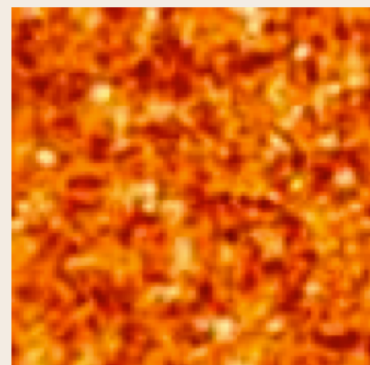
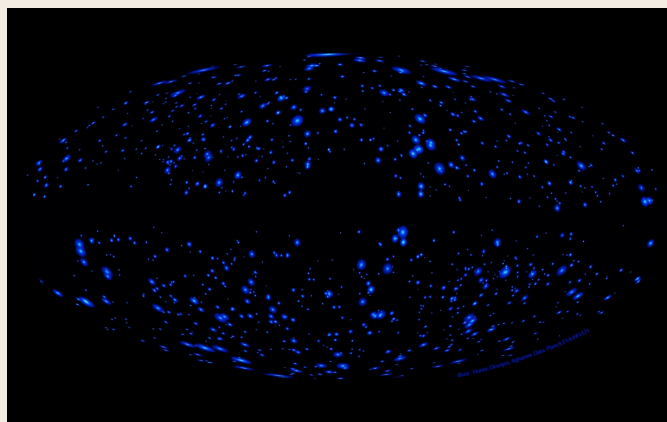


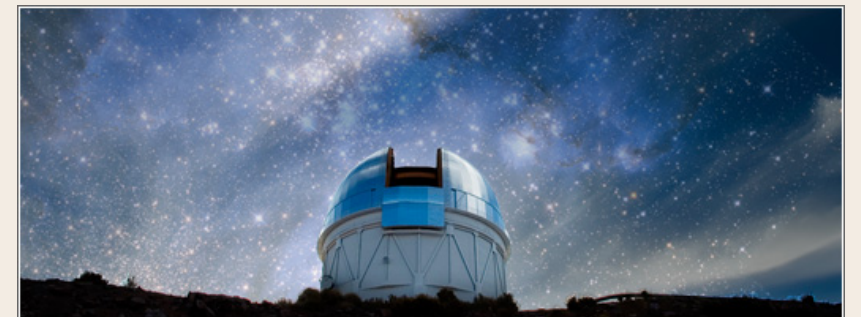
Non-Gaussian tools for the Large Scale Structure



k_1



k_2



k_3



Fabien Lacasa
São Paulo, Brasil



Outline

I. Non-Gaussianity with *Planck*

- A. Primordial NG and point-sources
- B. Cosmic Infrared Background
- C. thermal Sunyaev-Zel'dovich effect

II. Combining probes for *DES*

- A. Covariance of galaxy spectrum and cluster counts
- B. Joint likelihood

Planck non-Gaussianity

A. Primordial NG and point-sources

- primary CMB NG constrains inflations models
- NG of most models is well described as linear combination of 3 shapes : local, equilateral and orthogonal
- *Planck* puts quasi-optimal constraints on these amplitudes in temperature.

Polarization information further decreases error bars.

Constraints after subtraction of the lensing- ISW bias		T 2013	T+E 2014 (preliminary)
	$f_{\text{NL}}^{\text{local}}$	2.7 ± 5.8	0.71 ± 5.1
	$f_{\text{NL}}^{\text{equilateral}}$	-42 ± 75	-9.5 ± 44
	$f_{\text{NL}}^{\text{orthogonal}}$	-25 ± 39	-25 ± 22

A. Contamination of PNG estimation

- estimation of f_{NL} is biased by contaminants :
 - secondary anisotropies : lensing, iSW, SZ
 - Galactic emission : synchrotron, free-free, AME, dust
 - extragalactic point-sources :
 - radio sources
 - Cosmic Infrared Background

radio sources can be considered
as *white-noise* :
unclustered, simple NG given by counts and
flux cut

CIB is more complex :
strongly clustered enhancement of NG on
large scales
→ phenomenological prescription
by **Lacasa et al. 2011**

A. Point-source NG

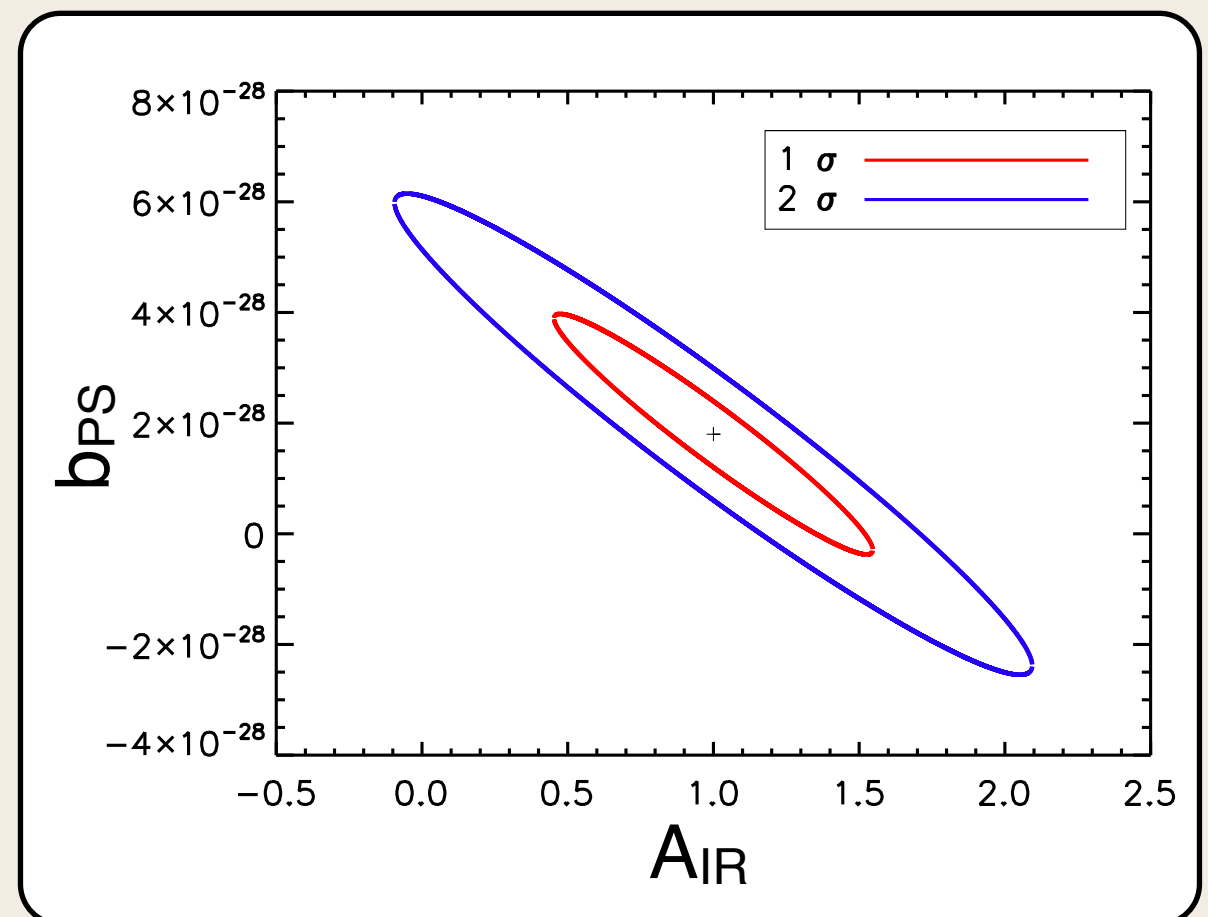
- contamination of $f_{\text{NL}}^{\text{local}}$ after masking bright sources

ν [GHz]	143	217	353
$\Delta f_{\text{NL}}^{\text{RAD}}$	0.269	0.089	0.046
$\Delta f_{\text{NL}}^{\text{IR}}$	0.006	0.108	19.3

- optimal estimator for A_{IR}
- joint estimation of f_{NL} , b_{PS} and A_{IR} with a coupling matrix

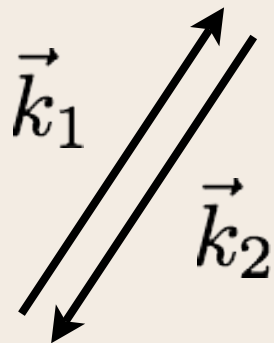
forecasts of $(b_{\text{PS}}, A_{\text{IR}})$
joint constraints at
217 GHz :
ellipses at 1σ and 2σ

Lacasa & Aghanim 2013



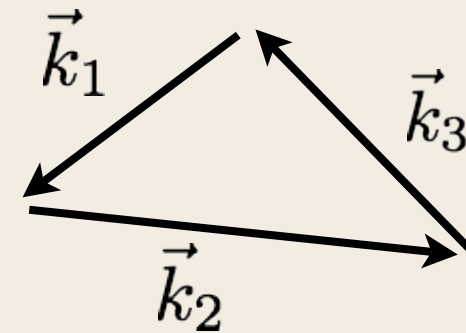
B. Interlude I : how to quantify NG

Power spectrum : 2-point correlation function in Fourier space



$P(k)$

Bispectrum : 3-point correlation function in Fourier space



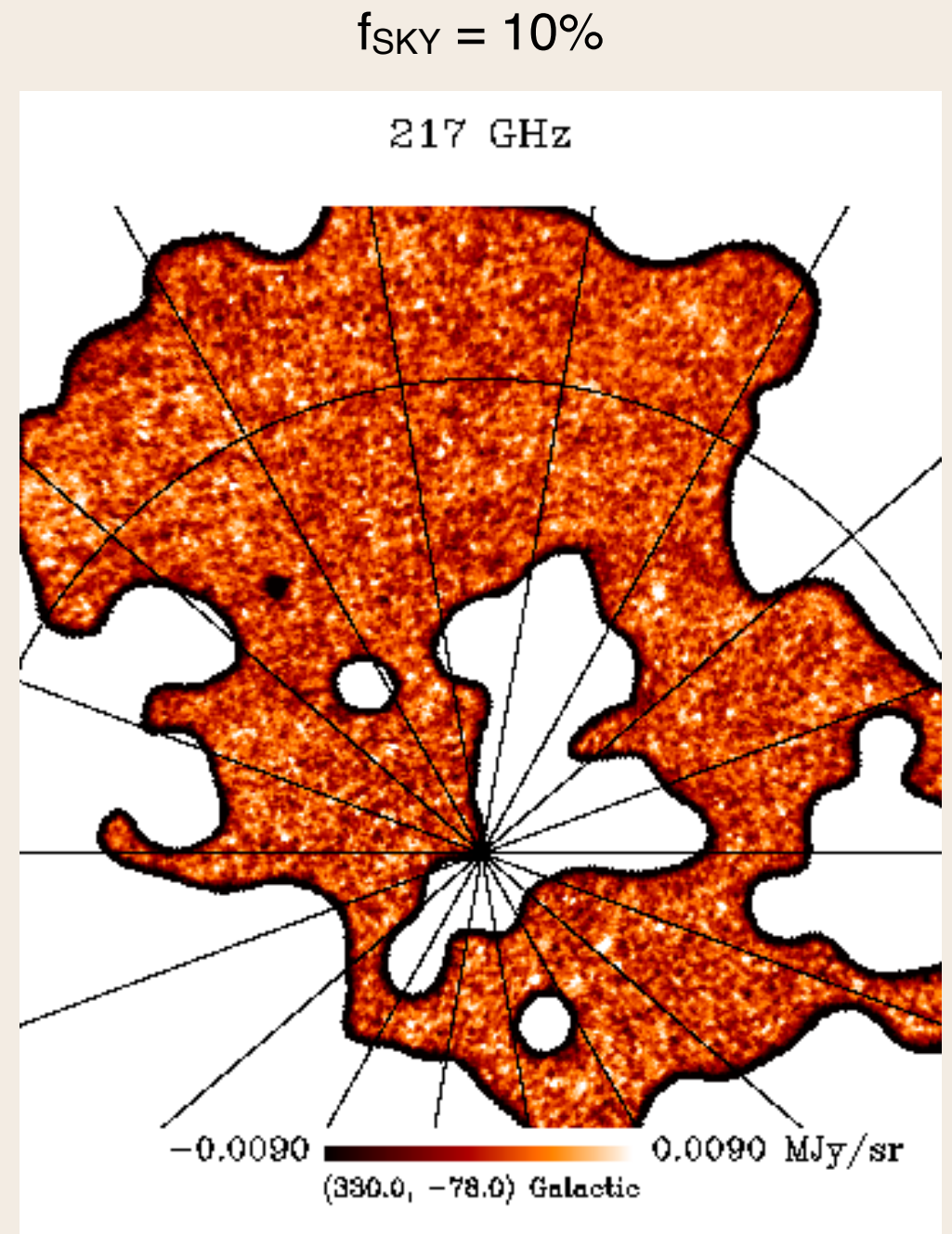
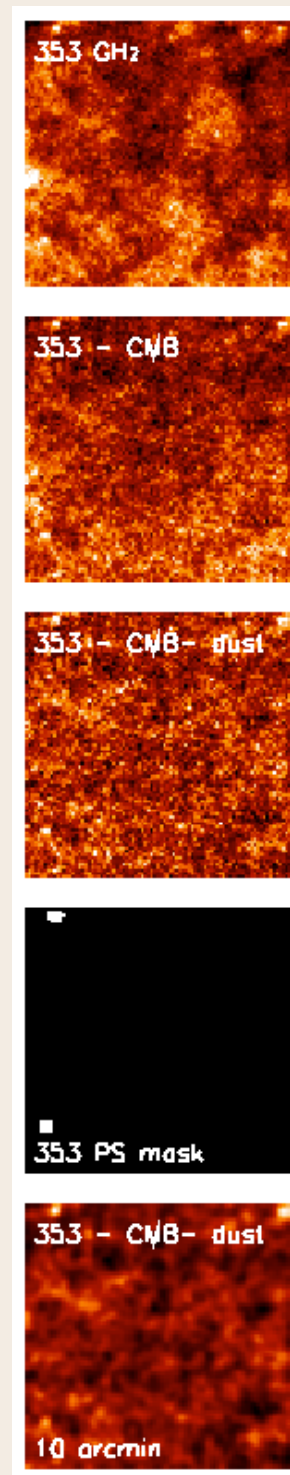
$B(k_1, k_2, k_3)$

Bispectrum : lowest order indicator of NG

Important triangle shapes :
equilateral, flat and mostly squeezed ($k_1 \ll k_2, k_3$)

B. Cosmic Infrared Background : *Planck* maps cleaning

- original map
- CMB subtraction
(Wiener filter of the 100 GHz map)
- Galactic dust subtraction
(model based on external HI data)
- Galactic and point-sources mask
- estimated CIB map



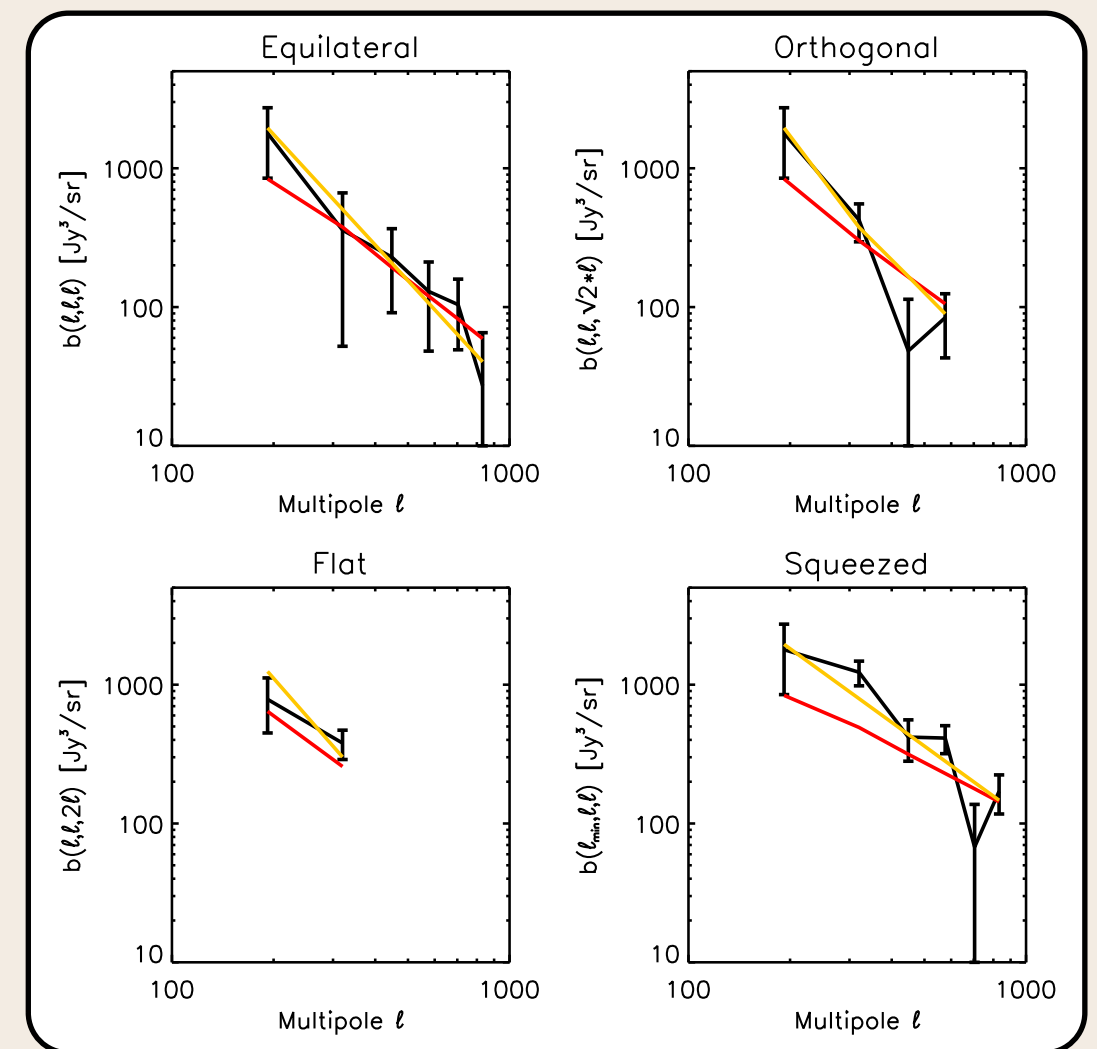
B. CIB : *Planck* bispectrum

- all configurations/triangles
- mask effect : optimisation (linear term) and debiasing (by simulations)
- correction of the tSZ leakage (due to 100 GHz)

ν [GHz]	SNR moyen par config	SNR total
217	1.24	5.83
353	2.85	19.27
545	4.59	28.72

**First detection of the CIB
bispectrum per configuration
and per frequency**

353 GHz



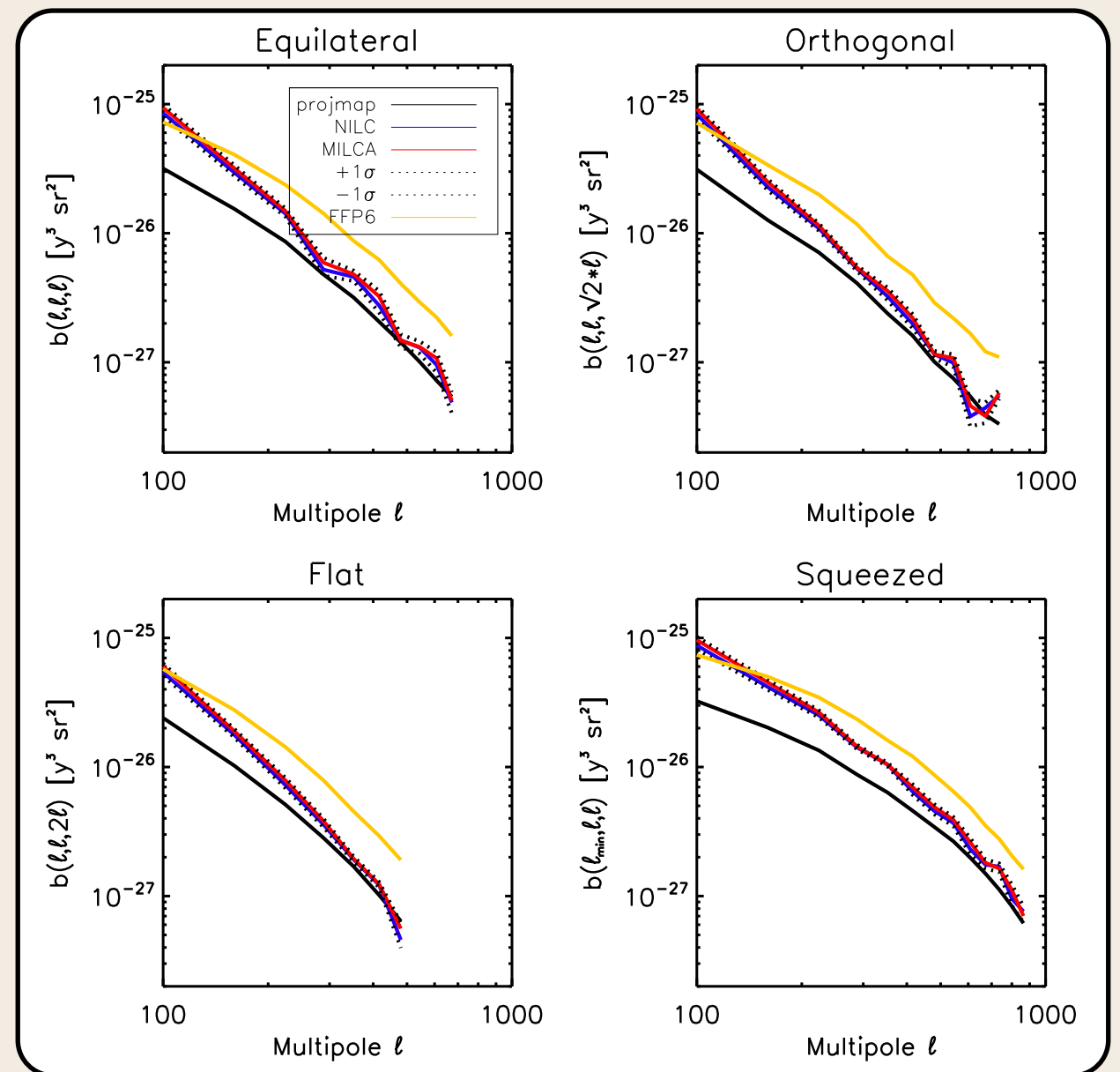
— data — prescription — power law

B. CIB bispectrum : teaching and perspectives

- frequency dependence consistent with spectrum
- high S/N → information there to extract
- data fit better by **power law** than by **prescription** or by previous model (from Lacasa et al. 2013, Pénin et al. 2013)
- necessity of a new model
cf Aurélie Pénin's talk

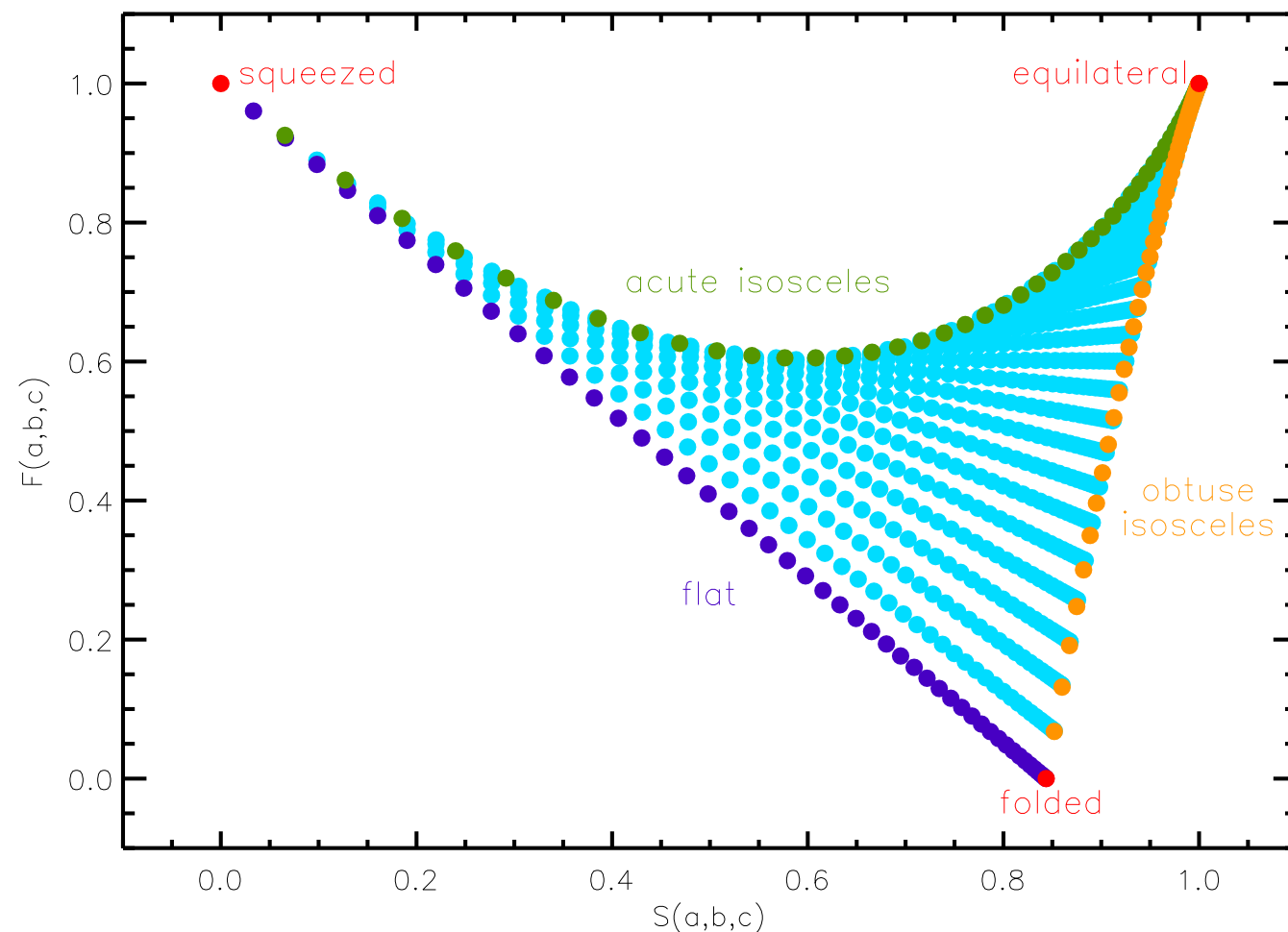
C. thermal Sunyaev-Zel'dovich NG

- tSZ traces ionised gas in clusters
- *Planck* can map it.
2 methods used : **NILC** and **MILCA** (resp. Remazeilles et al. 2011 and Hurier et al. 2013)
- Bispectrum estimated with ~60% of the sky
- Correction of the mask effect with a mean ratio computed on simulations



— NILC — MILCA error bars
— simulations — projected map of detected clusters

C. Interlude II : parametrisation to plot a bispectrum



All triangles of a given
perimeter

(in a parameter space related to
symmetric polynomials)

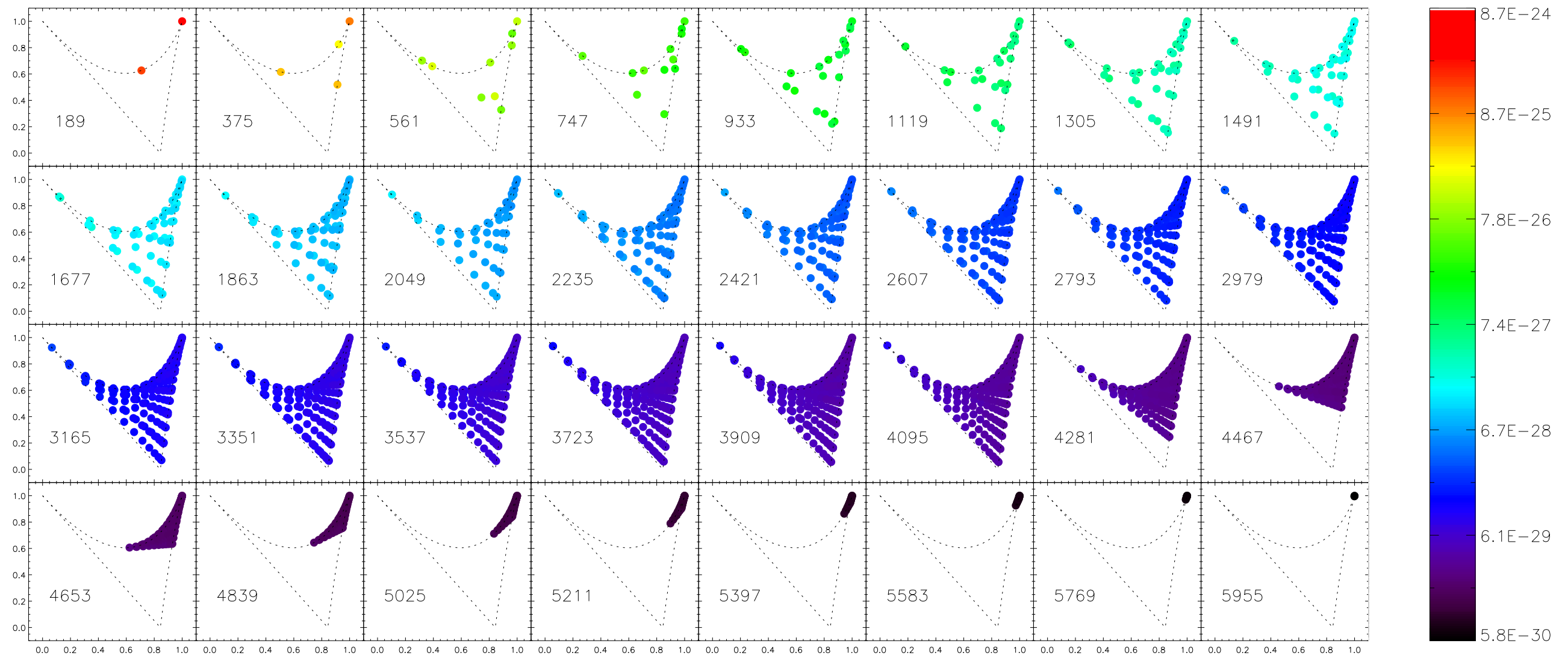
Allows to represent efficiently a
bispectrum :

- no redundancy
- separates geometric and scale
dependence

Lacasa et al. 2011

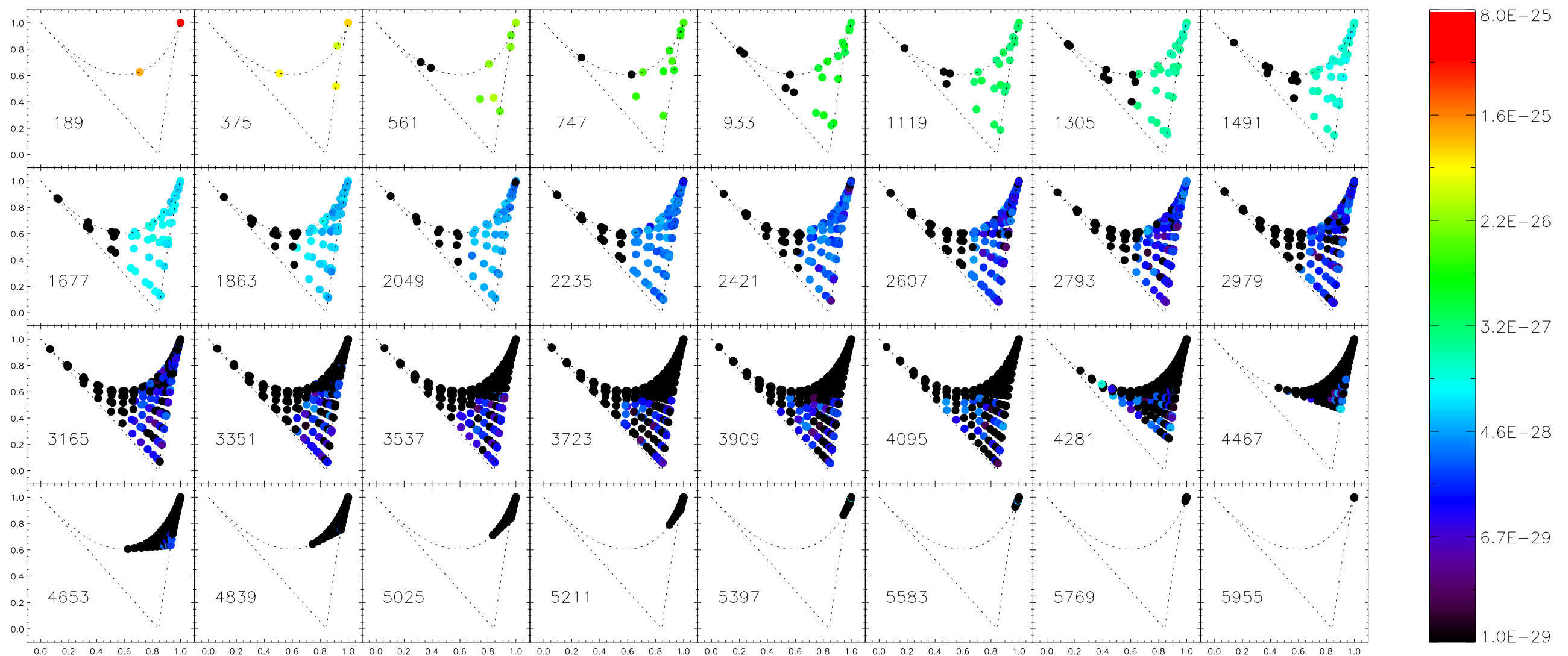
C. tSZ bispectrum in the parametrisation : simulations

Planck simulations



C. tSZ bispectrum in the parametrisation : NILC

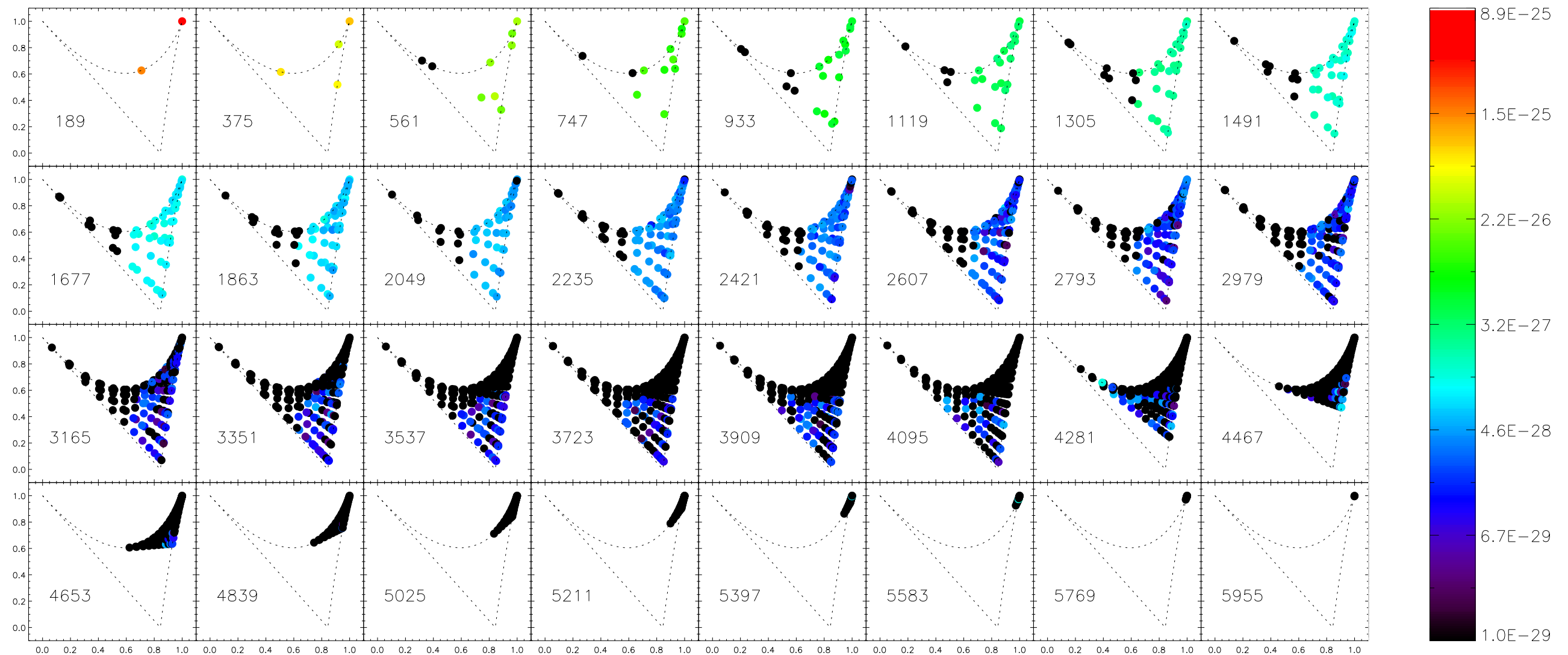
Planck 2013 results



Black points : configurations discarded conservatively to avoid systematics (foregrounds, mask effect)

C. tSZ bispectrum in the parametrisation : MILCA

Planck 2013 results



C. *Planck* tSZ bispectrum : conclusions/perspectives

- high consistency of the measurement :
 - internally : NILC/MILCA
 - externally : with simulations
- high detection signal-to-noise
there is valuable information to extract
- ‘2014’ results to come. Better, stronger
- improvements are even possible
- necessity of a model

C. tSZ bispectrum : model

Aim : a model predicting both the tSZ power spectrum and bispectrum

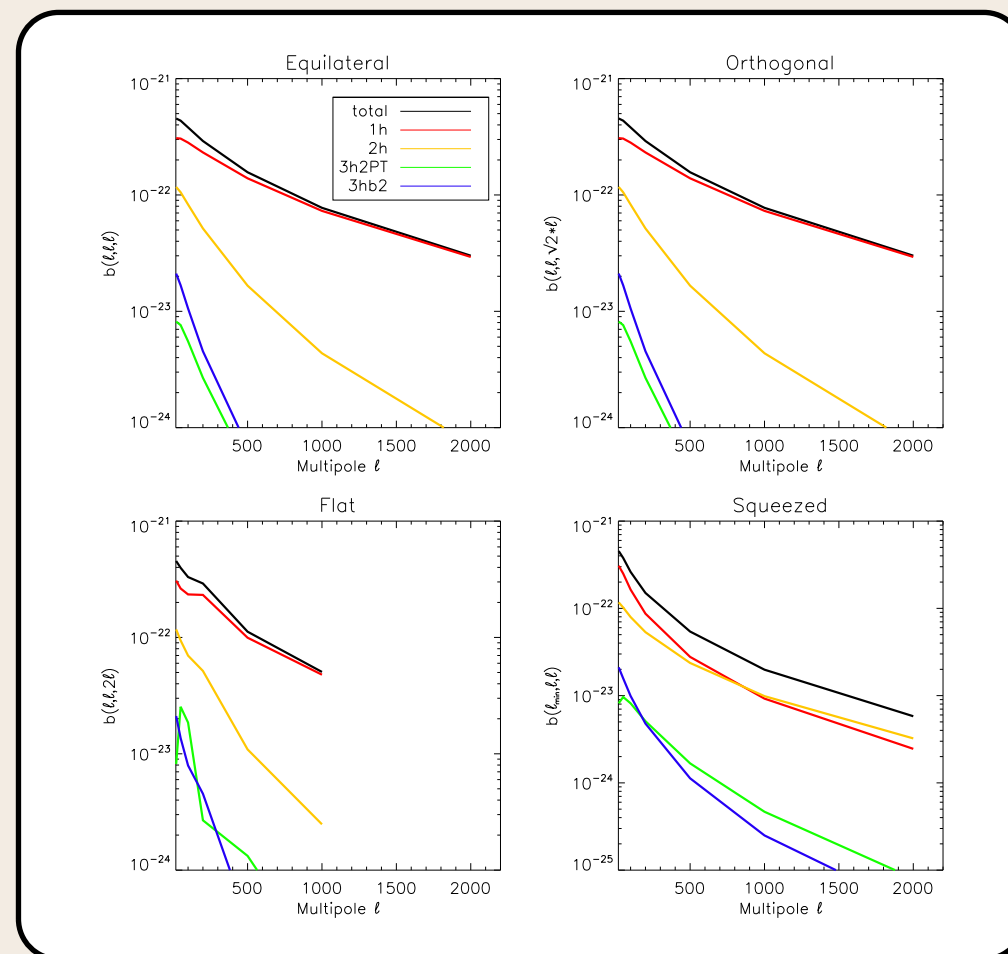
and later on their covariance matrix, to constrain tSZ physics and cosmology : σ_8 and Ω_m (possibly also neutrinos, MG, f_{NL})

Preliminary results

including 1-halo, 2-halo and 3-halo terms

Find that 1h dominates, except on large scales and in squeezed triangles

Consistent with expectations from power spectrum



in progress...

Combining probes for *DES*

A. Covariance of the galaxy spectrum and cluster counts

Cluster count is the monopole of the halo density field

$$\hat{N}_{\text{cl}}(i_M, i_z) = \overline{N}_{\text{cl}}(i_M, i_z) + \frac{1}{\Omega_S} \int dM d^2\hat{n} dz r^2 \frac{dr}{dz} \frac{d^2 n_h}{dM dV} \delta_{\text{cl}}(\mathbf{x}, z | M, z)$$

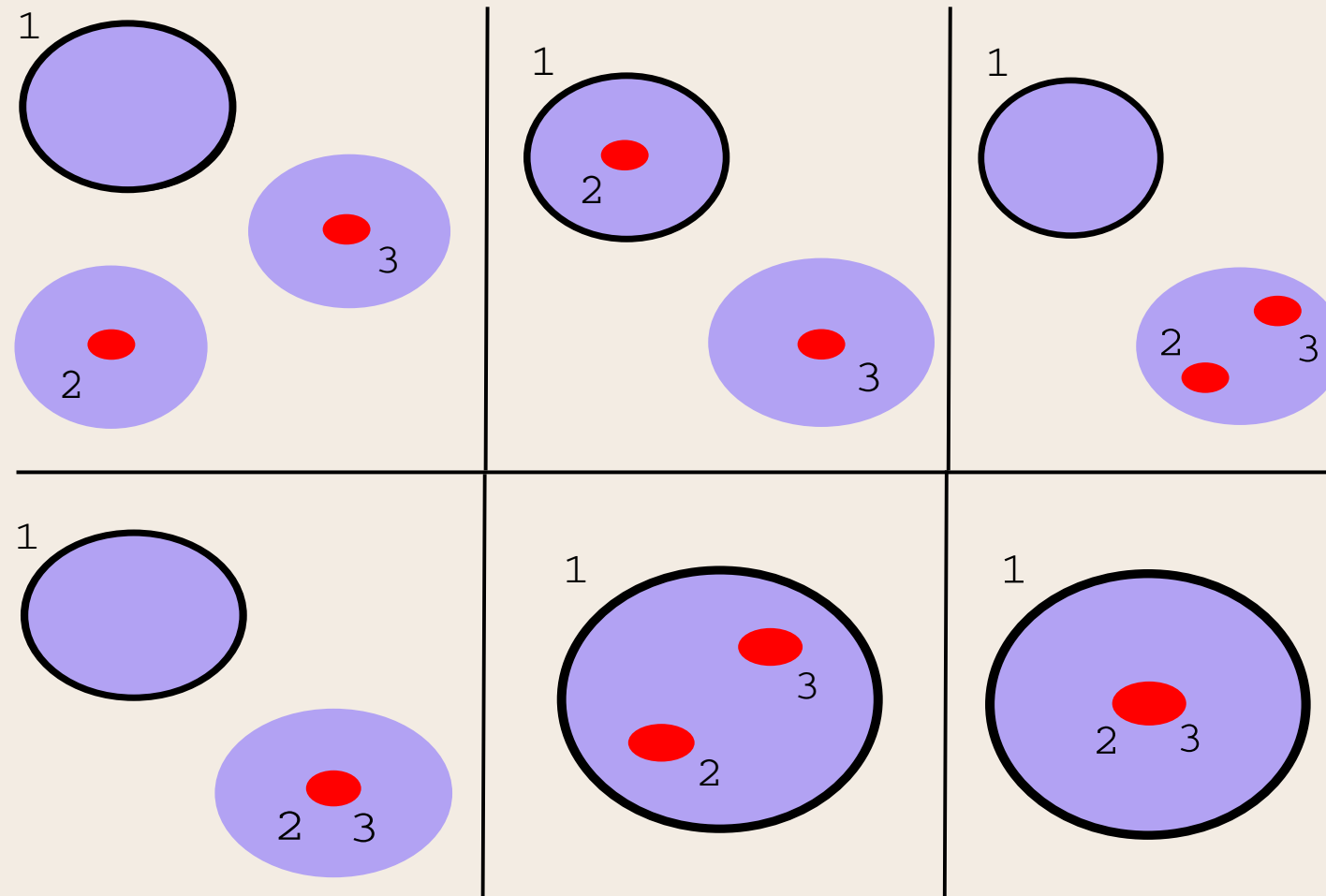
$$\text{Cov} \left(\hat{N}_{\text{cl}}(i_M, i_z), \hat{C}_{\ell}^{\text{gal}}(j_z, k_z) \right) = \int \frac{dM_1 dz_{123}}{4\pi} \frac{dV}{dz_1} \frac{d^2 n_h}{dM dV} \Big|_{M_1, z_1} b_{0\ell\ell}^{\text{hgg}}(M_1, z_{123})$$

Cluster count in a bin of mass (i_M) and redshift (i_z)

Galaxy angular power spectrum between two redshift bins

Halo-galaxy-galaxy angular bispectrum

A. Diagrammatic method for the hgg bispectrum

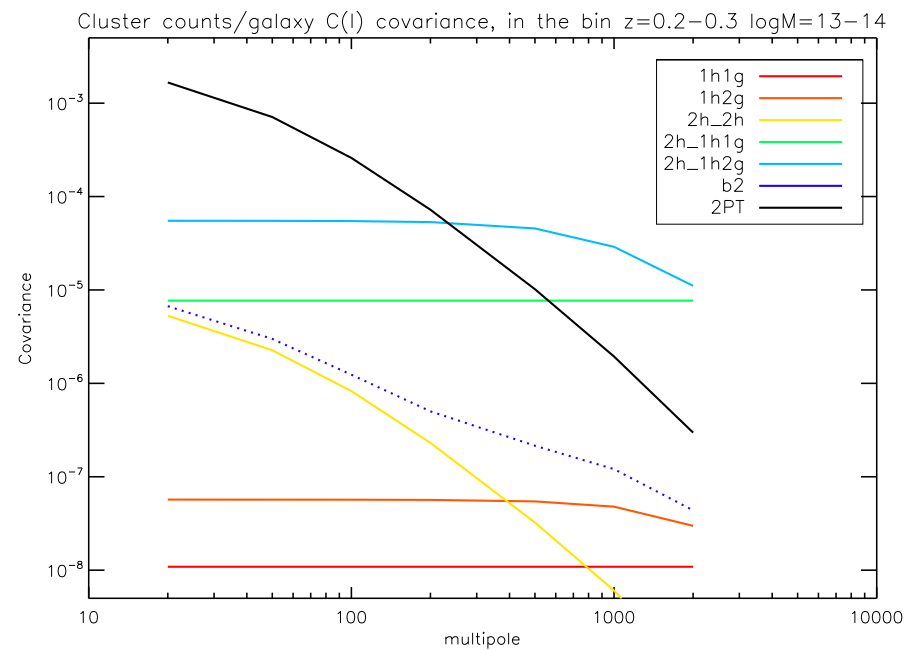


Ex:
$$B_{\text{hgg}}^{2h-1h2g}(k_{123}|M_1, z_{123}) = \frac{\delta_{z_2, z_3}}{\bar{n}_{\text{gal}}^2(z_2)} \int dM \frac{d^2 n_h}{dM dV} \langle N_{\text{gal}}(N_{\text{gal}} - 1)(M) \rangle u(k_2|M) u(k_3|M) P_{\text{halo}}(k_1|M_1, M, z_1, z_2)$$

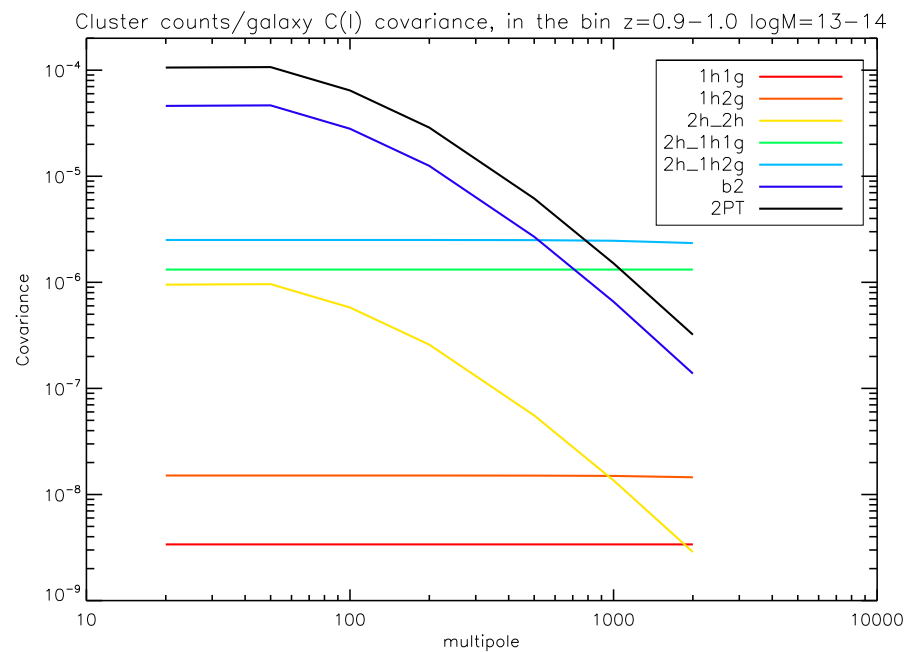
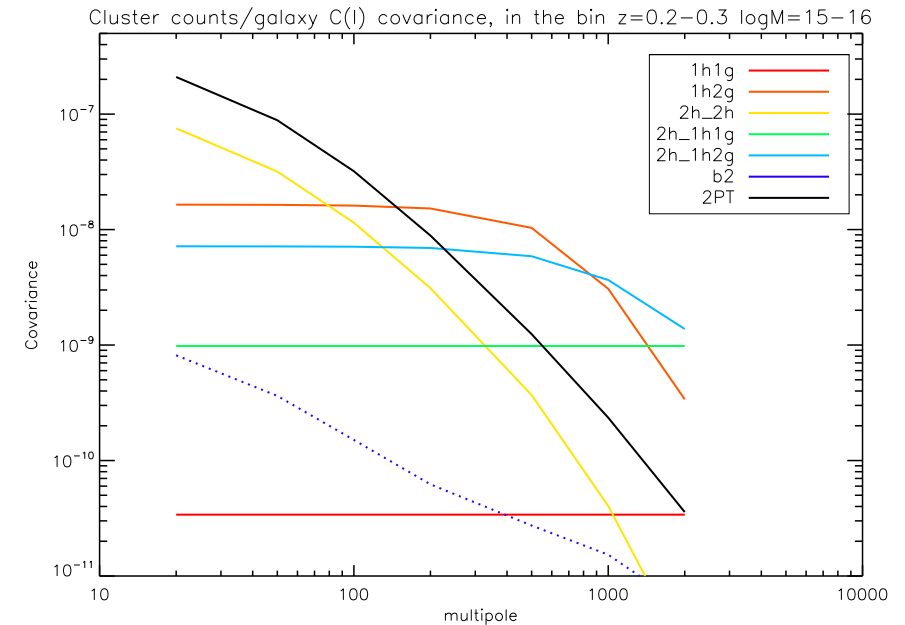
Ingredients : cosmology, halo model, Halo Occupation Distribution (HOD)

A. Ideal results I : scale dependence of the covariance

1



2



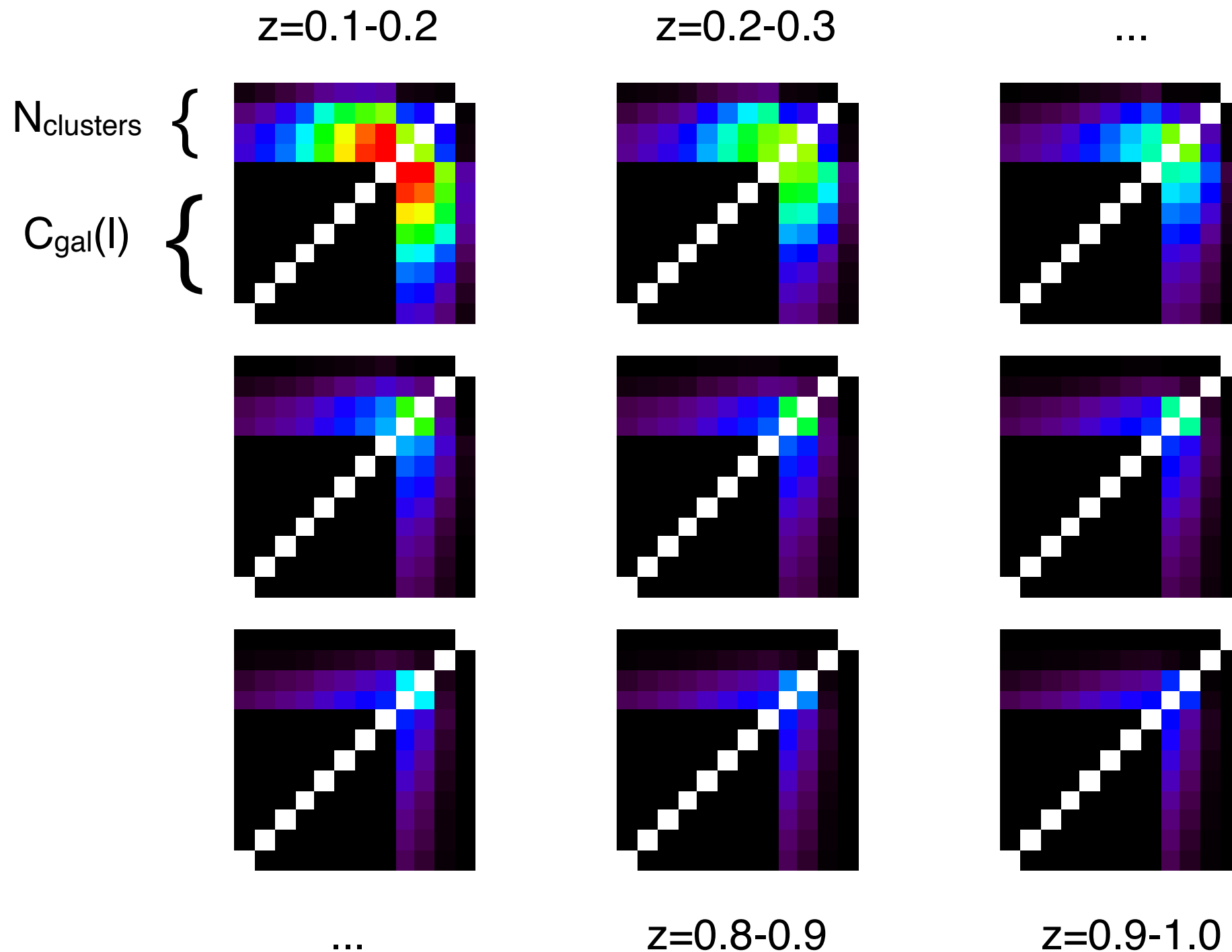
3

1 : $z=0.2-0.3$ and $\log(M/M_{\text{sun}}) = 13-14$

2 : $z=0.2-0.3$ and $\log(M/M_{\text{sun}}) = 15-16$

3 : $z=0.9-1.0$ and $\log(M/M_{\text{sun}}) = 13-14$

A. Ideal results II : joint covariance matrix



B. Joint likelihood

Cluster counts follow a Poisson distribution
Galaxy correlation is more Gaussian

→ How to mix their likelihood ?
(cannot assume that the joint likelihood is Gaussian)

Edgeworth / Gram-Charlier expansion

$$P(x) = \exp \left[\sum_{n=1}^{+\infty} \left(\kappa_n(P) - \kappa_n(P_{\text{fidu}}) \right) \frac{(-1)^n}{n!} \frac{d^n}{dx^n} \right] P_{\text{fidu}}(x)$$

Expand around independent case. Result :

$$\mathcal{L}(\text{counts}, C_\ell) = \exp \left[- \sum_{ij} \langle c_i C_{\ell_j} \rangle_c (\log \bar{c}_i - \Psi(c_i + 1)) ({}^T C_\ell C^{-1} e_j) \right] \mathcal{L}(\text{counts}) \mathcal{L}(C_\ell)$$

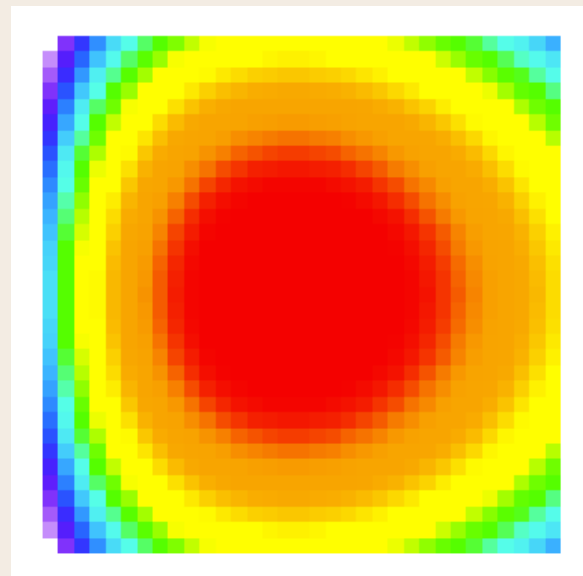
B. Joint likelihood : functional form

without correlation

with correlation

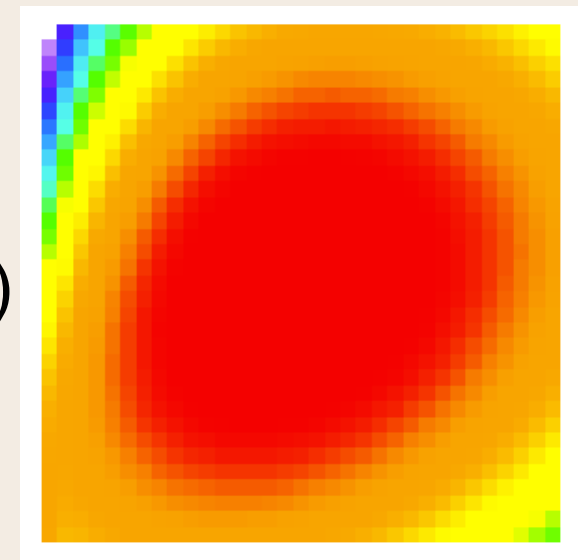
$\langle N_{\text{clust}} \rangle = 16$

$C(I)$



0 counts 20

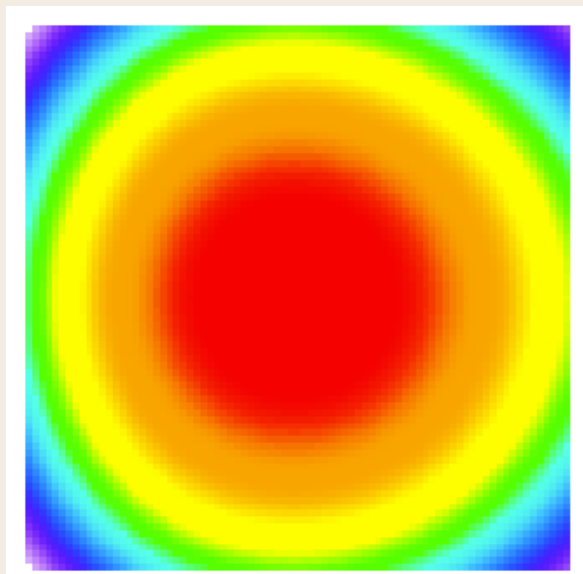
$C(I)$



0 counts 20

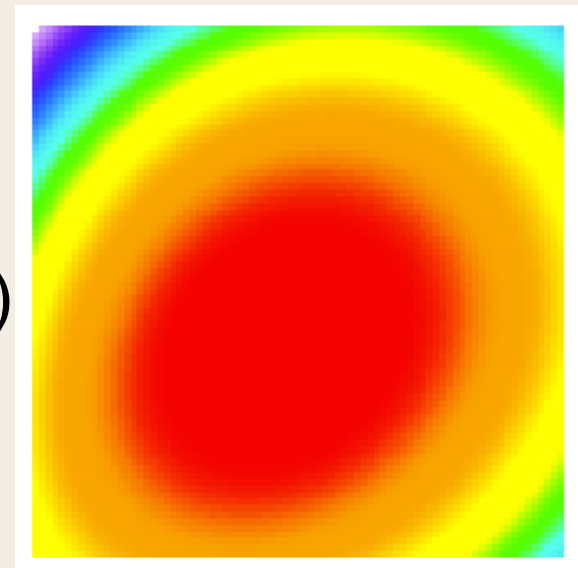
$\langle N_{\text{clust}} \rangle = 100$

$C(I)$



60 counts 140

$C(I)$



60 counts 140

B. Joint likelihood : conclusions / perspectives

- large counts and small cross-covariance limit :
Gaussian with correct covariance matrix
- is easily extended to include cluster sample covariance
- inclusion of Bayesian hyperparameters
→ robustness to tension and error estimates

- valid to combine any Gaussian and Poisson observables (e.g. weak-lensing/counts)
- Markov chain pipeline to be built
 - for realistic forecasts
 - application to *DES* data

Conclusions

- *Planck* yields high quality NG measurement for the Cosmic Infrared Background and for the thermal Sunayev-Zel'dovich effect
- Combining probes for galaxy surveys like *DES* also involves NG indirectly
- NG is an important tool to extract as much information as possible from Large Scale Structure observables in the present and for the future.

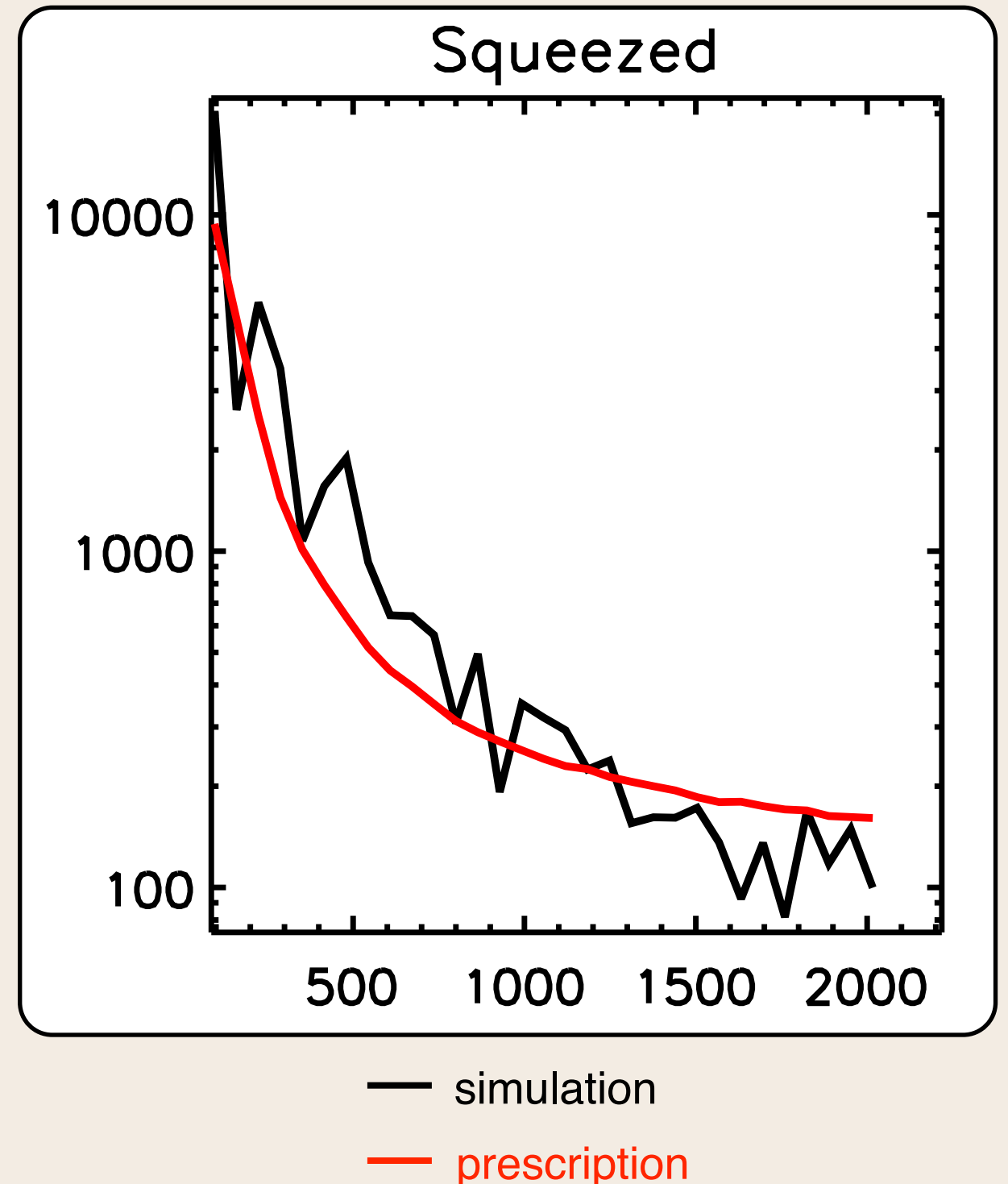
Thanks for your attention

CIB phenomenological NG prescription

- analytical prescription for clustered sources

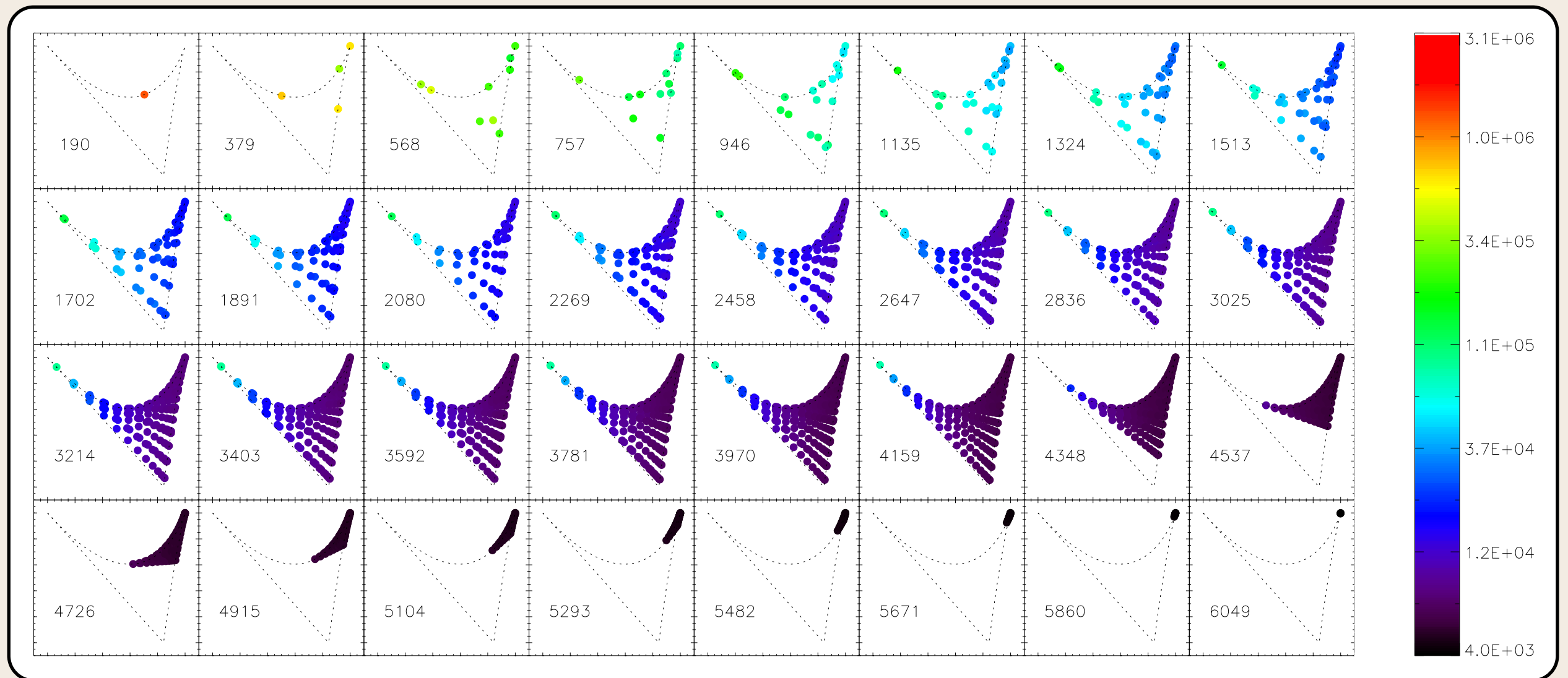
$$b_{\ell_1 \ell_2 \ell_3} = \alpha \sqrt{C_{\ell_1} C_{\ell_2} C_{\ell_3}} \quad \text{avec} \quad \alpha = \frac{\int S^3 \frac{dn}{dS} dS}{\left(\int S^2 \frac{dn}{dS} dS \right)^{3/2}}$$

- based on moment of order 1 (counts) and 2 (spectrum)
- reproduces the CIB bispectrum on simulations (Sehgal et al. 2010)
- CIB bispectrum maximal in squeezed



Configuration dependence of the CIB bispectrum : model

Total bispectrum

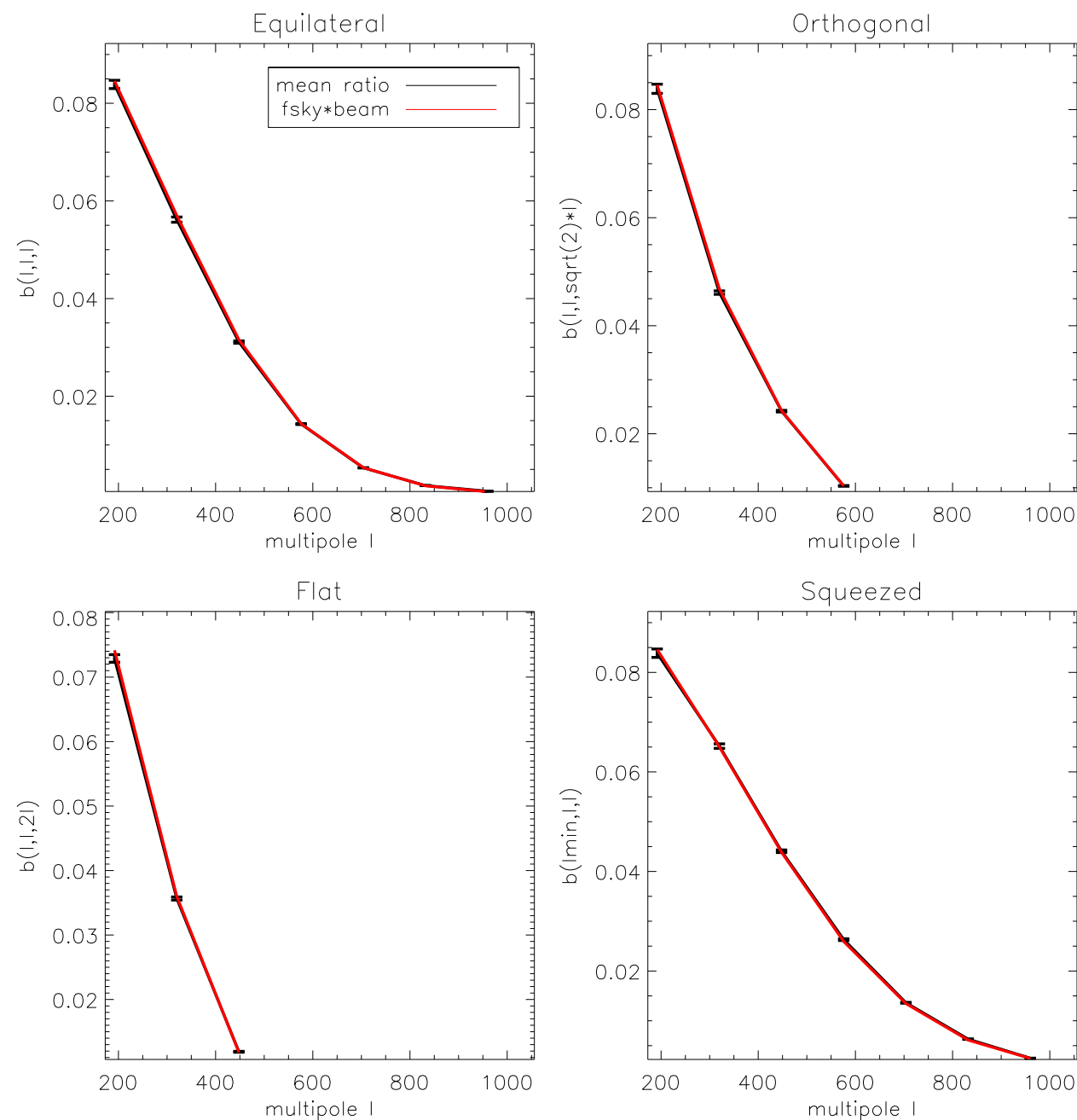


maximal in squeezed

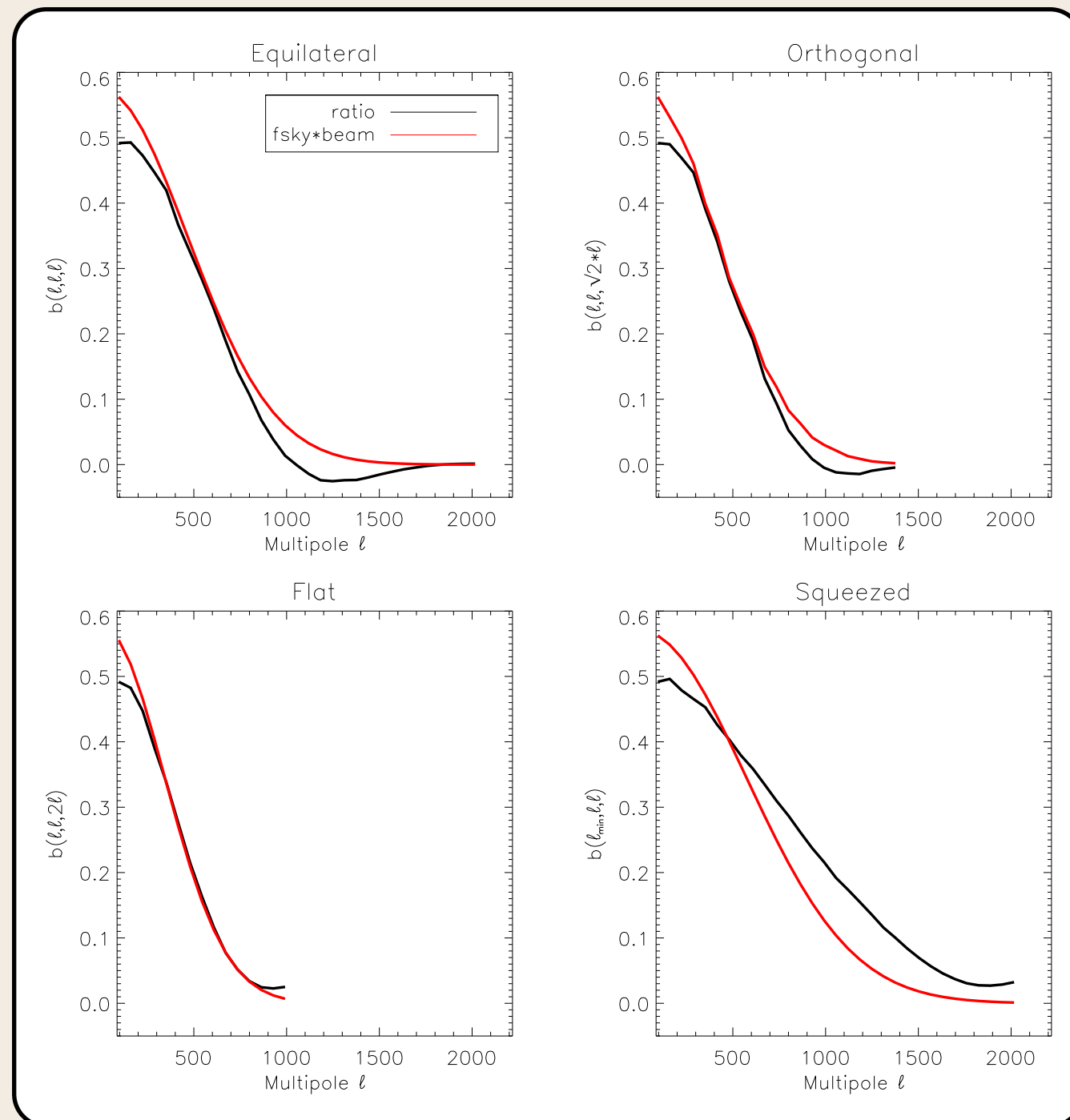
CIB bispectrum : debiasing the mask effect

ratio full-sky bispectrum/
partial sky bispectrum
on simulations with CIB
prescription

— simulations ratio
— $f_{\text{sky}} * \text{beam}$



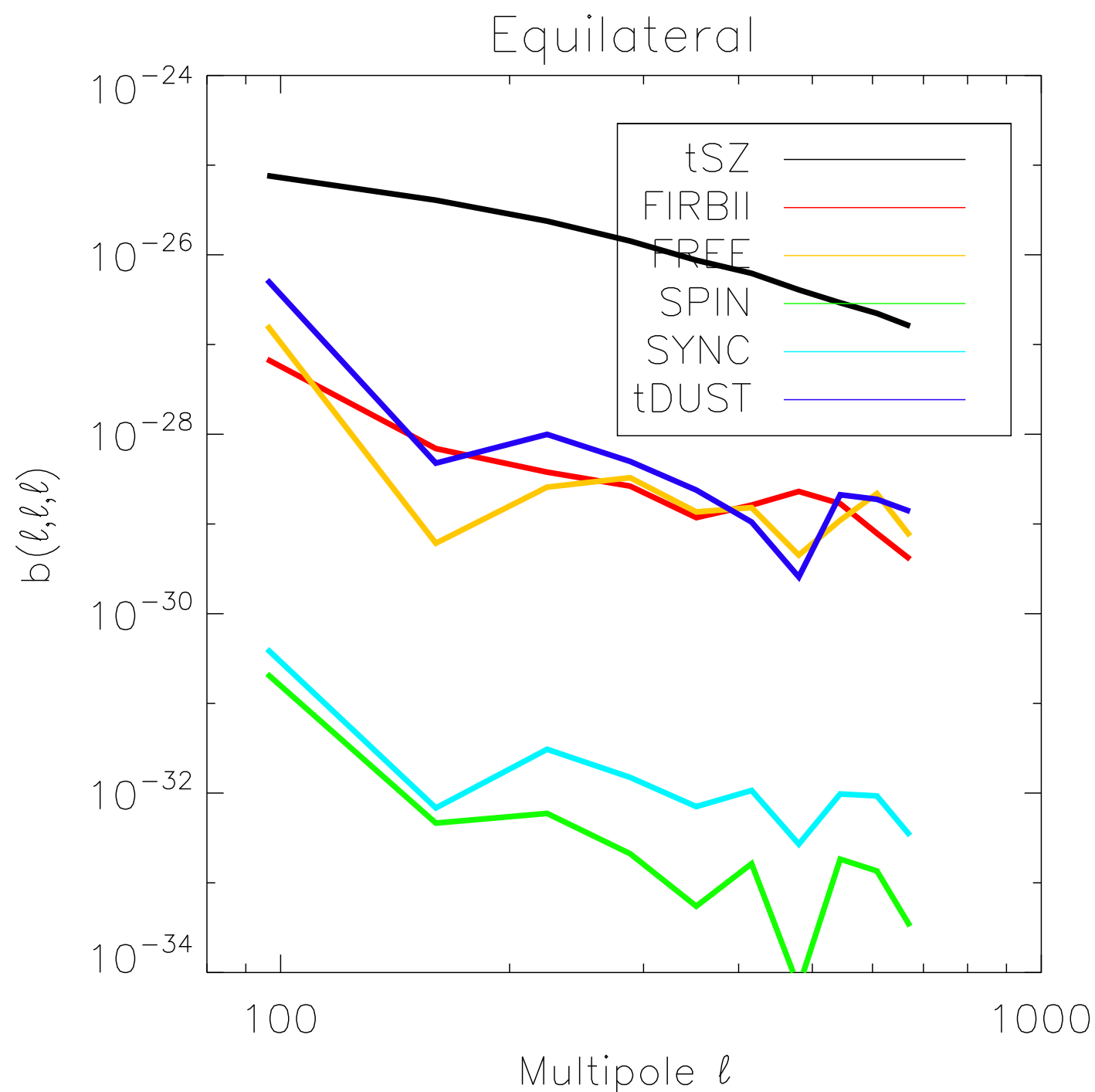
tSZ bispectrum : debiasing the mask effect



ratio full-sky bispectrum/
partial sky bispectrum
on NG simulations
reproducing the tSZ
bispectrum

— simulations ratio
— $f_{\text{sky}} * \text{beam}$

tSZ bispectrum : contamination by foregrounds



For MILCA :
Comparison of the tSZ
bispectrum and of
foreground residuals.
On *Planck* simulations

Halo-galaxy-galaxy bispectrum : from 3D to 2D

$$b_{0\ell\ell}^{\text{hgg}}(M_1, z_{123}) = \frac{\delta(z_2 - z_3)}{r_2^2 \frac{dr}{dz_2}} \frac{2}{\pi} \int k_1^2 dk_1 B_{\text{hgg}}(k_1, k_2^*, k_2^* | M_1, z_1, z_2, z_2) j_0(k_1 r_1) j_0(k_1 r_2) \quad \text{with} \quad k_2^* = \frac{\ell + 1/2}{r(z_2)}$$

angular bispectrum

3D bispectrum

Bessel functions

Limber's approximation on k_2 and k_3
(bispectrum varies slowly compared to bessel's oscillations)

Ingredients

$$B_{\text{hgg}}^{2h-1h2g}(k_{123}|M_1, z_{123}) = \frac{\delta_{z_2, z_3}}{\bar{n}_{\text{gal}}^2(z_2)} \int dM \frac{d^2 n_h}{dM dV} \langle N_{\text{gal}}(N_{\text{gal}} - 1)(M) \rangle u(k_2|M) u(k_3|M) P_{\text{halo}}(k_1|M_1, M, z_1, z_2)$$

halo mass function
(Sheth&Tormen or Tinker)

Halo Occupation Distribution
(Tinker&Wetzel 2010)

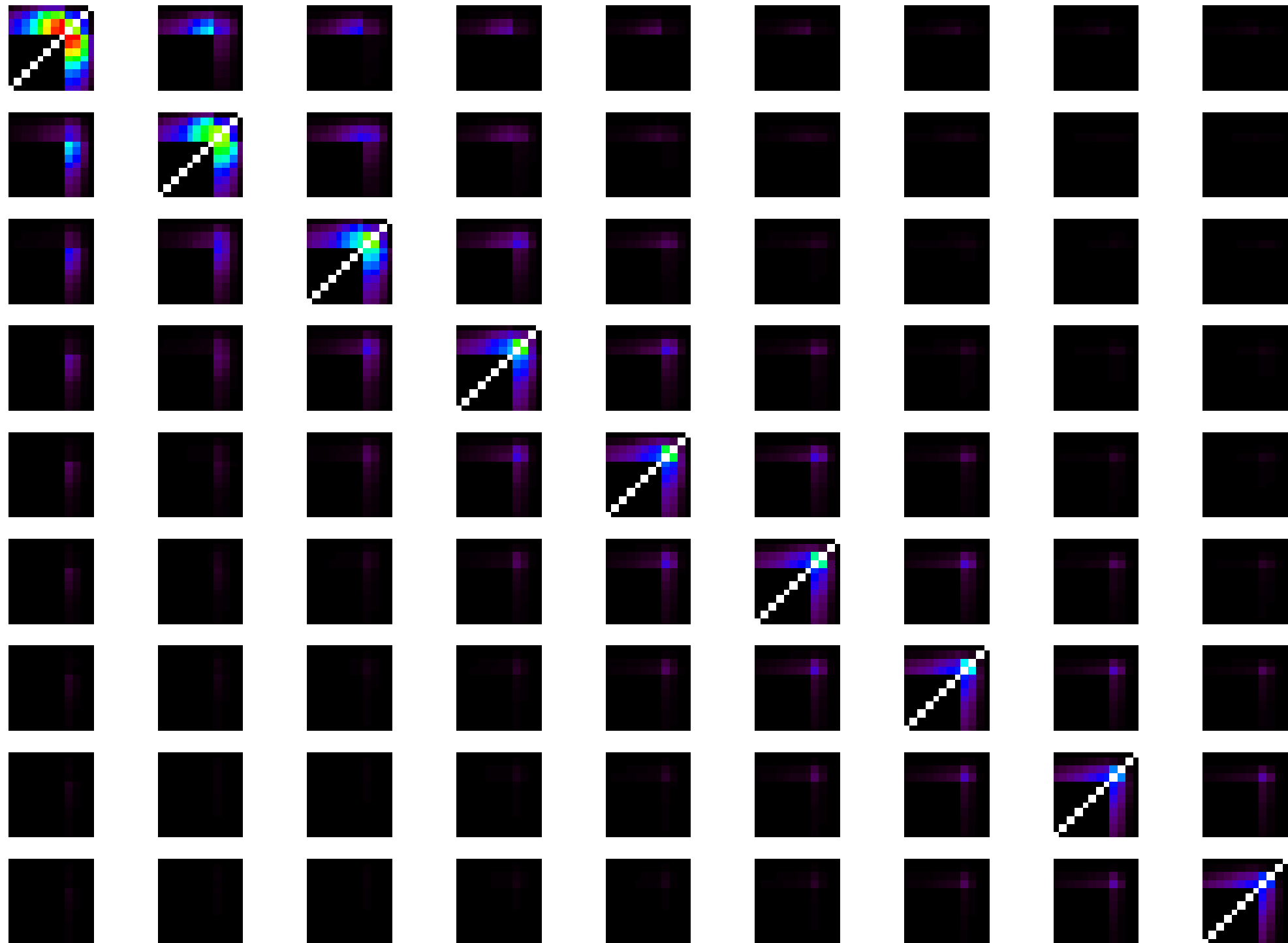
halo profile
(NFW)

halo bias
(Tinker : 1, Sheth&Tormen : 1&2)

+ cosmology
(growth function, comoving volume...)

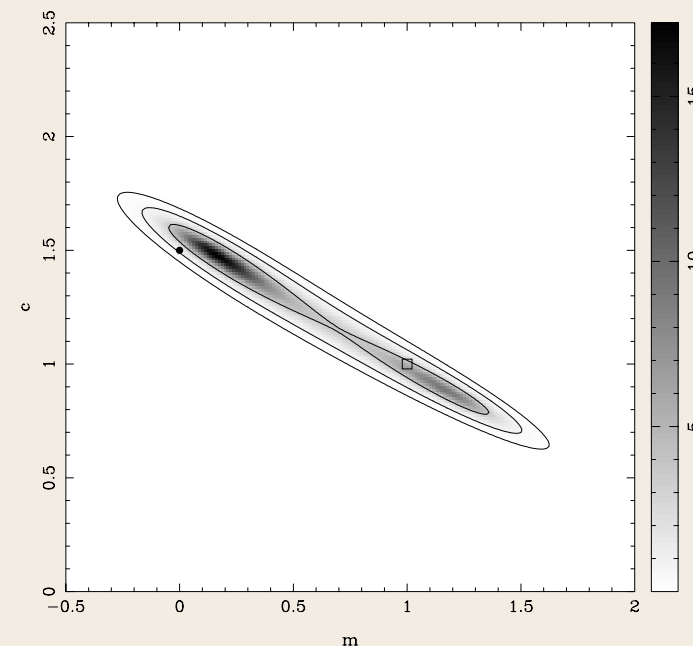
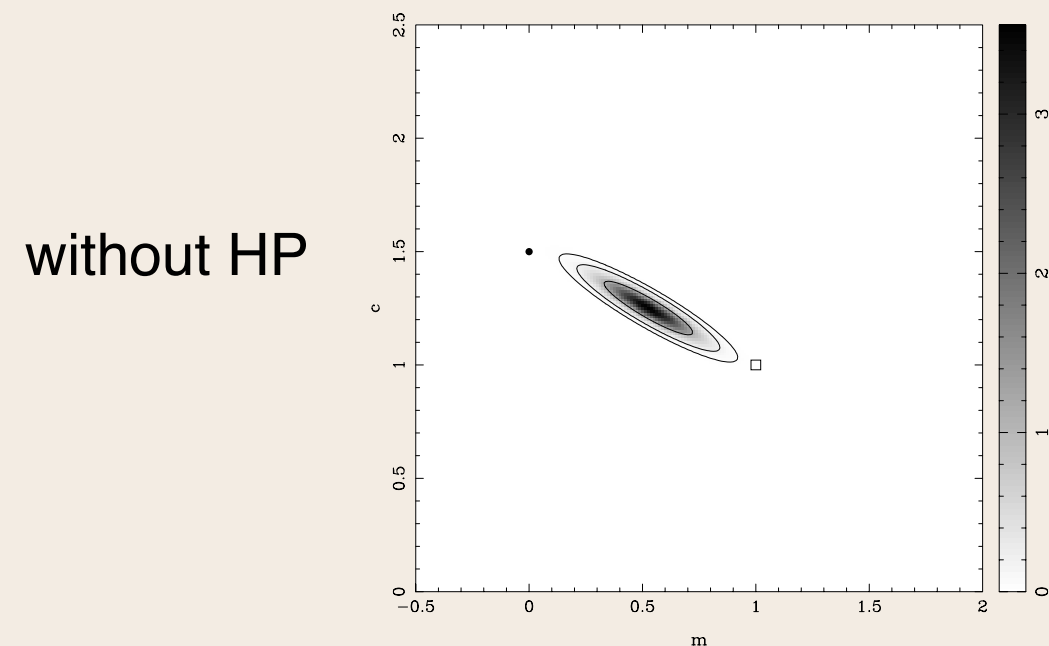
matter power spectrum
(Eisenstein&Hu 98)

Counts - gal spectrum covariance, with cross-redshifts



Bayesian hyperparameters

- Statistical method allowing to detect underestimation of error bars or inconsistencies between data sets



with HP

Hobson et al. 2002

- Idea : rescale error bars
one rescaling parameter per data set. These parameters are included in the MCMC exploration. Then marginalise over them.
- Only done for Gaussian distribution at the moment

Hyperparameters for a Poisson distribution

- not possible to satisfy all the properties of the Gaussian case
(i.e. keep the mean but rescale the variance)
- two possible approximate prescriptions with a good asymptotic behaviour

