

Precision cosmology with LSST : Development of an unbiased cosmic shear estimator

Enya Van den Abeele, Pierre Astier, Anna Niemiec

Action Dark Energy - October 30th 2024

Cosmic shear

Distortion applied to image coordinates
(linear approximation):

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

convergence

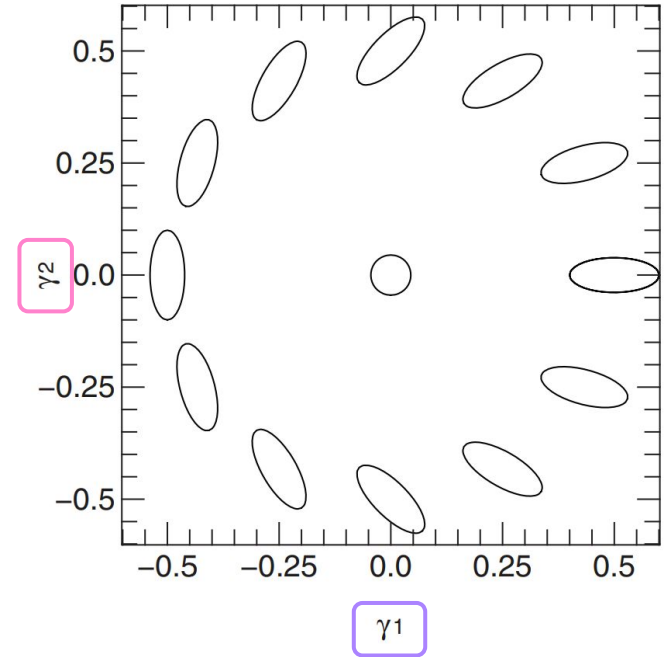
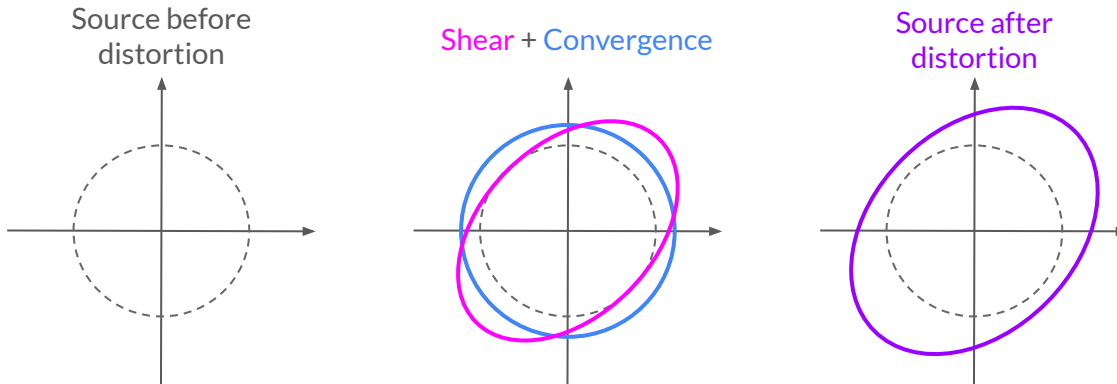


Figure : Martin Kilbinger

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convergence

Weak lensing limit \rightarrow *reduced shear*

$$|\gamma| \ll 1 \quad \longrightarrow \quad g_i = \frac{\gamma_i}{1 - \kappa}$$

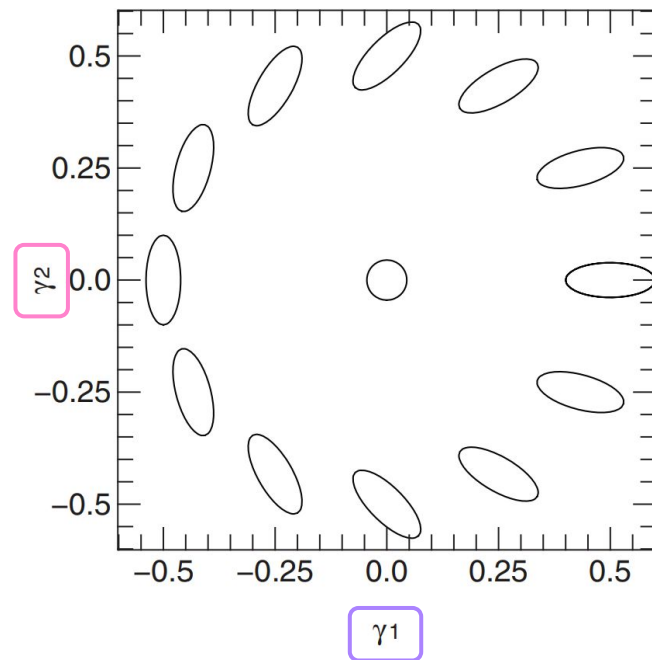
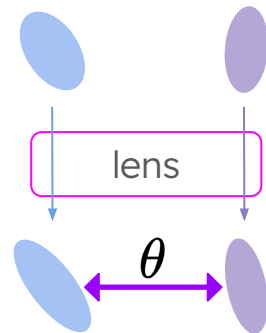


Figure : Martin Kilbinger

Cosmic shear and cosmological parameters

Two-point correlation function (2PCF) :

$$\xi_{\pm}(\theta) = \langle \gamma_t \gamma_t \rangle(\theta) \pm \langle \gamma_{\times} \gamma_{\times} \rangle(\theta) \quad \left\{ \begin{array}{l} \gamma_t = -\gamma_1 \\ \gamma_{\times} = -\gamma_2 \end{array} \right.$$

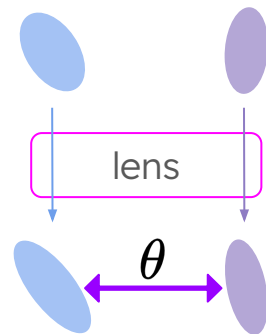


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$$\xi_{\pm}(\theta) = \frac{1}{2\pi} \int d\ell \ell J_{0/4}(\ell\theta) P_{\kappa}(\ell) \longrightarrow \kappa$$



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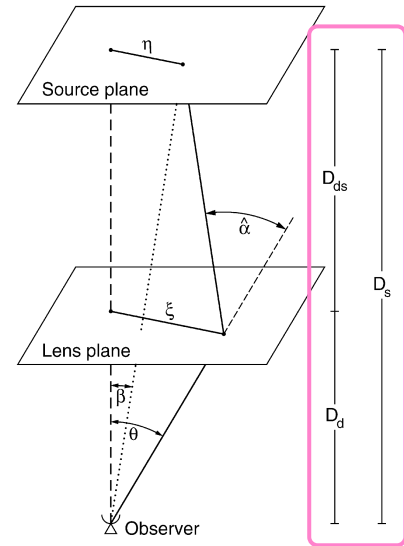
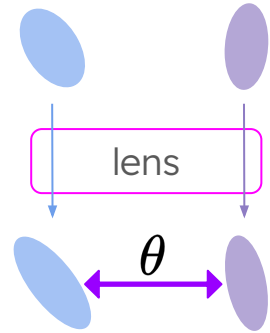
$$\xi_{\pm}(\theta) = \frac{1}{2\pi} \int dl l J_{0/4}(l\theta) P_{\kappa}(l) \longrightarrow \kappa$$

$$\kappa(z) = \frac{3}{2} \left(\frac{H_0}{c} \right)^2 \Omega_m \int \frac{\delta}{a} \frac{D_D D_{DS}}{D_S} dz$$

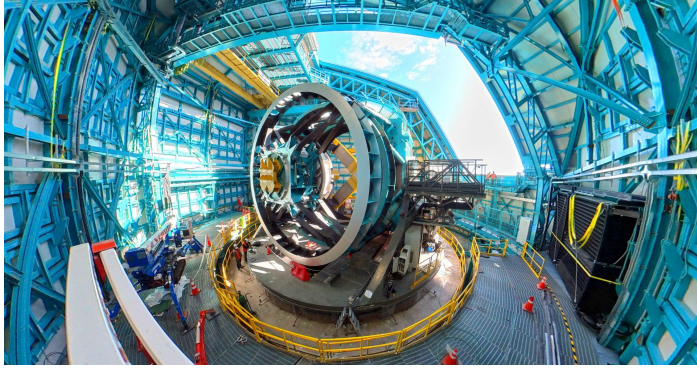
Density contrast :

- Matter density fluctuations
- Evolution with redshift

Distances



LSST : *Large Survey of Space and Time*



First ground-based telescope designed for weak lensing

- Primary mirror : 8.4m
- 3200 megapixels camera
- Scale : 0.2 arcsec/pixel



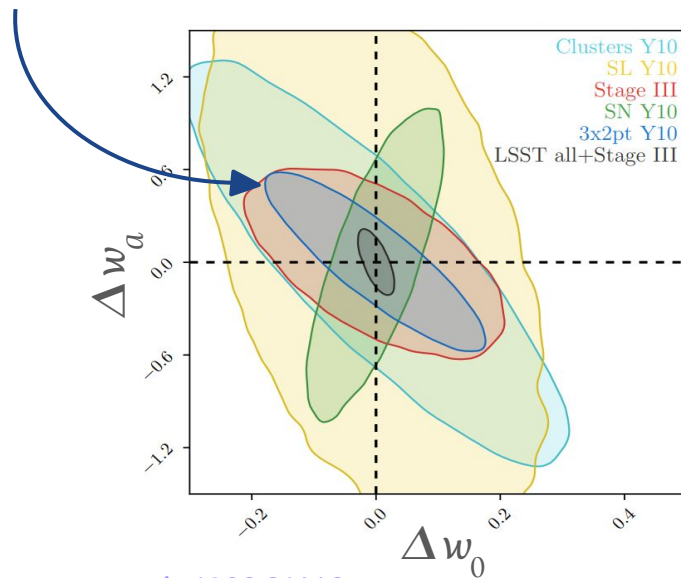
LSST : Large Survey of Space and Time



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Most sensitive probe for constraining **dark energy**



LSST : *Large Survey of Space and Time*



Unprecedented density of galaxies in images :

- Density : ~ 40 galaxies/arcmin²
(~ 26 after blending and masking)
- Total lensing : ~ 1.7 billion galaxies

DES : ~ 5.6 galaxies/arcmin²

HSC : ~ 15 galaxies/arcmin²

→ Precision cosmology !

Shear bias

Cosmological estimations with cosmic shear possible with LSST, but complex measurement associated with biases :

Shear bias

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$$m < 1.5 \times 10^{-3}$$
$$c < 1.3 \times 10^{-4}$$

(Cropper 2013)

Many sources :

- Poor estimator calibration
- PSF size
- Shot noise
- PSF anisotropy
- Astrometry
- Galaxy selection bias

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$$g^{obs} \rightarrow \xi_{+/-}^{ij}(\theta) \rightarrow P_{\kappa}(l) = f(\Omega_M, \sigma_8)$$

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bias on shear
=
bias on cosmology

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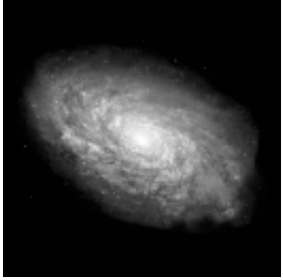
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bias on shear

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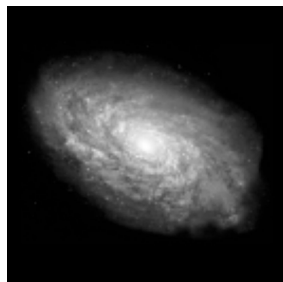
What shape do we measure ?



Galaxy with
intrinsic
ellipticity

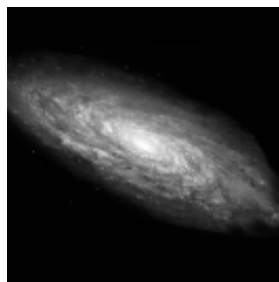
e^i

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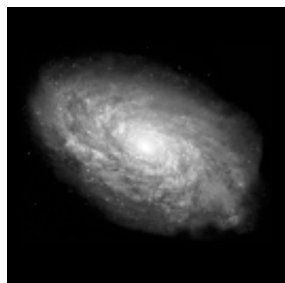
e^i



Cosmic shear
distortion

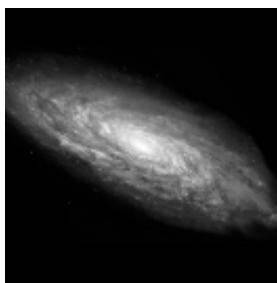
$e^i + \text{shear}$

What shape do we measure ?



Galaxy with
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e^i



Cosmic shear
distortion

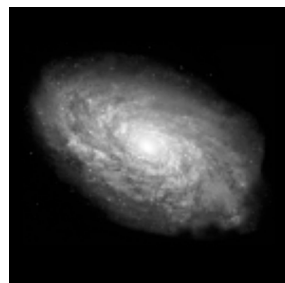
$e^i + \text{shear}$



PSF : atmosphere +
instrumental
response

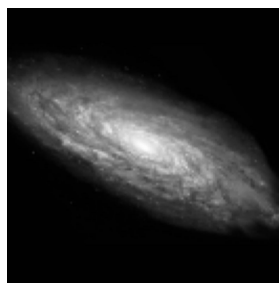
$e^i + \text{shear} + \text{PSF}$

What shape do we measure ?



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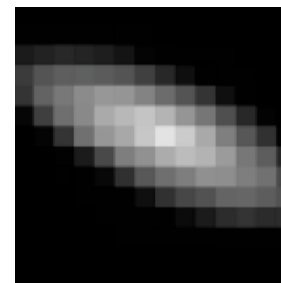
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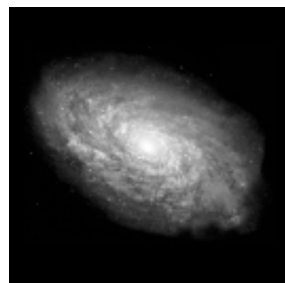
$e^i + \text{shear} + \text{PSF}$



Pixellisation

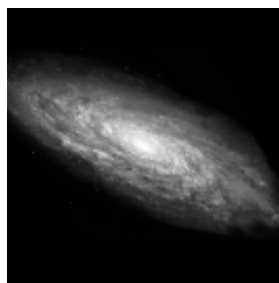
$e^i + \text{shear} + \text{PSF} +$
pixels

What shape do we measure ?



Galaxy with
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e^i



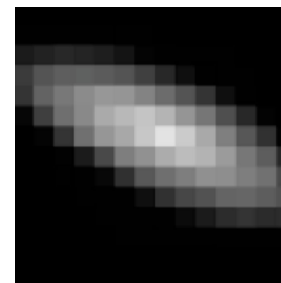
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$e^i + \text{shear}$



PSF : atmosphere +
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$e^i + \text{shear} + \text{PSF}$



Pixellisation

$e^i + \text{shear} + \text{PSF} + \text{pixels}$

Our observable (e)

Signal we want to
measure (g)

Estimator : Ellipticity Taylor's expansion

$$e = e|_{g=0} + g \frac{\partial e}{\partial g} |_{g=0} + \dots$$

Averaging over a large number of galaxies :

$$\langle e \rangle = \underbrace{\langle e|_{g=0} \rangle}_{= 0} + \langle g \frac{\partial e}{\partial g} |_{g=0} \rangle + \dots \simeq \langle \mathbf{R}g \rangle$$

Method

$$g^{obs} = (1 + m)g^{true} + c$$

$$m < 1.5 \times 10^{-3}$$

$$c < 1.3 \times 10^{-4}$$

(Cropper 2013)

Many sources :

- **Poor estimator calibration**
- PSF size
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$$\langle g \rangle = \langle R \rangle^{-1} \langle e \rangle$$

Self-calibrated shear
estimator

Method : Second moments and calibration

Shape measurement :

Second moments

Method : Second moments and calibration

Shape measurement :

Second moments

$$M = \int (X - X_0)(X - X_0)^T \overbrace{XX^T} \underbrace{W(X)} \underbrace{I(X)} dX^2$$

Pixel coordinates Weight function $I = I_0 \otimes \psi$

Method : Second moments and calibration

Shape measurement :
Second moments

$$M = \int (X - X_0)(X - X_0)^T \underbrace{XX^T}_{\text{Pixel coordinates}} \underbrace{W(X)}_{\text{Weight function}} \underbrace{I(X)}_{I = I_0 \otimes \psi} dX^2 \longrightarrow \mathbf{e} = \begin{pmatrix} e_1 \\ e_2 \end{pmatrix} = \frac{1}{F} \begin{pmatrix} M_{xx} - M_{yy} \\ 2M_{xy} \end{pmatrix}$$

Ellipticity

Normalized by the flux

Method : Second moments and calibration

Shape measurement :

Second moments

$$M = \int \overbrace{XX^T}^{(X-x_0)(X-x_0)^T} W(X) \underbrace{I(X)}_{I = I_0 \otimes \psi} dX^2 \longrightarrow \mathbf{e} = \begin{pmatrix} e_1 \\ e_2 \end{pmatrix} = \frac{1}{F} \begin{pmatrix} M_{xx} - M_{yy} \\ 2M_{xy} \end{pmatrix}$$

Pixel coordinates Weight function

Ellipticity

Calibration :

Second moments
derivatives with
respect to the
shear

Method : Second moments and calibration

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Second moments

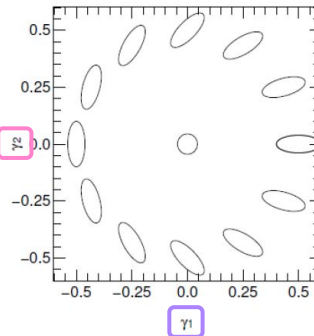
derivatives with

respect to the

shear

$$\frac{dM}{d\gamma} = \int \frac{dG(S(\gamma), X)}{d\gamma} I(X) dX^2$$

$$S = \frac{1}{\sqrt{1-g^2}} \begin{pmatrix} 1 + g_1 & g_2 \\ g_2 & 1 - g_1 \end{pmatrix}$$



Method : Second moments and calibration

Shape measurement :

Second moments

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Ellipticity

Calibration :

Second moments

derivatives with

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$$\frac{dM}{d\gamma} = \int \frac{dG(S(\gamma), X)}{d\gamma} I(X) dX^2 \longrightarrow \mathbf{R} = \begin{pmatrix} \frac{\partial e_1}{\partial g_1} & \frac{\partial e_2}{\partial g_1} \\ \frac{\partial e_1}{\partial g_2} & \frac{\partial e_2}{\partial g_2} \end{pmatrix}$$

$$S = \frac{1}{\sqrt{1-g^2}} \begin{pmatrix} 1+g_1 & g_2 \\ g_2 & 1-g_1 \end{pmatrix}$$

$$= \begin{pmatrix} \frac{\partial M_{xx}}{\partial g_1} & -\frac{\partial M_{yy}}{\partial g_1} & \frac{2\partial M_{xy}}{\partial g_1} \\ \frac{\partial M_{xx}}{\partial g_2} & -\frac{\partial M_{yy}}{\partial g_2} & \frac{2\partial M_{xy}}{\partial g_2} \end{pmatrix}$$

Auto-calibration factor

Method : Second moments and calibration

Shape measurement :

Second moments

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Ellipticity

Calibration :

Second moments

derivatives with respect to the shear

$$\frac{dM}{d\gamma} = \int \frac{dG(S(\gamma), X)}{d\gamma} I(X) dX^2 \longrightarrow \mathbf{R} = \begin{pmatrix} \frac{\partial e_1}{\partial g_1} & \frac{\partial e_2}{\partial g_1} \\ \frac{\partial e_1}{\partial g_2} & \frac{\partial e_2}{\partial g_2} \end{pmatrix}$$

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↑ ?

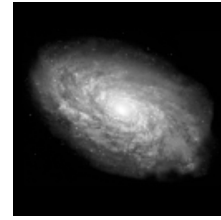
Auto-calibration factor

Method : Second moments and calibration

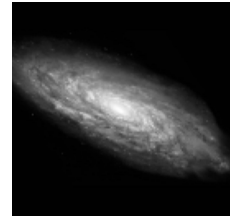
$$M(S) = \int (X X^T W(X)) \psi \otimes I_0(SX) dX^2$$

Metacalibration

arXiv:1702.02601



X



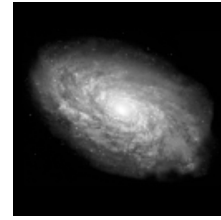
SX

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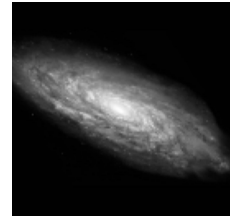
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X



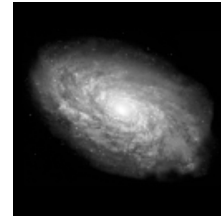
SX

→ Distorting original image introduces **correlated noise** !

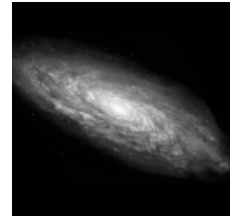
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Metacalibration
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X



SX

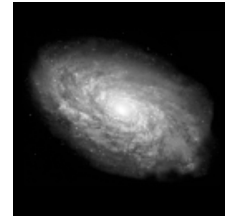
Parseval identity :

$$M(S) = \int \underbrace{(X X^T W(X) \circledast \psi_-)}_{F(X)} \overset{\circledast}{\curvearrowright} I_0(SX) dX^2$$

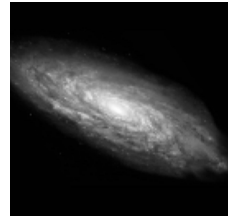
Method : Second moments and calibration

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X



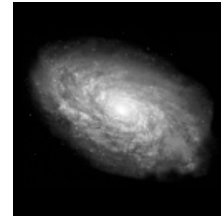
SX

$$\int F(X) I_0(SX) d^2 X = \int F(S^{-1} X) I_0(X) d^2 X$$

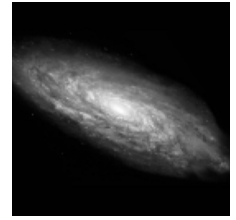
Method : Second moments and calibration

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Metacalibration
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X



SX

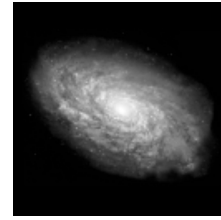
$$M(S) = \int F(Sk) I_0(k) d^2k \quad \longleftarrow \quad \text{Fourier space}$$

Method : Second moments and calibration

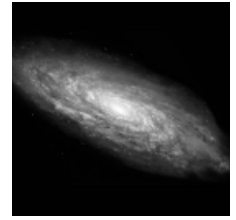
$$M(S) = \int (X X^T W(X)) \psi \otimes I_0(SX) dX^2$$

Metacalibration

arXiv:1702.02601



X



SX

$$M(S) = \int F(Sk) I_0(k) d^2k$$

$$= \int \frac{F(Sk)}{\psi^*(k)} \psi^*(k) I_0(k) d^2k$$



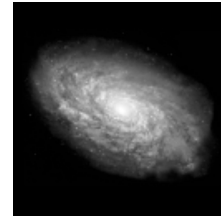
Division + Multiplication by the PSF

Method : Second moments and calibration

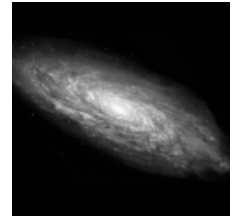
$$M(S) = \int (X X^T W(X)) \psi \otimes I_0(SX) dX^2$$

Metacalibration

arXiv:1702.02601



X



SX

$$\begin{aligned} M(S) &= \int F(Sk) I_0(k) d^2k \\ &= \int \frac{F(Sk)}{\psi^*(k)} \psi^*(k) I_0(k) d^2k \\ &= \int G(S, X) [\psi \otimes I_0] d^2X \end{aligned}$$



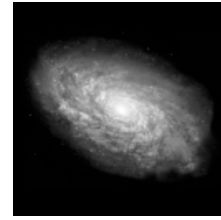
= Convolution in real space

Method : Second moments and calibration

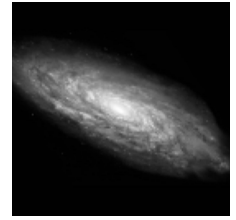
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X



SX

$$\begin{aligned} M(S) &= \int F(Sk) I_0(k) d^2 k \\ &= \int \frac{F(Sk)}{\psi^*(k)} \psi^*(k) I_0(k) d^2 k \\ &= \int G(S, X) [\psi \otimes I_0] d^2 X \\ &= \int G(S, X) I(X) d^2 X \end{aligned}$$



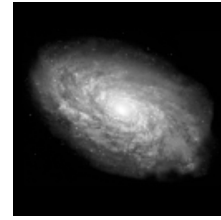
I(X) : original configuration

Method : Second moments and calibration

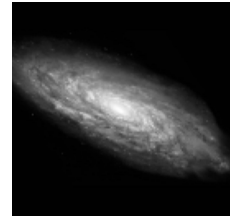
$$M(S) = \int (X X^T W(X)) \psi \otimes I_0(SX) dX^2$$

Metacalibration

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X



SX

$$\begin{aligned} M(S) &= \int F(Sk) I_0(k) d^2k \\ &= \int \frac{F(Sk)}{\psi^*(k)} \psi^*(k) I_0(k) d^2k \\ &= \int G(S, X) [\psi \otimes I_0] d^2X \\ &= \int G(S, X) I(X) d^2X \end{aligned}$$

- No distortion applied on original image !
- Allows shear estimations on under-sampled images

Method : Shear estimation

$$\mathbf{e} = \begin{pmatrix} e_1 \\ e_2 \end{pmatrix} = \frac{1}{F} \begin{pmatrix} M_{xx} - M_{yy} \\ 2M_{xy} \end{pmatrix}$$

$$\mathbf{R} = \begin{pmatrix} \frac{\partial e_1}{\partial g_1} & \frac{\partial e_1}{\partial g_2} \\ \frac{\partial e_2}{\partial g_1} & \frac{\partial e_2}{\partial g_2} \end{pmatrix} = \begin{pmatrix} \frac{\partial M_{xx}}{\partial g_1} - \frac{\partial M_{yy}}{\partial g_1} & \frac{2\partial M_{xy}}{\partial g_1} \\ \frac{\partial M_{xx}}{\partial g_2} - \frac{\partial M_{yy}}{\partial g_2} & \frac{2\partial M_{xy}}{\partial g_2} \end{pmatrix}$$



Shear estimation :

$$\langle g \rangle = \langle R \rangle^{-1} \langle e \rangle$$

(LSST : we aim to have biases less than ‰)

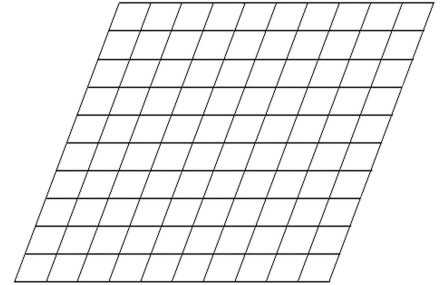
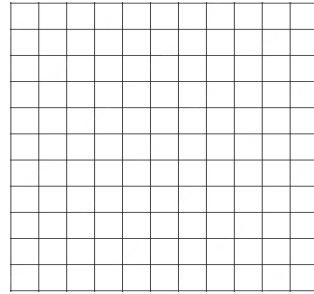
Method : Technical aspects

Application of shear variations ($\pm\varepsilon$) to calculate derivatives : distortion of the coordinate system (with \mathbf{S} matrix), then interpolation of the image (\mathbf{F} function) onto the new grid.

$$M(S) = \int [\psi^{-1} \circledast F(SX)] I(X) dX^2$$

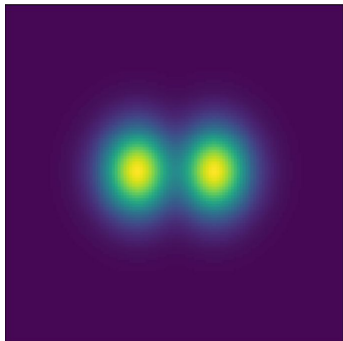
Galsim convention

$$S = \frac{1}{\sqrt{1-g^2}} \begin{pmatrix} 1+g_1 & g_2 \\ g_2 & 1-g_1 \end{pmatrix} \longrightarrow \begin{pmatrix} x' \\ y' \end{pmatrix} = S \begin{pmatrix} x \\ y \end{pmatrix}$$



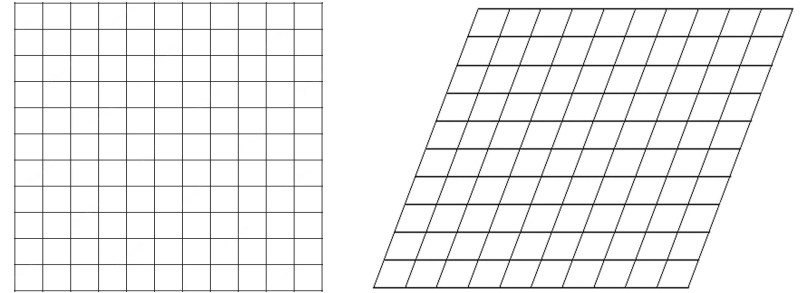
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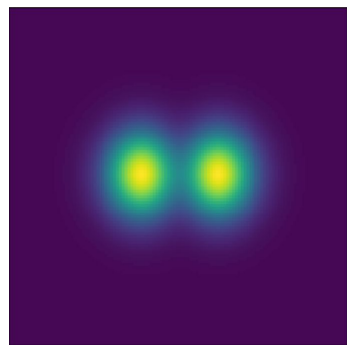
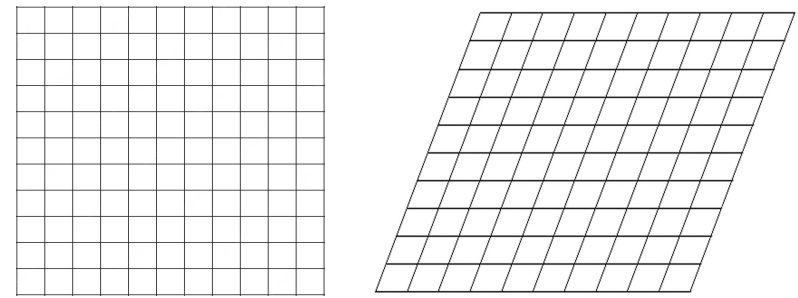


Method : Technical aspects

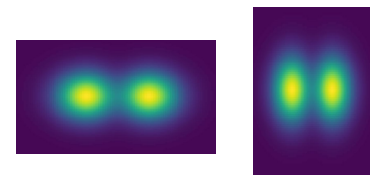
Application of shear variations ($\pm \varepsilon$) to calculate derivatives : distortion of the coordinate system (with S matrix), then interpolation of the image (F function) onto the new grid.

Galsim convention

$$S = \frac{1}{\sqrt{1-g^2}} \begin{pmatrix} 1+g_1 & g_2 \\ g_2 & 1-g_1 \end{pmatrix} \longrightarrow \begin{pmatrix} x' \\ y' \end{pmatrix} = S \begin{pmatrix} x \\ y \end{pmatrix}$$

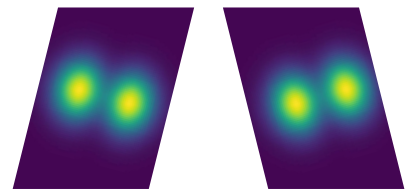


$g_1 = \pm \varepsilon ; g_2 = 0$



grid distortion + interpolation

$g_1 = 0 ; g_2 = \pm \varepsilon$

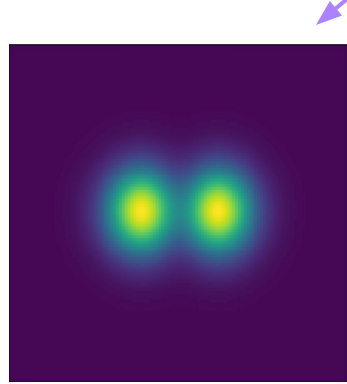
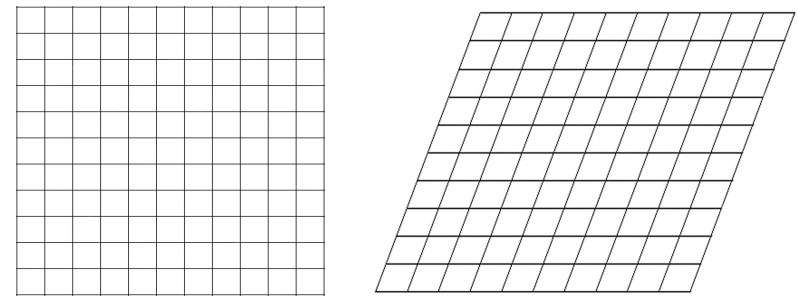


Method : Technical aspects

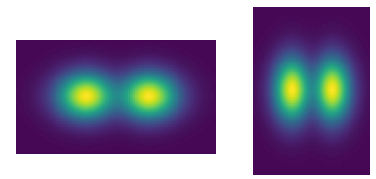
Application of shear variations ($\pm\epsilon$) to calculate derivatives : distortion of the coordinate system (with \mathbf{S} matrix), then interpolation of the image (\mathbf{F} function) onto the new grid.

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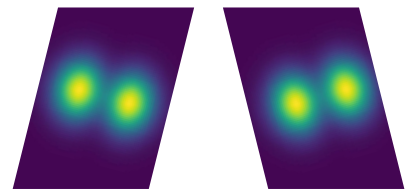


$$g_1 = \pm \epsilon ; g_2 = 0$$



grid distortion + interpolation

$$g_1 = 0 ; g_2 = \pm \epsilon$$



$$\frac{\partial M}{\partial g_1} = \frac{S M_{1+} - S M_{1-}}{2\epsilon}$$

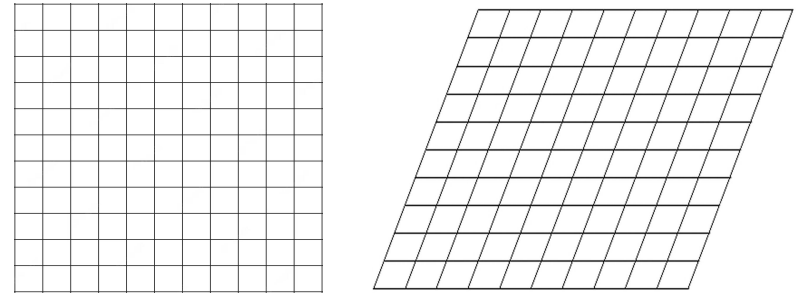
$$\frac{\partial M}{\partial g_2} = \frac{S M_{2+} - S M_{2-}}{2\epsilon}$$

Method : Technical aspects

Application of shear variations ($\pm\varepsilon$) to calculate derivatives : distortion of the coordinate system (with \mathbf{S} matrix), then interpolation of the image (\mathbf{F} function) onto the new grid.

↷ *Galsim convention*

$$S = \frac{1}{\sqrt{1-g^2}} \begin{pmatrix} 1+g_1 & g_2 \\ g_2 & 1-g_1 \end{pmatrix} \longrightarrow \begin{pmatrix} x' \\ y' \end{pmatrix} = S \begin{pmatrix} x \\ y \end{pmatrix}$$



Sampling :

$$I = I_c \otimes \Pi \Leftrightarrow M = M_c + M_{pix}$$

$$M = \int XX^T I(X) d^2X + \frac{JJ^T}{12}$$

with :

- J : the Jacobian involved in the affine transformation of coordinates (pixel \leftrightarrow physical)

- s : the image pixel scale (arcsec/pixel)

$$J = \begin{pmatrix} s & 0 \\ 0 & s \end{pmatrix}$$

Method : Technical aspects - shear & sampling cross-effect

As we measure the SM matrices on a distorted pixels grid, we should subtract a *distorted* pixels second moments matrix M_{pix} to recover the real object second moments.

New second moments formalism :

$$M(s, \epsilon) \propto \underbrace{\gamma + \alpha\epsilon + \alpha'\epsilon^2}_{\text{theoretical second moments}} + \underbrace{\beta s^2 + \beta' s^4}_{\text{sampling correction}} + \underbrace{\delta s^2 \epsilon}_{\text{sampling x shear correction}} + \underbrace{\delta' s^4 \epsilon}_{\text{sampling x shear correction}}$$

Method : Technical aspects - shear & sampling cross-effect

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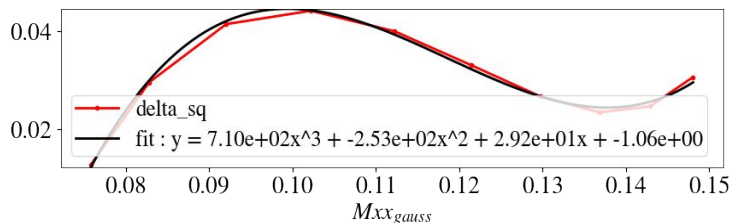
↓
No big impact of the galaxy profile on the estimation of δ' , but highly related to the size of the galaxy !

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Relation between δ' and the galaxy 2nd moments

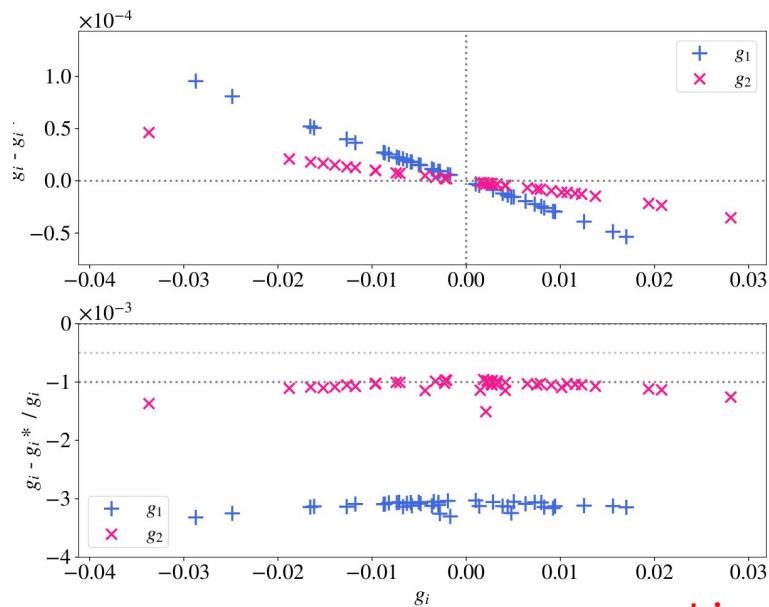


No big impact of the galaxy profile on the estimation of δ' , but highly related to the size of the galaxy !



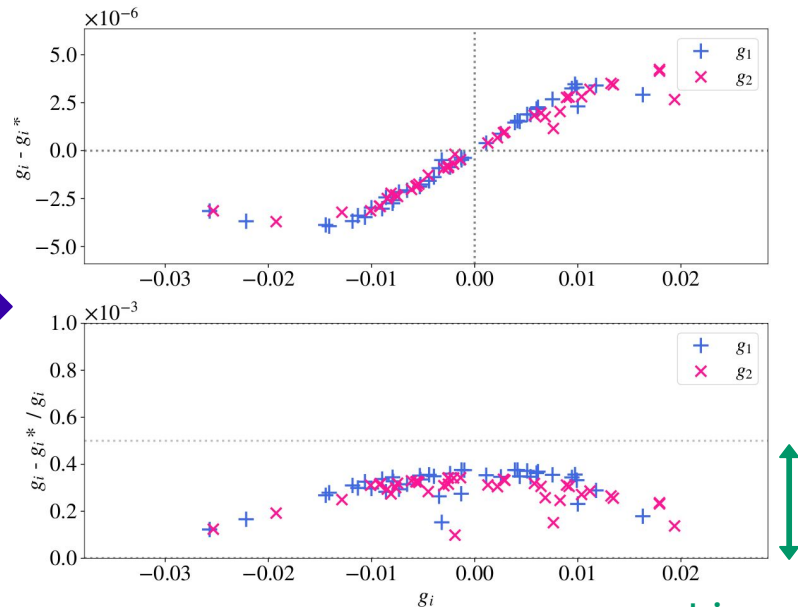
Method : Technical aspects - shear & sampling cross-effect

Before SSB correction



bias \sim 3‰

After SSB correction



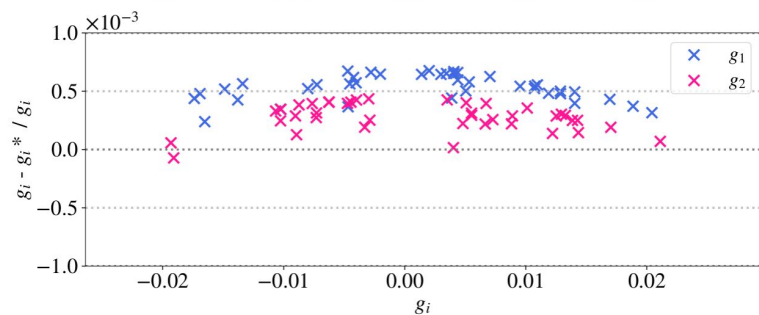
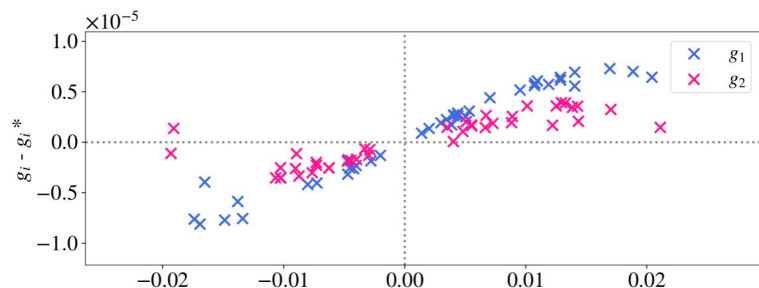
bias $<$ 1‰

Results on noise-free simulations

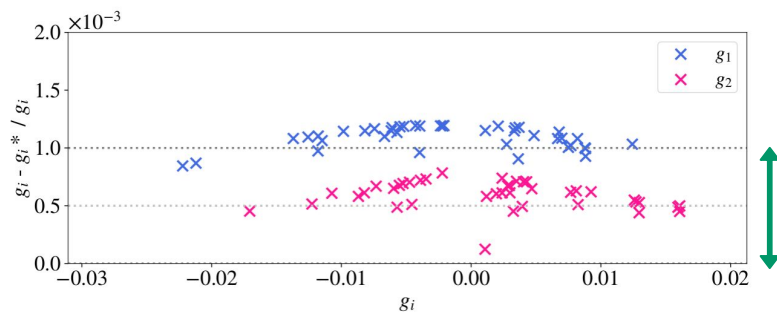
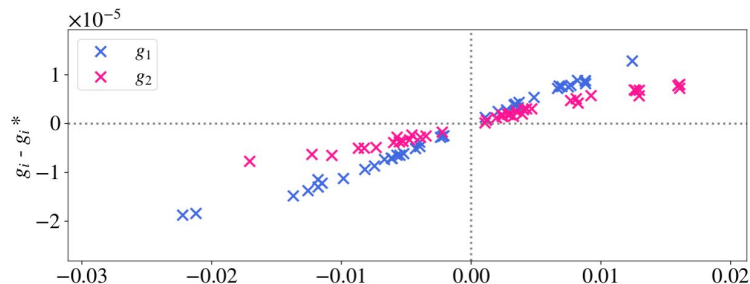
Bulge + Disk galaxies

$$TR = \frac{Tr(M_{image})}{Tr(M_{PSF})}$$

- Moffat PSF
- $n = 1.0$ (bulge)
- $r = 0.8''$ (disk)
- $TR = 1.57$

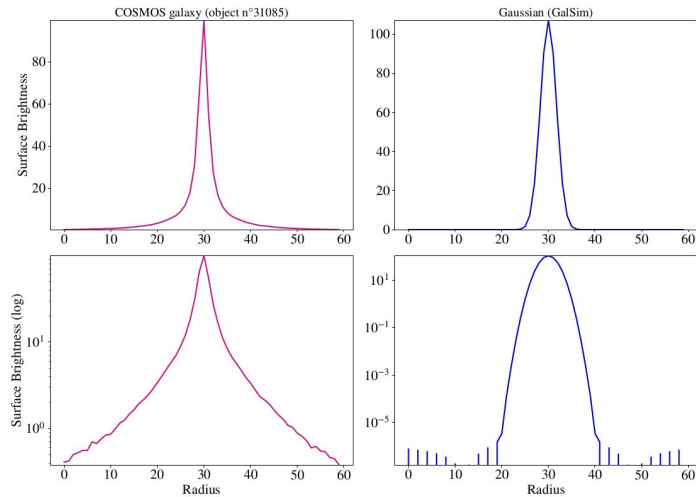
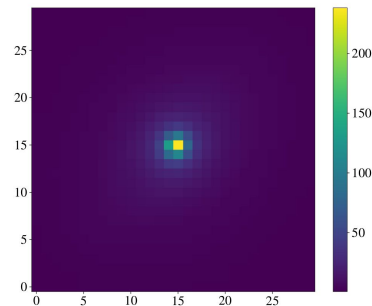
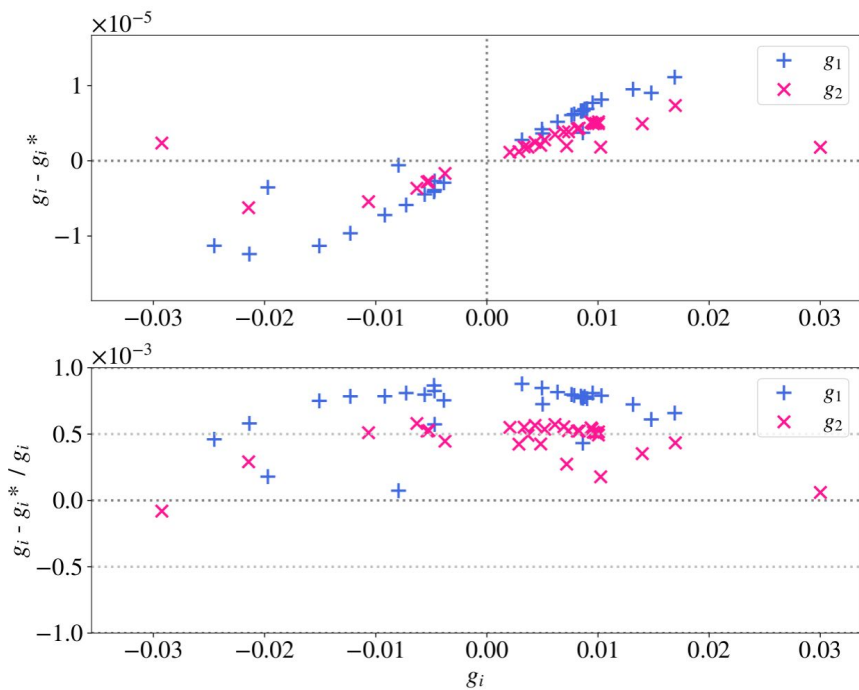


- Moffat PSF
- $n = 4.0$ (bulge)
- $r = 0.8''$ (disk)
- $TR = 1.52$



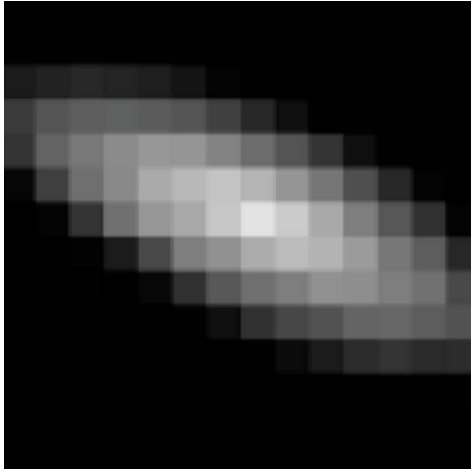
Realistic galaxies (COSMOS catalog)

- Moffat PSF
- TR = 1.62

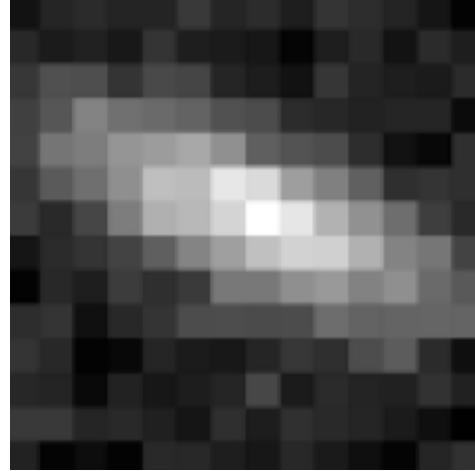


Noisy simulations

Received image

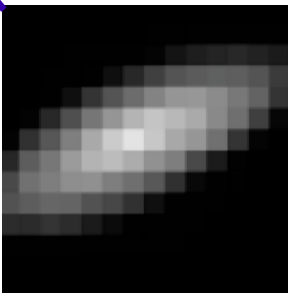
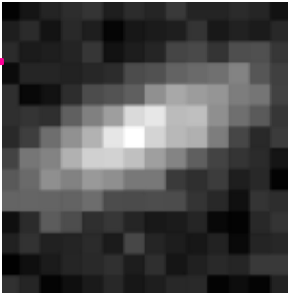
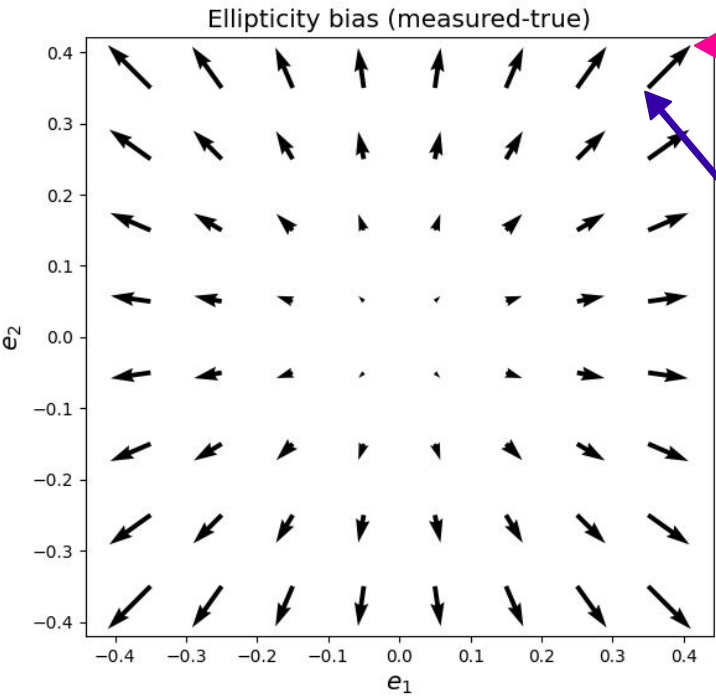


e^i + shear + PSF + pixels

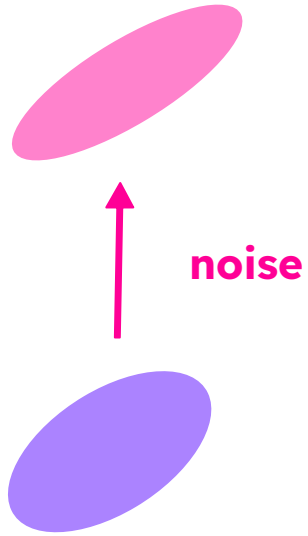


e^i + shear + PSF + pixels + **noise**

Ellipticity bias



Observed ellipticity :



Noise bias correction

$$M = \int \overbrace{X X^T} (x - x_0)(x - \boxed{x_0})^T W(X) I(X) dX^2$$

Galaxy position measured on a noisy image → **introduces a bias !**

Noise bias correction

$$M = \int \underbrace{(x - x_0)(x - \boxed{x_0})^T}_{XX^T} W(X) I(X) dX^2$$

Galaxy position measured on a noisy image → **introduces a bias !**

Idea : calculate an analytical formula of the second moments noise bias :

$$m_I(I + n) = m_I(I) + \sum_k \frac{dm_I}{dI_k} n_k + \frac{1}{2} \sum_{kl} \boxed{\frac{d^2 m_I}{dI_k dI_l}} n_k n_l$$

Noise bias correction

$$M = \int \underbrace{(x - x_0)(x - \boxed{x_0})^T}_{XX^T} W(X) I(X) dX^2$$

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Same for the flux : $F = \sum W(x - x_0) I$

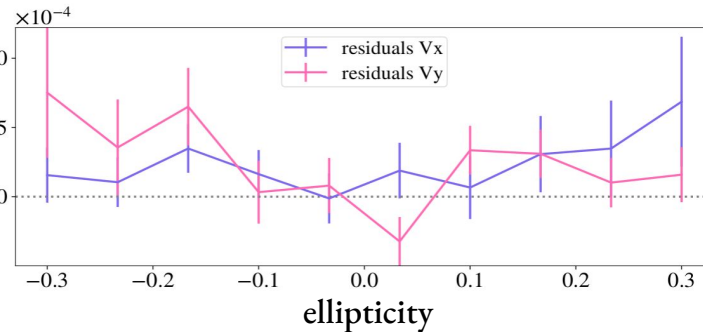
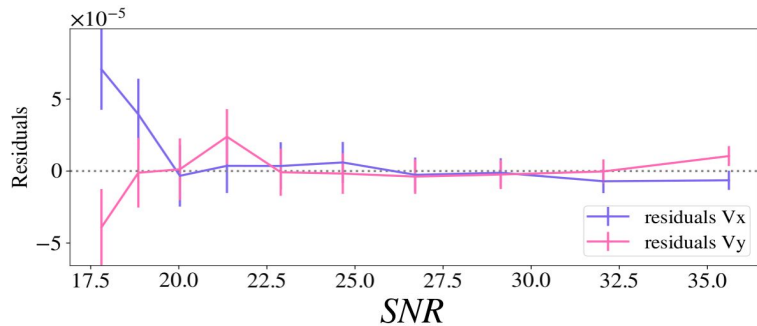
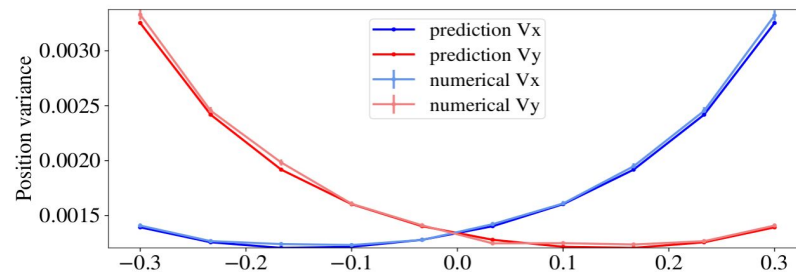
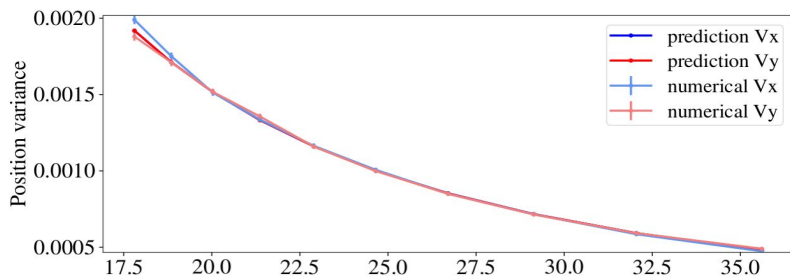
And the position variance : $V(x_0)$

Results : Position variance prediction

$$V = \sigma_{noise}^2 K M_W^2 K$$

$$K = \frac{1}{F} [1 - M_W^{-1} M_P]^{-1}$$

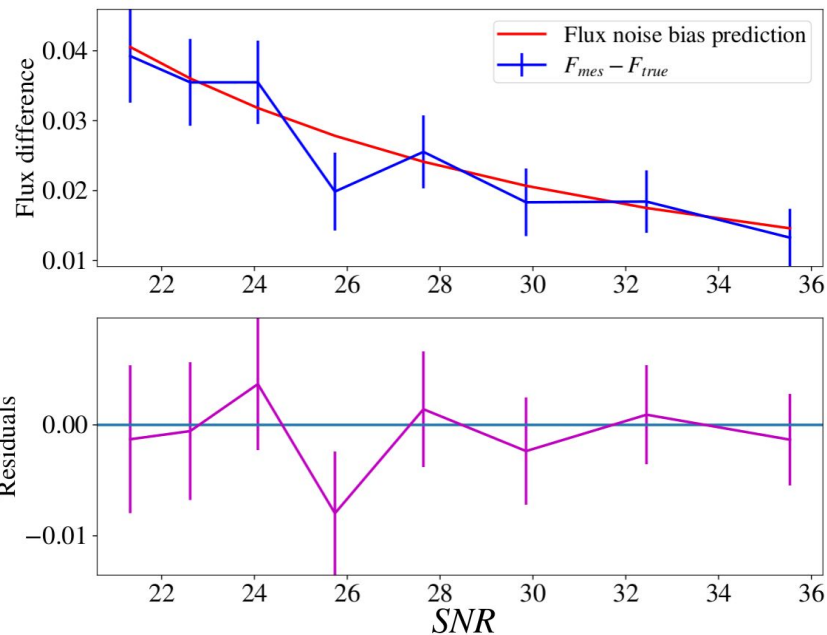
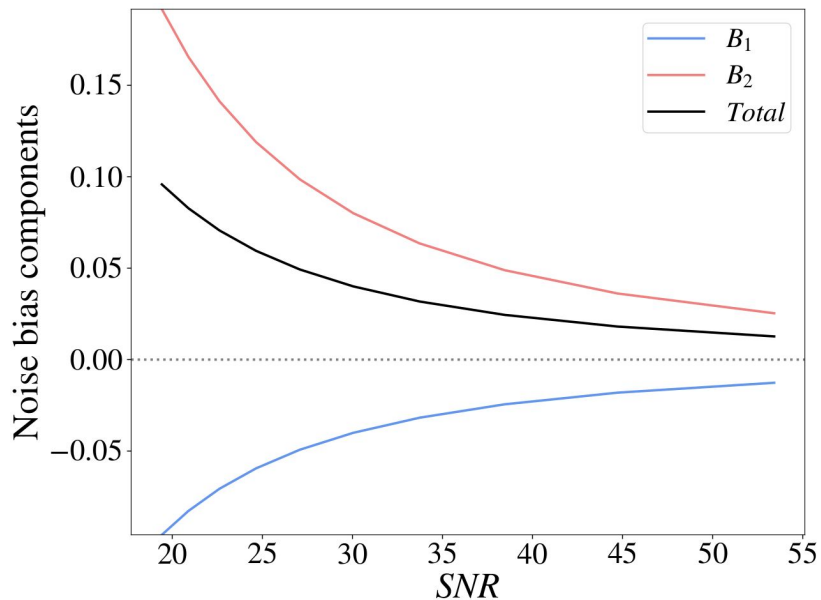
Flux P = I*W



Results : Flux bias prediction

$$F = \sum W(x - x_0)I$$

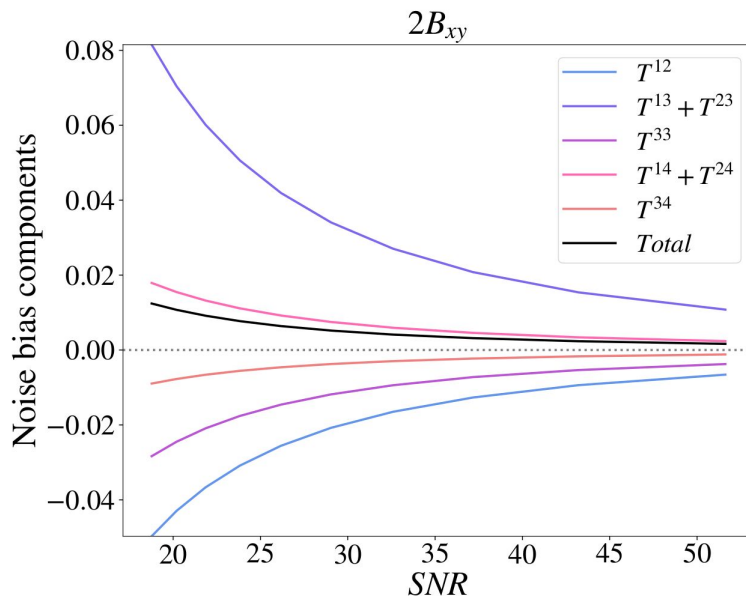
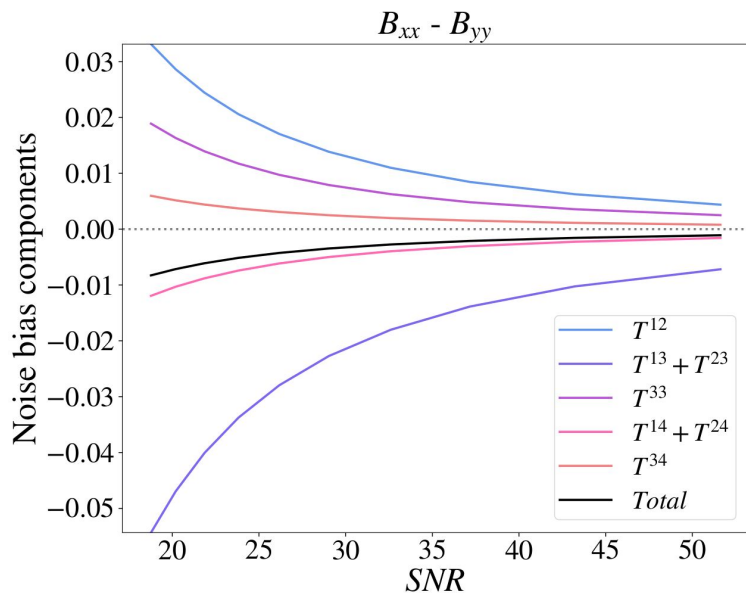
$$B_F = B_1 + B_2$$



Results : Second moments bias prediction

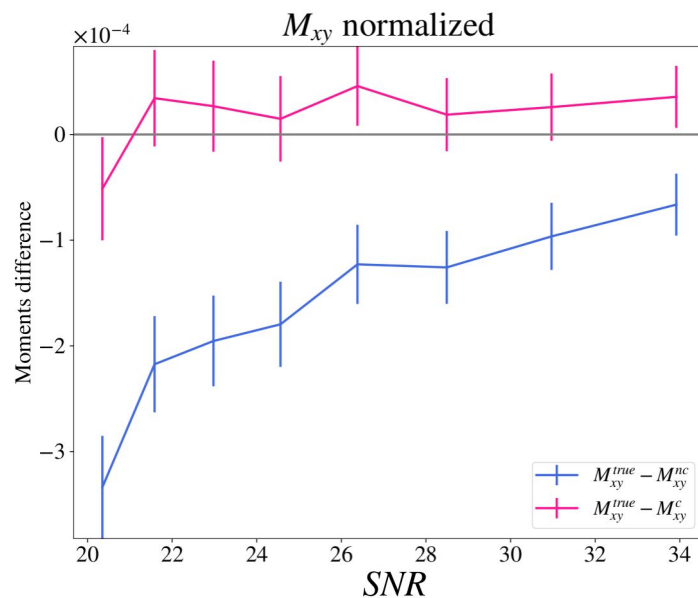
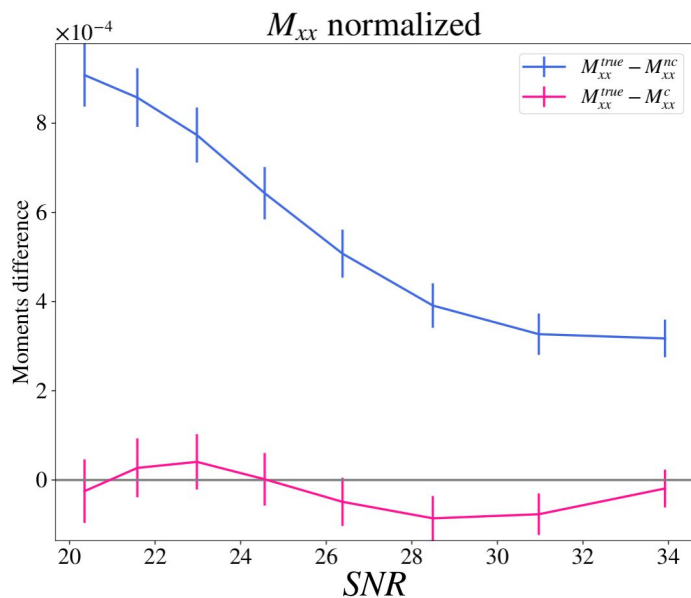
$$m_{\alpha\beta}^P = \sum_k \mathbf{x}_\alpha^k \mathbf{x}_\beta^k W_k I_k$$

$$B_M = \frac{1}{2}(T^{12} + 2T^{13} + T^{33}) + 2T^{14} + T^{34}$$



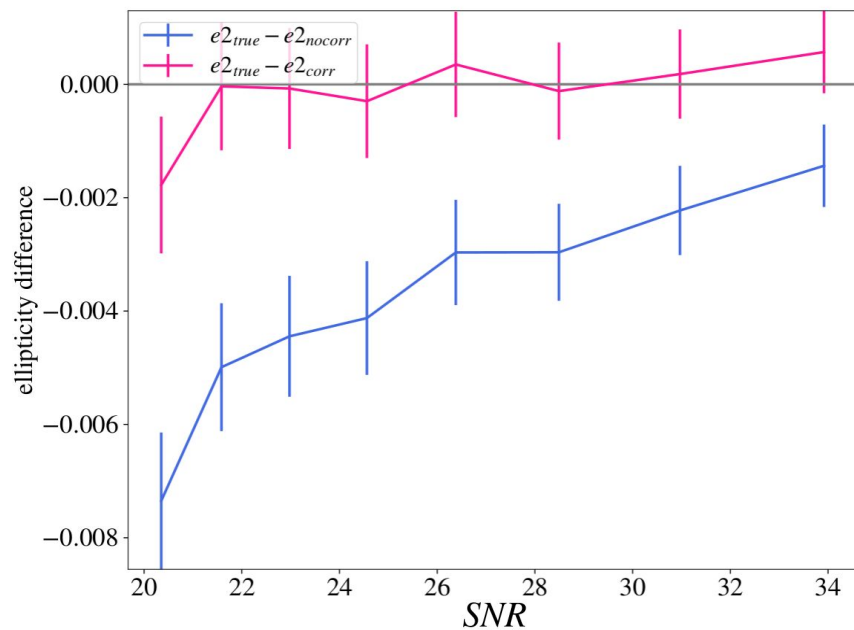
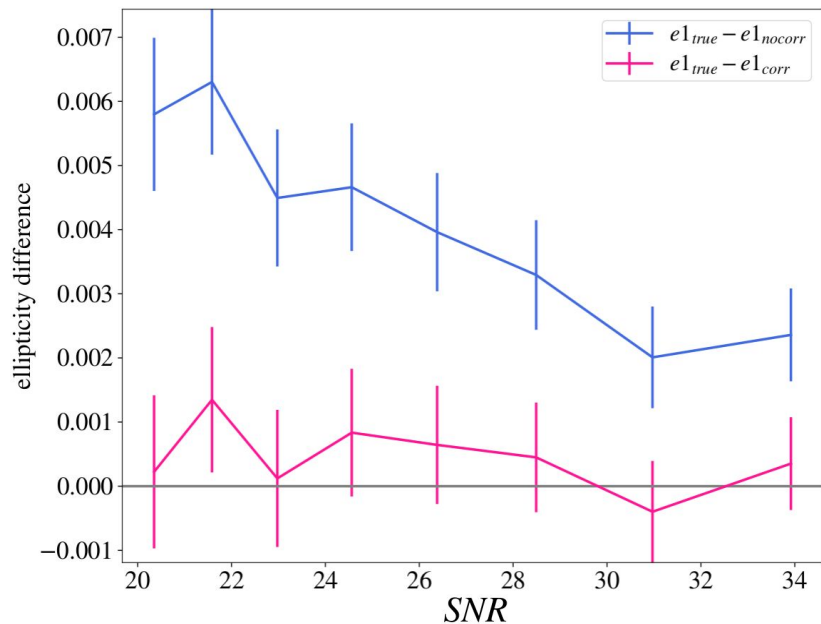
Results : Second moments bias prediction

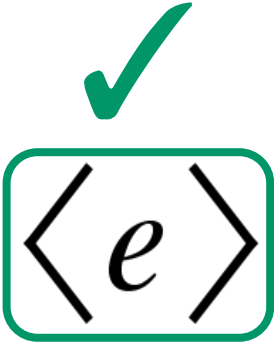
$$M_c = \frac{M_{mes} - B_M}{F_{mes} - B_F}$$



Results : Ellipticity bias correction

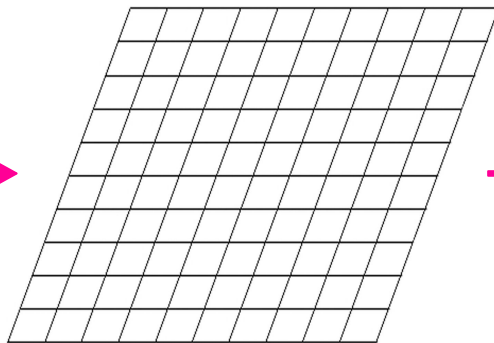
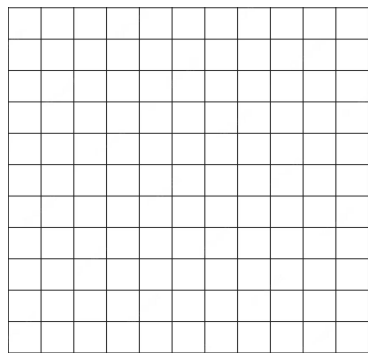
$$\mathbf{e} = \begin{pmatrix} e_1 \\ e_2 \end{pmatrix} = \frac{1}{F} \begin{pmatrix} M_{xx} - M_{yy} \\ 2M_{xy} \end{pmatrix}$$



$$\langle g \rangle = \langle R \rangle^{-1} \langle e \rangle$$


$$\langle g \rangle = \langle R \rangle^{-1} \langle e \rangle$$

Auto-calibration factor bias prediction



Second moments
and flux correlated !

~~$$M_c = \frac{M_{mes} - B_M}{F_{mes} - B_F}$$~~

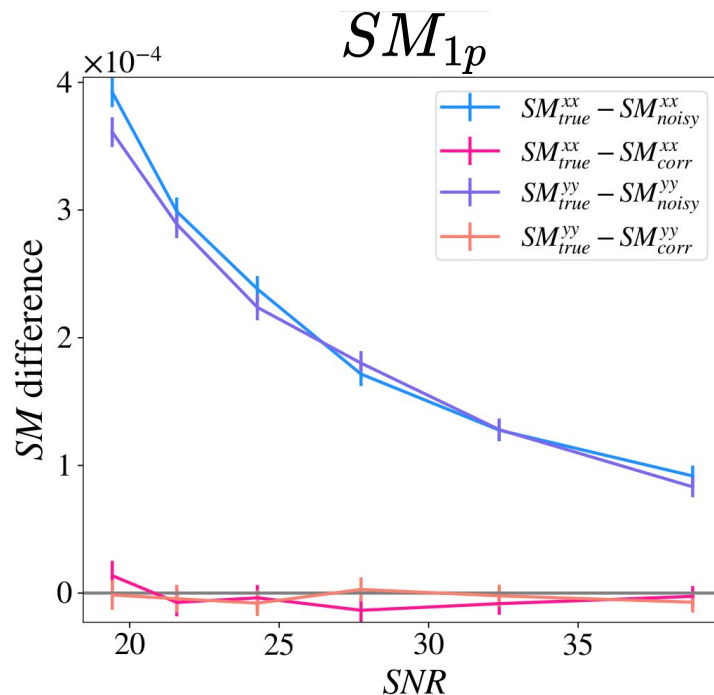
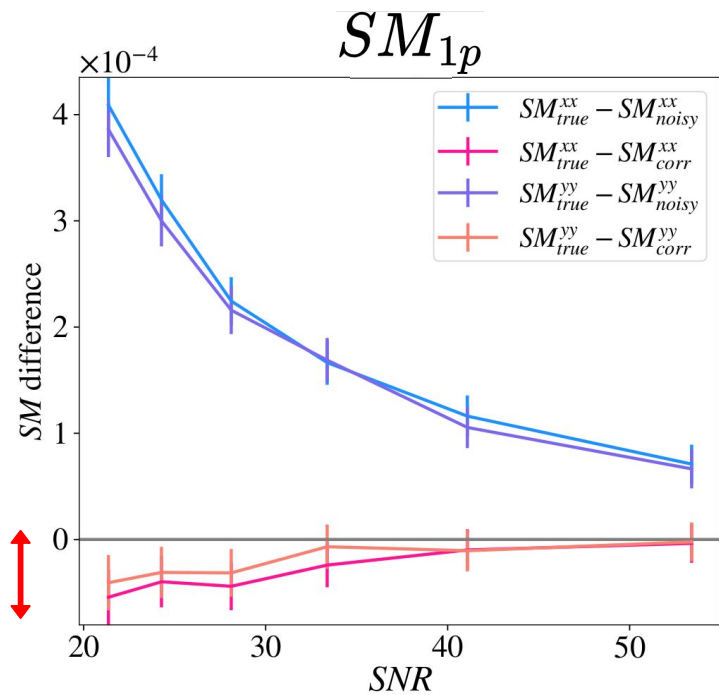


$$M_c = \left(\frac{M}{F} \right)_{mes} - B_{M/F}$$

Results : R correction

$$M_c = \frac{M_{mes} - B_M}{F_{mes} - B_F}$$

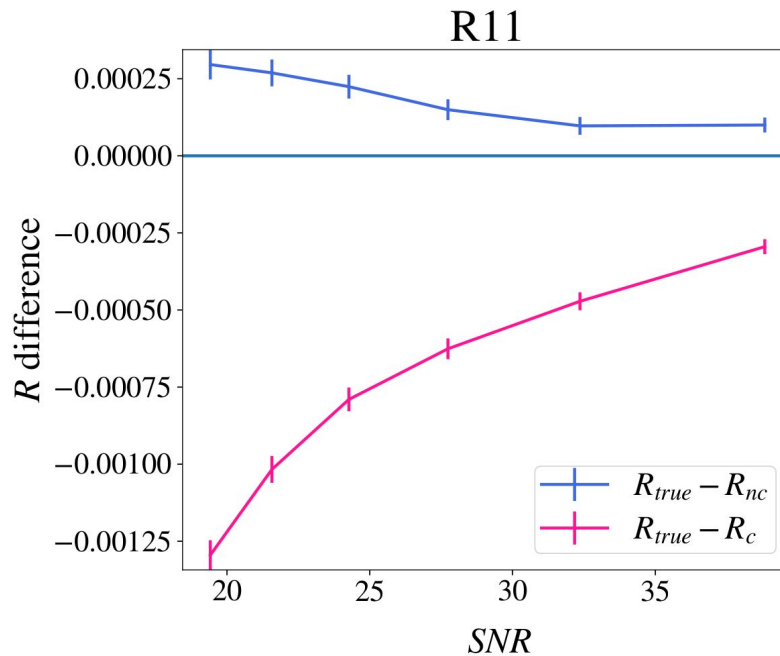
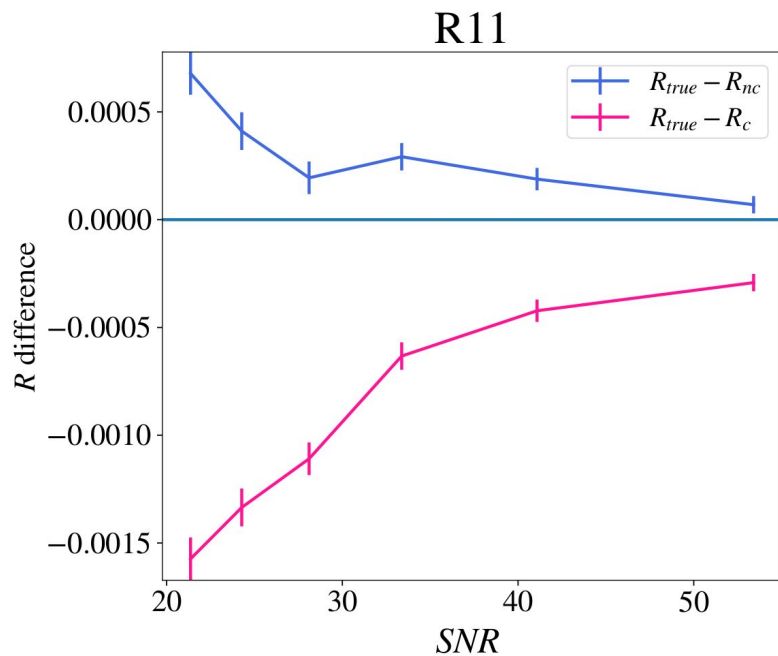
$$M_c = \left(\frac{M}{F} \right)_{mes} - B_{M/F}$$



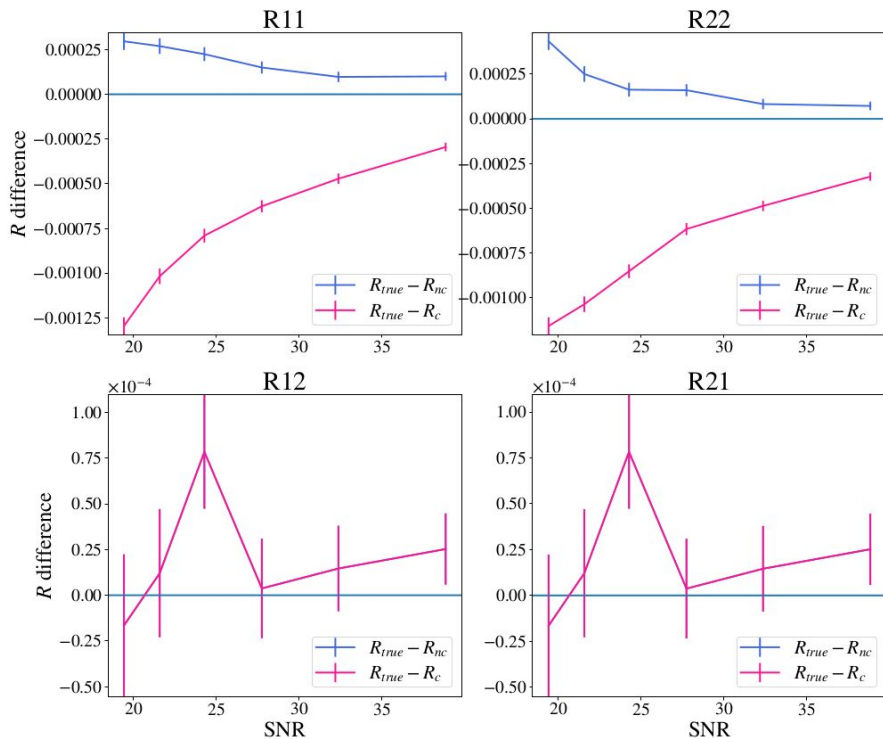
Results : R correction

$$M_c = \frac{M_{mes} - B_M}{F_{mes} - B_F}$$

$$M_c = \left(\frac{M}{F} \right)_{mes} - B_{M/F}$$



Results : R correction



Perspectives :

- Combining numerical and analytical derivatives
- M derivatives wrt shear in Fourier space (disappearance of SSB ?)
- Calculate the true noise bias prediction (JAX, Schoenholz and Cubuk, 2019)

Conclusion

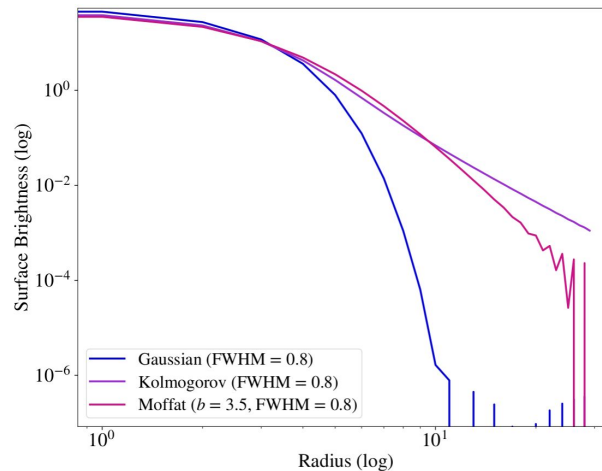
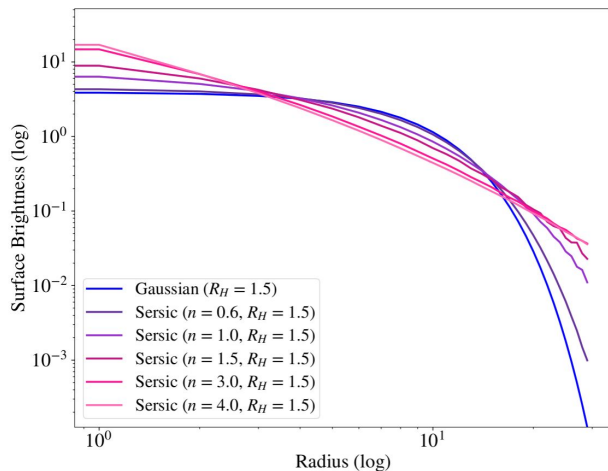
- LSST : future-generation survey → **precision cosmology**
- Bias in shear measurement : limit to be set on **m** ($< 10^{-3}$)
- Development of a **self-calibrated shear estimator**
 - Independent of galaxy profile
 - No distortion applied to the original image
 - Satisfactory results on basic (noise-free) tests
- Noisy simulations : **analytical formulas to correct noise bias**
 - Good correction of ellipticity bias
 - Auto-calibration factor : promising initial results, work in progress

Thank you

Backup

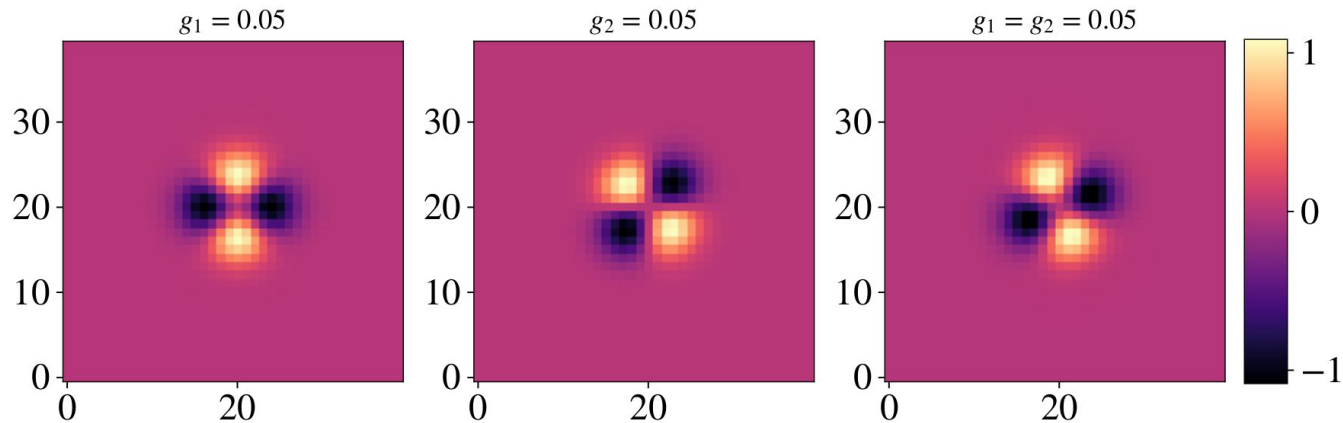
Simulations

- Images simulated with the *GalSim* package (Rowe et al. 2015)
- Used profiles :
 - Galaxies : Gaussian, Sersic, Bulge + Disk
 - PSF : Gaussian, Kolmogorov, Moffat



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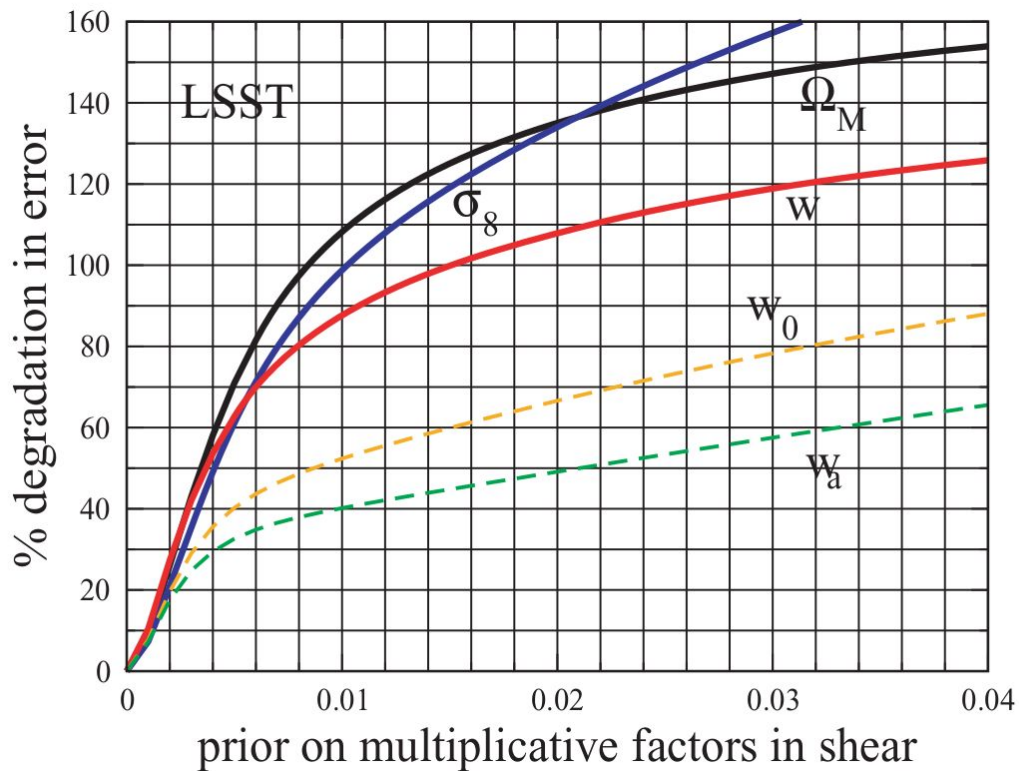
Simulations

- Images simulated with the *GalSim* package (Rowe et al. 2015)
- Used profiles :
 - **Galaxies** : Gaussian, Sersic, Bulge + Disk
 - **PSF** : Gaussian, Kolmogorov, Moffat
- Sizes choices : Trace Ratio (TR) :
 - Fixed PSF size (FWHM = 0.8 arcsec)
 - Fixed W size (Gaussienne FWHM = 1 arcsec)
 - Simulations with **TR ~ 1.5**

$$TR = \frac{\text{Tr}(M_{image})}{\text{Tr}(M_{PSF})}$$
$$M_{image} = \sum_{i,j}^N X_i X_j^T W_i [I_0 \otimes \psi]_i$$
$$M_{PSF} = \sum_{i,j}^N X_i X_j^T W_i \psi_i$$

→ Mandelbaum et al. 2018, Zuntz et al. 2018

Impact of multiplicative bias on cosmological parameters



Shear correlation functions

2 shear components for each galaxy \rightarrow one tangential and one crossed :

$$\gamma_t = -\Re(\gamma e^{-2i\Phi})$$

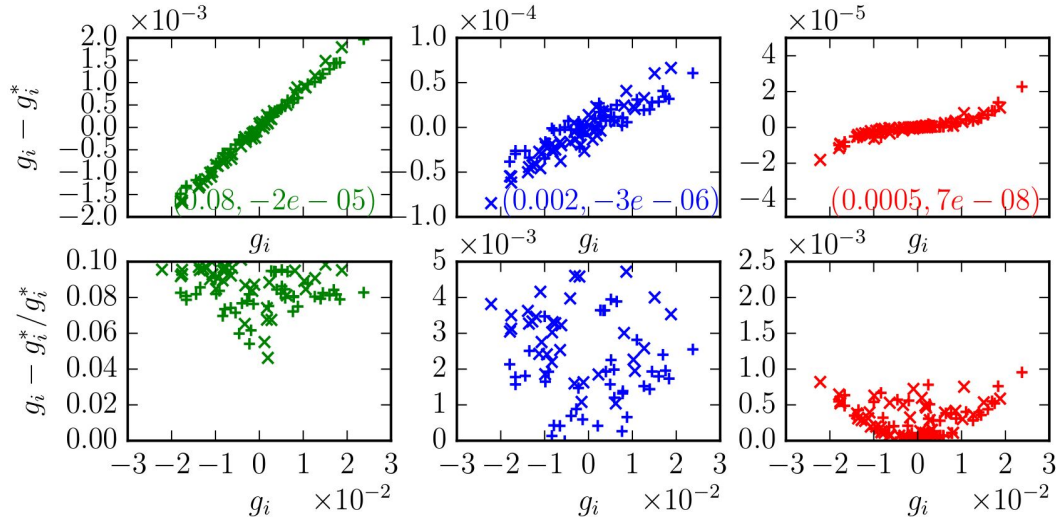
$$\gamma_{\times} = -\Im(\gamma e^{-2i\Phi})$$

Then the 2 non-zero components of the 2pt correlation function :

$$\xi_+ = \langle \gamma \gamma^* \rangle(\theta) = \langle \gamma_t \gamma_t \rangle + \langle \gamma_{\times} \gamma_{\times} \rangle$$

$$\xi_- = \Re \left[\langle \gamma \gamma \rangle(\theta) e^{-4i\Phi} \right] = \langle \gamma_t \gamma_t \rangle - \langle \gamma_{\times} \gamma_{\times} \rangle$$

Shear estimation formalism



Upper panel : absolute difference between input and output shear

Lower panel : relative difference

$$\langle \mathbf{R}^{-1} e \rangle$$

$$\langle \mathbf{R}^{-1} \rangle \langle e \rangle$$

$$\langle \mathbf{R} \rangle^{-1} \langle e \rangle$$

$$\langle g \rangle = \langle R \rangle^{-1} \langle e \rangle$$

Other shear estimation methods : Moments

KSB (1995) :

- Weighted quadrupole moments (with iteratively adjusted W)
- Strong PSF assumptions
- Setup adjustment with data

DEIMOS (2011) :

- Same idea as KSB
- **Mathematically exact PSF deconvolution** (no assumption on galaxy or PSF profiles)

Other shear estimation methods : Model fitting

lensfit (2007) :

- 7-parameter galaxy model fit : galaxy position, flux, scale-length, bulge-to-disc ratio, galaxy ellipticity
- Bayesian method : each exposure fitted independently

HSM (2003) :

- Adaptive moments : Gaussian weight matched to the image

- Minimise $\rightarrow E = \frac{1}{2} \int_{\mathbf{R}^2} \left| I(\mathbf{x}) - A \underbrace{\exp \left[-\frac{1}{2} (\mathbf{x} - \mathbf{x}_0)^T \mathbf{M}^{-1} (\mathbf{x} - \mathbf{x}_0) \right]}_w \right|^2 d^2 \mathbf{x}$

- In practice : Gaussian model fitting

Metacalibration (Sheldon et al. 2017)

Image after shear application : $I(s) = P \otimes [\mathbf{s}(P^{-1} \otimes I)]$

with s the shear operator and P the atmospheric seeing + PSF + pixel response function.

To remove noise amplified by deconvolution, creation of a dilated PSF Γ : $\Gamma(x) = P((1 + 2|\gamma|)x)$ ← shear distortion

New sheared image : $I(s) = \Gamma \otimes [\mathbf{s} * (P^{-1} \otimes I)]$

→ This procedure introduces correlated anisotropic noise, which can lead to a systematic multiplicative bias.

Estimation method :

$$\langle \gamma \rangle \approx \langle \mathbf{R}_\gamma \rangle^{-1} \langle e \rangle \approx \langle \mathbf{R}_\gamma \rangle^{-1} \langle \mathbf{R}_\gamma \gamma \rangle.$$

$$\left\{ \begin{aligned} \langle e \rangle &\approx \int de \frac{\partial P(e)e}{\partial \gamma} \Big|_{\gamma=0} \gamma de = \langle \mathbf{R}_\gamma \gamma \rangle \\ \langle \mathbf{R}_\gamma \rangle &= \int \frac{\partial P(e)e}{\partial \gamma} \Big|_{\gamma=0} de \approx \int de \left(\frac{P^+ e_i^+ - P^- e_i^-}{\Delta \gamma_j} \right) de \\ &= \frac{\langle e_i^+ \rangle - \langle e_i^- \rangle}{\Delta \gamma_j}, \end{aligned} \right. \quad (10)$$

Shear application to seconds moments

Theoretical second moments (after shear application) :

$$M(S) = SMS^T = A^2 \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

$$S = \overset{= A}{\frac{1}{\sqrt{1-g^2}}} \begin{pmatrix} 1+g_1 & g_2 \\ g_2 & 1-g_1 \end{pmatrix}$$

If $g_1 \neq 0$ and $g_2 = 0$:

If $g_2 \neq 0$ and $g_1 = 0$:

$$a = (1 + g_1)^2 M_{xx}$$

$$b = (1 - g_1^2) * M_{xy}$$

$$c = (1 - g_1^2) * M_{yx}$$

$$d = (1 - g_1)^2 * M_{yy}$$

$$a = M_{xx} + g_2 * (M_{xy} + M_{yx} + g_2 * M_{yy})$$

$$b = M_{xy} + g_2 * (M_{xx} + M_{yy} + g_2 * M_{yx})$$

$$c = M_{yx} + g_2 * (M_{xx} + M_{yy} + g_2 * M_{xy})$$

$$d = M_{yy} + g_2 * (M_{xy} + M_{yx} + g_2 * M_{xx})$$

Pixel second moments calculation

$$M_{pix} = \int_{pixel} (\vec{X} - \vec{X}_c)(\vec{X} - \vec{X}_c)^T d^2\vec{X} / \int_{pixel} d^2\vec{X}$$

X_c : pixel center
M : Jacobian

$$\begin{aligned} (\vec{X} - \vec{X}_c) &= M(\vec{i} - \vec{i}_c) \\ d^2\vec{X} &= |\det(M)| d^2\vec{i} \end{aligned} \quad \longrightarrow \text{change of variable}$$

$$\begin{aligned} M_{pix} &= \int_{pixel} (\vec{X} - \vec{X}_c)(\vec{X} - \vec{X}_c)^T d^2\vec{X} / \int_{pixel} d^2\vec{X} \\ &= |\det(M)| \int_{pixel} M(\vec{i} - \vec{i}_c)(\vec{i} - \vec{i}_c)^T M^T d^2\vec{i} / |\det(M)| \int_{pixel} d^2\vec{i} \\ &= M \left[\int_{pixel} (\vec{i} - \vec{i}_c)(\vec{i} - \vec{i}_c)^T d^2\vec{i} \right] M^T \\ &= MM^T / 12 \end{aligned} \quad \curvearrowright \int_0^1 (x - 1/2)^2 dx = 1/12$$

Simulation parameters

- **General parameters :**
 - 40 random shear values (between -0.03 et 0.03)
 - Averaging each over 20 pairs of random and opposite ellipticities (between -0.3 et 0.3)
 - Pixel scale : 0.2 arcsec/pixel
- **Noisy simulations :**
 - Noise image simulated with a normal law (mean = 0)
 - SNR values between ~17 and ~35
 - Mean of each point over 10^4 realisations

Position estimation on images

Estimation of x_0 and y_0 by minimizing the implicit function :

$$f(x_0, I) = \frac{\sum_i (x_i - x_0) W(x_i - x_0) I_i}{\sum_i W(x_i - x_0) I_i}$$

→ Iterative calculation until reaching 10^{-4}

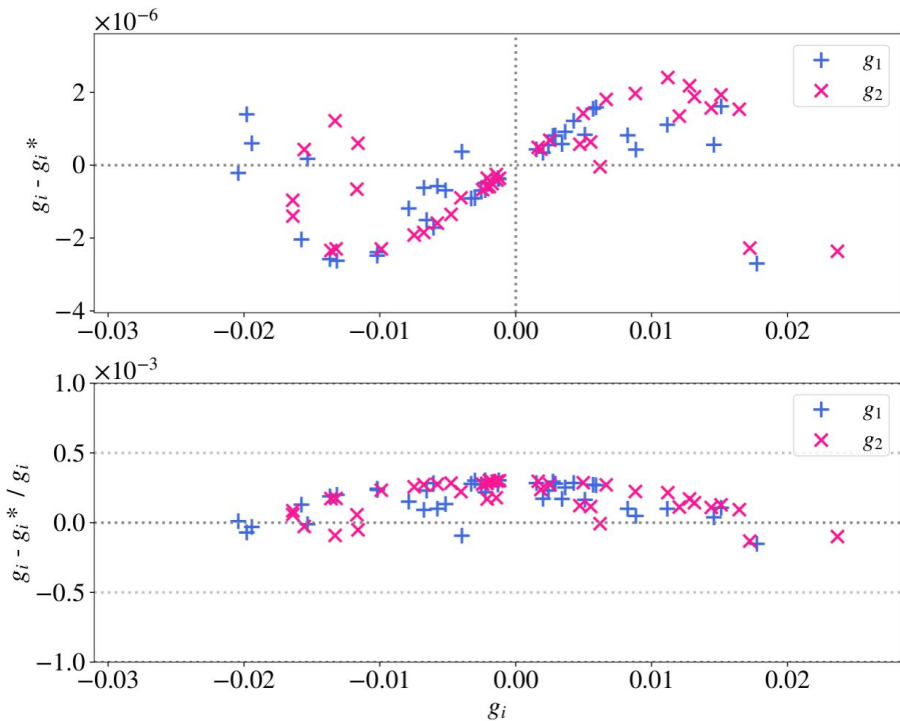
Gaussians galaxies

- Gaussian PSF
- TR = 1.5

$$M_{image} = \sum_{i,j}^N X_i X_j^T W_i [I_0 \otimes \psi]_i$$

$$TR = \frac{Tr(M_{image})}{Tr(M_{PSF})}$$

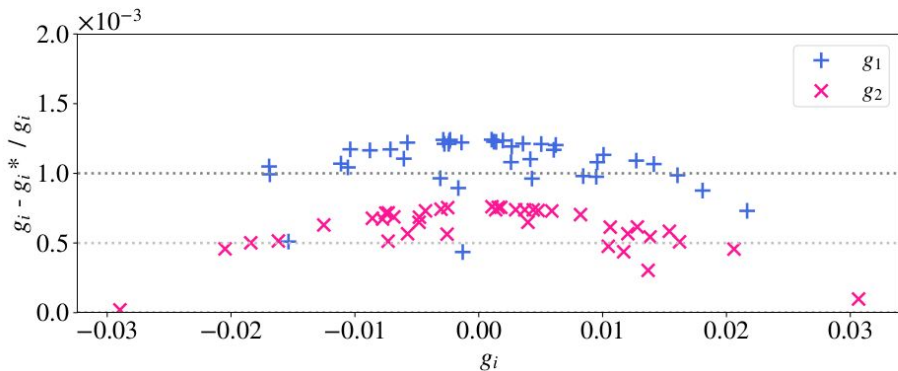
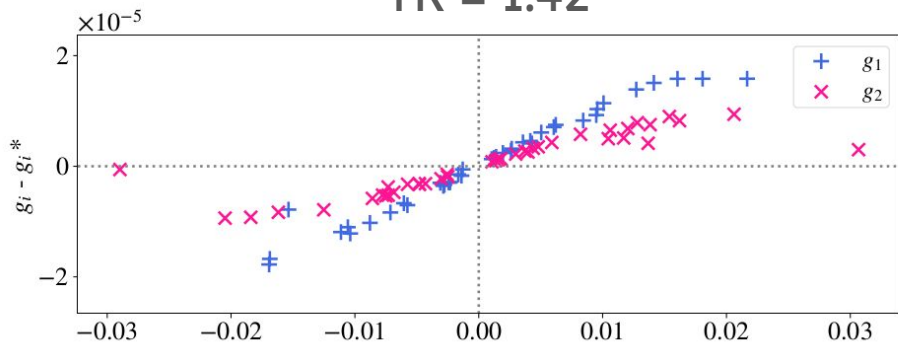
$$M_{PSF} = \sum_{i,j}^N X_i X_j^T W_i \psi_i$$



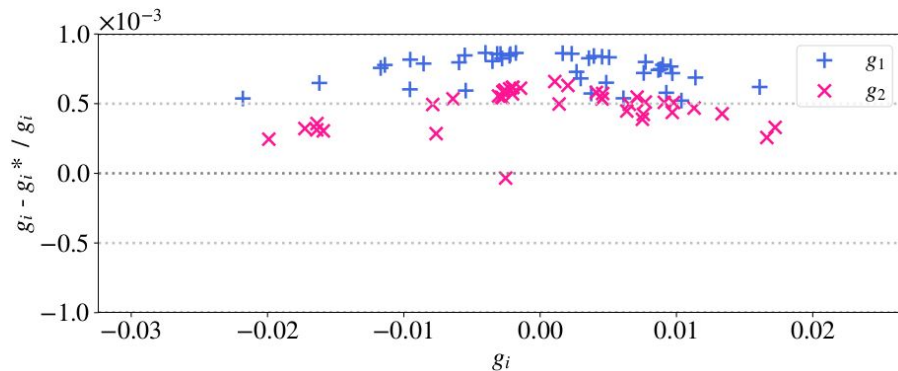
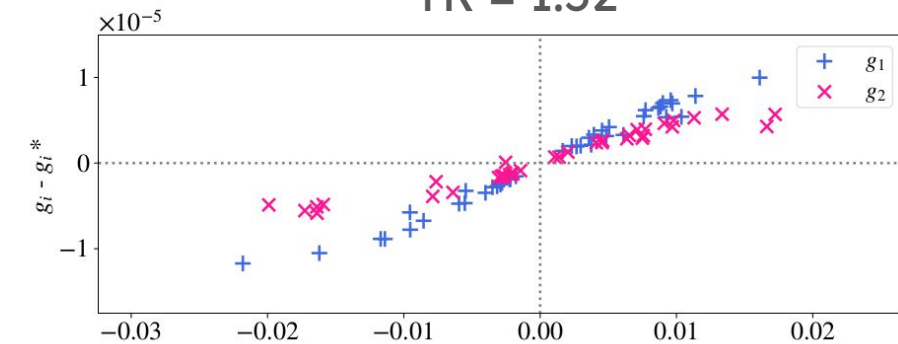
bias < 1%

Estimations : TR limits (Sersic n=1.0)

TR = 1.42



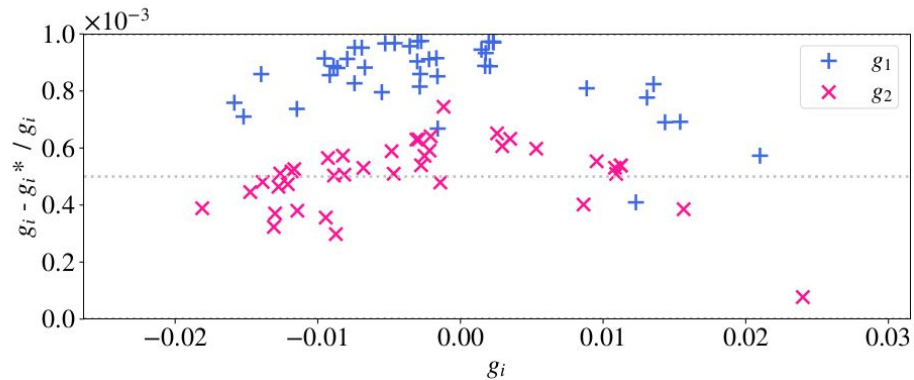
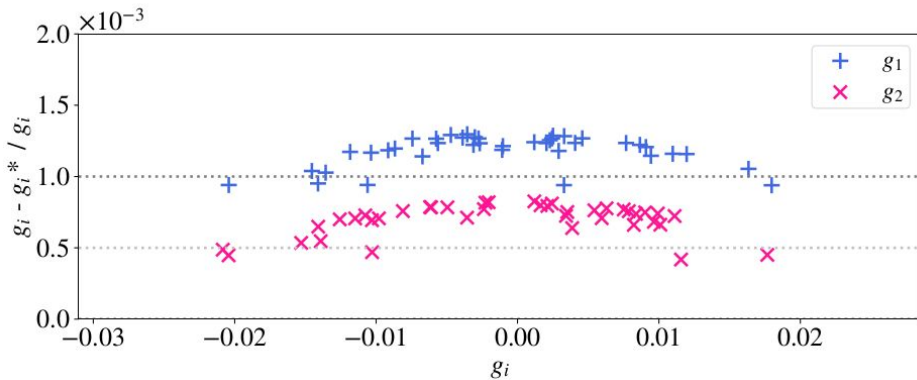
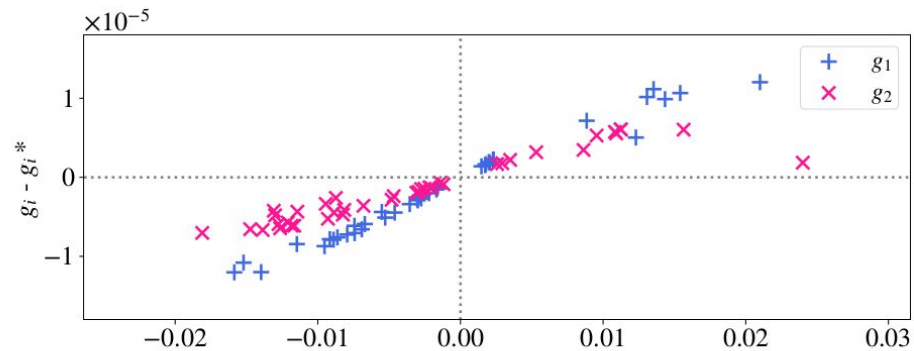
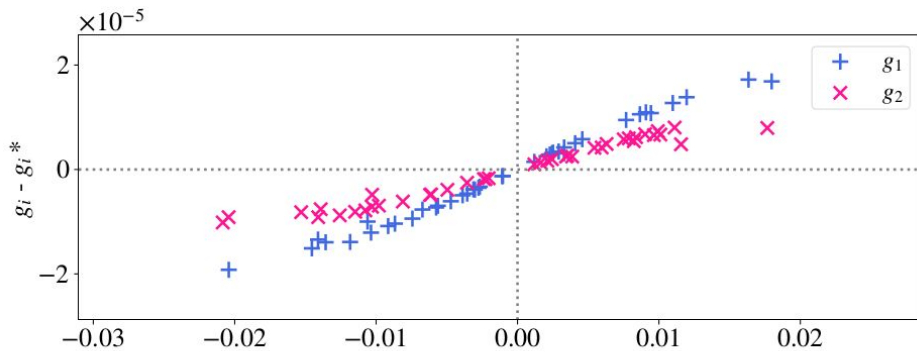
TR = 1.52



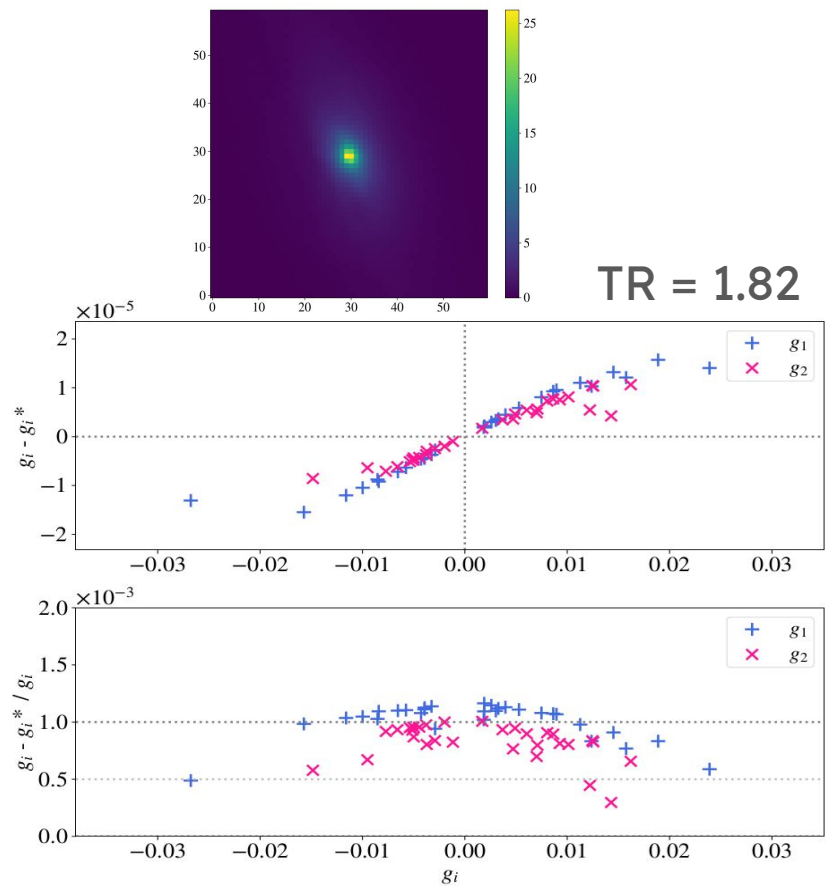
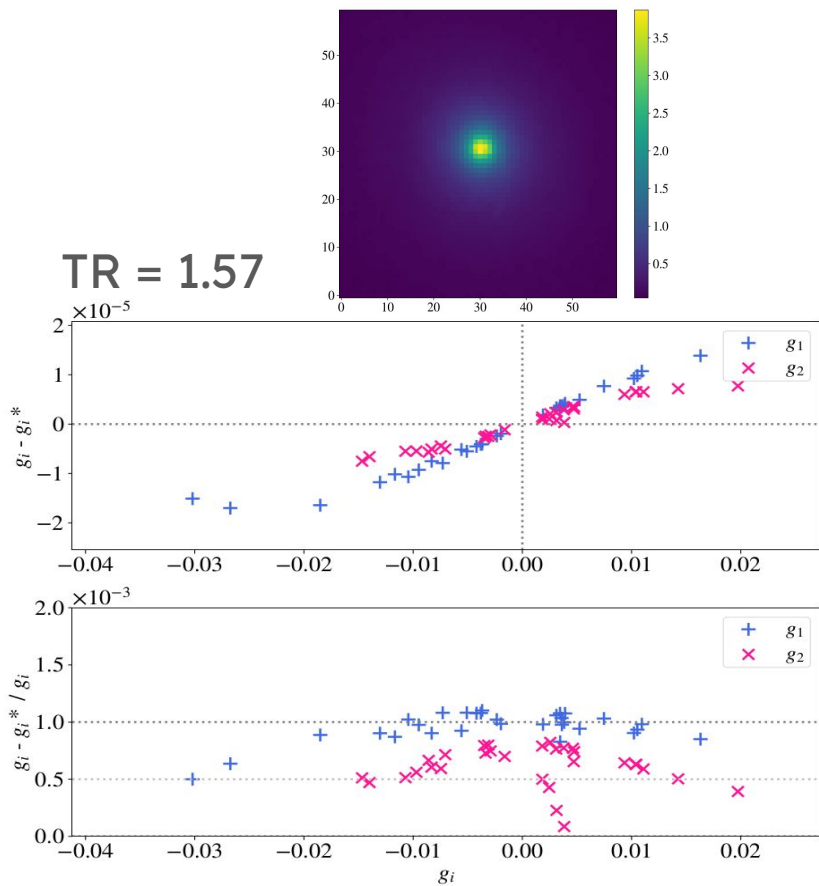
Estimations : TR limits (Sersic n=4.0)

TR = 1.54

TR = 1.63

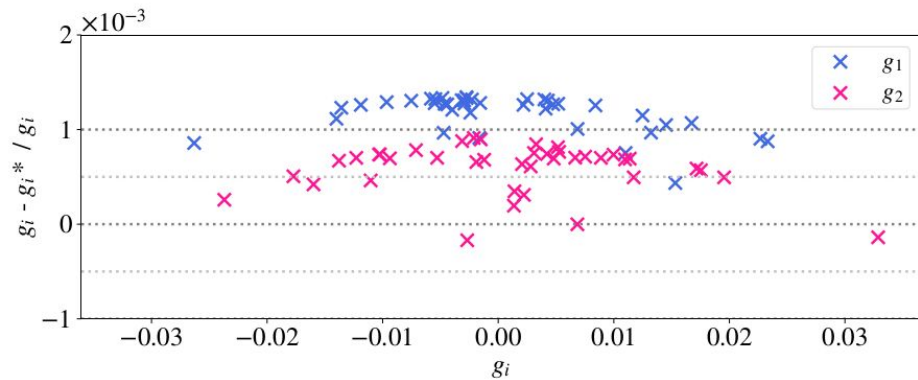
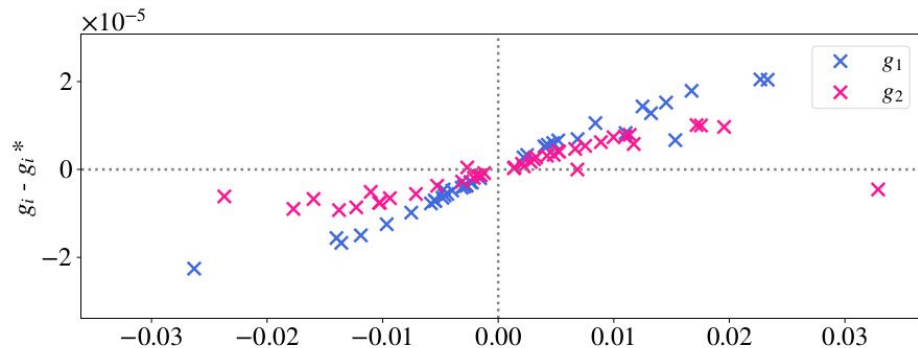


Estimations : TR limits (*COSMOS*)

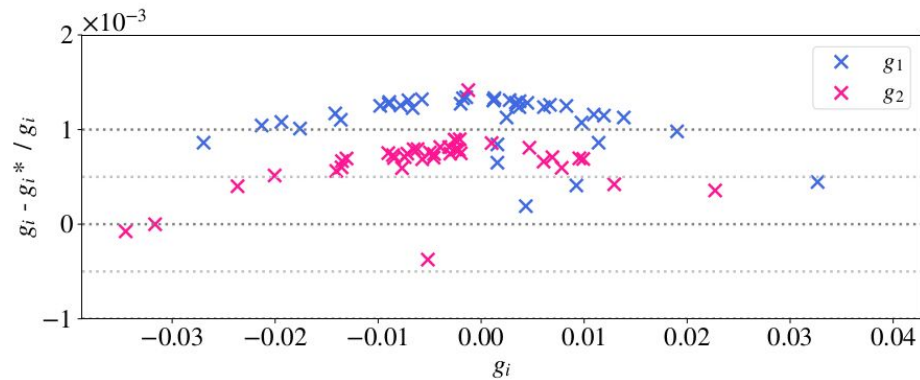
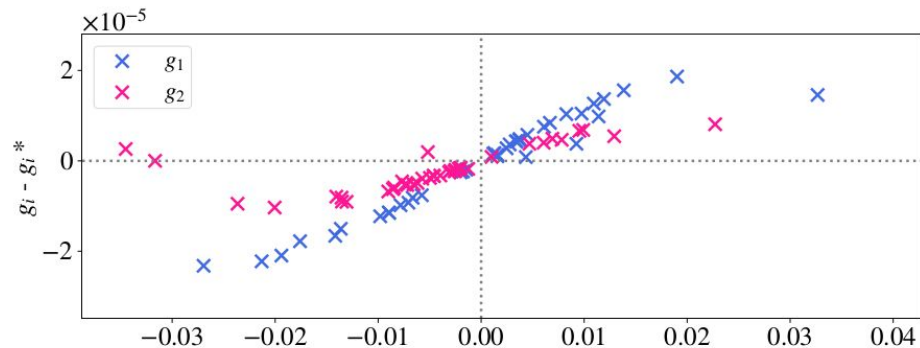


Sersic estimation : different stamp sizes (TR = 1.5)

N = 100



N = 60



Position variance calculation

$$f(x_0, I) = \sum (x_i - x_0) W (x_i - x_0) I_i$$

$$\frac{\partial x_0}{\partial I_i} = - \frac{\frac{\partial f}{\partial I_i}}{\frac{\partial f}{\partial x_0}} \left\{ \begin{array}{l} \frac{\partial f}{\partial I_i} = (x_i - x_0) W_i \\ \frac{\partial f}{\partial x_0} = - \sum W I + M_W^{-1} \sum (x_i - x_0)(x_i - x_0)^T W I \\ \quad = -F \mathbf{1} + M_W^{-1} M_P^* F \end{array} \right.$$

→ $\frac{\partial x_0}{\partial I_i} = \frac{1}{F} [\mathbf{1} - M_W^{-1} M_P]^{\dagger} (x_i - x_0) W_i$

When noise is added (ϵ) :

$$\begin{aligned} \delta x_0 &= \frac{\partial x_0}{\partial I_i} \epsilon_i \\ &= K (x_i - x_0) W_i \epsilon_i \end{aligned}$$

$$\begin{aligned} \sigma(x_0)^2 &= \sum \left(\frac{\partial x_0}{\partial \epsilon_i} \right)^2 \sigma(\epsilon_i)^2 \\ &= \sum \left(\frac{\partial x_0}{\partial I_i} \right)^2 \sigma_{noise}^2 \\ &= K^2 \sigma_{noise}^2 \sum (x_i - x_0)(x_i - x_0)^T W_i^2 \\ &= K^2 \sigma_{noise}^2 \frac{M_W}{2} \sum W_i^2 \end{aligned}$$

Analytical prediction of flux noise bias

$$F = \sum W(x - x_0)I$$

$$t_1 = \frac{1}{2} \frac{d^2 F}{dx_0 dx_0} V = -F \text{tr}(M_W^{-1} V_x) + \text{tr}(M_W^{-1} m_I M_W^{-1} V_x)$$

$$t_2 = \frac{d^2 F}{dx_0 dI_k} C = \text{tr}(M_W^{-1} K M_W^{-2})$$

Noise bias on unnormalized 2nd moment analytical calculation

$$m_I(I + n) = m_I(I) + \underbrace{\sum_k \left(\frac{dm_I}{dx_0} + \frac{dm_I}{dI_k} \right) n_k}_0 + \frac{1}{2} \sum_k \left(\underbrace{\frac{d^2m_I}{dx_0^2}}_{\substack{\downarrow \\ \text{4 terms}}} + 2 \underbrace{\frac{d^2m_I}{dx_0 dI_k}}_{\substack{\downarrow \\ \text{2 terms}}} + \underbrace{\frac{d^2m_I}{dI_k^2}}_0 \right) n_k^2$$

$$t1 = 2FV_x$$

$$t2 = -2(V_x M_W^{-1} m_i)$$

$$t3 = t2$$

$$t4 = -m_i \text{Tr}(V_x M_W^{-1}) + M_{4i}(M_W^{-1} V_x M_W^{-1})$$

$$t5 = -4K M_W^2$$

$$t6 = M_{4W^2}(K M_W^{-1})$$

Noise bias on normalized 2nd moment analytical calculation

$$m_I F^{-1}(I+n) = m_I F^{-1}(I) + \underbrace{\sum_k \frac{d(m_I F^{-1})}{dI_k} n_k}_{0} + \frac{1}{2} \sum_{kl} \frac{d^2(m_I F^{-1})}{dI_k dI_l} n_k n_l$$



5 terms

$$t1 = B[m_I]/F$$

$$t2 = -B[F]m_I/F^2$$

$$t3 = t4 = -2M(W^2)/F^2$$

$$t5 = 2(m_I/F^3) \sum W_k^2$$

$B[m_I]$ = noise bias on unnormalized second moments

$B[F]$ = noise bias on flux

Auto calibration factor

$$\mathbf{R} = \frac{1}{2\epsilon} \begin{pmatrix} R_{11} & R_{12} \\ R_{21} & R_{22} \end{pmatrix}$$

$$R_{11} = (M_{xx}^{1+} - M_{xx}^{1-}) - (M_{yy}^{1+} - M_{yy}^{1-})$$

$$R_{12} = 2(M_{xy}^{1+} - M_{xy}^{1-})$$

$$R_{21} = (M_{xx}^{2+} - M_{xx}^{2-}) - (M_{yy}^{2+} - M_{yy}^{2-})$$

$$R_{22} = 2(M_{xy}^{2+} - M_{xy}^{2-})$$