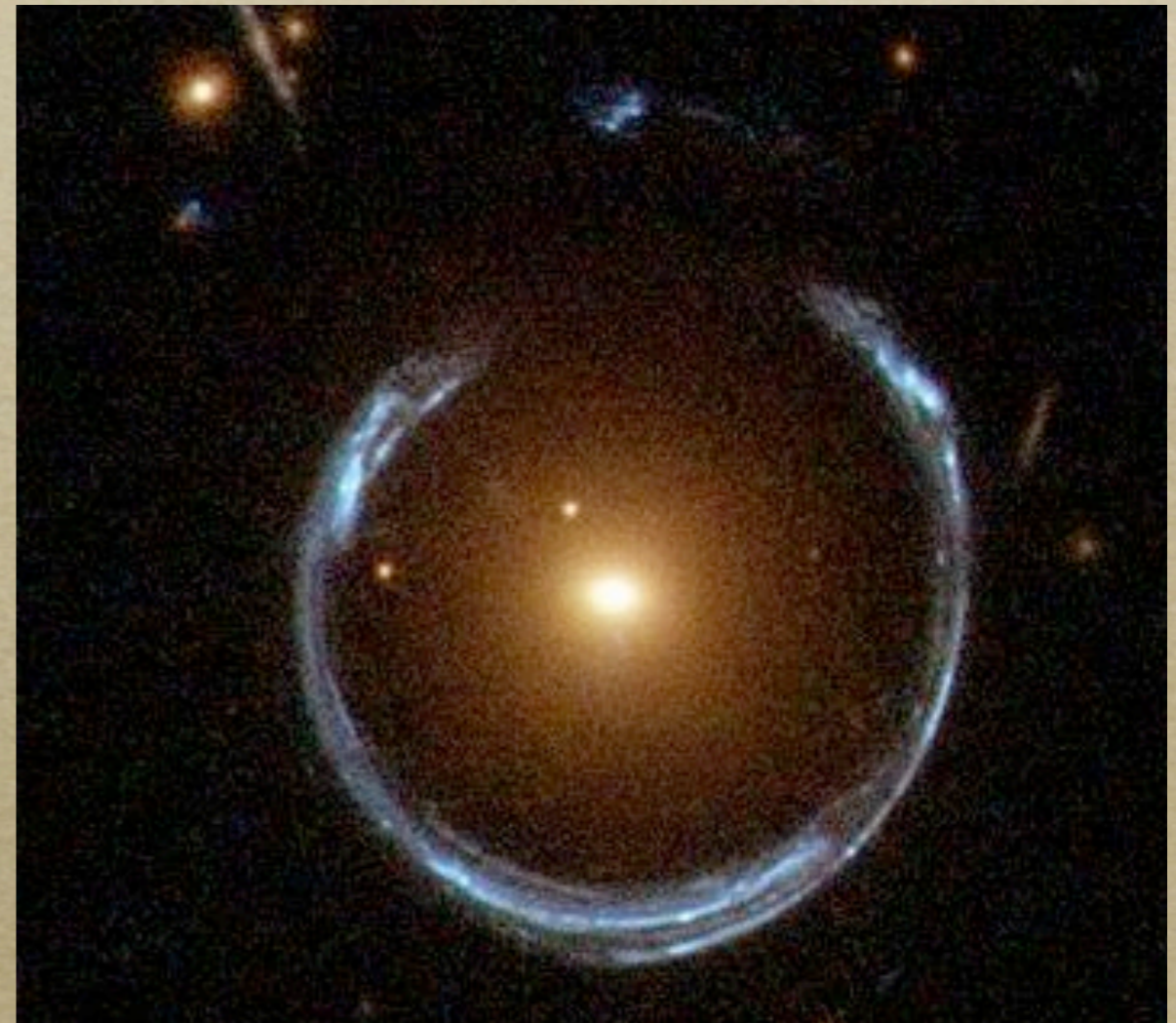
The physics of lensing

- *Everyone familiar with single object lensing*



The physics of lensing

- *Everyone familiar with single object lensing*



The physics of lensing

source
 $s(\theta)$

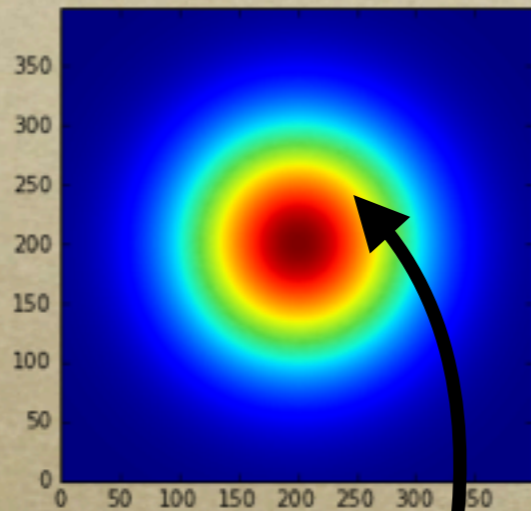
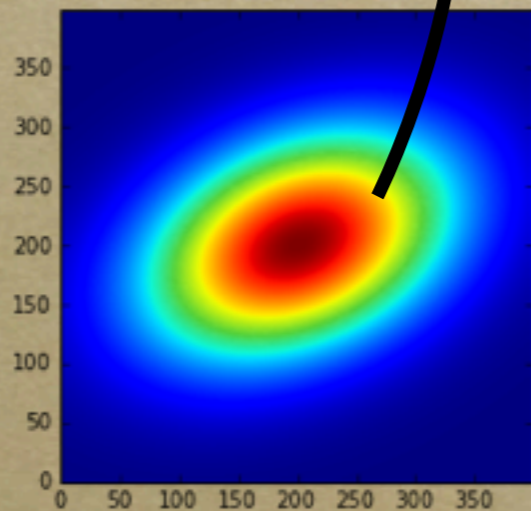


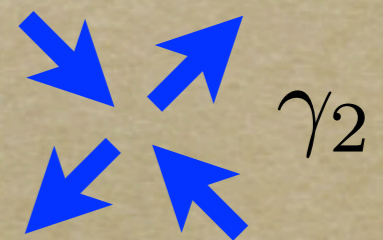
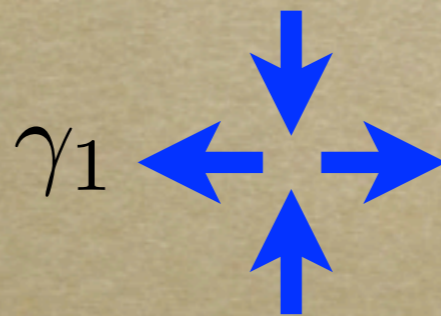
image
 $f(\theta)$



$$f(\theta_i) = s(A_{ij}\theta_j)$$

$$A = \delta_{ij} - \frac{\partial^2 \Psi}{\partial \theta_i \partial \theta_j}$$

$$= \begin{pmatrix} 1 - \kappa - \gamma_1 & -\gamma_2 \\ -\gamma_2 & 1 - \kappa + \gamma_1 \end{pmatrix}$$

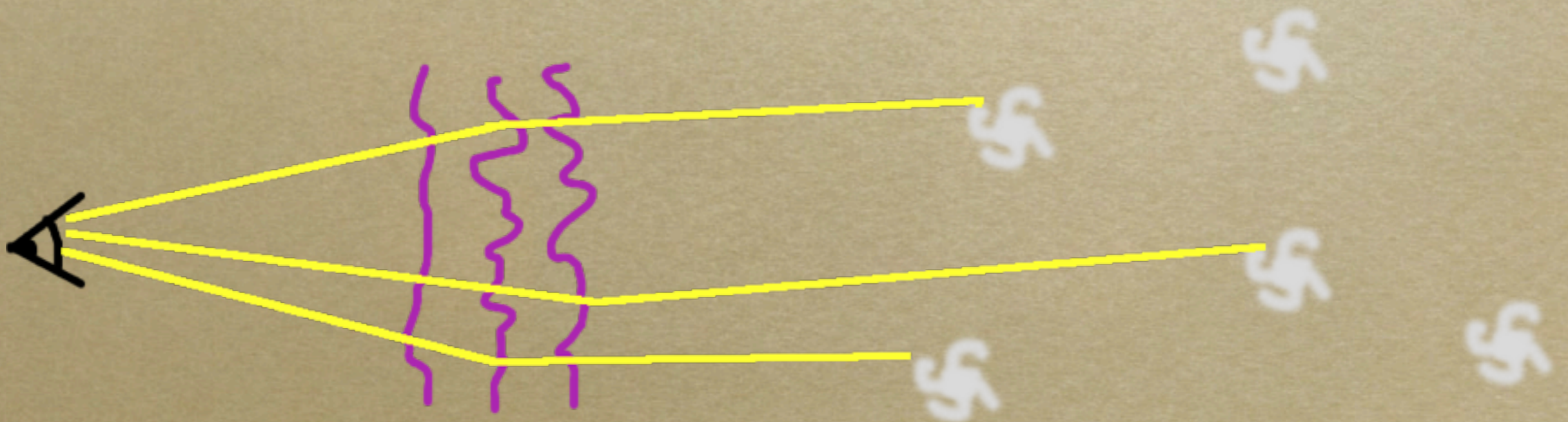


- o see Hoekstra & Jain for introduction

The cosmology of lensing

$$\frac{1}{2}k^2\Psi = \int_0^\infty d\chi W(\chi)P(\chi, \chi\theta)$$

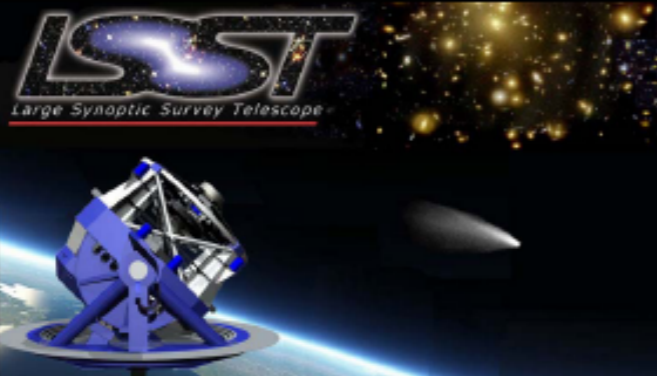
$$W(\chi) = \frac{3}{2}\Omega_m H_0^2 \chi a^{-1} \int_\chi^\infty d\chi_s n(\chi_s) \frac{\chi_s - \chi}{\chi_s}$$



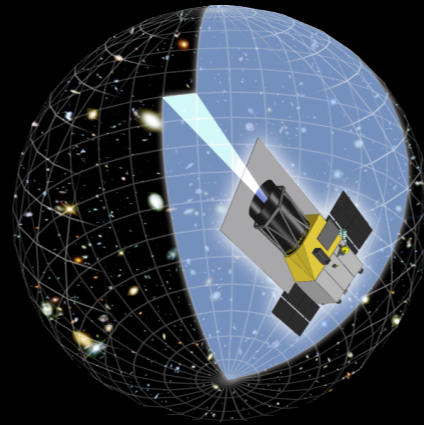
The cosmology of lensing

The Dark Energy Survey

KIDS



HYPER SUPRIME-CAM



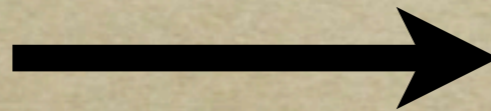
Pan-STARRS



EUCLID



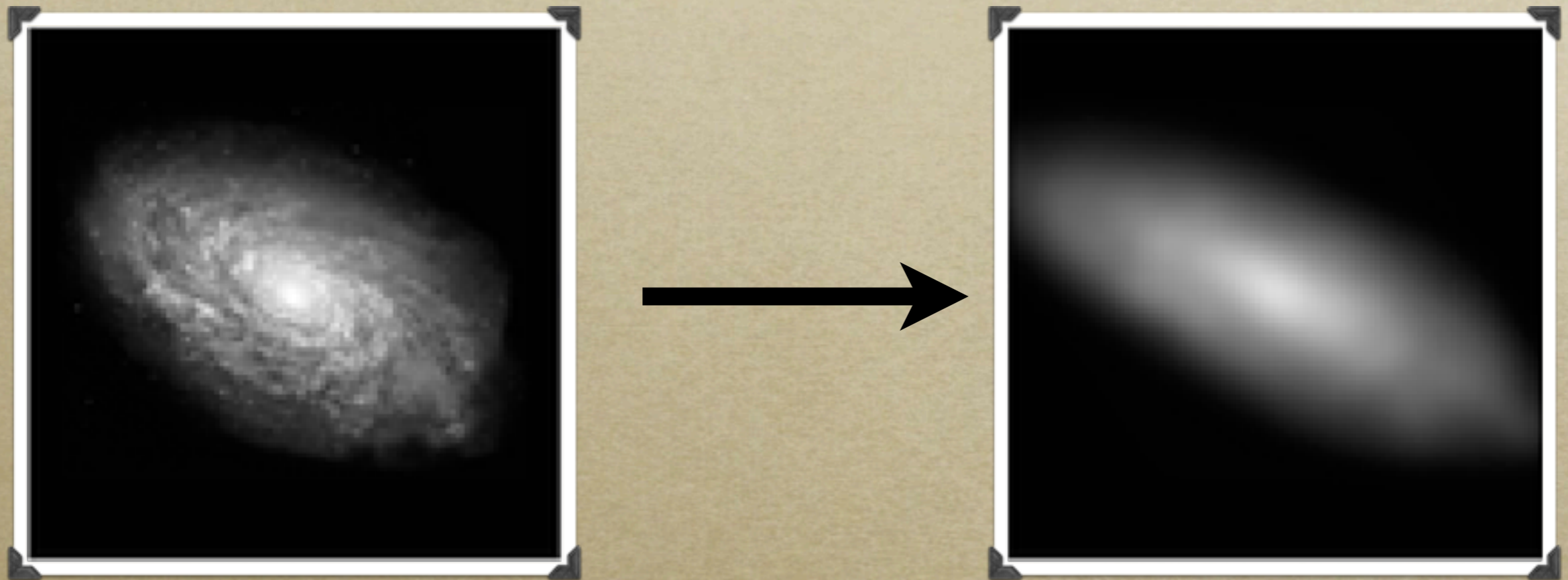
Measuring Shear



Shear

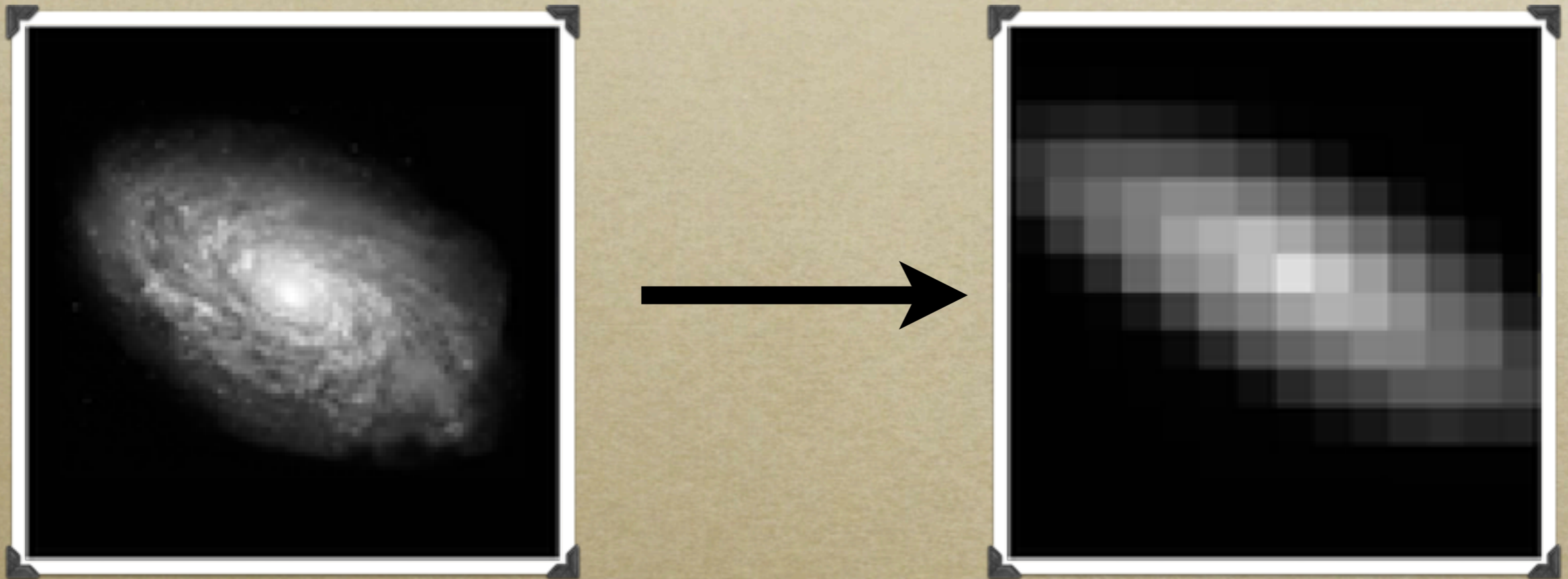
Images: Bridle et al 2008

Measuring Shear



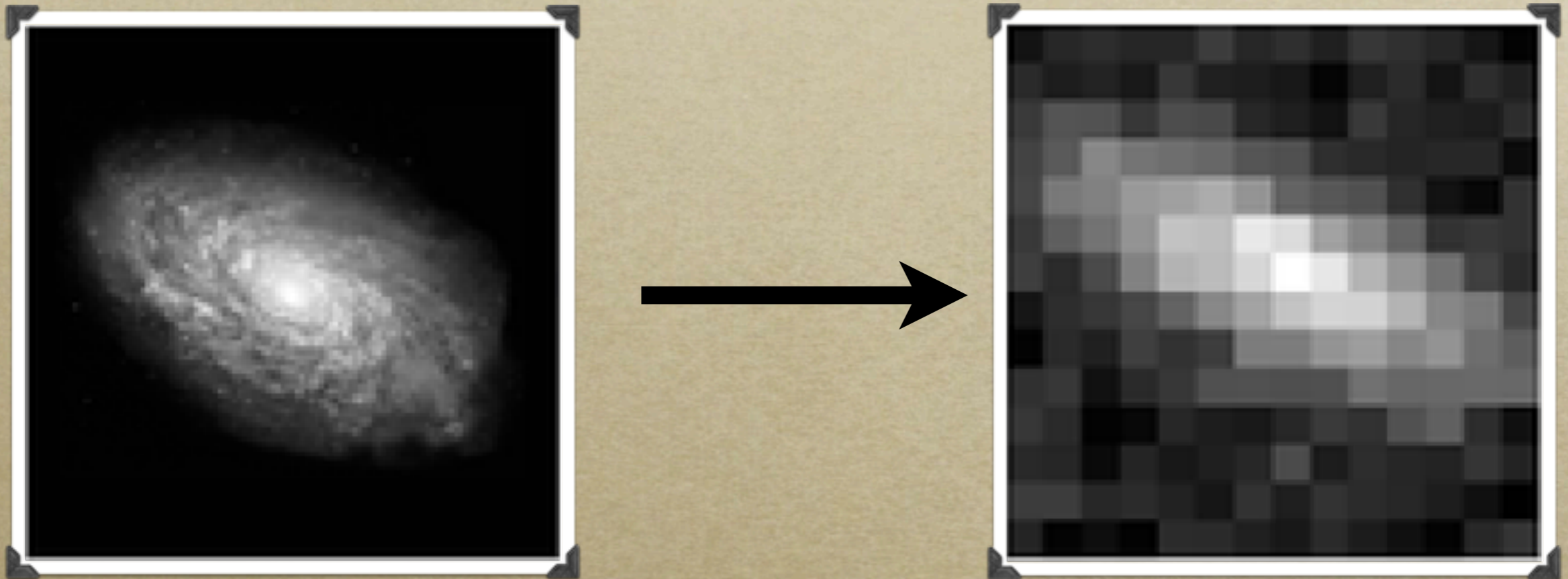
Shear + Point-spread

Measuring Shear



Shear + point-spread + pixelization

Measuring Shear



Shear + point-spread + pixelization + noise

Requirements

- *How well do we have to do this? Taylor expand our estimates:*

$$\hat{\gamma}_i \approx (1 + m_i)\gamma_i + c_i$$

- *Then we require for upcoming experiments:*

$$m_i < 4 \cdot 10^{-3}$$

$$c_i < 6 \cdot 10^{-4}$$

Some approaches

- *Estimators - quadrupole moments, KSB*
- *Shapelets and other simple fits*
- *Other methods*
- *Modelling methods*

Modelling Galaxies

- *Forward modelling methods: fit parameters (including ellipticity)*
- *A maxim:*
“Don’t model your data. Model the process that led to your data”

Basic Galaxy Model



Basic Galaxy Model

*Exponential
Disc*



$$\exp - (x^T M x)^{\frac{1}{2}}$$

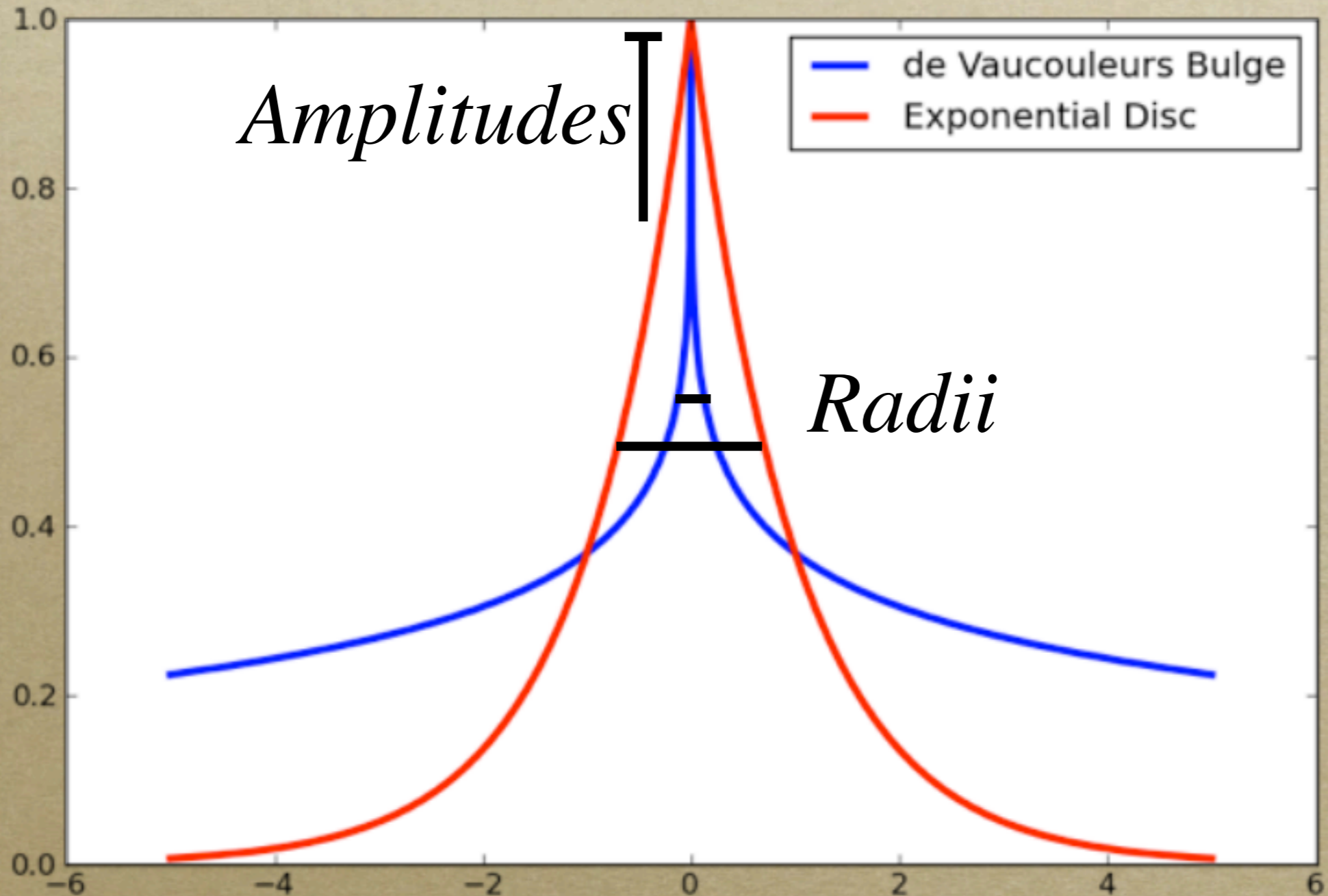
Basic Galaxy Model

*De
Vaucouleurs
Bulge*



$$\exp - (x^T M x)^{\frac{1}{8}}$$

Basic Galaxy Model



Basic Galaxy Model

Parameter	Meaning	Fixed
x_0	Horizontal centroid	
y_0	Vertical centroid	
e_1	x-y shear	
e_2	45° shear	
r_d	Disc half-light radius	
A_b	Bulge peak flux	
A_d	Disc peak flux	
R_r	Bulge-disc size ratio	✓
n_d	Disc Sérsic index	✓
n_b	Bulge Sérsic index	✓
Δe	Bulge-disc ellipticity	✓
$\Delta\theta$	Bulge-disc angle	✓

Im3Shape

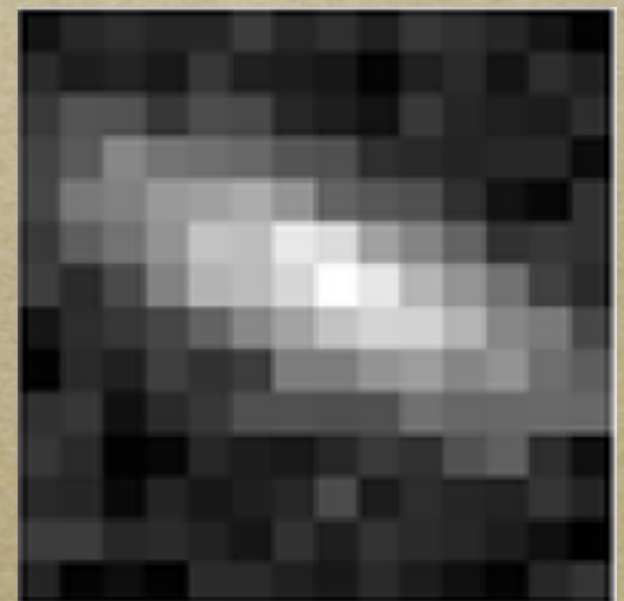
- *Reimagining of Bridle im2shape code*
- *Forward-model ML method*
- *Optimized but pleasant C code*
- *Flexible: Very easy to add components*

Known Biases

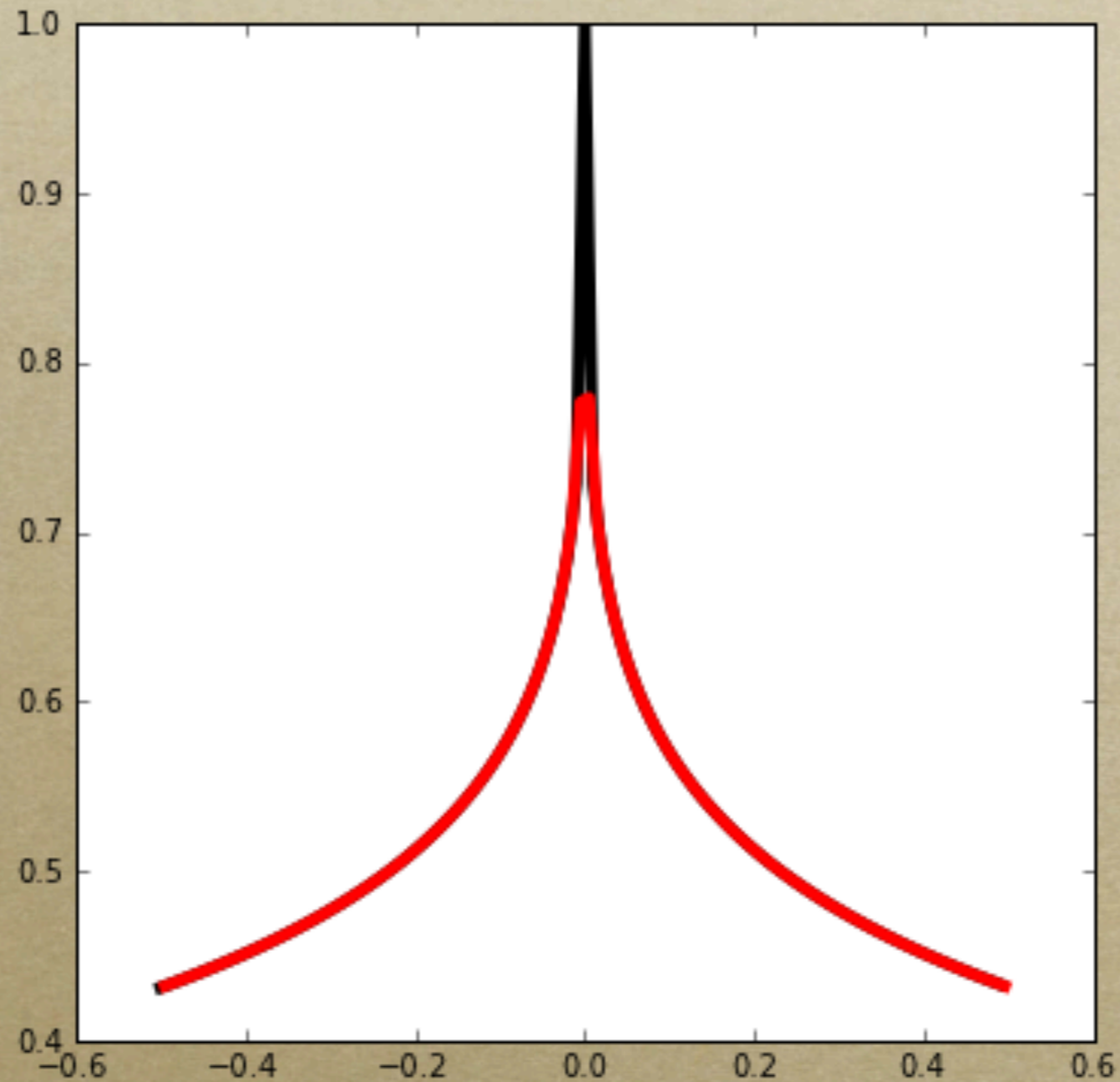
- *Resolution bias*
- *Model bias*
- *Noise bias*

Resolution Bias

- *True model has infinite resolution*
- *Matter most for sharp bulges*
- *Central pixel double upsampling*



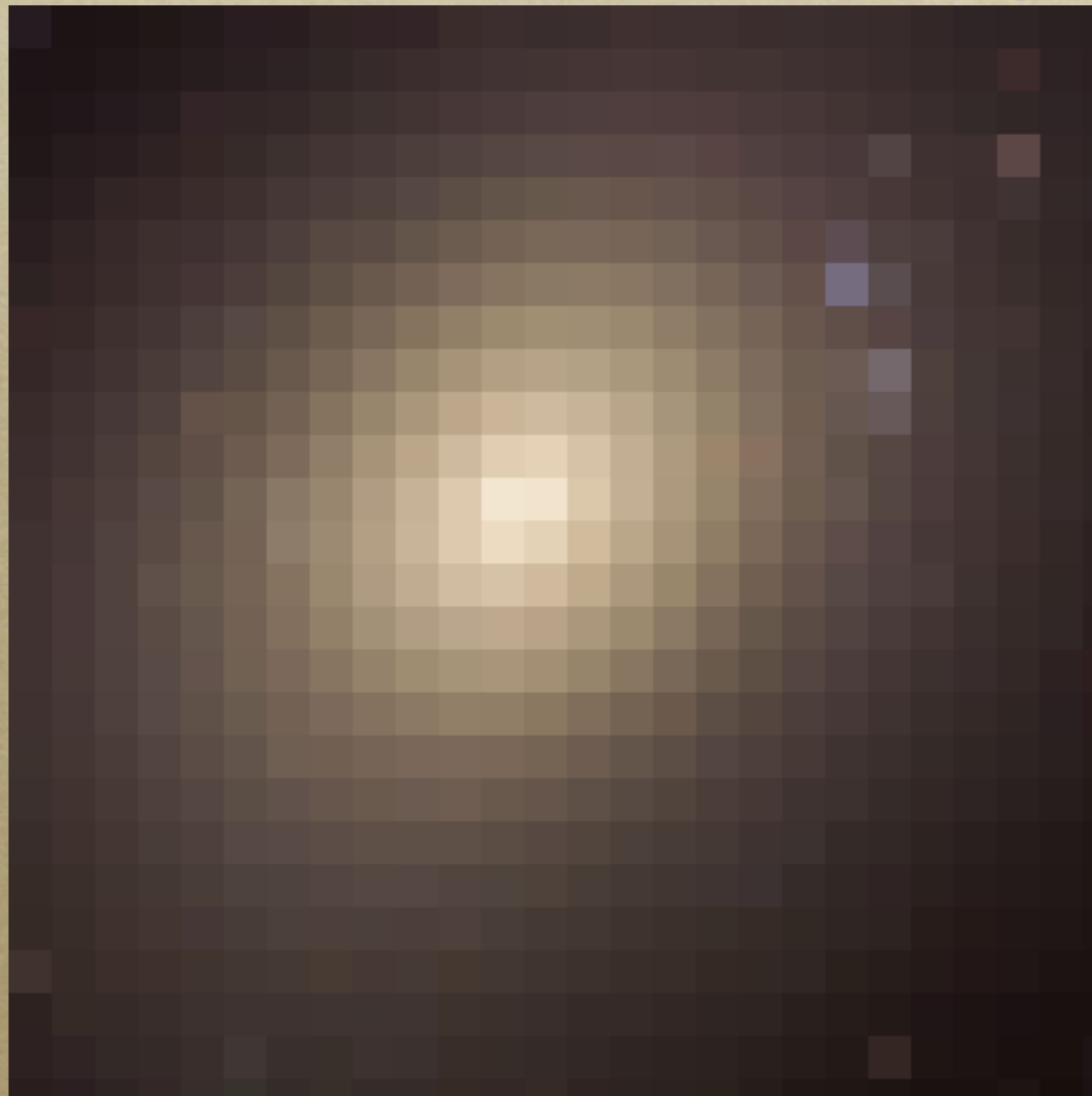
Resolution Bias: Sharp Centers



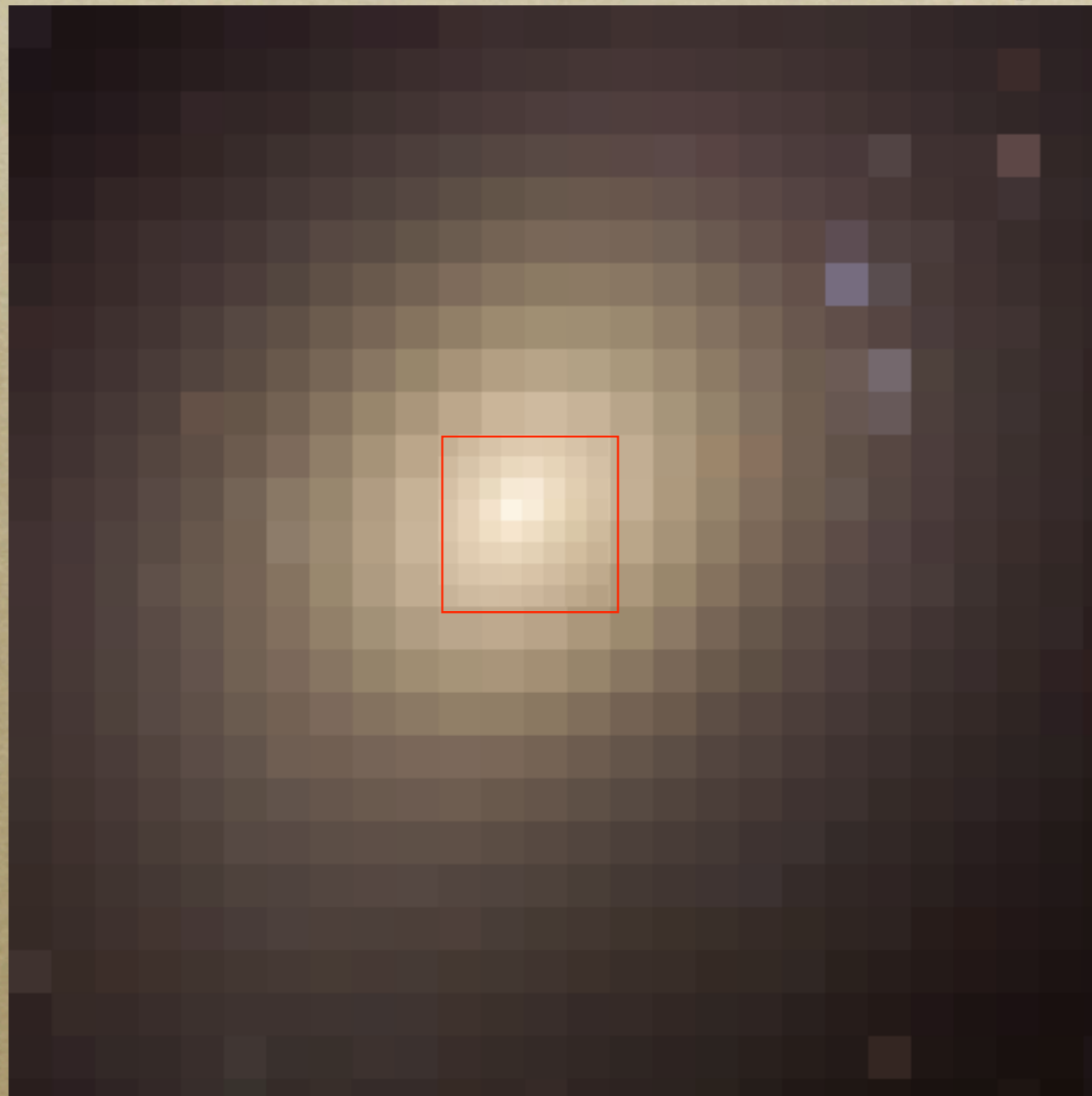
Central Pixel Upsampling



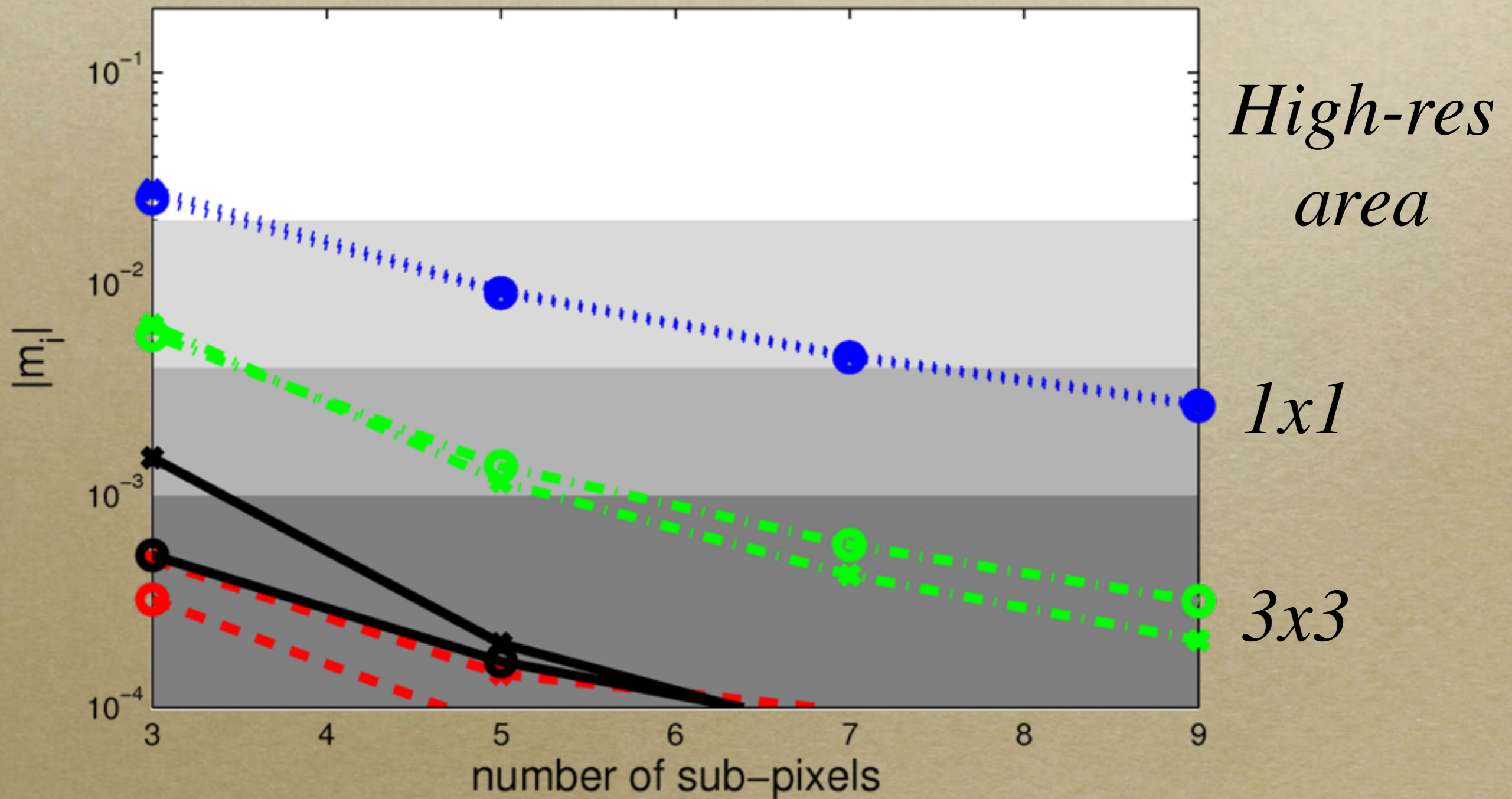
Central Pixel Upsampling



Central Pixel Upsampling

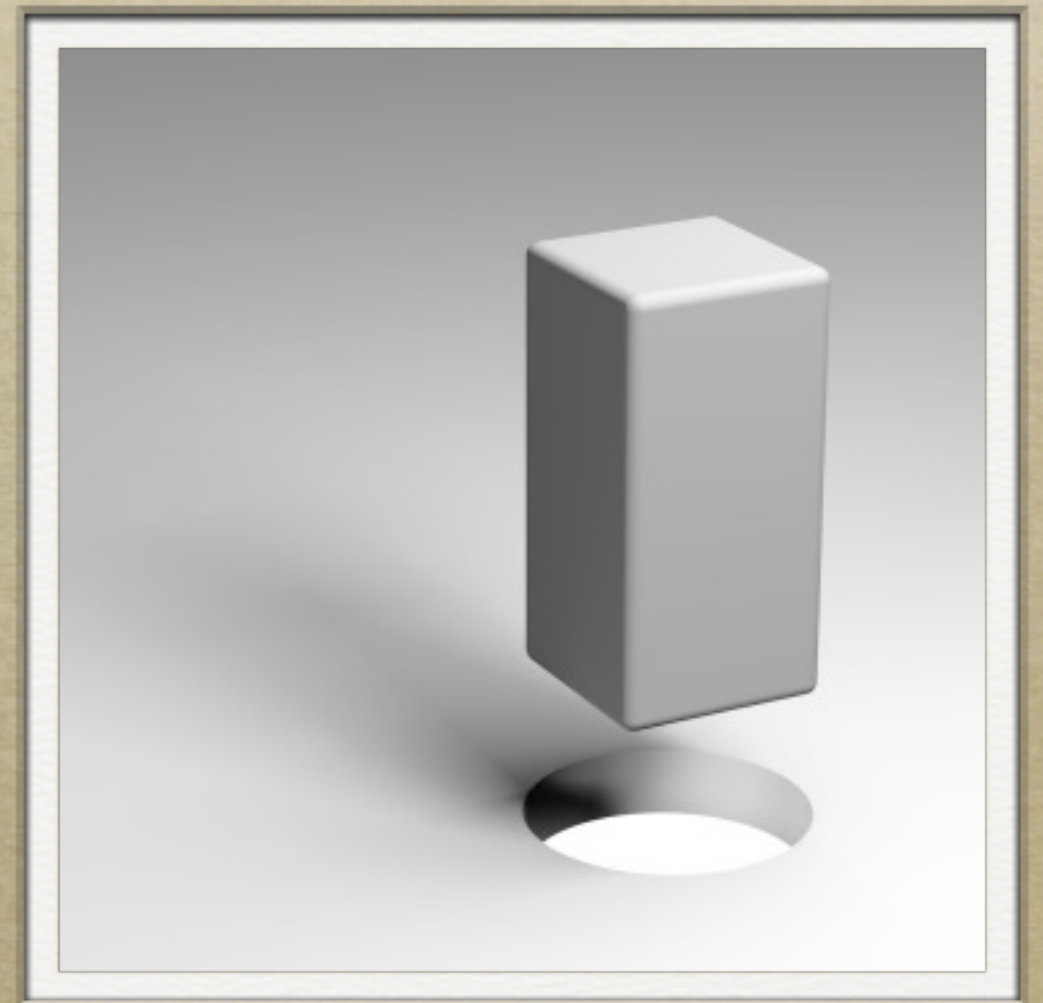


How much area is higher res?

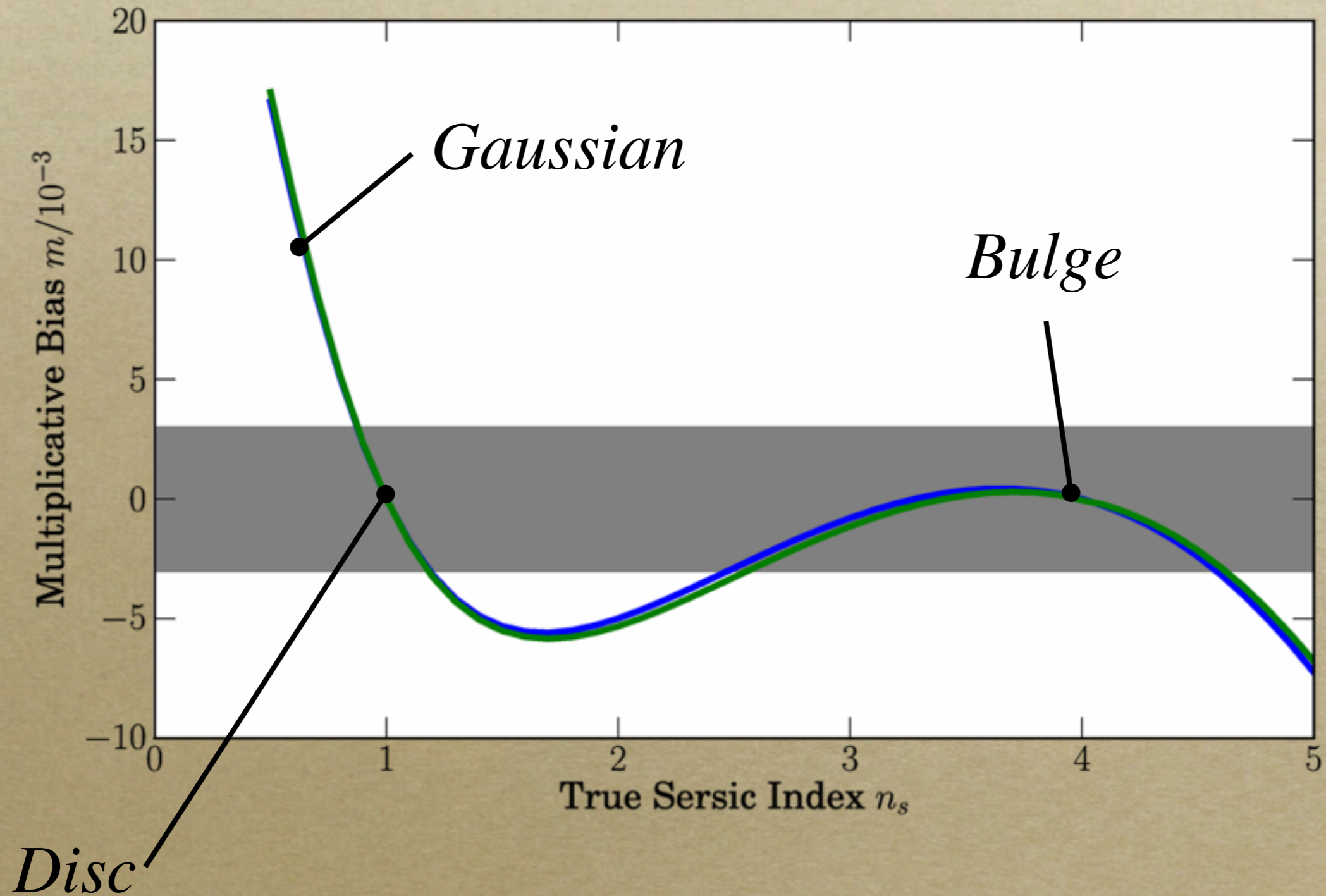


Model bias

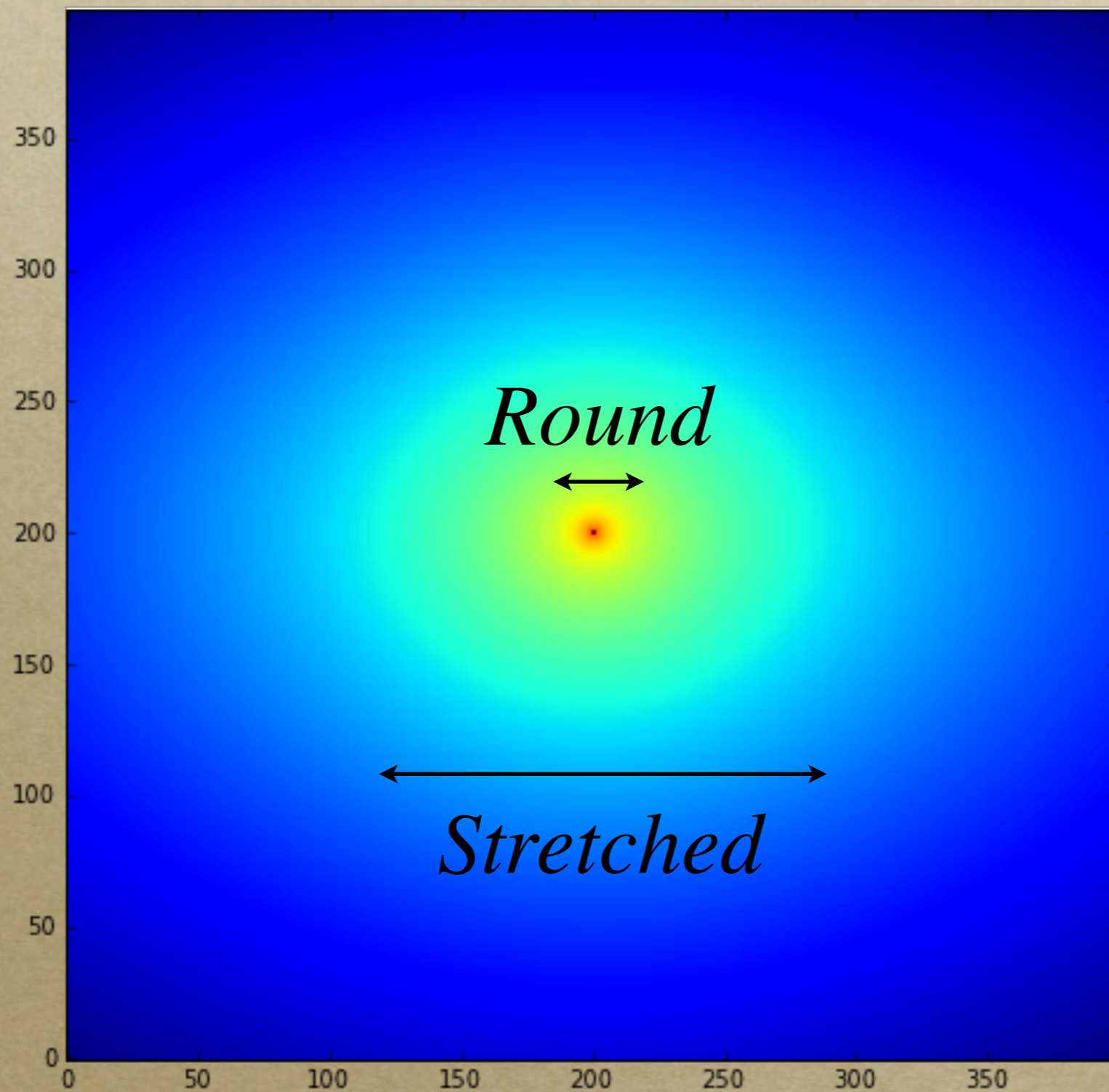
- *When you fit an “incorrect” model to the data*
- *e.g.*
 - *Different profiles*
 - *Off-centering*
 - *Radius ratio*



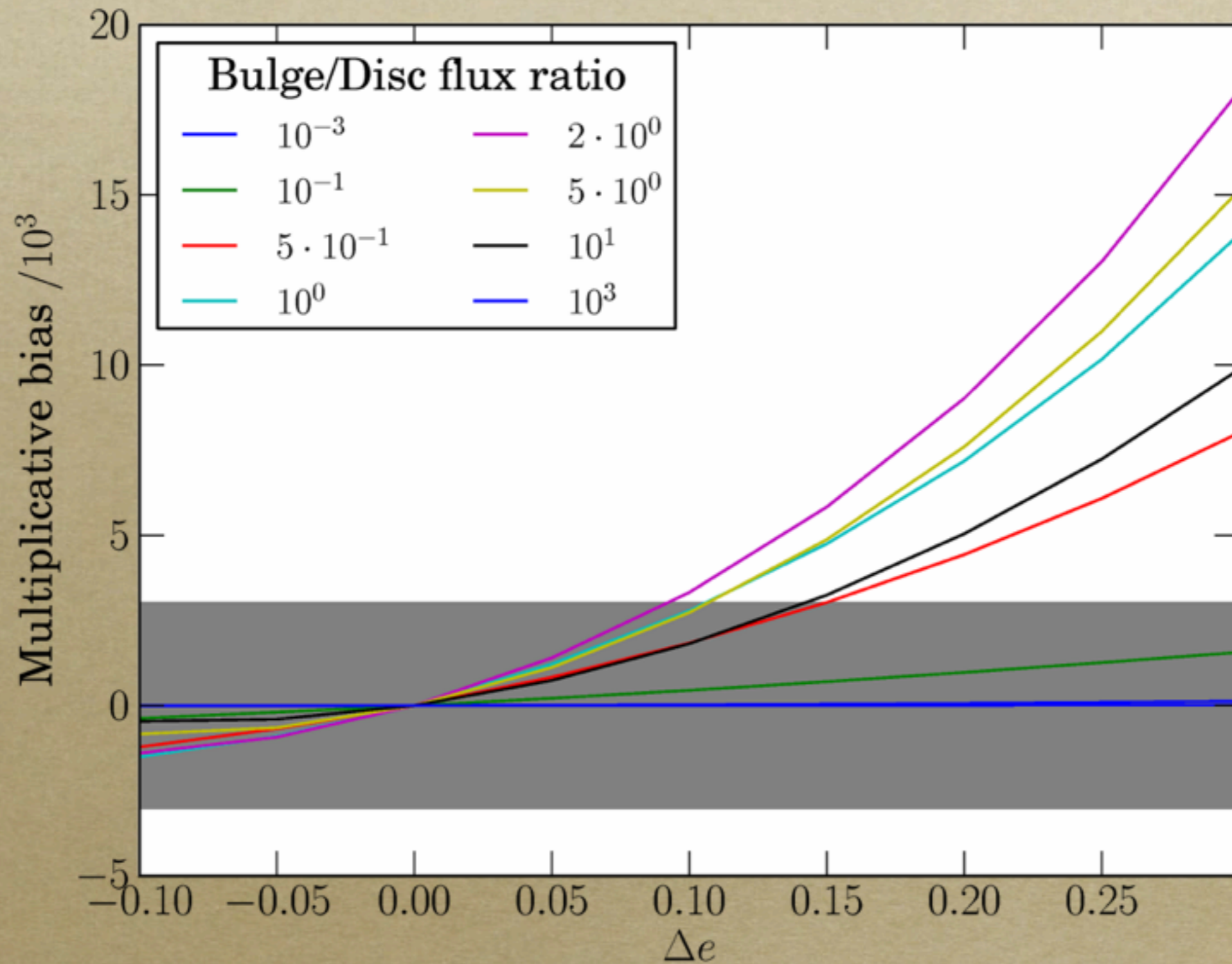
Model Bias:



Model bias: Component e_i

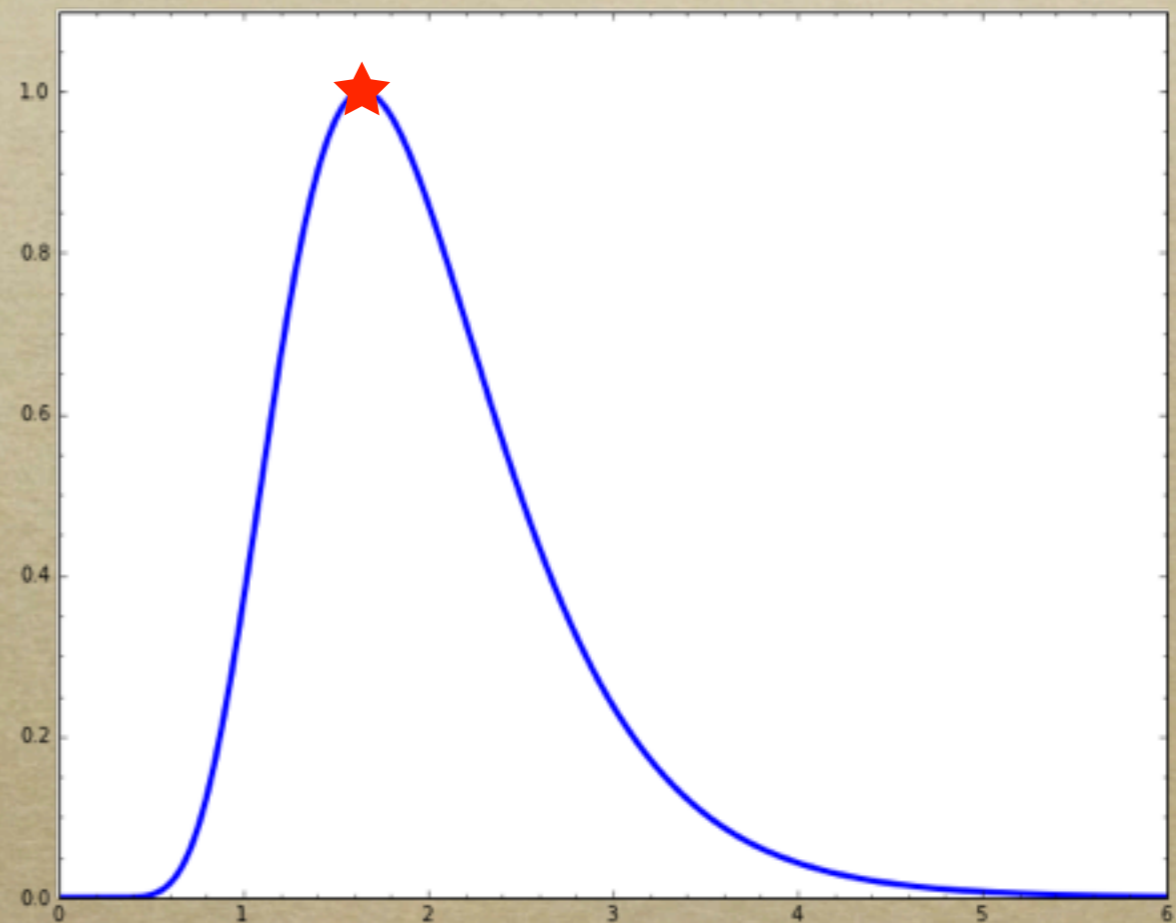


Model bias: Component e_i

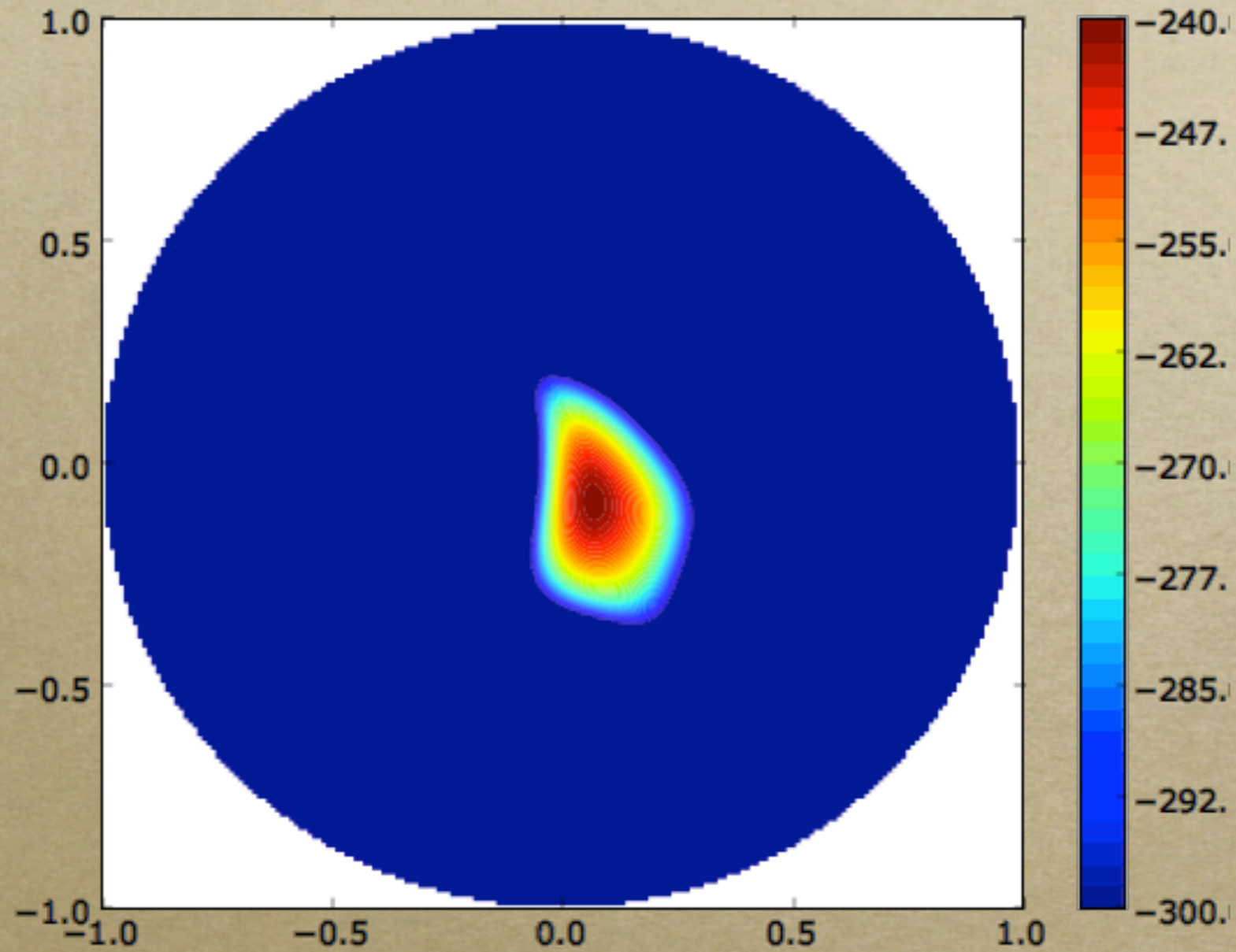


Noise bias

- *Mean of ML \neq ML of Mean*
- *Combining galaxies we should multiply PDFs*
- *Nonlinear dependence of pixels on ellipticity*



Noise Bias: Origin



Noise Bias: Origin

- *Refriger et al 2012*
- *Kacprzak et al 2012*

$$b[\hat{a}_i] = -\frac{1}{2}(F^{-1})_{ij}(F^{-1})_{kl}B_{jkl} + O(\rho^{-4})$$

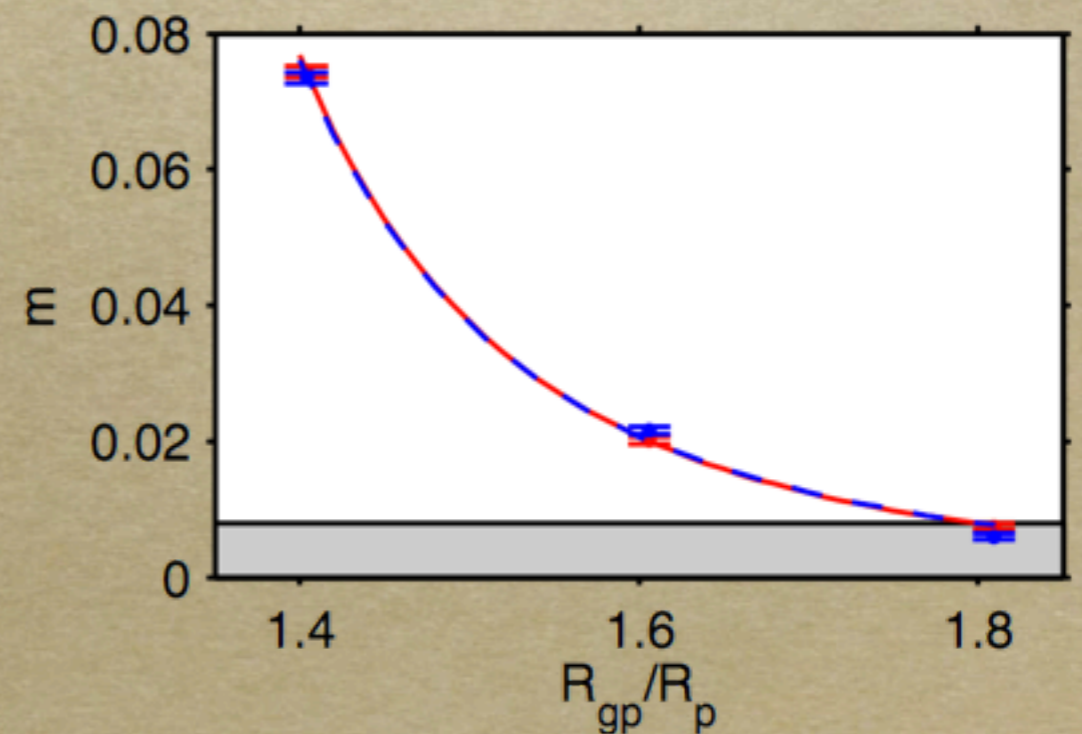
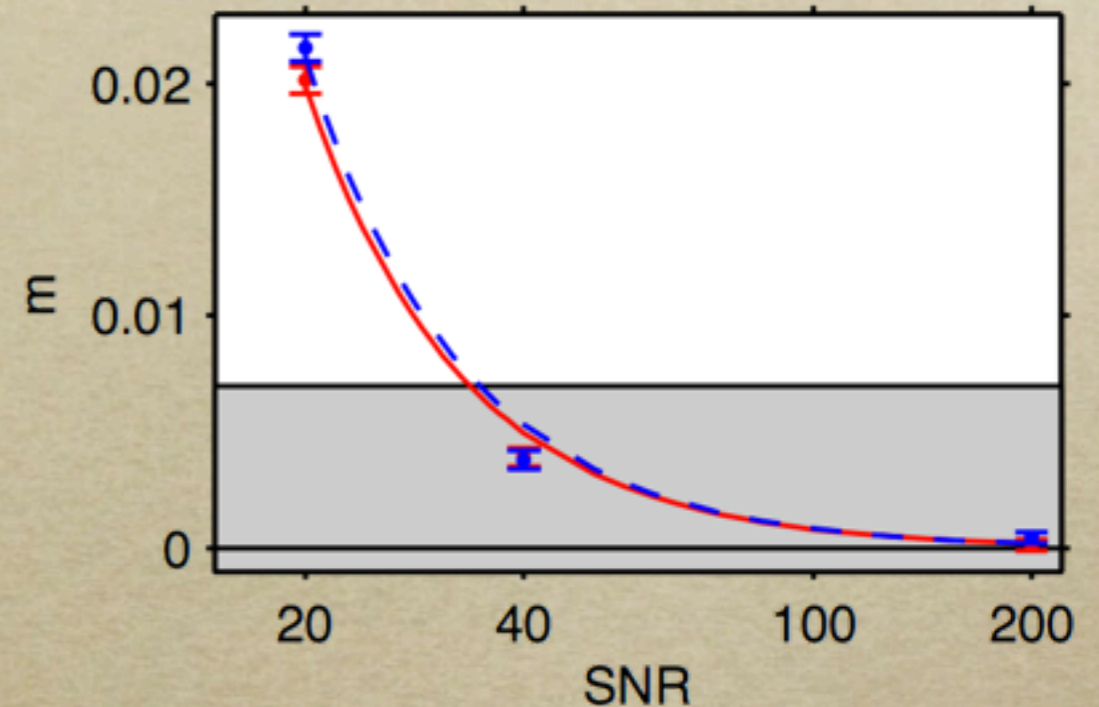
$$B_{ijk} = \left\langle -\frac{1}{2} \frac{\partial^3 \ln L}{\partial a_i \partial a_j \partial a_k} + \frac{\partial \ln L}{\partial a_j} \frac{\partial^2 \ln L}{\partial a_i \partial a_k} \right\rangle$$

Some fun facts

<i>To minimize:</i>	error ²	$\prod(w * \text{error})$	abs(error)
<i>Take from your posterior the...</i>	Mean	Mode (ML)	Median

Calibrating noise bias

- *Calibrate with simulations over relevant parameters*
- *Apply with polynomial fit*
- *“Bias-on-bias” problem*



Dodging noise bias?

- *Can we avoid noise bias altogether?*
- *Samples from $P(e|I)$*
- *Requires prior information and power spectrum estimation that can cope*

Summary

- We are about to release a shear-measurement code called *Im3shape* which fits bulge+disc galaxy models to images
- Resolution bias pushes us to high model resolution requirements, especially at image centers
- Model biases are not quite enough to worry us yet
- Noise bias is very significant to ML methods but can be calibrated

Unanswered Question

- *How do the different biases interact?*