

# Optimized Cosmic Shear Statistics

Tim Eifler, AlfA University Bonn  
Martin Kilbinger, IAP Paris  
Peter Schneider, AlfA University Bonn

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## Goal

Find an improved cosmic shear data vector

- high information content  $\rightarrow$  tight constraints on cosmological parameters
- robust against contamination of the signal (B-modes)
- small correlation between data points of different angular scales (covariance matrices)

## Two-point correlation function

$$\xi = (\xi_+(\vartheta_1), \dots, \xi_+(\vartheta_m), \xi_-(\vartheta_1), \dots, \xi_-(\vartheta_m))$$

## Aperture mass dispersion

$$\langle \mathbf{M}_{\text{ap}}^2 \rangle = (\langle M_{\text{ap}}^2 \rangle(\theta_1), \dots, \langle M_{\text{ap}}^2 \rangle(\theta_n))$$

## Two-point correlation function (2PCF)

$$\xi_{\pm}(\vartheta) = \langle \gamma_t \gamma_t \rangle(\vartheta) \pm \langle \gamma_{\times} \gamma_{\times} \rangle(\vartheta)$$

## Relation to the power spectrum $P_{\kappa} = P_E + P_B$

$$\begin{aligned}\xi_{+}(\vartheta) &= \frac{1}{2\pi} \int_0^{\infty} d\ell \ell J_0(\ell\vartheta) [P_E(\ell) + P_B(\ell)] \\ \xi_{-}(\vartheta) &= \frac{1}{2\pi} \int_0^{\infty} d\ell \ell J_4(\ell\vartheta) [P_E(\ell) - P_B(\ell)]\end{aligned}$$

## Important

$\xi_{\pm}$  are filtered versions of the power spectrum  $P_{\kappa}$ . The filter functions are the Bessel functions  $J_0$  and  $J_4$

## Aperture mass dispersion ( $\langle M_{\text{ap}}^2 \rangle$ )

$$\langle M_{\text{ap}}^2 \rangle(\theta) = \frac{1}{2\pi} \int_0^\infty d\ell \ell P_E(\ell) W_{\text{ap}}(\theta\ell)$$

$$\langle M_{\perp}^2 \rangle(\theta) = \frac{1}{2\pi} \int_0^\infty d\ell \ell P_B(\ell) W_{\text{ap}}(\theta\ell)$$

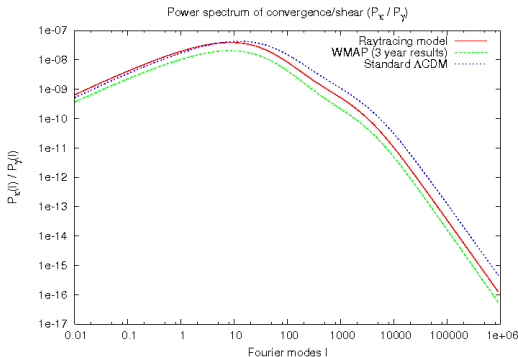
## Important

- 1  $\langle M_{\text{ap}}^2 \rangle$  is also a filtered version of the power spectrum  $P_\kappa$
- 2  $\langle M_{\text{ap}}^2 \rangle$  can be calculated from the 2pcf  $\rightarrow$  2pcf can be seen as the basic quantity

$$\langle M_{\text{ap}}^2 \rangle(\theta) = \int_0^{2\theta} \frac{d\vartheta}{\theta^2} \frac{\vartheta}{2} \left[ \xi_+(\vartheta) T_+ \left( \frac{\vartheta}{\theta} \right) + \xi_-(\vartheta) T_- \left( \frac{\vartheta}{\theta} \right) \right]$$

$\xi_{\pm}$

- very broad filter  $\rightarrow \xi_{+}$  probes a wide range of  $P_{\kappa}$
- includes information on angular scales larger than the size of the survey



$\langle M_{ap}^2 \rangle$

- narrowest filter function
- gives highly localized measure of  $P_{\kappa}$
- low correlation between data points of different scales
- no large scale information of the power spectrum due to its narrow filter
- difficult to measure from data field due to gaps, holes, stars

## Combined data vector

idea: add one value of  $\xi_+(\theta_0)$  to a  $\langle M_{\text{ap}}^2 \rangle$  data vector to include the large scale information of  $P_{\kappa}$

$$\mathcal{N} = (\langle M_{\text{ap}}^2 \rangle(\theta_1), \dots, \langle M_{\text{ap}}^2 \rangle(\theta_n), \xi_+(\theta_0))$$

## Two-point correlation function

$$\xi = (\xi_+(\vartheta_1), \dots, \xi_+(\vartheta_m), \xi_-(\vartheta_1), \dots, \xi_-(\vartheta_m))$$

## Aperture mass dispersion

$$\langle \mathbf{M}_{\text{ap}}^2 \rangle = (\langle M_{\text{ap}}^2 \rangle(\theta_1), \dots, \langle M_{\text{ap}}^2 \rangle(\theta_n))$$

## Method

Information content of a cosmic shear data vector → ability of constraining cosmological parameters → Bayesian likelihood analysis

## Bayes theorem - cosmic shear data

$$P(\boldsymbol{\pi}|\boldsymbol{\xi}_{\pm}, \Lambda\text{CDM}) = \frac{P(\boldsymbol{\xi}_{\pm}|\boldsymbol{\pi}, \Lambda\text{CDM}) P(\boldsymbol{\pi}|\Lambda\text{CDM})}{P(\boldsymbol{\xi}_{\pm}|\Lambda\text{CDM})}$$

## Likelihood

$$P(\boldsymbol{\xi}|\boldsymbol{\pi}, \Lambda\text{CDM}) = \frac{1}{(2\pi)^{n/2} \sqrt{\det \mathbf{C}_{\xi}}} \exp \left[ -\frac{1}{2} (\boldsymbol{\xi}(\boldsymbol{\pi}) - \boldsymbol{\xi}^f)^t \mathbf{C}_{\xi}^{-1} (\boldsymbol{\xi}(\boldsymbol{\pi}) - \boldsymbol{\xi}^f) \right]$$

## Marginalization

$$P(\boldsymbol{\pi}_{12}|\boldsymbol{\xi}_{\pm}, \Lambda\text{CDM}) = \int d\pi_3 \int d\pi_4 P(\boldsymbol{\pi}_{1234}|\boldsymbol{\xi}_{\pm}, \Lambda\text{CDM})$$



## First guess

$$\hat{\mathbf{C}}_*^{-1} = \left( \hat{\mathbf{C}}^{ML} \right)^{-1}$$

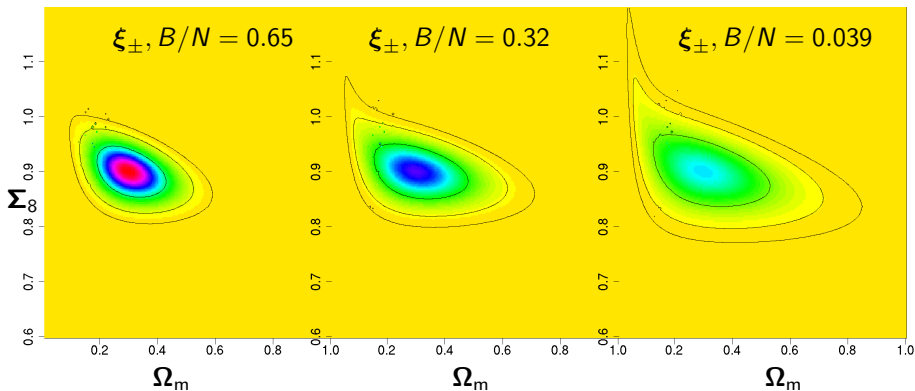
- estimator is consistent but it is *biased* due to noise in  $\hat{\mathbf{C}}$ .
- only linear transformations preserve “unbiasedness”

The amount of bias/size of the likelihood contours vary dependent on the relation

$$\frac{\text{number of bins (B)}}{\text{number of independent realizations (N)}}$$

- more realisations  $\rightarrow$  larger contours
- more bins  $\rightarrow$  smaller contours
- for  $B \geq N - 2$  the covariance matrix becomes singular

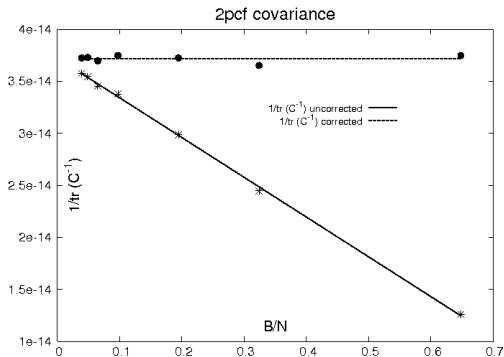
# Without correction factor...one example



Correction-Factor ( Hartlap et al. 2006; Anderson 2003)

An unbiased estimator for the inverted covariance is given by

$$\hat{\mathbf{C}}^{-1} = \frac{N - B - 2}{N - 1} \hat{\mathbf{C}}_*^{-1} \quad \text{for } B < N - 2$$

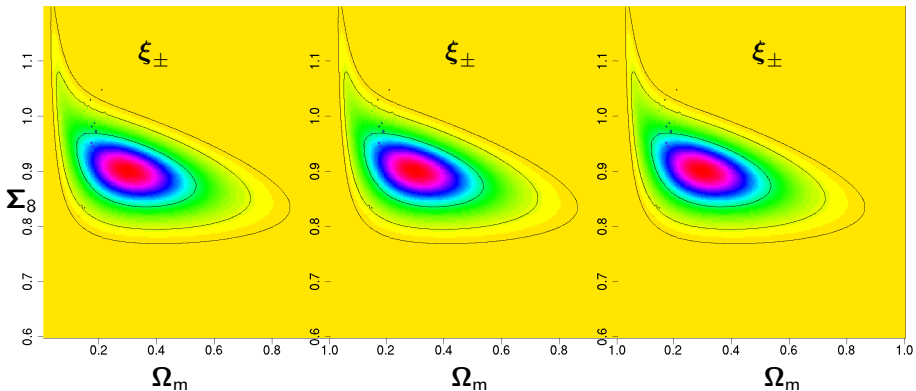


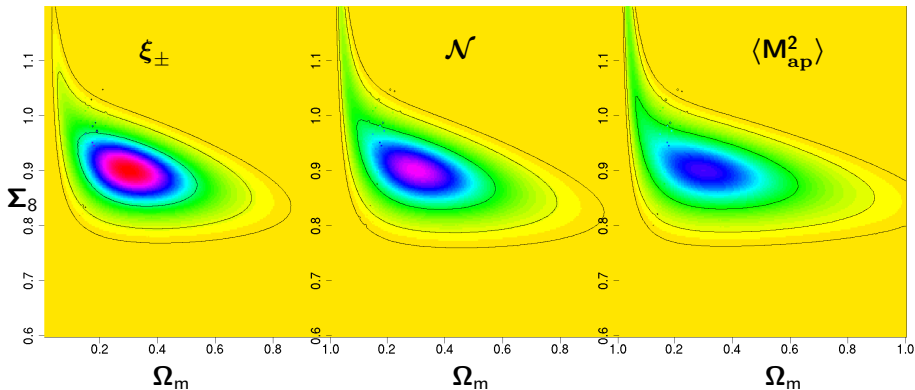
- ray-tracing simulations provide 36 independent realisations
- multiply the number of independent realisations by adding different Gaussian noise to the galaxy ellipticities ( $N = 108, 216, 360, 720, 1080, 1440, 1800$ )

- estimate the covariance from every sample
- consider the trace of the inverted covariance depending on the ratio bins/realisations

# Now it works...

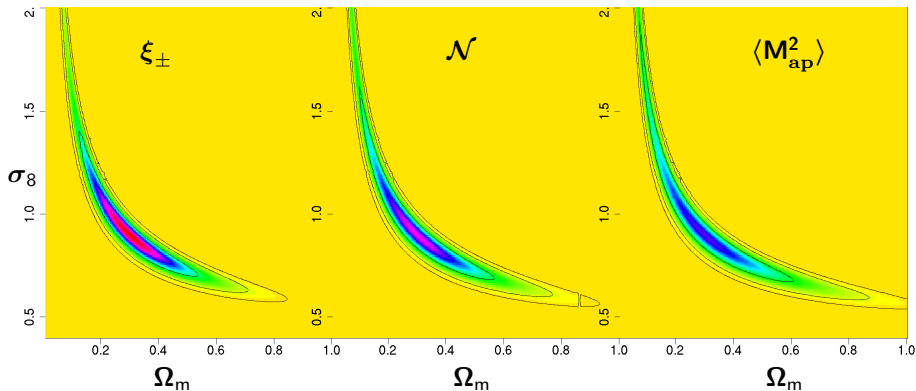
- 1 correction factor = 0.34
- 2 correction factor = 0.67
- 3 correction factor = 0.96





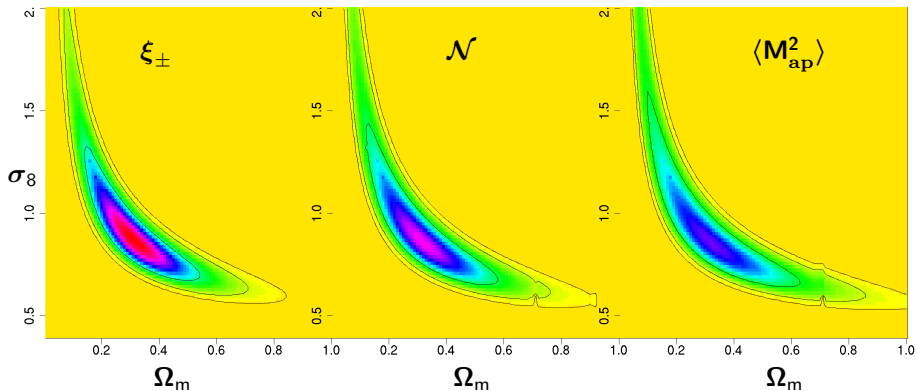
- 2pcf:  $\vartheta$ -range=0.2' – 199', 35 bins each for  $\xi_{\pm}$
- $\mathcal{N}$ : added data point  $\xi_+(5')$ , 21 bins
- $\langle M_{\text{ap}}^2 \rangle$ :  $\theta$ -range=2.2' – 99', 20 bins

# More results



parameter space	$\langle M_{\text{ap}}^2 \rangle$	$\mathcal{N}$	$\xi$	$\Delta \mathcal{N}$	$\Delta \xi$
$\Gamma$ vs. $\Omega_m$	14.7	11.7	9.1	20.4 %	38.1 %
$\sigma_8$ vs. $\Gamma$	23.1	19.0	14.6	17.8 %	36.8 %
$\sigma_8$ vs. $\Omega_m$	427.1	314.5	220.1	26.4 %	48.5 %
$z_0$ vs. $\Omega_m$	46.4	41.0	32.9	11.6 %	29.1 %

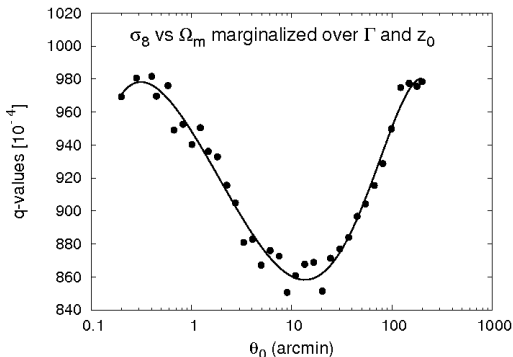
# Marginalized over $z_0$



parameter space	$\langle M_{\text{ap}}^2 \rangle$	$\mathcal{N}$	$\xi$	$\Delta \mathcal{N}$	$\Delta \xi$
$\sigma_8$ vs. $\Omega_m$ ( $z_0$ )	416.9	313.4	230.0	25.8 %	44.8 %
$\sigma_8$ vs. $\Omega_m$ ( $\Gamma$ )	780.5	720.9	527.0	7.6 %	32.5 %
$\sigma_8$ vs. $\Omega_m$ ( $\Gamma, z_0$ )	983.8	850.6	623.5	13.5 %	36.6 %

$$Q_{ij} \equiv \frac{\int d^2\pi P(\pi_1, \pi_2)(\pi_i - \pi_i^f)(\pi_j - \pi_j^f)}{\int d^2\pi P(\pi_1, \pi_2)}$$

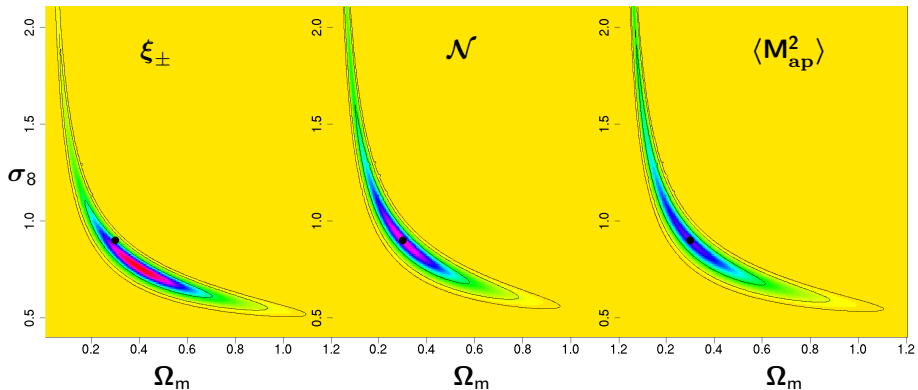
$$q = \sqrt{\det Q_{ij}}$$
$$= \sqrt{Q_{11}Q_{22} - Q_{12}^2}$$



Idea: Vary the added  $\xi_+(\theta_0)$  in  $\mathcal{N}$  in order to optimize the information content



# Contamination with B-modes



- The likelihood maximum of the 2PCF data vector is far away from the “true” cosmological parameters
- The combined data vector is hardly affected by the contamination
- It still gives better constraints on cosmological parameters than the aperture mass dispersion

## The combined data vector

- The new data vector  $\mathcal{N}$  is a strong improvement in information content compared to  $\langle M_{\text{ap}}^2 \rangle$
- $\mathcal{N}$  can be optimized by varying  $\xi_+(\theta_0)$
- its covariance matrix is much more diagonal compared with the 2PCF  $\rightarrow$  more robust against numerical problems during the inversion
- it is hardly affected by a B-mode contamination on small angular scales

## Future work

- Improve information content further by looking into other cosmic shear statistics (e.g. ring statistics)
- higher order statistics
- Find better estimates for covariances (robust, unbiased)

## Data vectors and covariances

- calculate power spectrum  $P_\delta$  for our fiducial model according to Smith et al.2003
- calculate  $P_\kappa$  and the data vectors  $\xi^f, \langle \mathbf{M}_{\text{ap}}^2 \rangle^f, \mathcal{N}^f$
- vary parameters and recalculate the data vectors  $\xi(\pi), \langle \mathbf{M}_{\text{ap}}^2 \rangle(\pi), \mathcal{N}(\pi)$  for every variation
- derive the covariance matrix  $\mathbf{C}_\xi$  by field to field variation from ray-tracing simulations (Jenkins et al. 2001 for simulation details and Hamana & Mellier 2001 for details of the ray-tracing algorithm)
- calculate therefrom the covariances of  $\langle \mathbf{M}_{\text{ap}}^2 \rangle$  and  $\mathcal{N}$

## Fiducial model

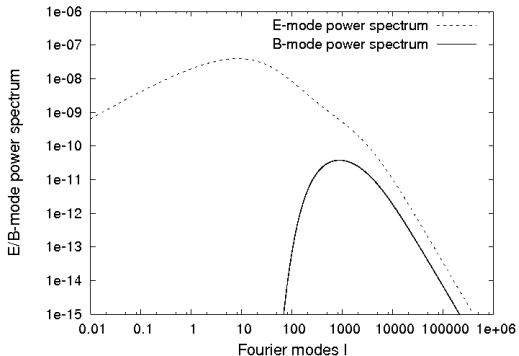
$\Omega_m$	$\Omega_\Lambda$	$h$	$\Gamma$	$\sigma_8$	$\Omega_b$
0.3	0.7	0.7	0.1723	0.9	0.04

## Recall

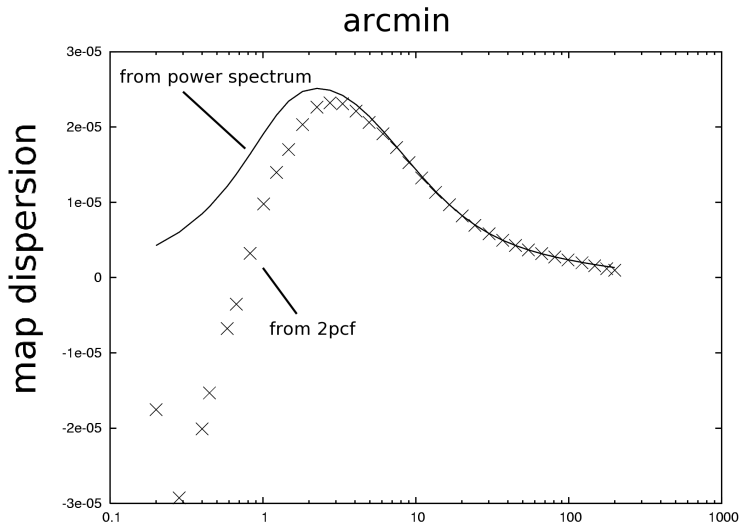
$$\langle \Delta P_E(\bar{\ell}) \Delta P_B(\bar{\ell}') \rangle = 0 \rightarrow \mathbf{C}_{tot} = \mathbf{C}_E + \mathbf{C}_B$$

$$\begin{aligned} \langle \Delta \xi_{\pm}(\theta_1) \Delta \xi_{\pm}(\theta_2) \rangle &= \int_0^{\infty} \frac{d\ell \ell}{\pi A} J_{0/4}(\ell\theta_1) J_{0/4}(\ell\theta_2) \\ &\quad \times \left\{ \left( P_E(\ell) + \frac{\sigma_{\epsilon}^2}{2n} \right)^2 + \left( P_B(\ell) + \frac{\sigma_{\epsilon}^2}{2n} \right)^2 \right\} \\ \langle \Delta \xi_{+}(\theta_1) \Delta \xi_{-}(\theta_2) \rangle &= \int_0^{\infty} \frac{d\ell \ell}{\pi A} J_0(\ell\theta_1) J_4(\ell\theta_2) \\ &\quad \times \left\{ \left( P_E(\ell) + \frac{\sigma_{\epsilon}^2}{2n} \right)^2 - \left( P_B(\ell) + \frac{\sigma_{\epsilon}^2}{2n} \right)^2 \right\} \end{aligned}$$

- currently there is no model available for B-modes
- we know, B-modes occur mainly on small scales
- Our model:  
$$P_B = 0.2 P_E \exp\left(\frac{-\ell_B}{\ell}\right)$$



For small  $\theta$ -values  $\text{Cov}_{\mathcal{M}}$  cannot be calculated properly from the  $\text{Cov}_{\xi}$   $\rightarrow$   
data vectors  $\mathcal{M}$  and  $\mathcal{N}$  start at a given  $\theta_{\min}$



- Recalculate all data vectors taking  $P_B$  into account
- Calculate the covariance matrix for the correlation function where
  - 1  $\mathbf{C}_E$  is taken from the ray-tracing simulations
  - 2  $\mathbf{C}_B$  is calculated directly from  $P_B$
- Carefull:  $\mathbf{C}_E$  needs correction factor for the inversion,  $\mathbf{C}_B$  does not  $\rightarrow$

$$\begin{aligned}\mathbf{C}_{\text{tot}}^{-1} &= (\mathbf{C}_E + \mathbf{C}_B)^{-1} \\ &= \mathbf{C}_E^{-1} - \frac{1}{\text{trace}(\mathbf{C}_E^{-1}\mathbf{C}_B)} \mathbf{C}_B \mathbf{C}_E^{-1} \mathbf{C}_B\end{aligned}$$

- calculate covariances for the two other data vectors from  $\mathbf{C}_\xi$
- calculate the posterior likelihood and plot contours