

Astronomy in SJTU, and Weak Lensing Measurement

Jun Zhang (张骏) (SJTU)

Collaborators:

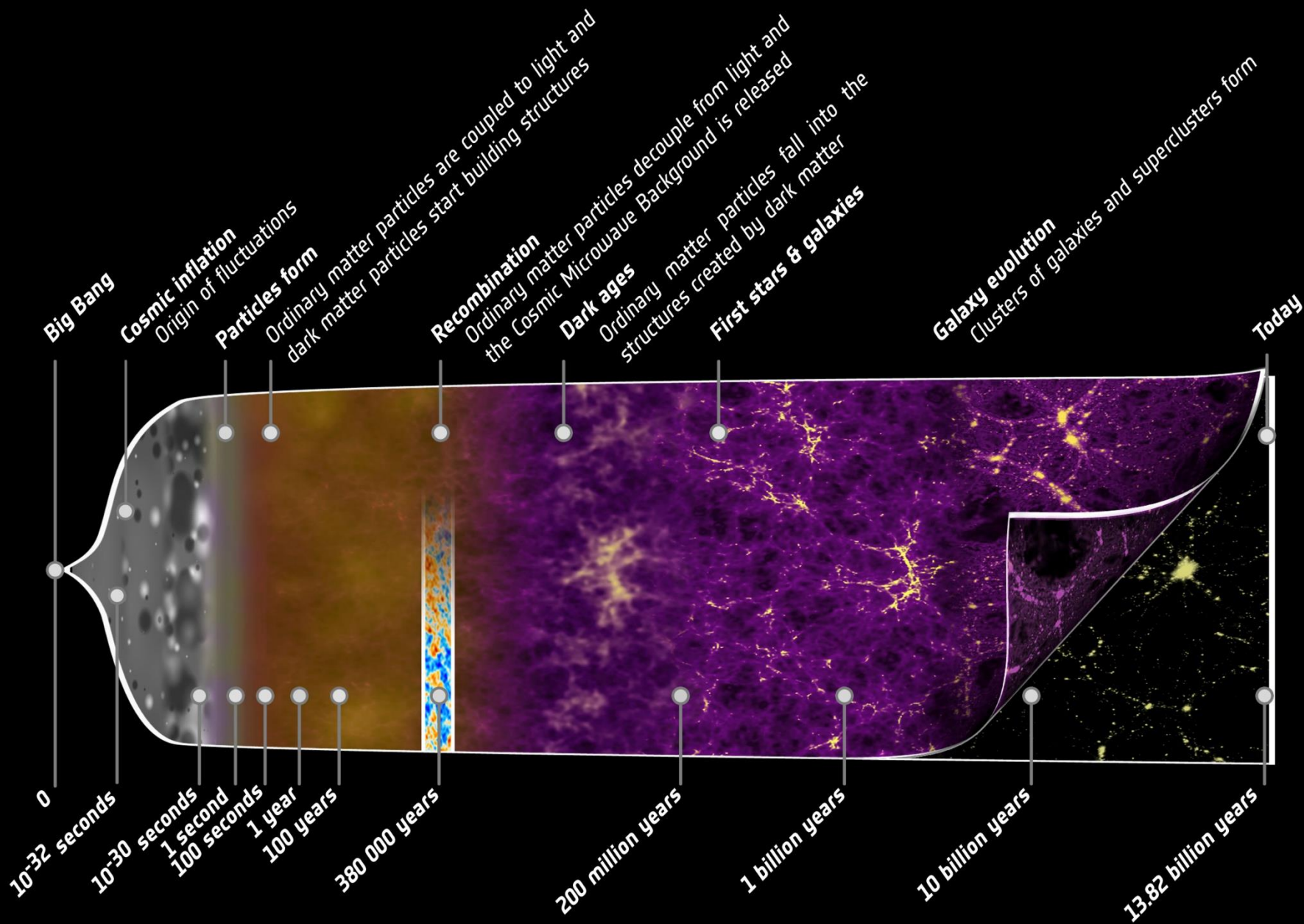
Fuyu Dong, Yingke Li, Xiangchong Li, Tianhuan Lu, Xiaohu Yang, Pengjie Zhang, Wentao Luo (SJTU); Nobuhiko Katayama (IPMU); Dezi Liu, Zuhui Fan (PKU); Liping Fu (SHNU); Guoliang Li (PMO). .

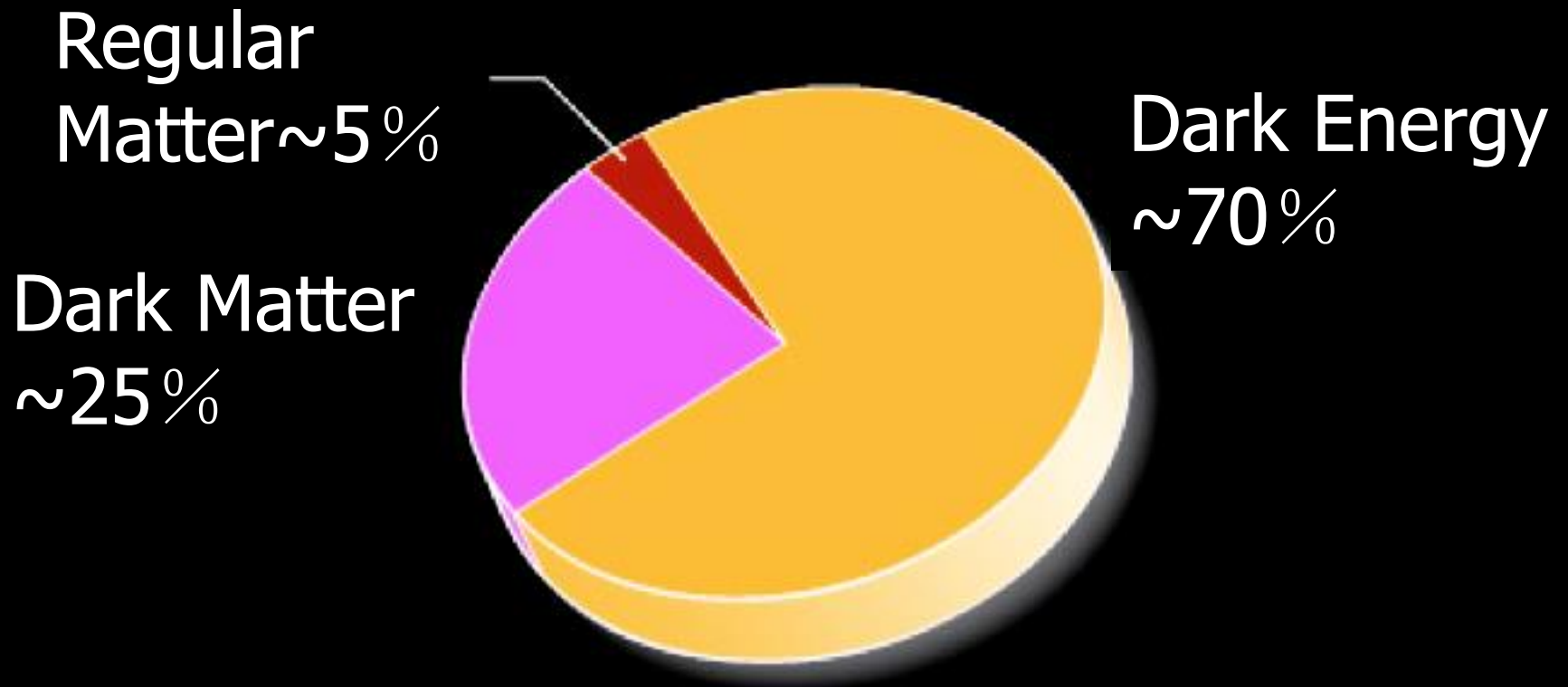
Outline:

- Introduction & Motivation
- The Fourier_Quad method
- Application on the CFHTlens data

Outline:

- Introduction & Motivation
- The Fourier_Quad method
- Application on the CFHTlens data



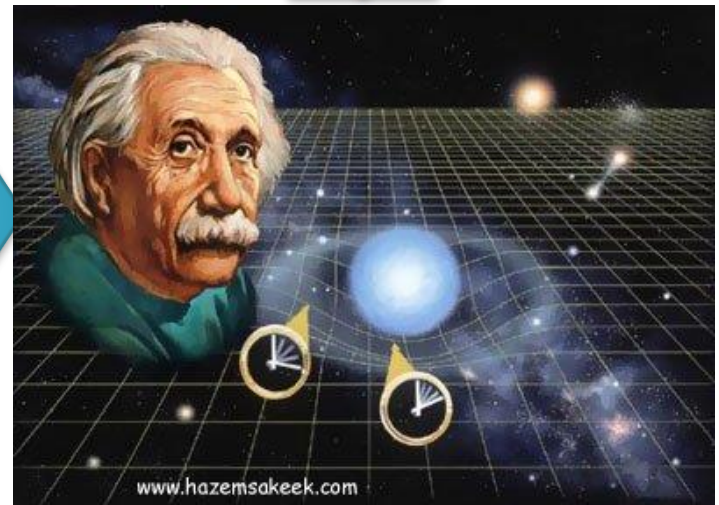
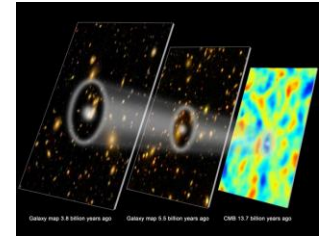
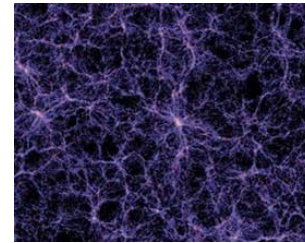
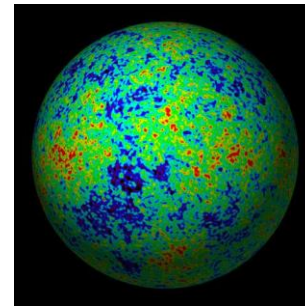
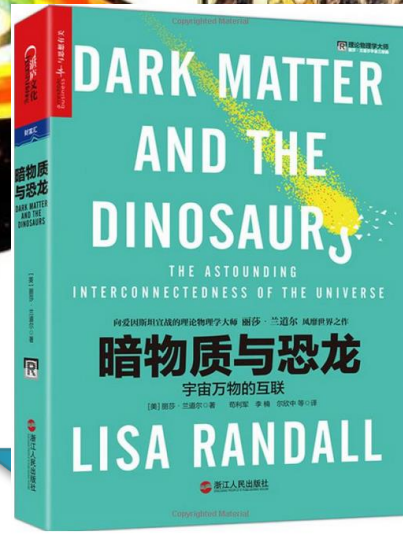


Cosmic Energy Budget

Credit: NASA/WMAP



元90~168)



Department of Astronomy of SJTU

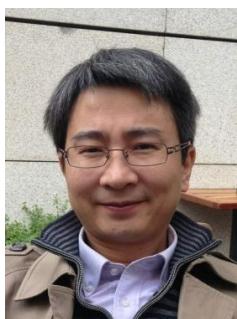
- Full Professors



Yipeng Jing



Bin Wang



Xiaohu Yang



Haiguang Xu



Pengjie Zhang

- Junior Faculty Members



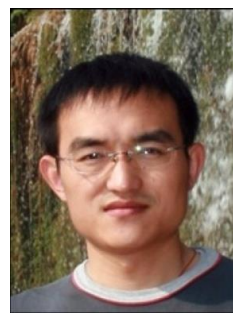
Le Zhang



Ying Zu



Dangbo Liu



Chengze Liu



Dechang Dai



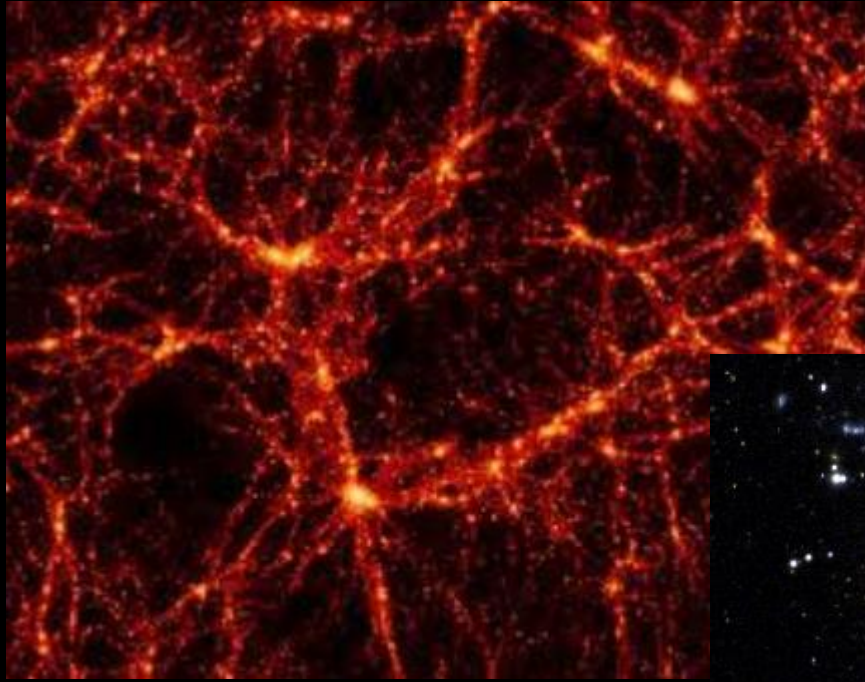
Jun Zhang



Yu Yu

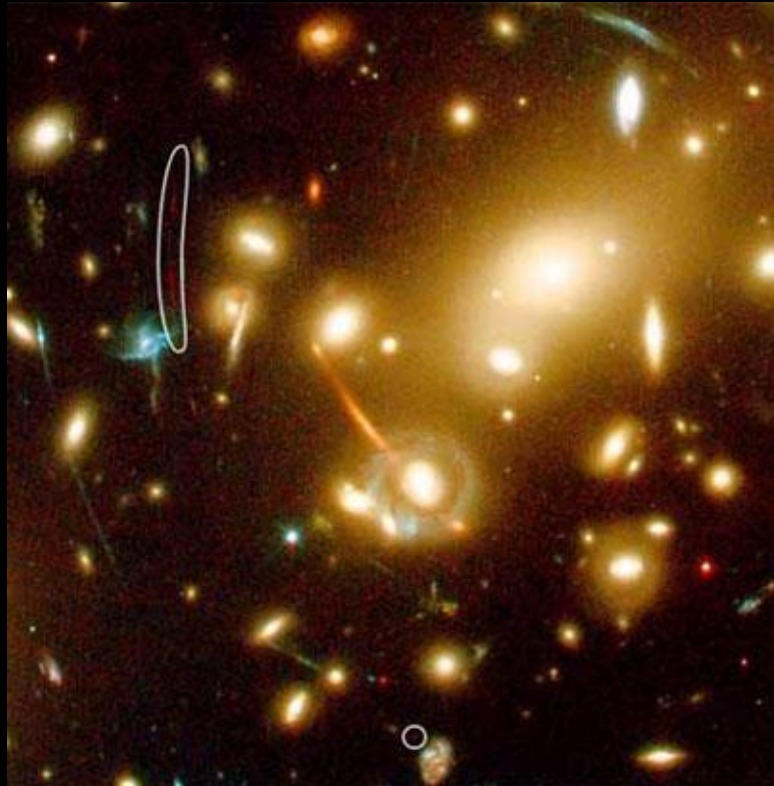
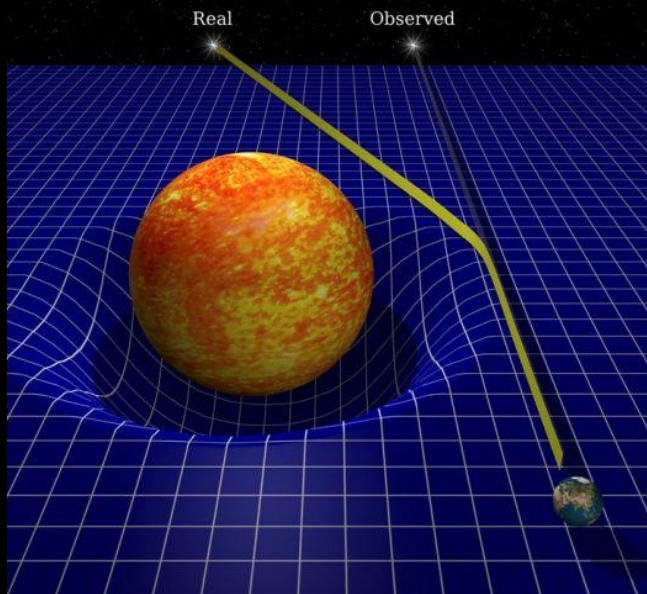


Zhigang Li



**Large Scale
Structure**

Gravitational Lensing



A Schematic Outline of the Cosmic History

Time since the Big Bang (years)

~ 300 thousand

~ 500 million

~ 1 billion

~ 9 billion

~ 13 billion



←The Big Bang

The Universe filled with ionized gas

←The Universe becomes neutral and opaque

The Dark Ages start

Galaxies and Quasars begin to form
The Reionization starts

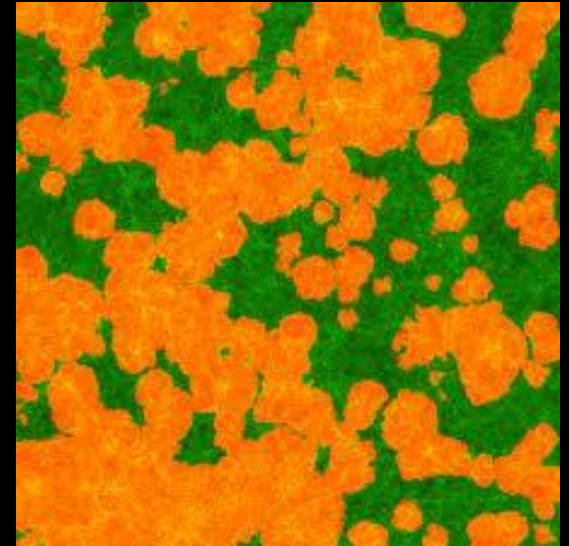
The Cosmic Renaissance
The Dark Ages end

←Reionization complete, the Universe becomes transparent again

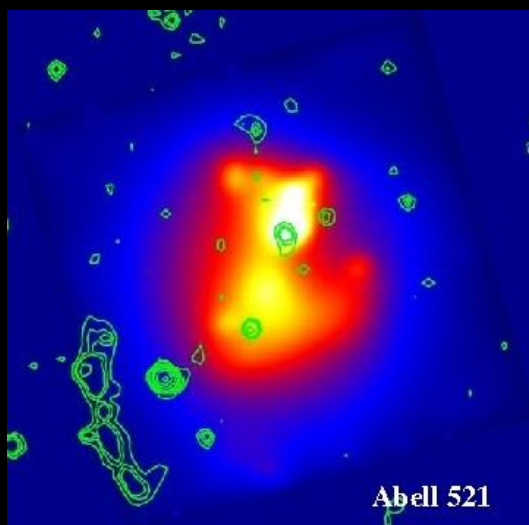
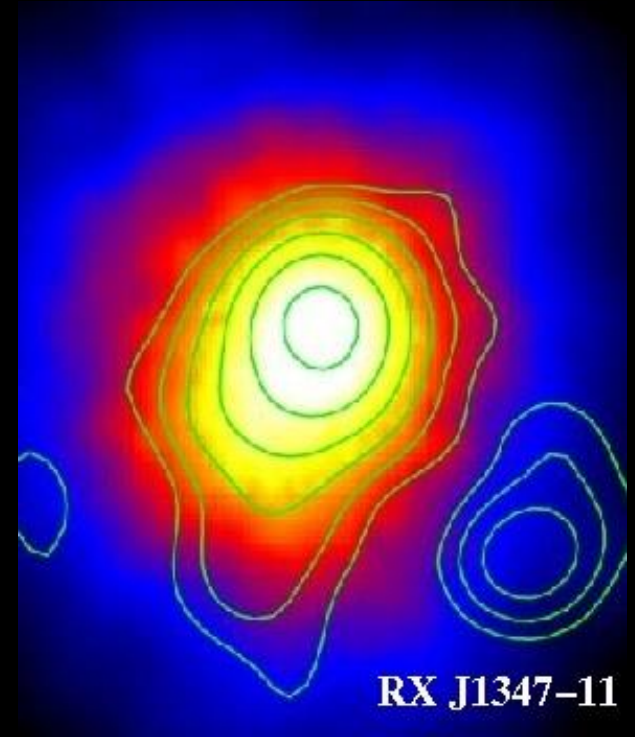
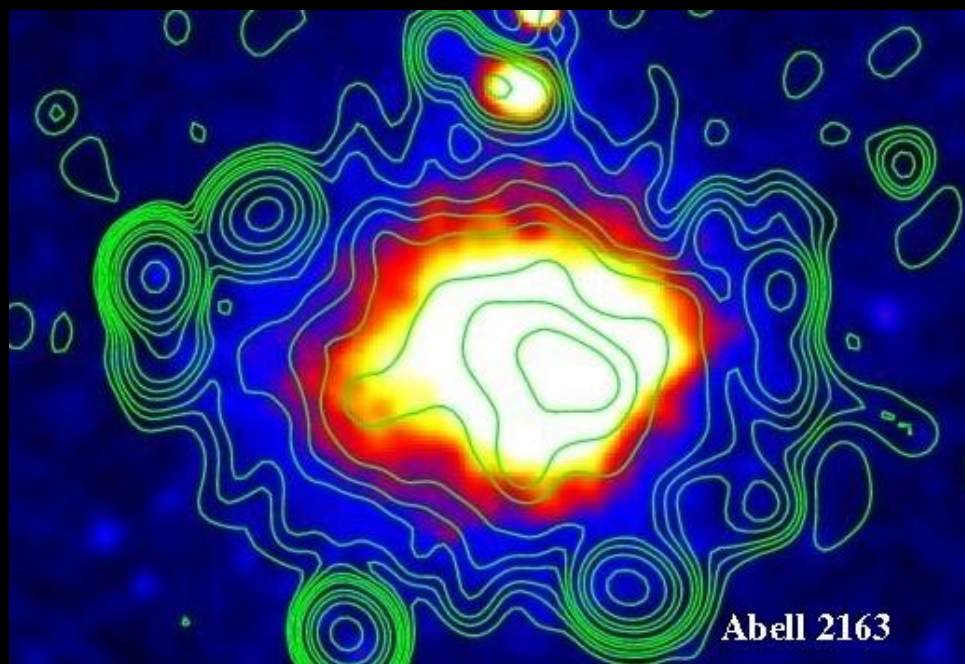
Galaxies evolve

The Solar System forms

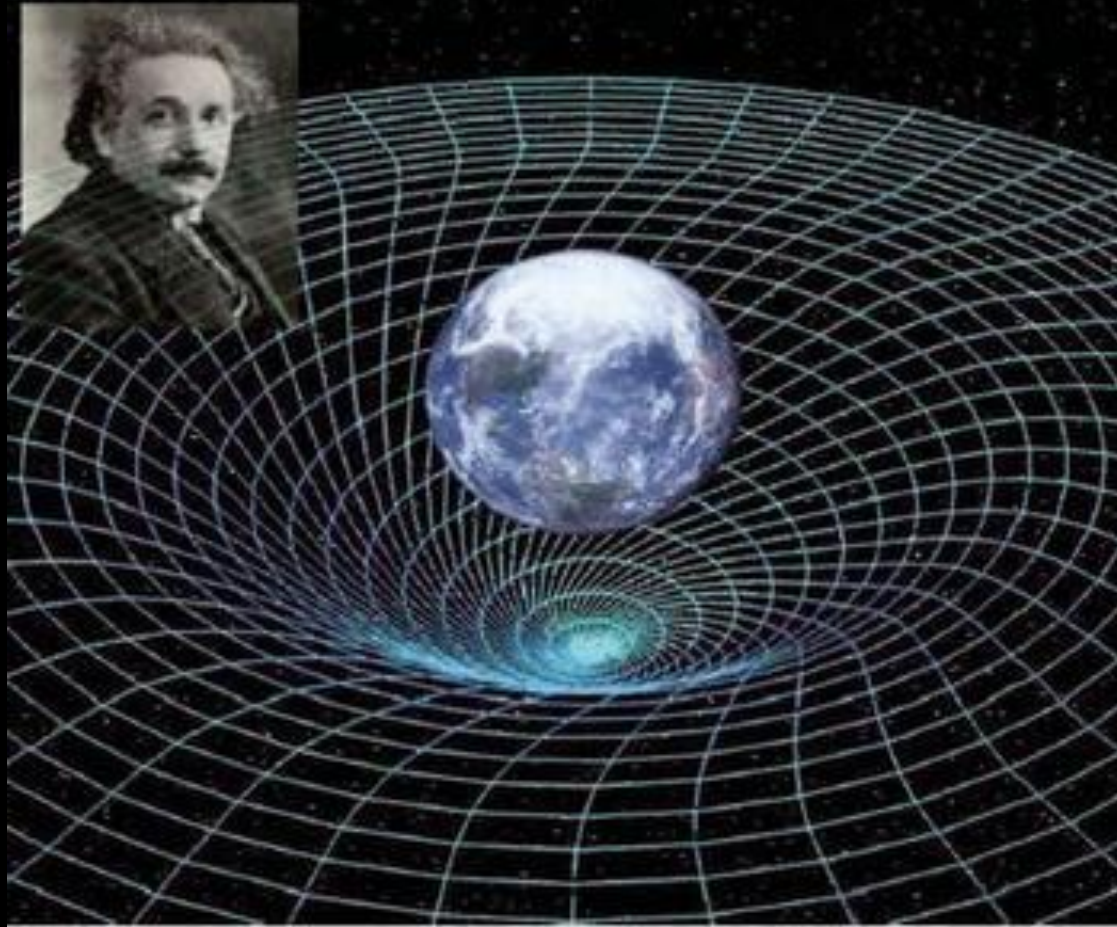
Today: Astronomers figure it all out!



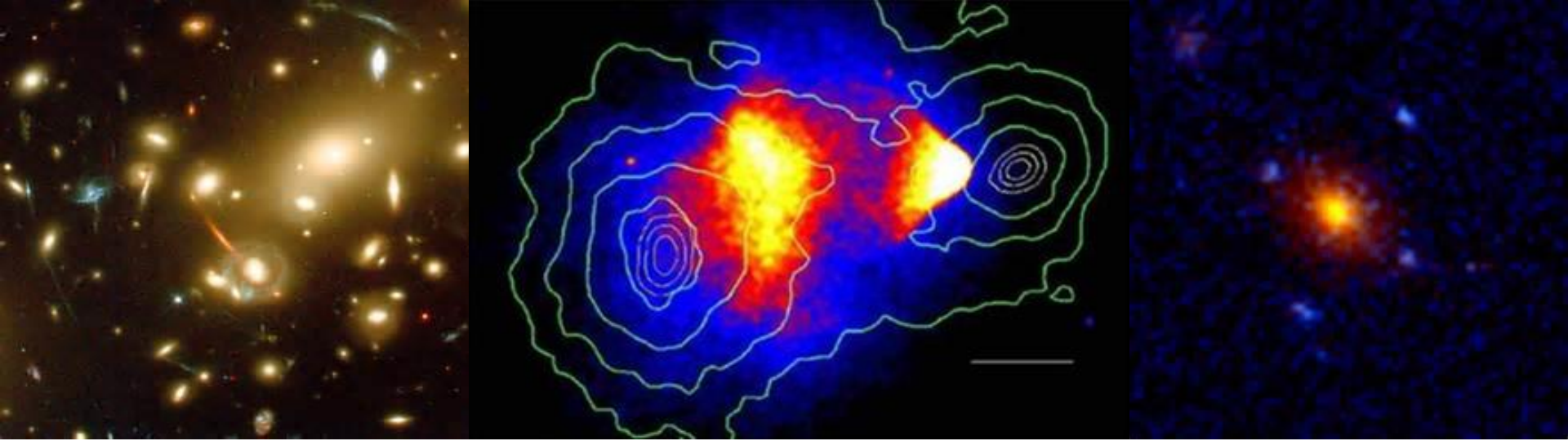
Cosmic Reionization



Cluster Physics



Gravity Theories



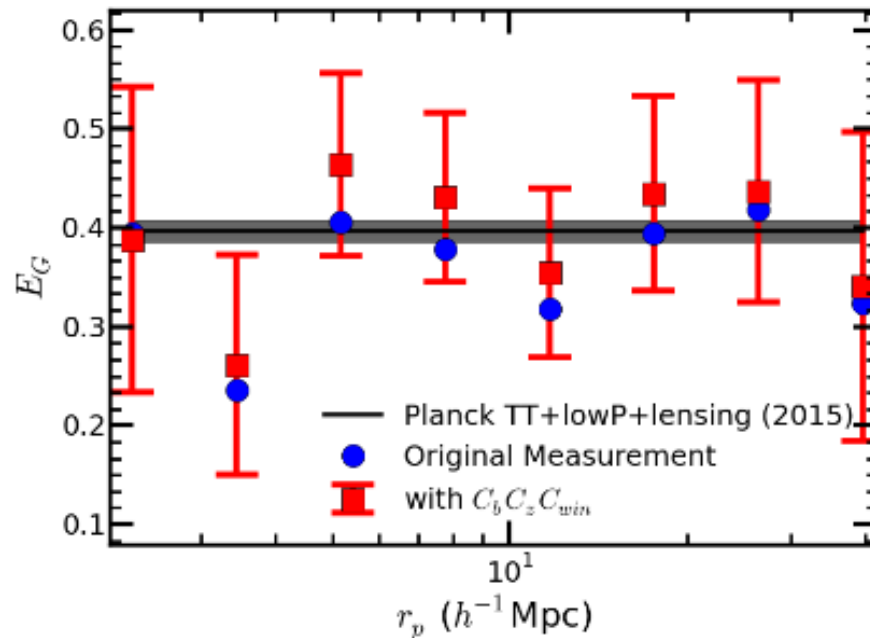
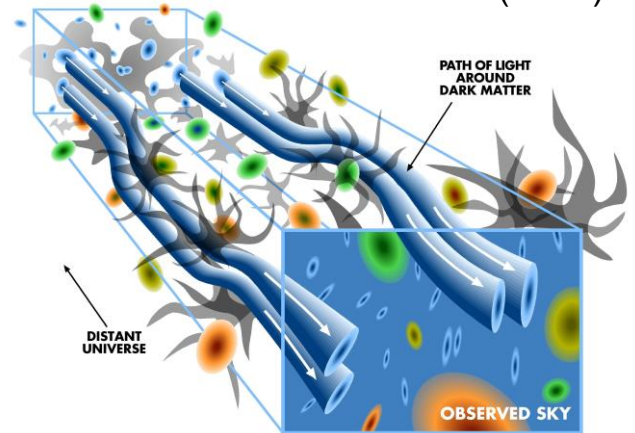
According to GR , lights from distant sources are deflected by gravity, causing the lensing effect. This visible effect can be used to probe the cosmic structures on large scales, leading to clues for possible new physics !!



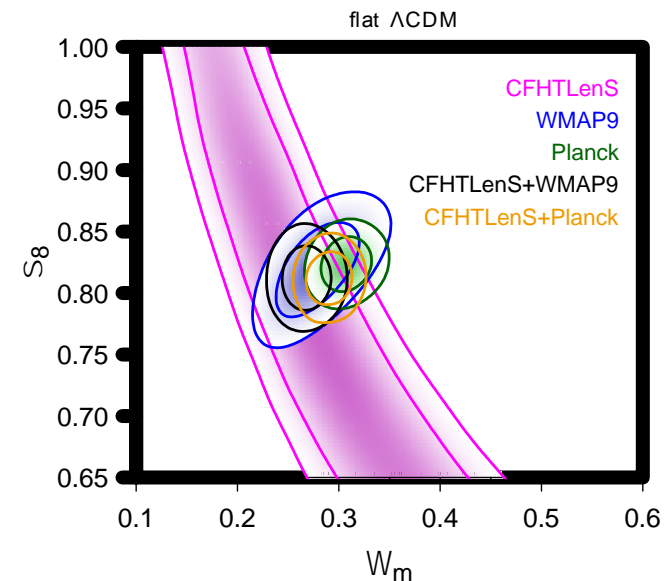
Application of Weak Lensing

- probing the **density growth history**
- probing the **distance-redshift relation**
- probing the nature of **dark matter/energy**
- **Testing the theory of GR on cosmic scales**

Image credit:
Wittman et al. (2000)



Zhang et al. (2017)



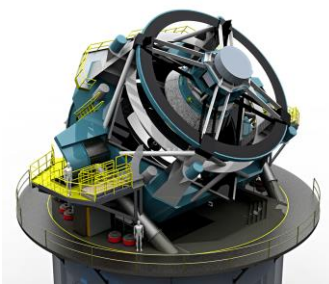
Credit: Fu et al. (2014)



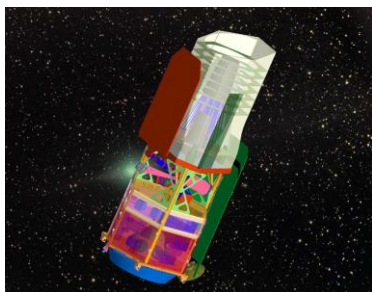
DES



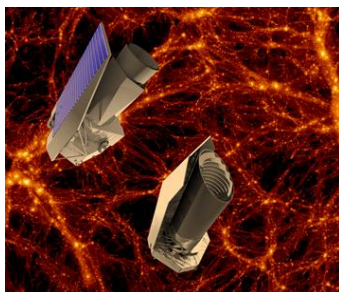
Subaru HSC



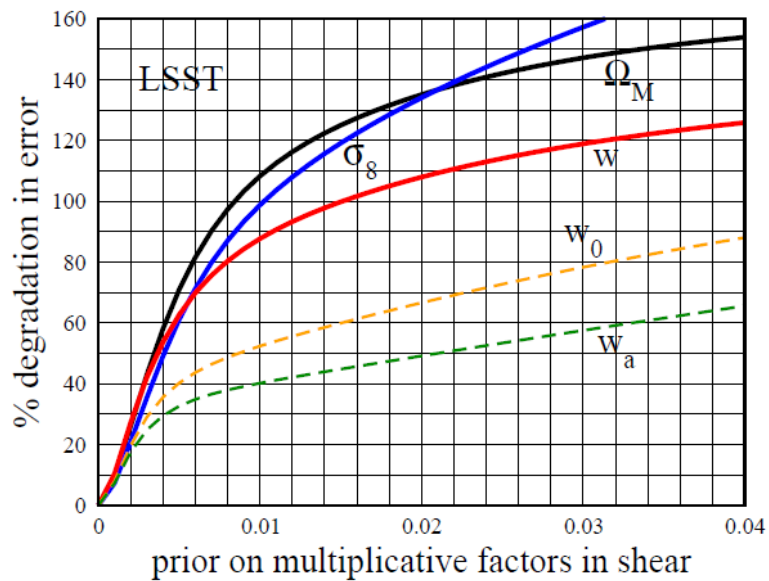
LSST



WFIRST



EUCLID



Huterer et al. (2006)



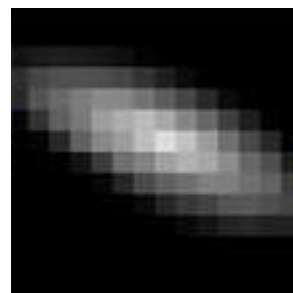
Lensing



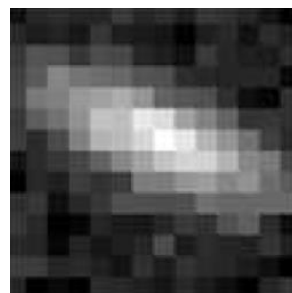
PSF



Pixelation



Noise



Shear TESTING Program I

Author	Key	Method
Bridle & Hudelot	SB	Im2shape (Bridle et al. 2001)
Brown	MB	KSB+ [Bacon et al. (2000) pipeline]
Clowe	C1 & C2	KSB+
Dahle	HD	K2K (Kaiser 2000)
Hetterscheidt	MH	KSB+ [Erben et al. (2001) pipeline]
Heymans	CH	KSB+
Hoekstra	HH	KSB+
Jarvis	MJ	Bernstein & Jarvis (2002) Rounding kernel method
Kuijken	KK	Shapelets to 12 th order Kuijken (2006)
Margoniner	VM	Wittman et al. (2001)
Nakajima	RN	Bernstein & Jarvis (2002) Deconvolution fitting method
Schrabback	TS	KSB+ [Erben et al. (2001) + modifications]
Van Waerbeke	LV	KSB+

Heymans et al., 2006, MNRAS, 368, 1323

Massey et al., 2007, MNRAS, 376, 13

Bridle et al., 2009, Annals of Applied Statistics, 3, 6

Kitching et al., 2011, Annals of Applied Statistics, 5, 2231

Mandelbaum et al., 2014, ApJS, 212, 5

Shear TESTING Program II

Author	Key	Method
Bergé	JB	Shapelets (Massey & Refregier 2005)
Clowe	C1	KSB+ (same PSF model used for all galaxies)
Clowe	C2	KSB+ (PSF weight size matched to galaxies')
Hetterscheidt	MH	KSB+
Hoekstra	HH	KSB+
Jarvis	MJ	Bernstein & Jarvis (2002)
Jarvis	MJ2	Bernstein & Jarvis (2002) (new weighting scheme)
Kuijken	KK	Shapelets (Kuijken 2006)
Mandelbaum	RM	Reglens (Hirata & Seljak 2003)
Nakajima	RN	Bernstein & Jarvis (2002) (deconvolution fitting)
Paulin-Henriksson	SP	KSB+
Schirmer	MS1	KSB+ (scalar shear susceptibility)
Schirmer	MS2	KSB+ (tensor shear susceptibility)
Schrabback	TS	KSB+
Semboloni	ES1	KSB+ (shear susceptibility fitted from population)
Semboloni	ES2	KSB+ (shear susceptibility for individual galaxies)



KSB+ Method:

$$Q_{ij}(x^o, y^o) = \frac{\int d^2\vec{\theta} \cdot W(\vec{\theta}) f^o(\vec{\theta}) \theta_i \theta_j}{\int d^2\vec{\theta} \cdot W(\vec{\theta}) f^o(\vec{\theta})}$$

$$\begin{pmatrix} \varepsilon_1 \\ \varepsilon_2 \end{pmatrix} = \frac{1}{Q_{11} + Q_{22}} \begin{pmatrix} Q_{11} - Q_{22} \\ 2Q_{12} \end{pmatrix}$$

$$\Gamma_\alpha = \left(P^\gamma \right)_{\alpha\beta}^{-1} \left[\varepsilon_\beta - P_{\beta\mu}^{sm} p_\mu \right]$$

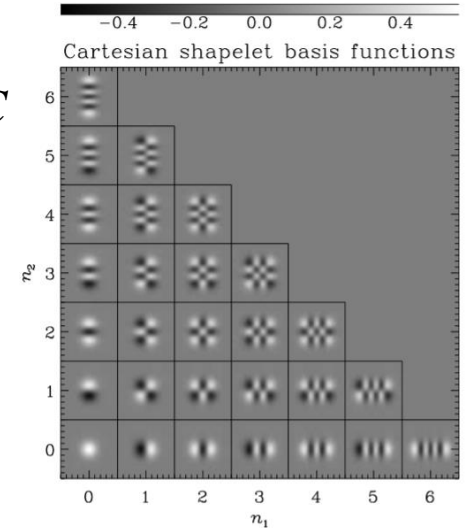
Heymans et al. (2006)

Shapelets Method:

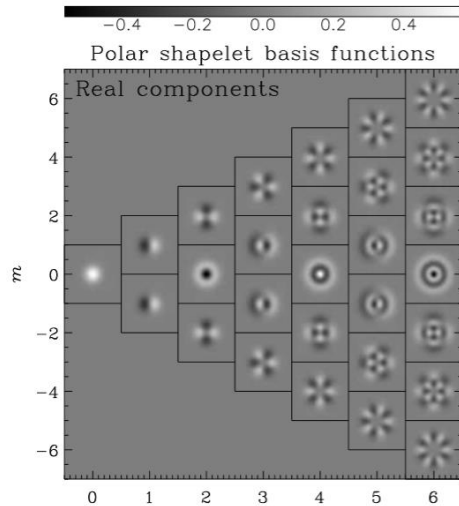
$$G_{model} = P \cdot (1 + \varepsilon_1 S_1 + \varepsilon_2 S_2) \cdot C$$

$$\gamma_i = \frac{\langle \varepsilon_i \rangle}{1 - \langle \varepsilon_1^2 + \varepsilon_2^2 \rangle}$$

Kuijken (2006)



BJ02 Method:



Bernstein & Jarvis (2002)

Bayesian Galaxy Shape Measurement Method: (Lensfit)

$$\chi^2 = \sum_i \left[\frac{y_i - SBg_i - S(1-B)f_i}{\sigma_i} \right]^2$$

Miller et al. (2013)

Challenges & Opportunities

Lensing is Low: Cosmology, Galaxy Formation, or New Physics?

Alexie Leauthaud^{1,2}, Shun Saito³, Stefan Hilbert^{4,5}, Alexandre B...
 Martin White⁶, Shadab Alam^{7,8}, Peter Behroozi^{6,9}, Kevin Bundy...
 Thomas Erben¹¹, Catherine Heymans⁸, Hendrik Hildebrandt¹¹, R...
 Lance Miller¹², Bruno Moraes¹³, Maria E. S. Pereira¹⁴, Sergio A...
 Fabian Schmidt³, Huan-Yuan Shan¹⁸, Matteo Viel^{19,20}, Francisco

¹Department of Astronomy and Astrophysics, University of California, Santa Cruz, 1156 High Street, San

²Kavli IPMU (WPI), UTIAS, The University of Tokyo, Kashiwa, Chiba, 277-8583, Japan

³Max-Planck-Institut für Astrophysik, Karl-Schwarzschild-Strasse 1, D-85740 Garching bei München, Ger

⁴Exzellenzcluster Universe, Boltzmannstr. 2, 85748 Garching, Germany

⁵Ludwig-Maximilians-Universität, Universitäts-Sternwarte, Scheinerstr. 1, 81679 München, Germany

⁶Department of Physics, University of California, Berkeley, CA 94720

⁷McWilliams

⁸The Scottish

⁹Hubble Fell

Problems with KiDS

George Efstathiou and Pablo Lemos

Kavli Institute for Cosmology Cambridge and Institute of Astronomy, Madingley Road,

4 July 2017

ABSTRACT

The Kilo-Degree Survey (KiDS) has been used to provide tight constraints on the amplitude of the matter power spectrum at low redshift. Some of these analyses have claimed a $\sim 2 - 3\sigma$ tension with the Planck cosmology at the $\sim 2 - 3\sigma$ level, perhaps indicating a departure from Λ CDM. This is consistent with other low redshift probes of galaxy clustering, redshift space distortions and the combined lensing and galaxy clustering spectra. Here we perform consistency tests of the KiDS results for various cuts of the data at $\gtrsim 3\sigma$ significance. If understood, we argue that it is premature to conclude that there are problems with KiDS.

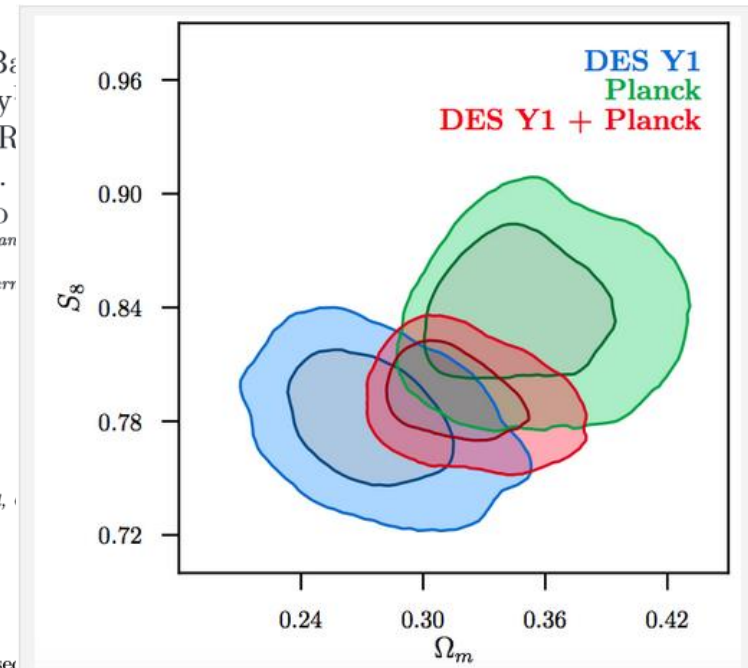


Figure 4. A plot of fluctuation scale S_8 (a robust functional form of σ_8) vs. matter density Ω_m from the DES Cosmology results – Figure 10 of the paper. Assumed here is a likelihood model with Λ CDM ($w=-1$), with darker contours depicting 68% confidence level intervals. DES Y1 results (weak lensing, clustering, blue) are a 'slight' departure from the Planck results (CMB, no lensing, green), but with combined datasets one can see stronger constraints on the cosmological parameters.

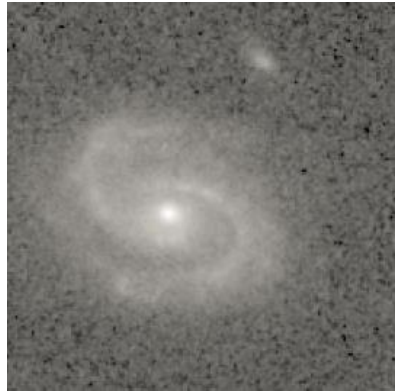
2017 Nov 25

2017 Jul 3

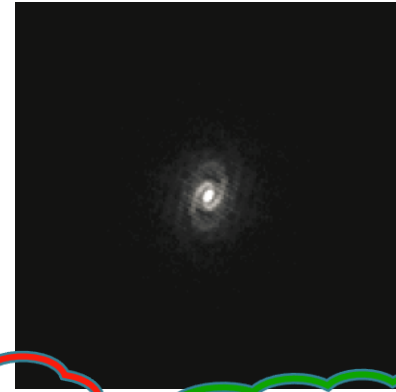
Outline:

- Introduction & Motivation
- **The Fourier_Quad method**
- Application on the CFHTlens data

The Fourier_Quad Method



$$\tilde{f}(\vec{k}) = \int d^2\vec{x} e^{i\vec{k}\cdot\vec{x}} f(\vec{x}).$$



Galaxy

Noise

$$G_1 = -\frac{C}{2} \sum_{j=1}^{N_T} [(\vec{k}_j)_x^2 - (\vec{k}_j)_y^2] T(\vec{k}_j) M(\vec{k}_j)$$

$$G_2 = -C \sum_{j=1}^{N_T} (\vec{k}_j)_x (\vec{k}_j)_y T(\vec{k}_j) M(\vec{k}_j)$$

$$N = C \sum_{j=1}^{N_T} \left[|\vec{k}_j|^2 - \frac{\beta^2}{2} |\vec{k}_j|^4 \right] T(\vec{k}_j) M(\vec{k}_j)$$

$$M(\vec{k}) = \frac{|\tilde{f}^O(\vec{k})|^2}{F^O} - \frac{|\tilde{f}^B(\vec{k})|^2}{F^B}$$

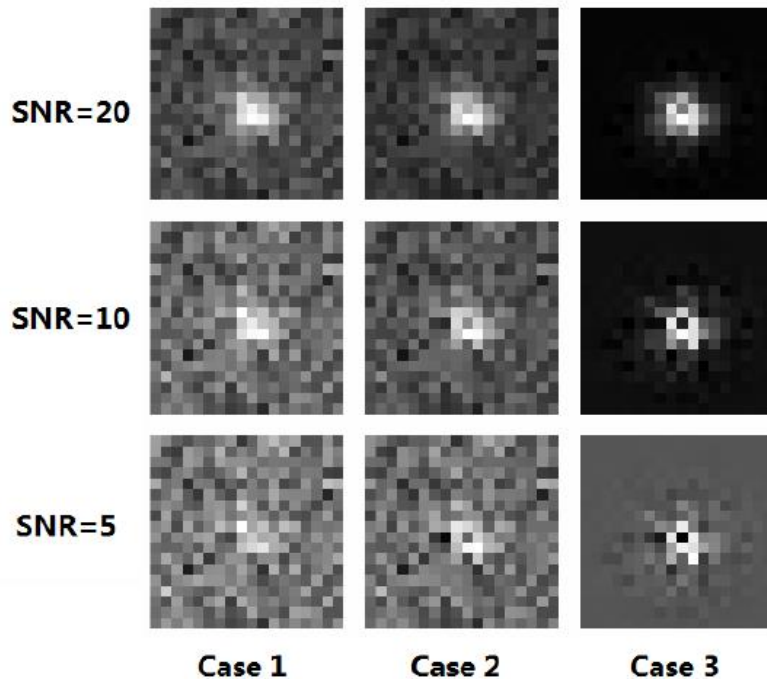
$$F^O = \frac{\sum_{|\vec{k}_j| > k_c} |\tilde{f}^O(\vec{k}_j)|^2}{\sum_{|\vec{k}_j| > k_c} 1}, \quad F^B = \frac{\sum_{|\vec{k}_j| > k_c} |\tilde{f}^B(\vec{k}_j)|^2}{\sum_{|\vec{k}_j| > k_c} 1}$$

$$\frac{\langle G_1 \rangle}{\langle N \rangle} = g_1, \quad \frac{\langle G_2 \rangle}{\langle N \rangle} = g_2$$

Test Results

$$g_1^{measured} = (1 + m_1)g_1^{input} + c_1$$

$$g_2^{measured} = (1 + m_2)g_2^{input} + c_2$$

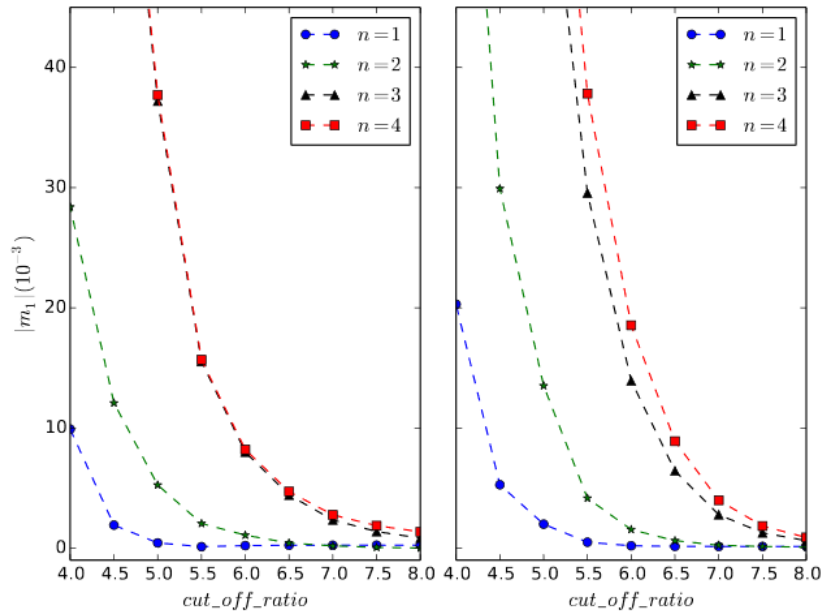


	Case 1	Case 2	Case 3
SNR= 20	$m_1(10^{-3}) : -0.2 \pm 0.2$	-0.2 ± 0.1	-0.08 ± 0.05
	$c_1(10^{-5}) : -1.0 \pm 0.7$	-0.6 ± 0.4	0.0 ± 0.2
SNR= 10	-0.6 ± 0.6	-0.4 ± 0.4	-0.1 ± 0.1
	-2.6 ± 2.1	-1.6 ± 1.2	0.0 ± 0.4
SNR= 5	-2.2 ± 2.3	-1.1 ± 1.2	0.0 ± 0.2
	-7.9 ± 7.9	-4.6 ± 4.1	0.1 ± 0.8
SNR= 20	$m_2(10^{-3}) : -0.2 \pm 0.2$	-0.3 ± 0.1	-0.3 ± 0.05
	$c_2(10^{-5}) : 1.4 \pm 0.7$	1.1 ± 0.4	0.3 ± 0.2
SNR= 10	-0.3 ± 0.6	-0.4 ± 0.4	-0.4 ± 0.1
	3.7 ± 2.1	2.6 ± 1.2	0.6 ± 0.4
SNR= 5	-0.7 ± 2.3	-0.9 ± 1.2	-0.6 ± 0.2
	10.6 ± 7.9	6.6 ± 4.1	1.2 ± 0.8

Team	Class	Weighting scheme	Calibration philosophy	Limitations	N_{branch}	Rank	Exact PSF?	New software	Time per galaxy
Amalgam@IAP	Maximum likelihood	Inverse variance	Ellipticity penalty	None	16	2	Yes	Some	0.1–1 s
BAMPenn	Bayesian Fourier	Implicit	$p(\epsilon)$ from deep data	Variable shear	2	-	Yes	Yes	< 1 s
EPFL_gfit	Maximum likelihood	Constant + rejection	None	None	8	6	Yes	Yes	1–3 s
CEA-EPFL	Maximum likelihood	Various	None	None	20	3	Yes	Yes	1–3 s
CEA_denoise	Moments	Constant	None	None	8	-	Yes	No	0.03 s
CMU experimenters	Stacking	Constant	External simulations	Variable shear	2	N/A	Yes	Some	0.03 s
COGS (IM3SHAPE)	Maximum likelihood	Constant	External simulations	None	12	N/A	Yes	Yes	1 s
E-HOLICS	Moments	Constant + rejection	External simulations	None	12	8	Yes	No	1–3 s
EPFL_HNN	Neural network	Constant	None	None	7	-	Yes	Yes	2–3 s
EPFL_KSB	Moments	Inverse variance	None	None	4	-	Yes	No	0.001–0.002 s
EPFL_MLP / EPFL_MLP_FIT	Neural network	Constant	None	None	5	-	Yes	Yes	2–3 s
FDNT	Fourier moments	Inverse variance	External simulations	None	12	N/A	Yes	Some	~ 1 s
Fourier_Quad	Fourier moments	Various	None	None	6	5	Yes	No	0.001–0.002 s
HSC/LSST-HSM	Moments	Inverse variance	External simulations	None	4	N/A	Yes	Some	0.05 s
MBI	Bayesian hierarchical	Implicit	Inferred $p(\epsilon)$	Variable shear, PSF	4	9	No	Some	10 s
MaltaOx (LENSFIT)	Partially Bayesian	Inverse variance	Self-calibration	None	3	7	Yes	Some	0.05 s
MegaLUT	Supervised ML	Constant + rejection	External simulations	None	16	4	Yes	Some	0.02 s
MetaCalibration	Moments + self-calibration	Inverse variance	Self-calibration	Variable shear	1	N/A	Yes	Yes	0.3 s
Wentao_Luo	Moments	Inverse variance	None	None	4	-	Yes	Yes	1–2 s
ess	Bayesian model-fitting	Implicit	$p(\epsilon)$ from deep data	Variable shear	2	-	No	Yes	1 s
sFIT	Maximum likelihood	Inverse variance	External simulations (iterative)	None	20	1	Yes	Yes	0.8 s

Some Recent Progress

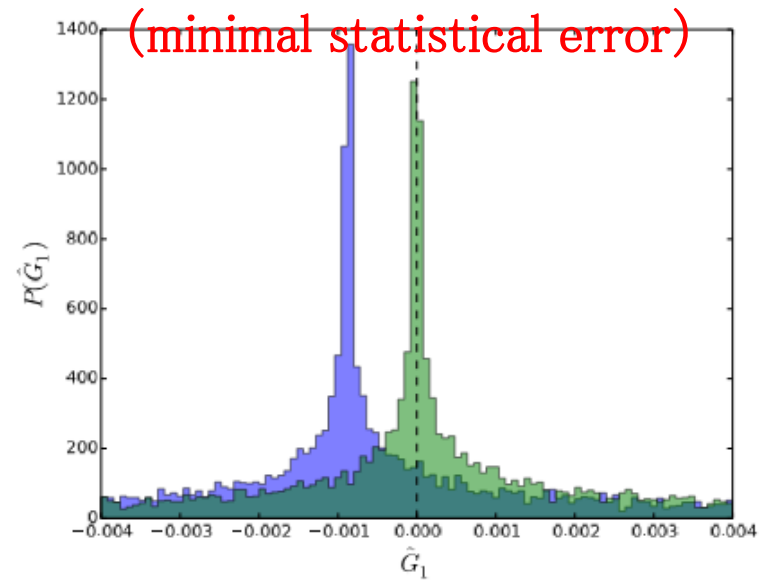
About Image Size



$$f(\vec{x}) = \frac{1}{a(n)r_e^2} e^{-b(n)(|\vec{x}|/r_e)^{\frac{1}{n}}}$$

Li & Zhang, 2016, ApJ, 830, 116

New Statistics Approaching the C-R Limit



Method	$g_1(-0.018)$	$g_2(0.011)$
Direct Averaging	-0.017(2)	0.011(2)
PDF-SYM (2 bins)	-0.01798(3)	0.01100(2)
PDF-SYM (4 bins)	-0.01799(2)	0.01102(2)
PDF-SYM (8 bins)	-0.01800(2)	0.01101(2)

JZ, Zhang, Luo, 2017, ApJ, 834, 8

APPROACHING THE CRAMER-RAO BOUND IN WEAK LENSING WITH PDF SYMMETRIZATION

C-R BOUND

$$0 = \frac{d}{d\hat{g}} \sum_i \ln P(x_i - \hat{g})$$

$$\sigma_{\hat{g}}^{-2} = - \sum_i \frac{\partial^2 \ln P(x_i - \hat{g})}{\partial \hat{g}^2}$$

Examples

$$P_1(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right)$$

$$P_2(x) = \frac{2}{\pi} (1 + x^2)^{-2}$$

$$P_3(x) = \frac{|x|^{-2/3}}{3\sqrt{2\pi}} \exp\left(-\frac{|x|^{2/3}}{2}\right)$$

$$N_T \sigma_1^2(\text{Ave}) = 1, \quad N_T \sigma_1^2(\text{CR}) = 1,$$

$$N_T \sigma_2^2(\text{Ave}) = 1, \quad N_T \sigma_2^2(\text{CR}) = 0.5,$$

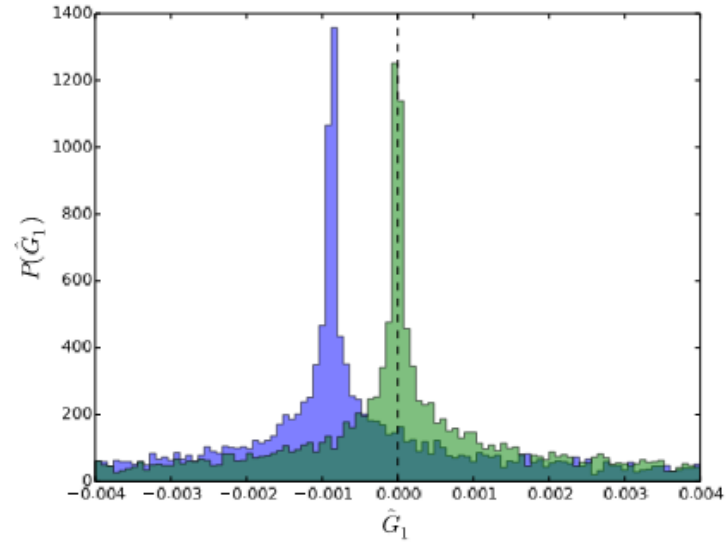
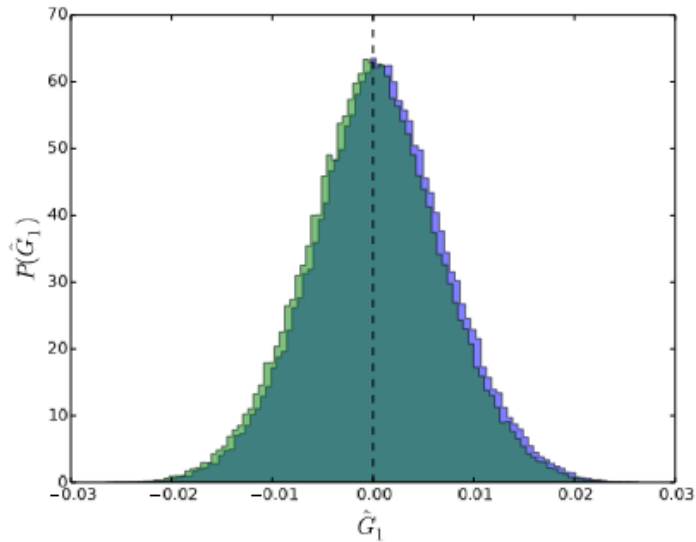
$$N_T \sigma_3^2(\text{Ave}) = 15, \quad N_T \sigma_3^2(\text{CR}) \rightarrow 0.$$

The results of signal recovery (input value is 0.01) for 10^7 data points of three types of PDF's

Results:	Averaging	PDF-SYM (2 bins)	PDF-SYM (8 bins)	PDF-SYM (16 bins)	PDF-SYM (32 bins)
P_1	0.0102(3)	0.0104(4)	0.0101(3)	0.0100(4)	0.0102(3)
P_2	0.0099(3)	0.0101(2)	0.0101(2)	0.0100(2)	0.0101(2)
P_3	0.011(1)	0.0099999998(2)	0.0099999998(1)	0.0099999998(1)	0.0099999999(2)

$N_T \sigma^2$:	Averaging	PDF-SYM (2 bins)	PDF-SYM (8 bins)	PDF-SYM (16 bins)	PDF-SYM (32 bins)
P_1	1.0	1.6	1.1	1.2	0.96
P_2	0.99	0.61	0.52	0.50	0.57
P_3	15	5×10^{-13}	2×10^{-13}	2×10^{-13}	3×10^{-13}

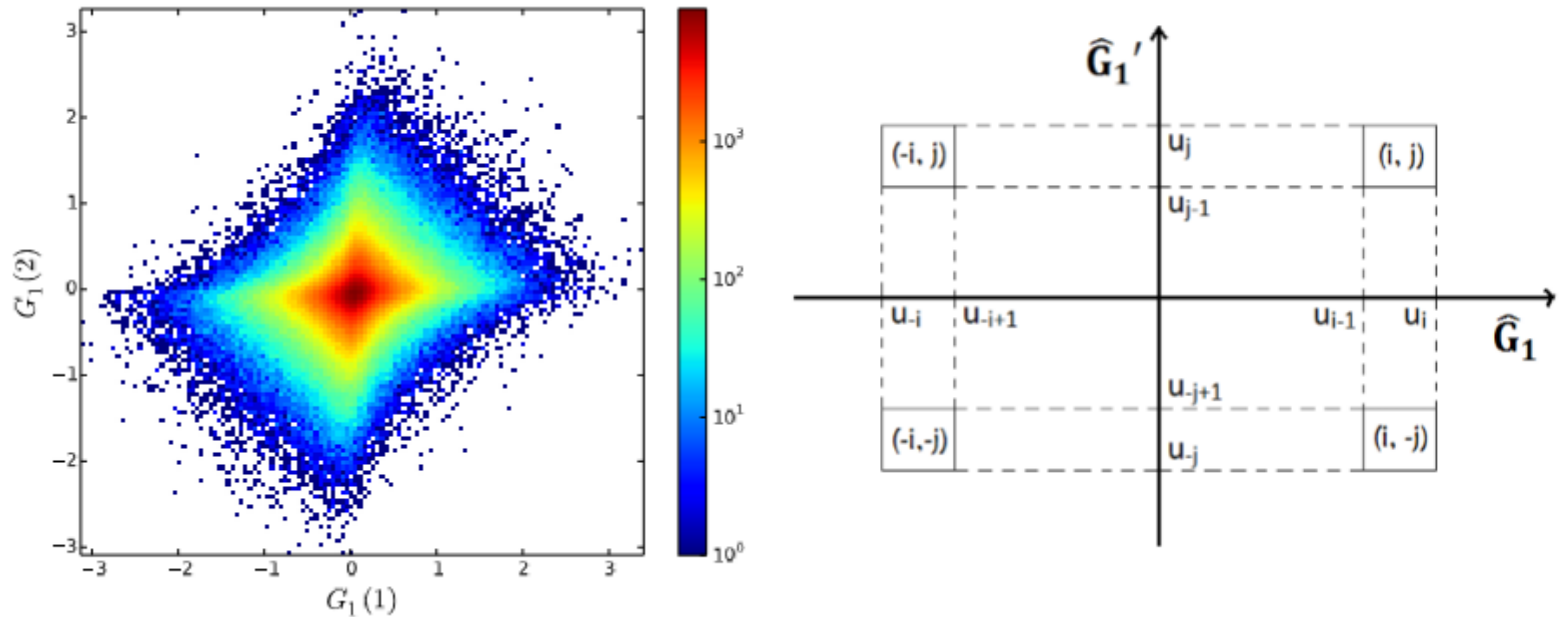
APPROACHING THE CRAMER-RAO BOUND IN WEAK LENSING WITH PDF SYMMETRIZATION



$$g_1 = 0.02277, g_2 = -0.01386$$

Results of $[g_1, g_2]$:	10^5 RW Galaxies	9×10^4 RW + 10^4 Ring
Averaging	$[0.0226(6), -0.0130(6)]$	$[0.0231(6), -0.0132(6)]$
PDF-SYM (8 bins)	$[0.0225(7), -0.0129(6)]$	$[0.02278(8), -0.01392(7)]$

APPROACHING THE CRAMER-RAO BOUND IN WEAK LENSING WITH PDF SYMMETRIZATION



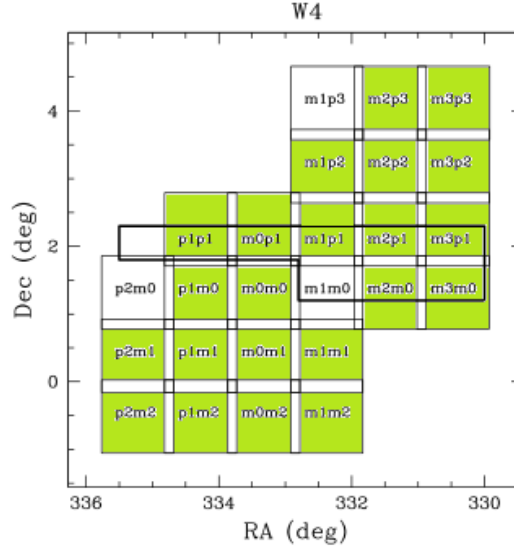
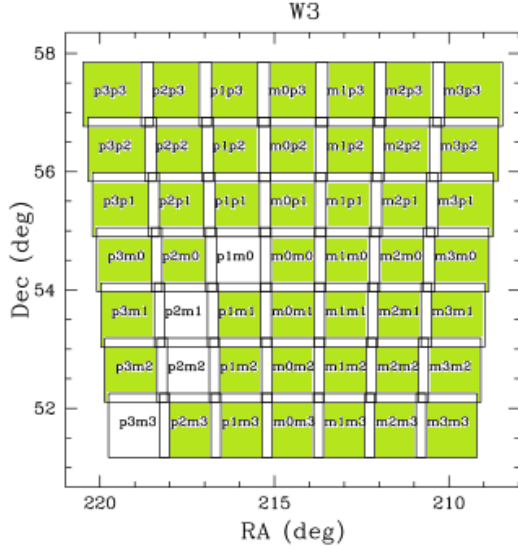
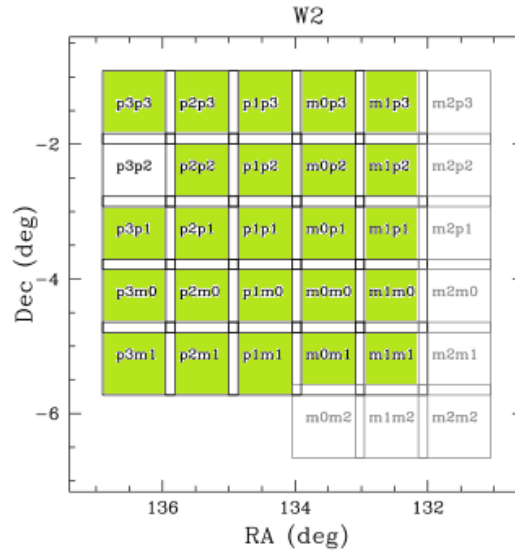
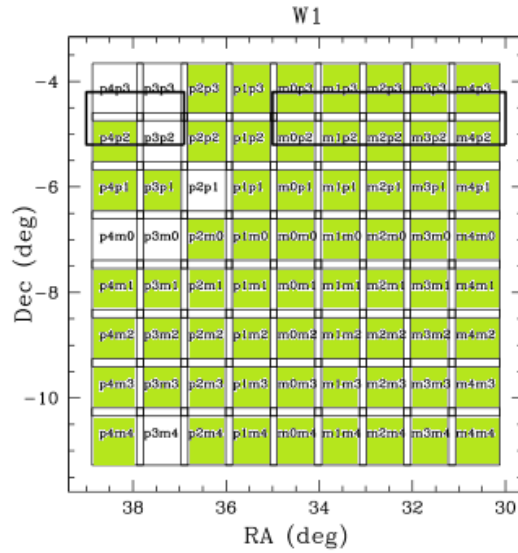
The recovered shear-shear correlations. The inputs are $\langle g_t^{(1)} g_t^{(2)} \rangle = 10^{-4}$ and $\langle g_{\times}^{(1)} g_{\times}^{(2)} \rangle = -10^{-4}$.

Results of $[\langle g_t^{(1)} g_t^{(2)} \rangle, \langle g_{\times}^{(1)} g_{\times}^{(2)} \rangle](10^{-4})$:	Averaging	PDF-SYM (8x8 bins)
4×10^7 RW Gal. Pairs	[1.09(8), -1.00(8)]	[1.09(8), -1.01(9)]
4×10^7 Ring Gal. Pairs	[1.05(7), -1.08(7)]	[1.002(5), -1.002(5)]
4×10^7 Gal. Pairs with 90% RW and 10% Ring	[1.09(8), -1.02(8)]	[0.99(3), -1.00(3)]
1.6×10^8 Ring Gal. Pairs with noise and 10% stars	[0.97(4), -1.05(4)]	[1.000(3), -1.001(3)]

Outline:

- Introduction & Motivation
- The Fourier_Quad method
- Application on the CFHTlens data

Application on the CFHTLenS data



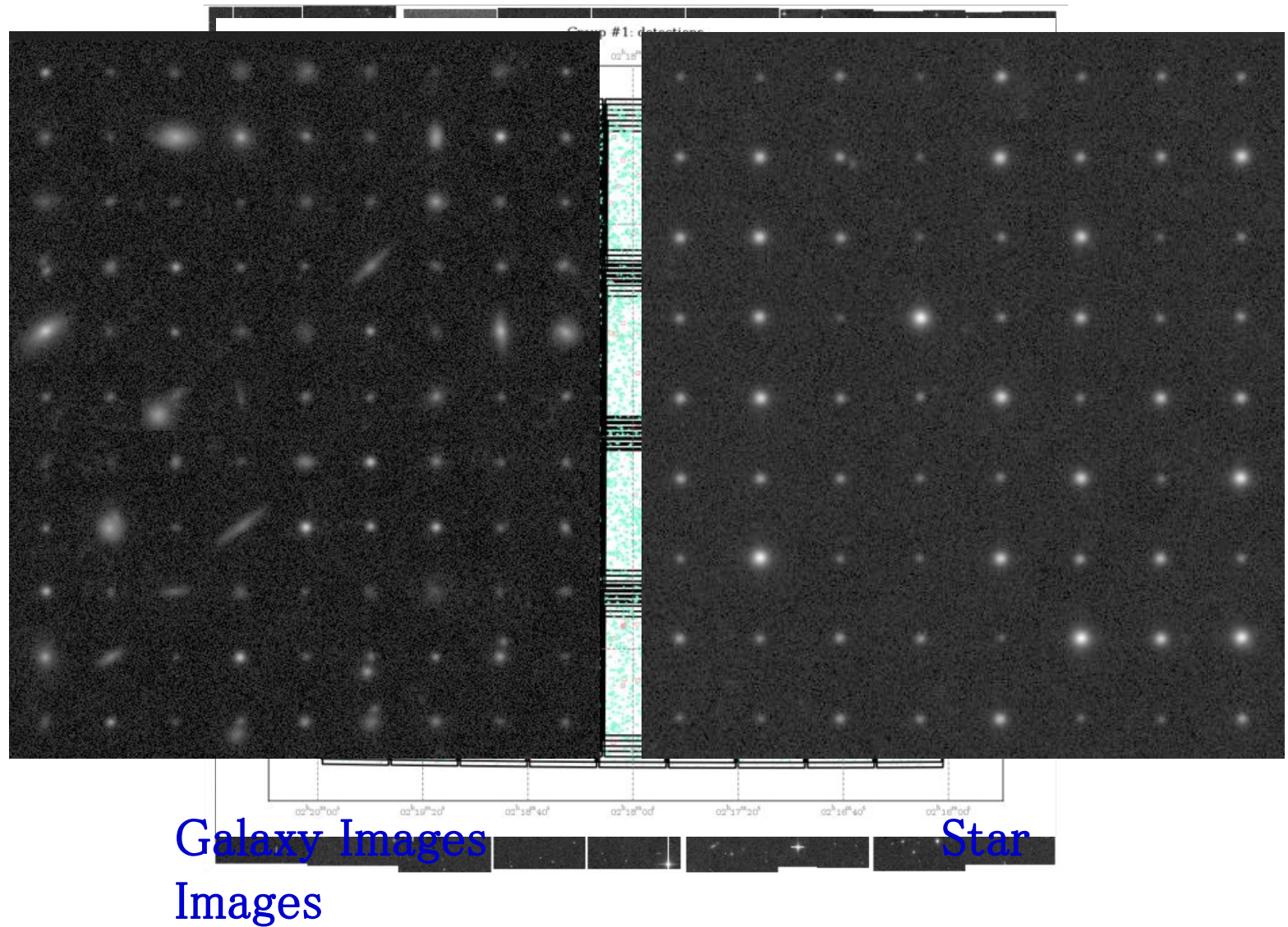
CFHTLenS

Table 1. Characteristics of the final CFHTLenS co-added science data (see the text for an explanation of the columns).

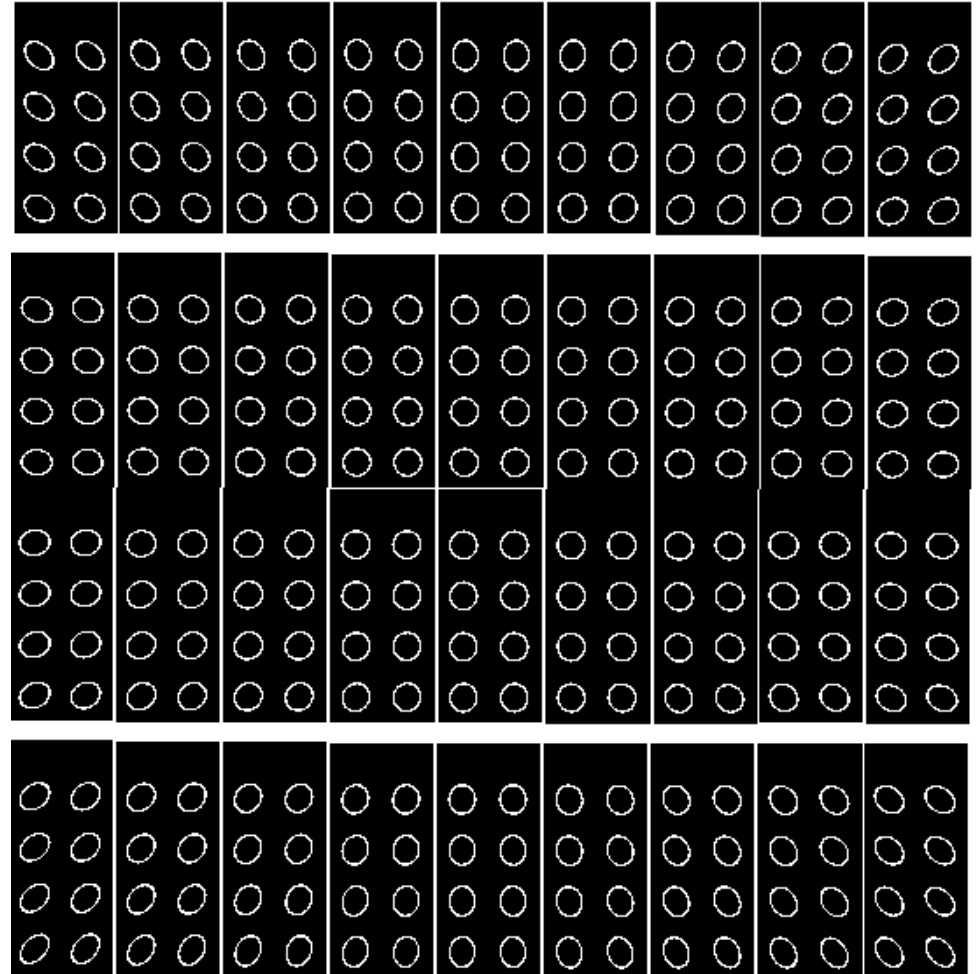
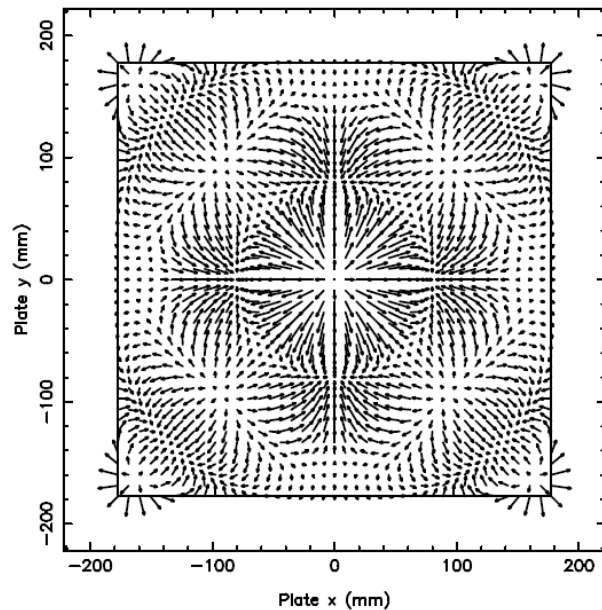
Filter	expos. time [s]	m_{lim} [AB mag] 5- σ lim. mag. in a $2''$ aperture	seeing [$''$]
u^* (<i>u.MP9301</i>)	5×600 (3000)	25.24 ± 0.17	0.88 ± 0.11
g' (<i>g.MP9401</i>)	5×500 (2500)	25.58 ± 0.15	0.82 ± 0.10
r' (<i>r.MP9601</i>)	4×500 (2000)	24.88 ± 0.16	0.72 ± 0.09
i' (<i>i.MP9701</i>)	7×615 (4305)	24.54 ± 0.19	0.68 ± 0.11
y' (<i>y.MP9702</i>)	7×615 (4305)	24.71 ± 0.13	0.62 ± 0.09
z' (<i>z.MP9801</i>)	6×600 (3600)	23.46 ± 0.20	0.70 ± 0.12

- <http://www.cfhtlens.org>
- Erben et al (2012)
- Heymans et al (2012)
- Miller et al (2012)

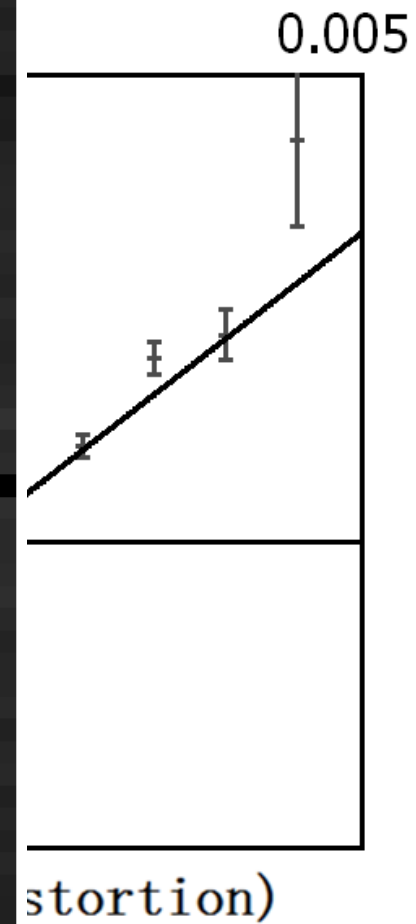
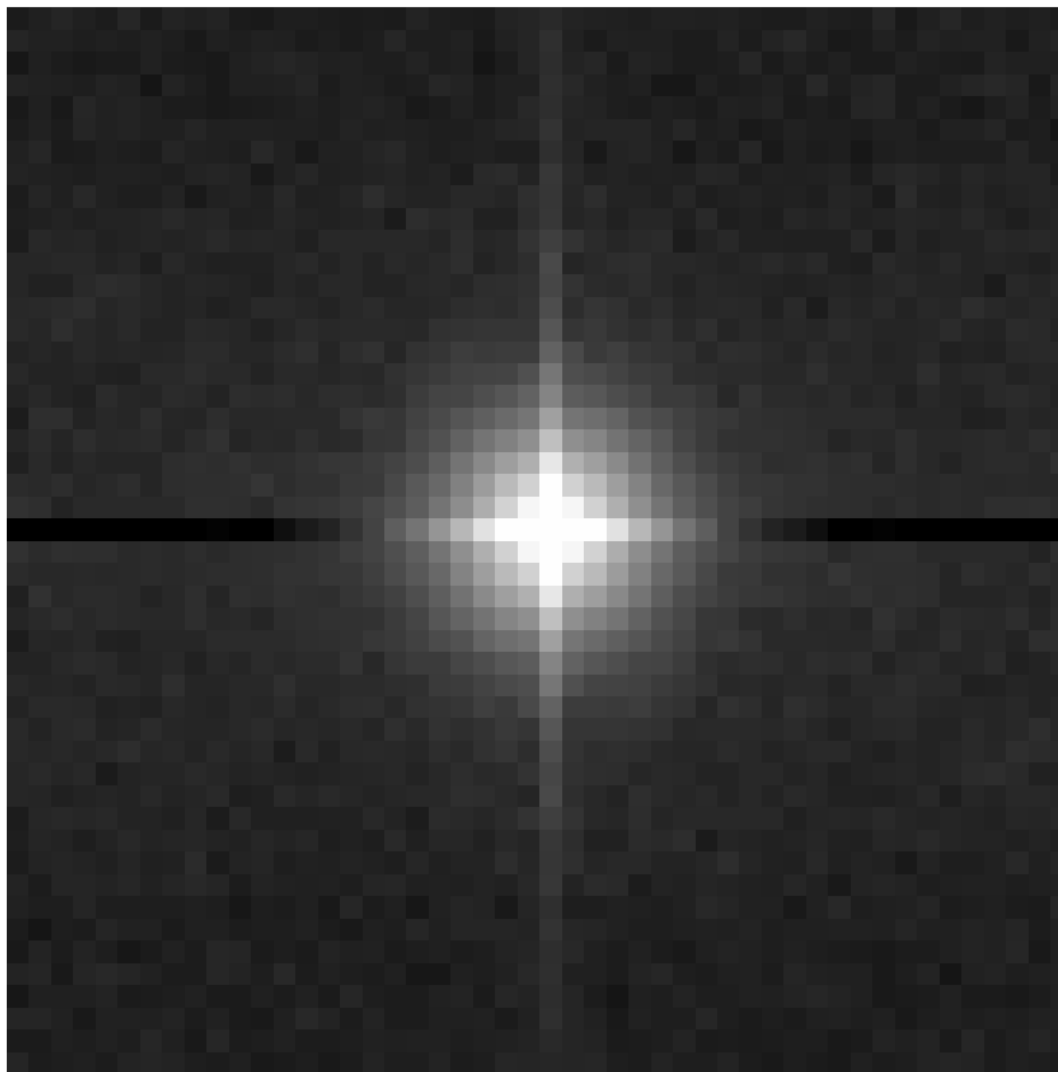
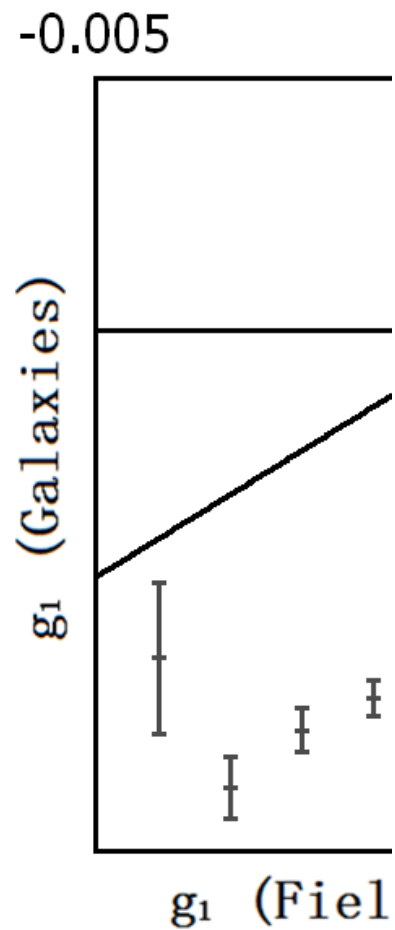
Application on the CFHTlens data



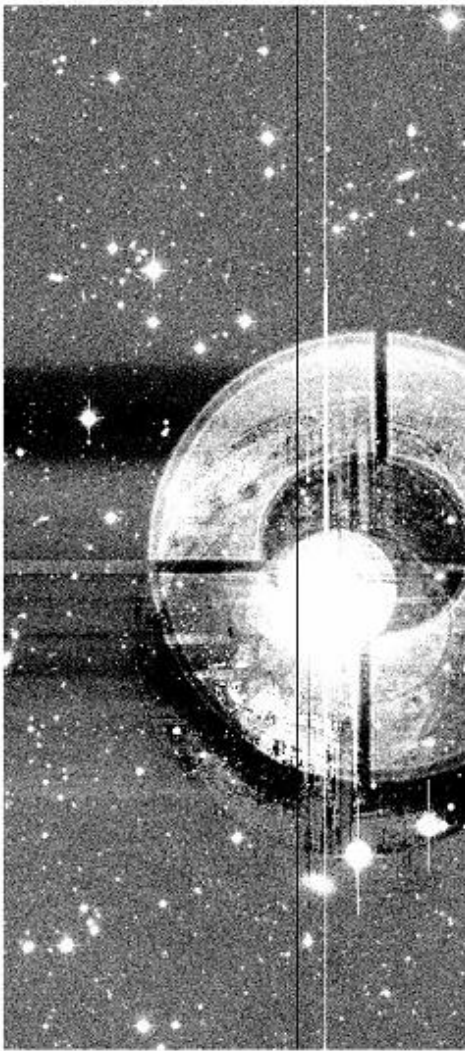
Test with Field Distortion



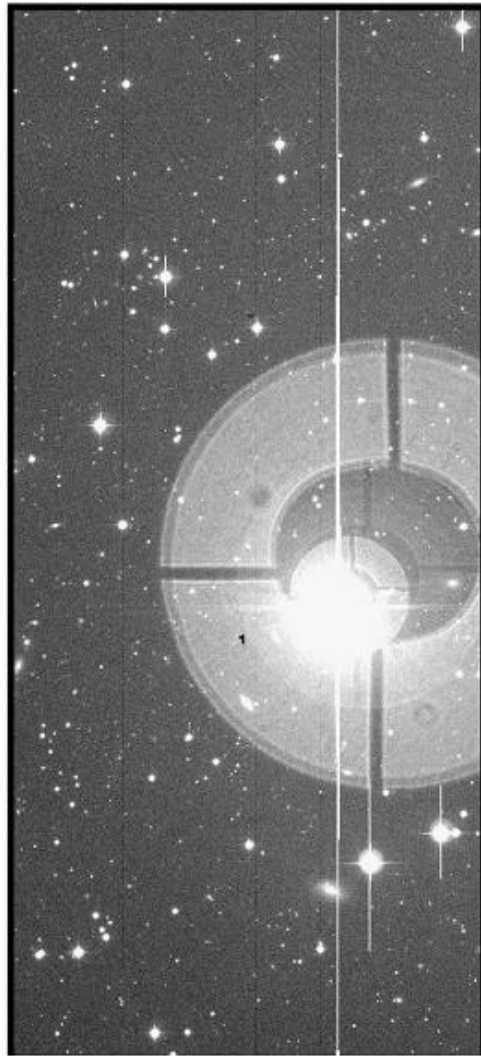
Test with Field Distortion



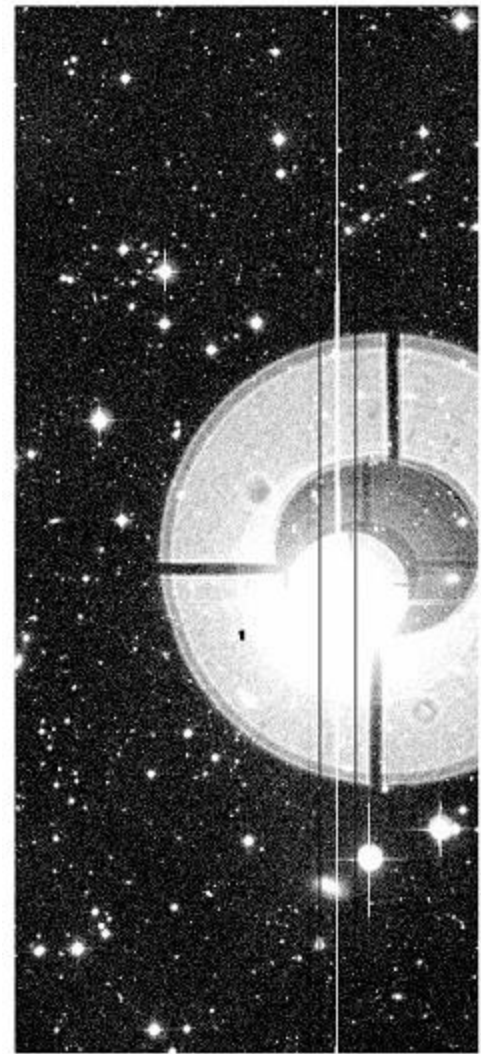
Application on the CFHTlens data



THELI Processed

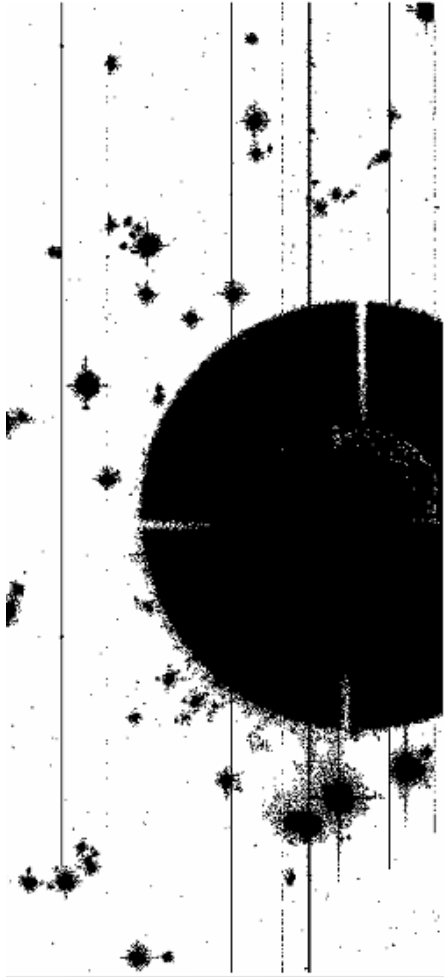


Original Image

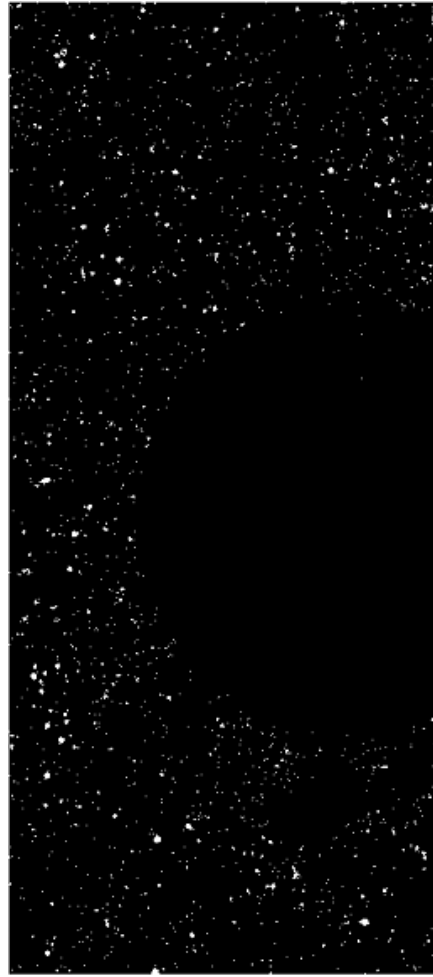


We Processed

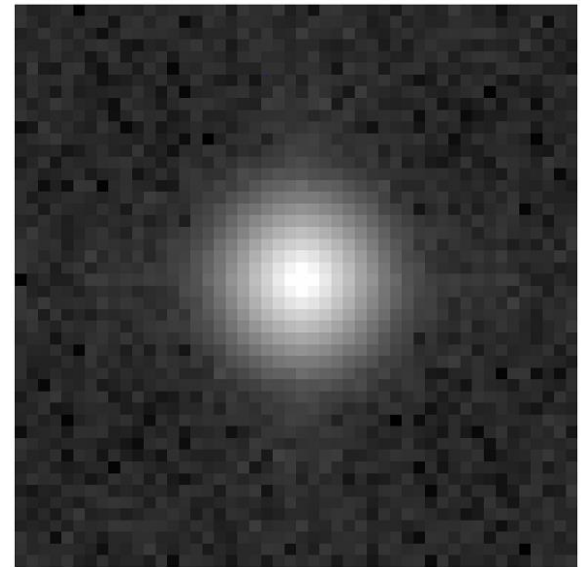
Application on the CFHTlens data



Mask Image

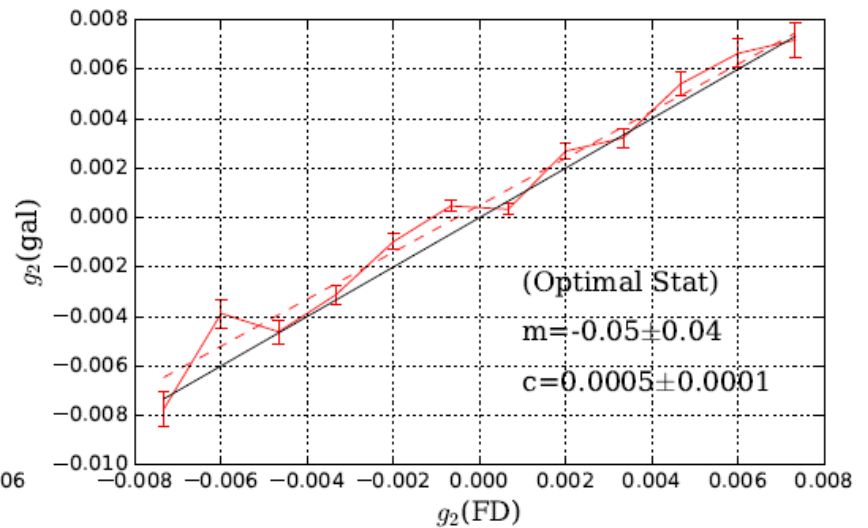
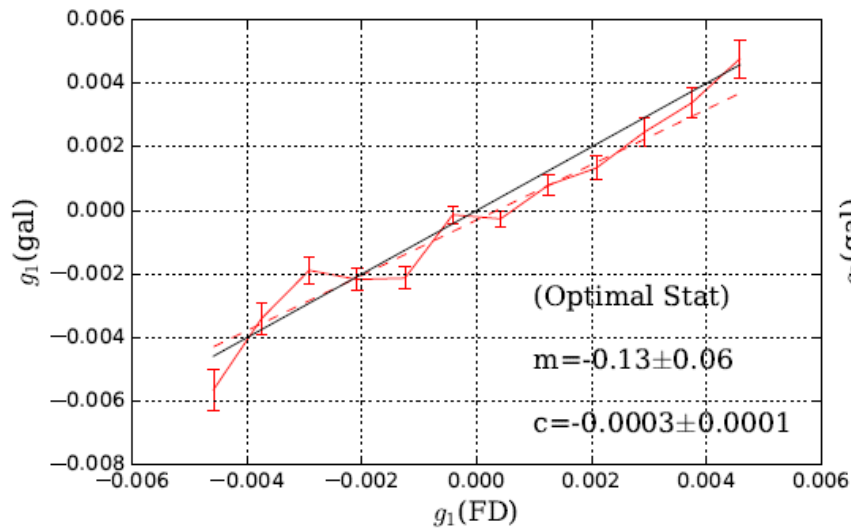
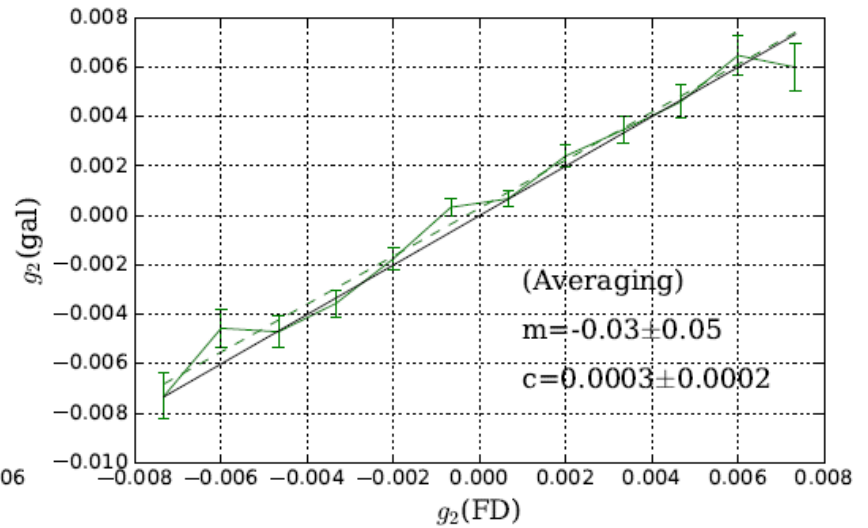
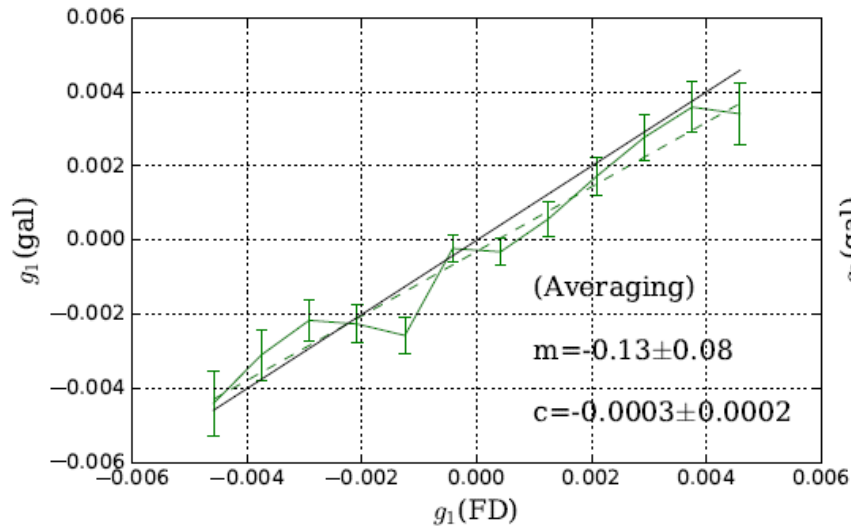


Source Distribution

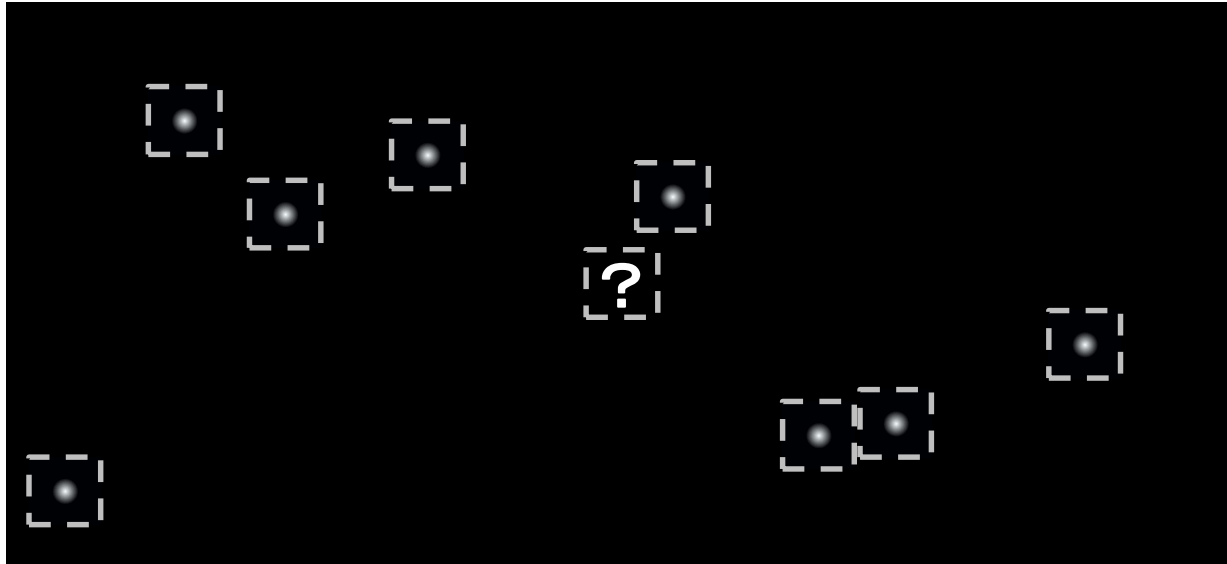
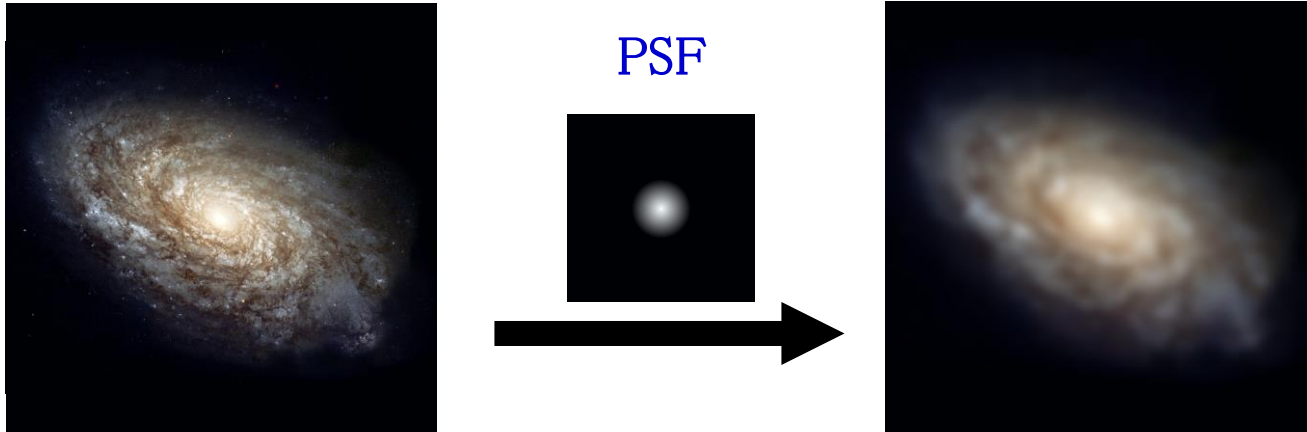


Stacked Source PS

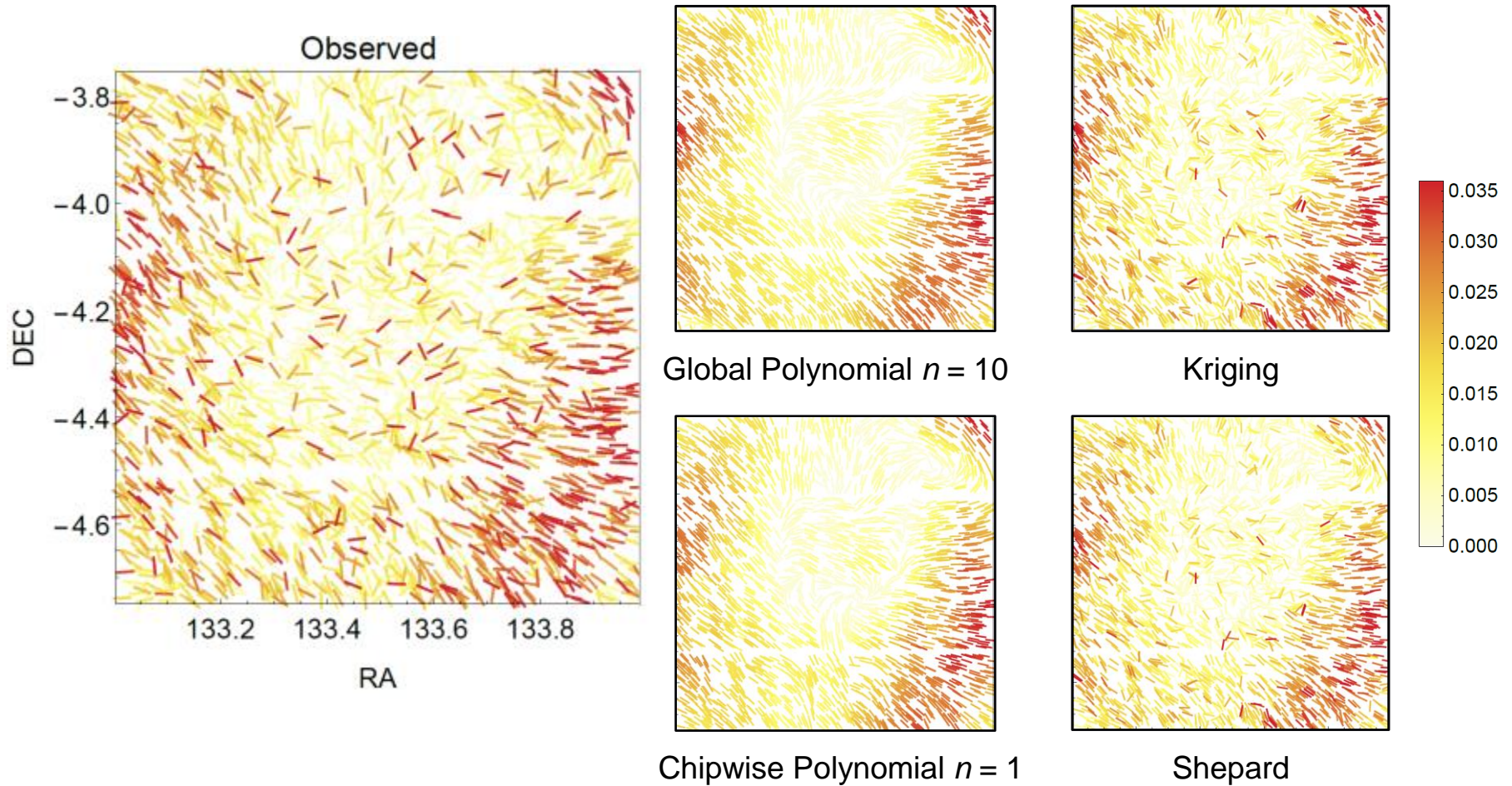
Application on the CFHT lens data



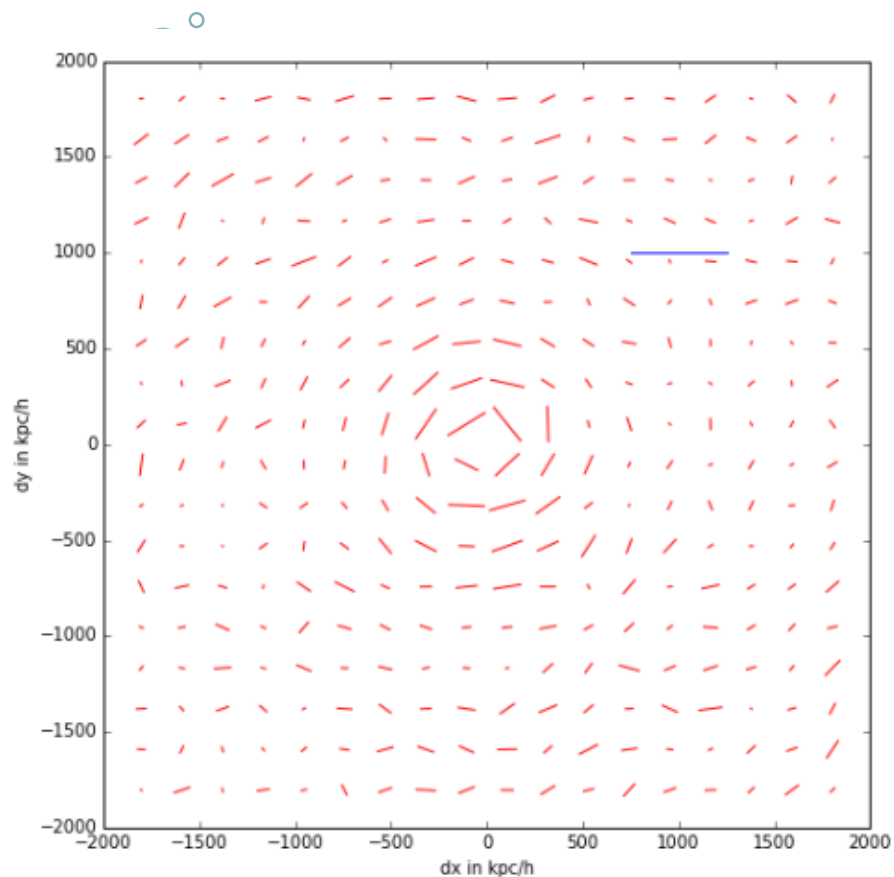
About PSF Interpolation



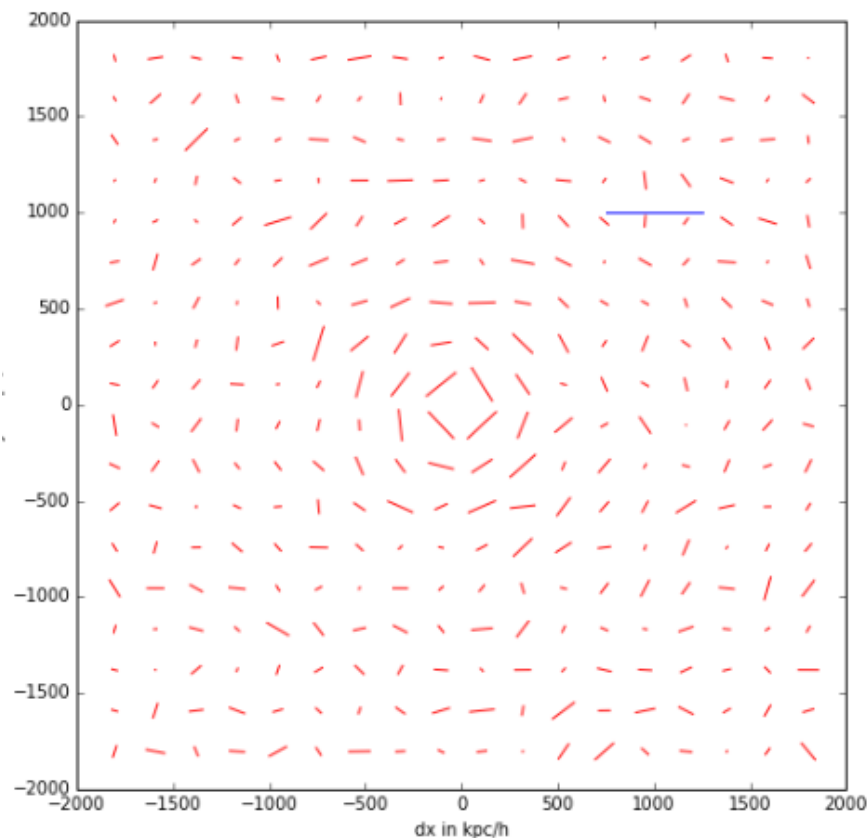
About PSF Interpolation



Preliminary Results on Cluster Lensing

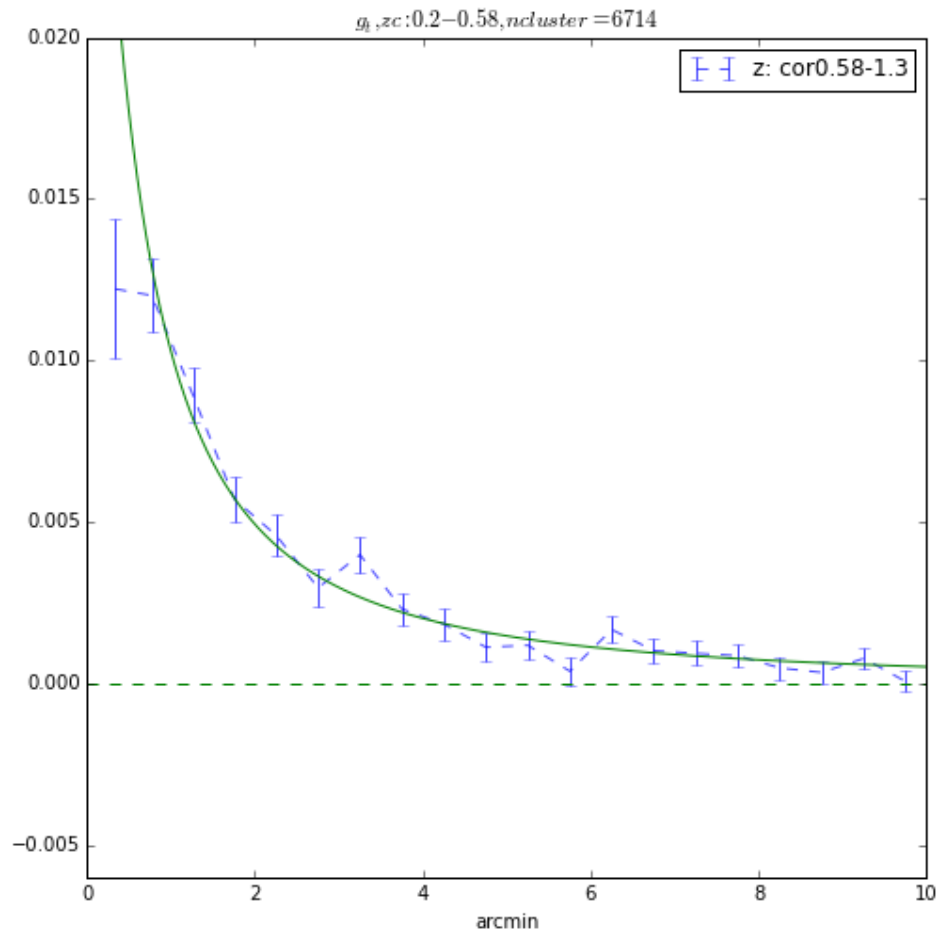


From Lensfit Method



From Our Method

Preliminary Results on Cluster Lensing



$$\lg m_{\text{vir}}(M_{\odot}/h) = 13.3552^{+0.0301}_{-0.0304}, c = 4.8900^{+0.9807}_{-0.7608}$$

Stay Tuned !
Thank You !