

Cosmology with galaxy clustering

A joint analysis of the power spectrum and bispectrum

Chiara Moretti



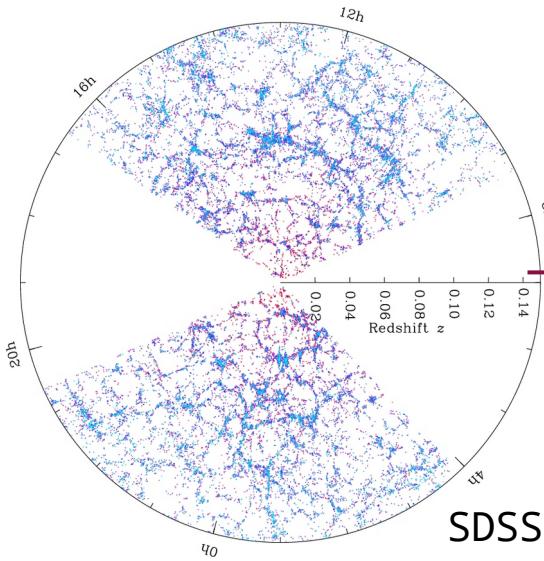
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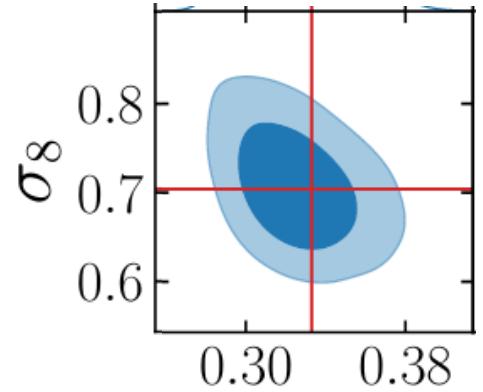
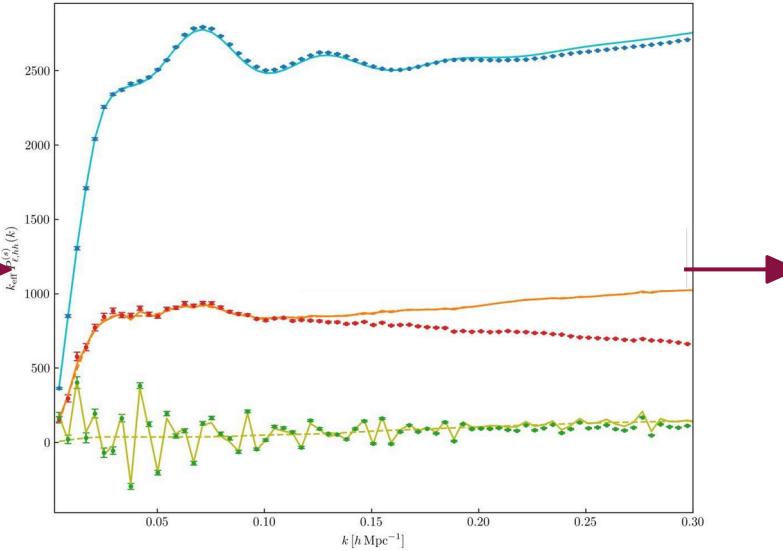
UK Research
and Innovation



From galaxy surveys to cosmology

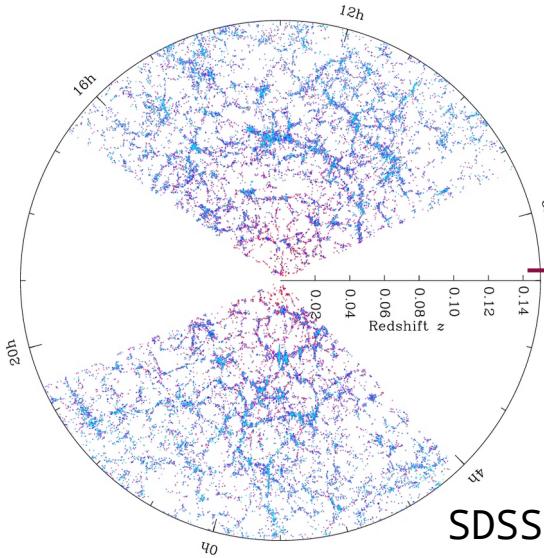


Galaxy distribution

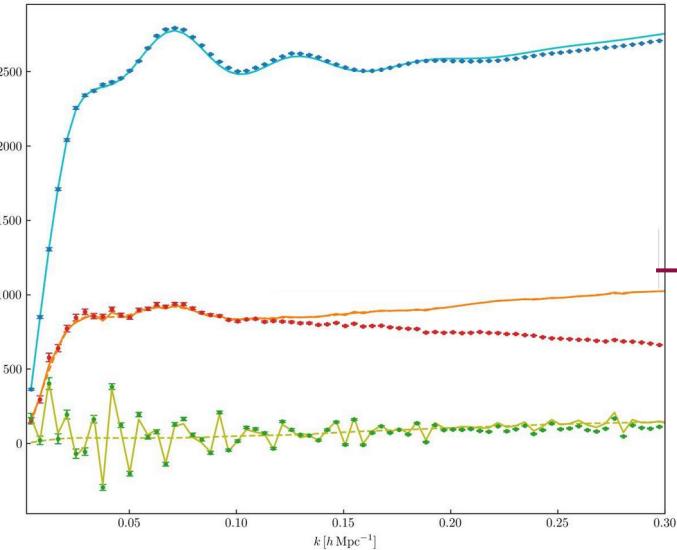


Cosmological parameters

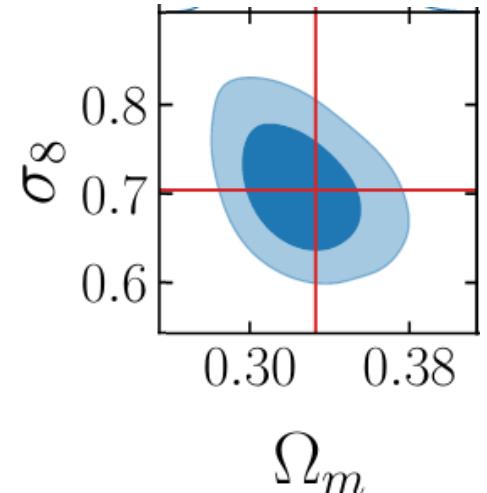
From galaxy surveys to cosmology



Galaxy distribution



Power spectrum multipoles



Cosmological parameters

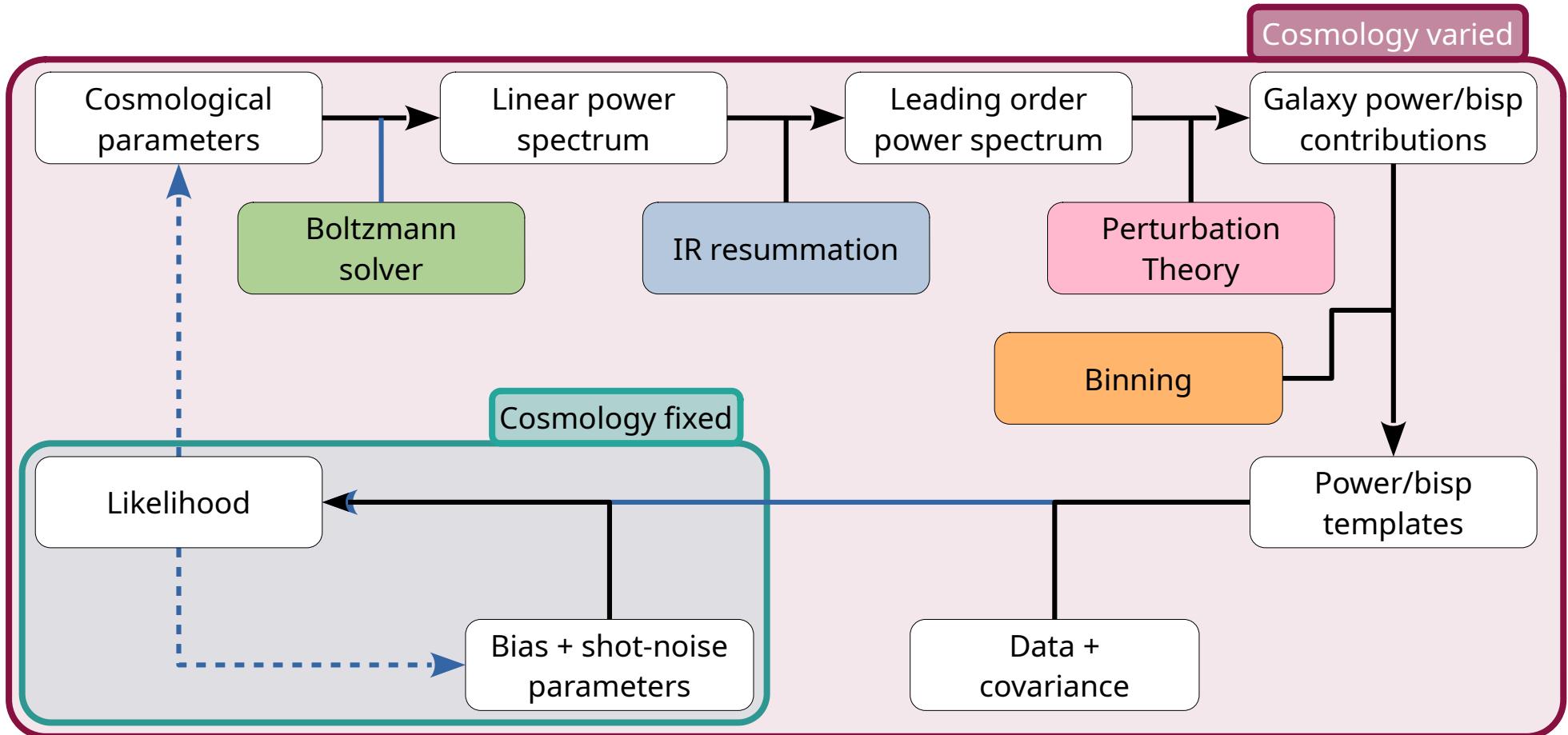
Accurate & fast theoretical model → Likelihood analysis

PBJ: Power spectrum & Bispectrum Joint analysis

- 1-loop power spectrum + tree-level bispectrum with non-linear galaxy bias (all in Python)
- IR-resummation routine ($w - nw$ split)
- samplers: emcee (affine invariant & Metropolis-Hastings), Multinest (nested sampling)
- Binning effects
- Different likelihood functions for noise in the covariance matrix

CM, F. Rizzo, K. Pardede, A. Oddo, E. Sefusatti, A. Eggemeier, C. Porciani
[1908.01774] [2108.03204] [2204.13628]

PBJ: Power spectrum & Bispectrum Joint analysis



Theoretical model – power spectrum

- EFTofLSS model (1loop + counterterms)
- Non-linear bias expansion
- Infra-red resummation routine (Eisenstein-Hu P_{smooth})
- FastPT → model evaluation (28 loop integrals!) in $\sim 30 \text{ ms}$

$$P_{gg}(\mathbf{k}) = Z_1^2(\mathbf{k})P_L(k) + 2 \int d^3\mathbf{q} [Z_2(\mathbf{q}, \mathbf{k} - \mathbf{q})]^2 P_L(q)P_L(|\mathbf{k} - \mathbf{q}|) \\ + 6Z_1(\mathbf{k})P_L(k) \int d^3\mathbf{q} Z_3(\mathbf{k}, \mathbf{q}, -\mathbf{q})P_L(q) + P_{\text{ctr}}(\mathbf{k}) + P_{\text{noise}}(\mathbf{k})$$

Nuisance parameters

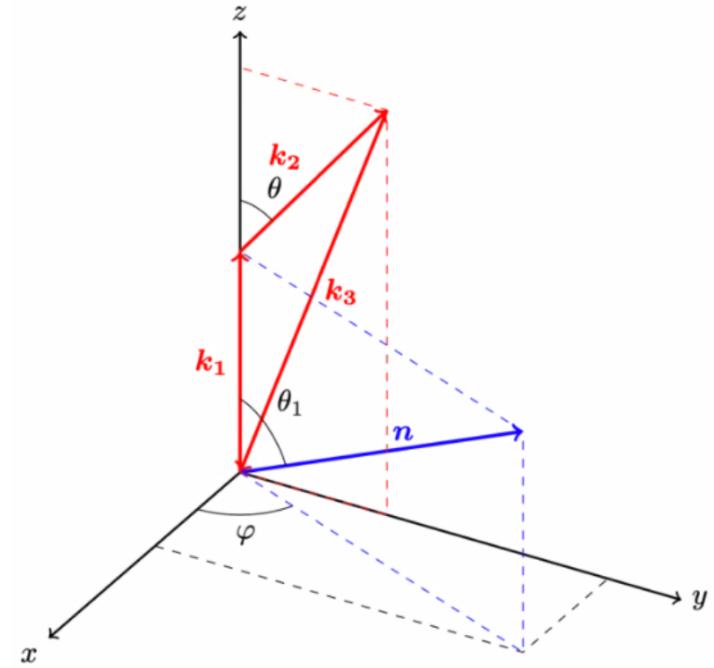
$b_1, b_2, b_{\gamma 2}, b_{\Gamma 3},$ bias

$\alpha_P, \epsilon_{1,k^2}, \epsilon_{2,k^2},$ noise

c_0, c_2, c_4, c_{k^4} EFT counterterms

Theoretical model – bispectrum

- Tree-level model
- Bias parameters: $b_1, b_2, b_{\gamma_2}, \alpha_1, \alpha_2, \alpha_3$
- Higher order multipoles
- Model evaluation in ~ 0.1 s



$$B_s(\vec{k}_1, \vec{k}_2, \hat{n}) = 2Z_1(\vec{k}_1)Z_1(\vec{k}_2)Z_2(\vec{k}_1, \vec{k}_2)P_L(k_1)P_L(k_2)$$

$$B_{stoch}(k_1, \vec{k}_2, \hat{n}) = \frac{1}{\bar{n}} \left[(1 + \alpha_1)b_1 + (1 + \alpha_3)f\mu^2 \right] Z_1(\vec{k}_1)P_L(k_1) + \frac{1 + \alpha_2}{\bar{n}^2}$$

Theoretical model – bispectrum

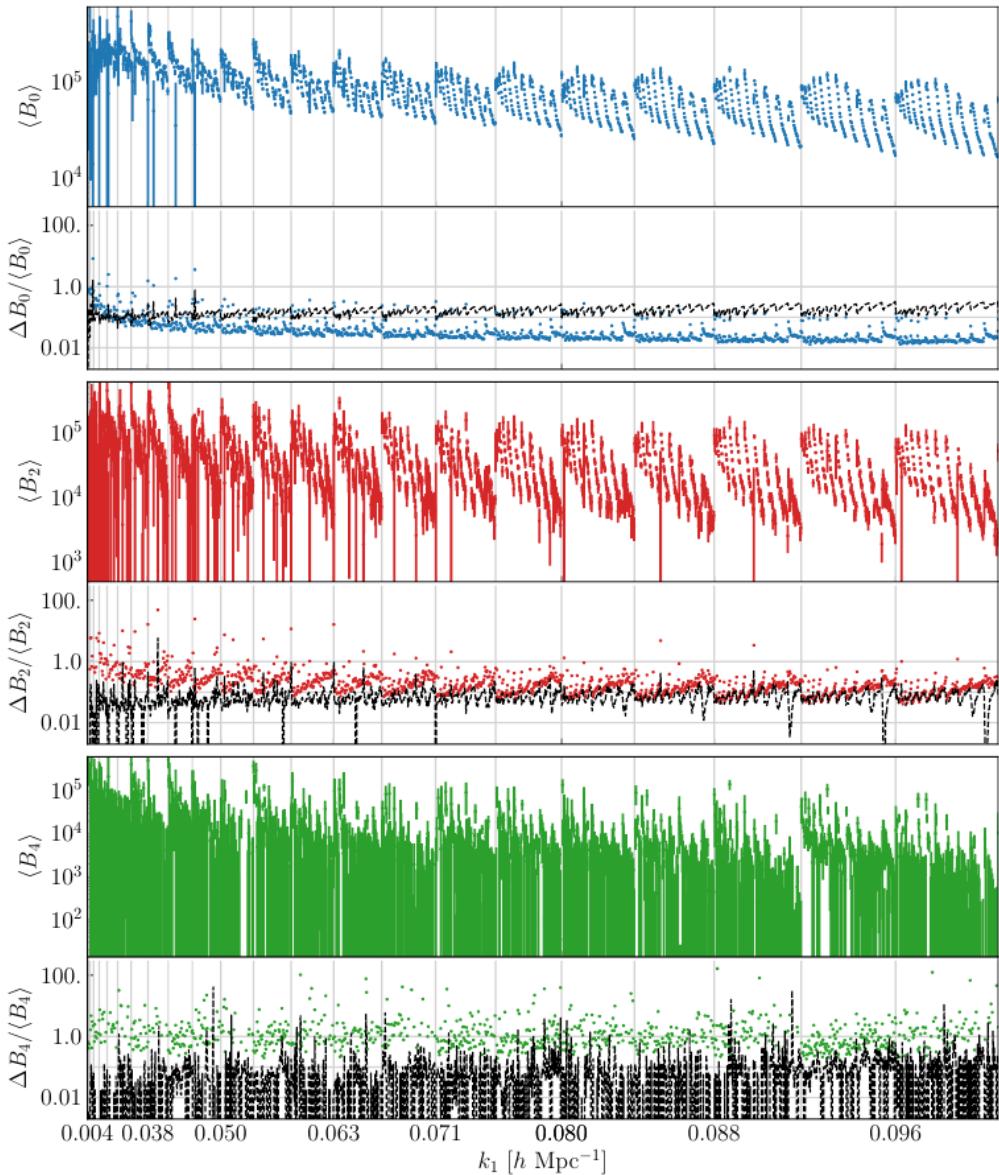
Alcock-Paczynski: expansion around $\alpha_{\parallel} \approx 1$ and $\alpha_{\perp} \approx 1$:

$$B_{\ell}(k_1, k_2, k_3) = \frac{2\ell + 1}{\alpha_{\perp}^4 \alpha_{\parallel}^2} \sum_{n_1, n_2} \int_{-1}^1 \frac{d\mu_1}{2} \int_0^{2\pi} \frac{d\varphi}{2\pi} \mathcal{L}_{\ell}(\mu_1) \mu_1^{n_1} \mu_2^{n_2} B_{n_1, n_2}(k_1, k_2, k_3) \times \\ \left\{ 1 + [n_1(\mu_1^2 - 1) + n_2(\mu_2^2 - 1)] (F - 1) + \right. \\ \left. \sum_{i=1}^3 [1 - \alpha_{\perp} + (\alpha_{\perp} - \alpha_{\parallel}) \mu_i^2] \frac{\partial \ln B_{n_1, n_2}}{\partial \ln k_i} (k_1, k_2, k_3) \right\}$$

→ we can factor out the dependence on $\alpha_{\parallel}, \alpha_{\perp}$ and treat them as bias parameters

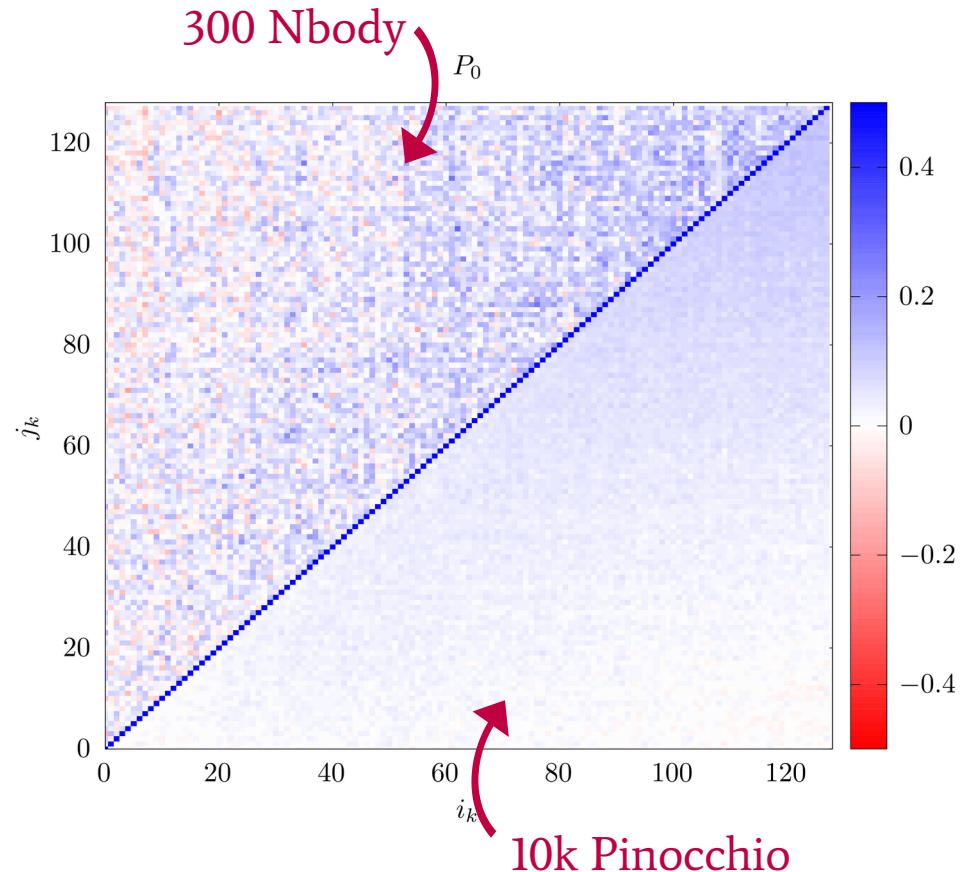
Validation dataset

- 300 DM-only N-body simulations
(Minerva), Λ CDM cosmology
- $L=1500 \text{ Mpc}/h \rightarrow V_{\text{tot}} \approx 1000 \text{ Gpc}^3/h^3$



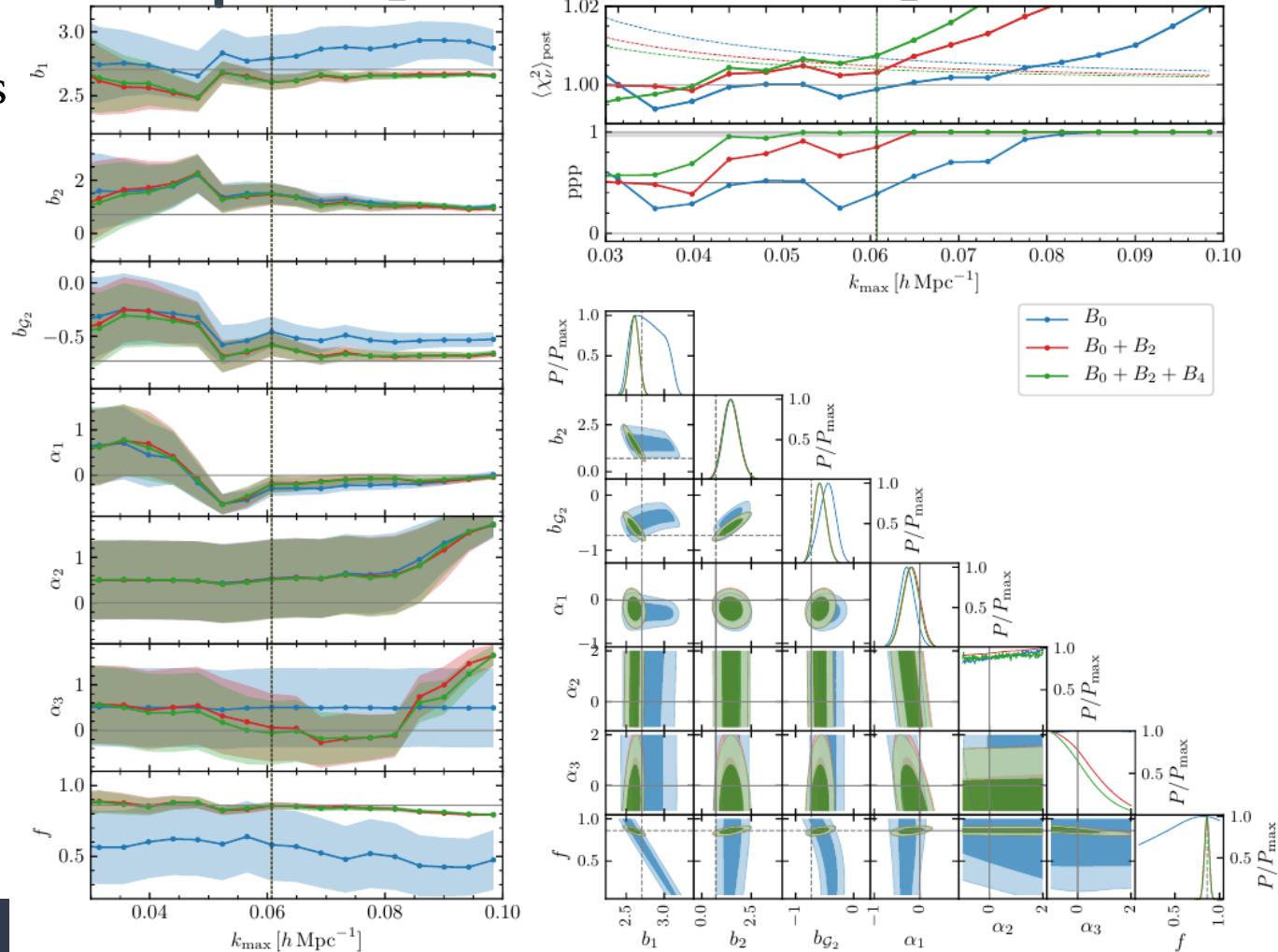
Validation dataset

- 300 DM-only N-body simulations (Minerva), Λ CDM cosmology
- $L=1500 \text{ Mpc}/h \rightarrow V_{\text{tot}} \approx 1000 \text{ Gpc}^3/h^3$
- Covariance from 10000 Pinocchio mocks



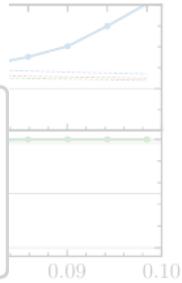
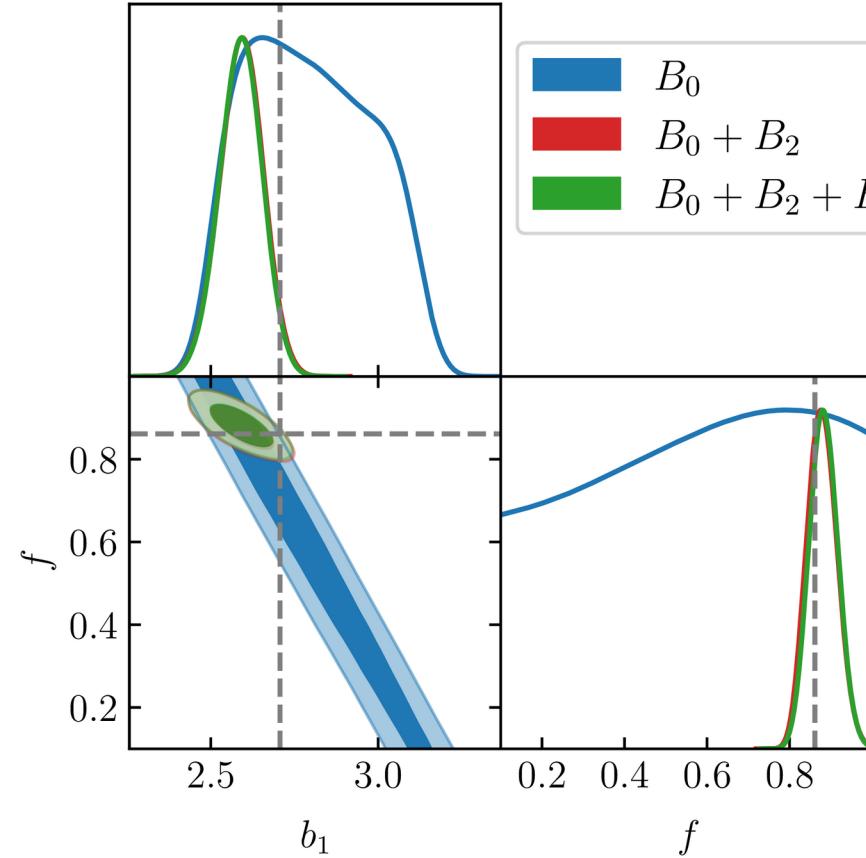
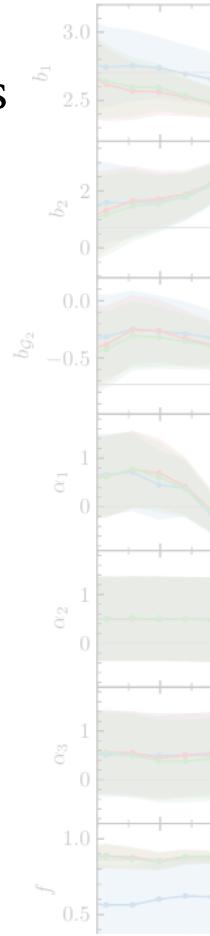
Bispectrum in redshift space [Rizzo,CM+ 2204.13628]

- Higher order multipoles



Bispectrum in redshift space [Rizzo,CM+ 2204.13628]

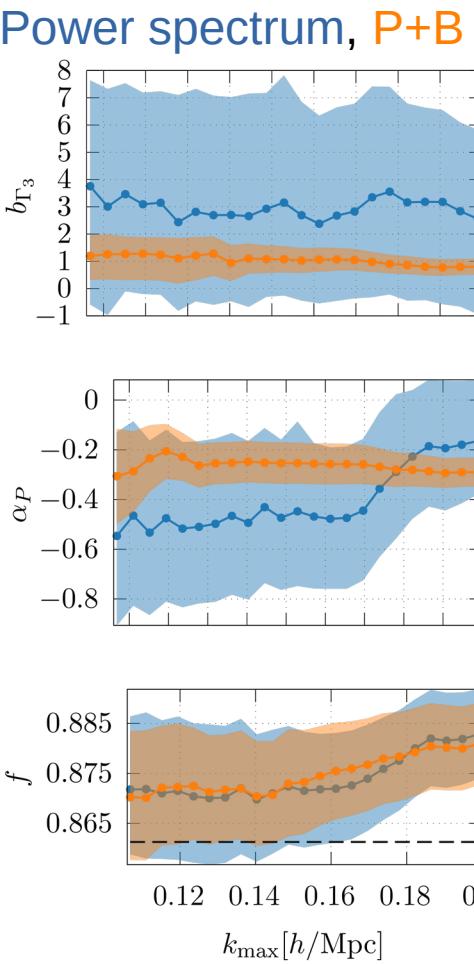
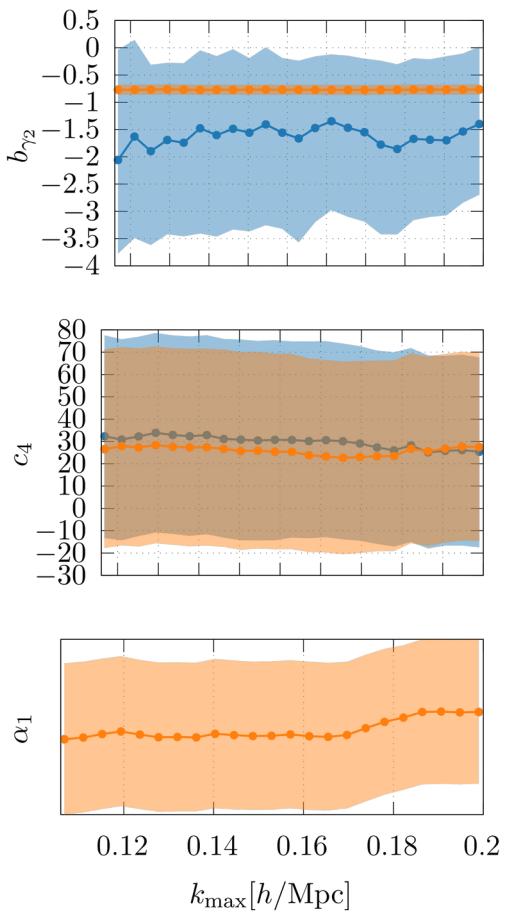
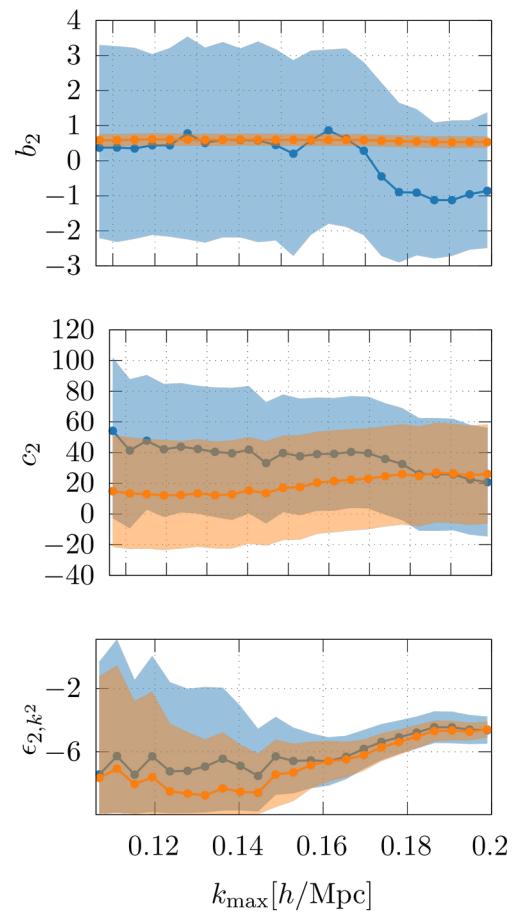
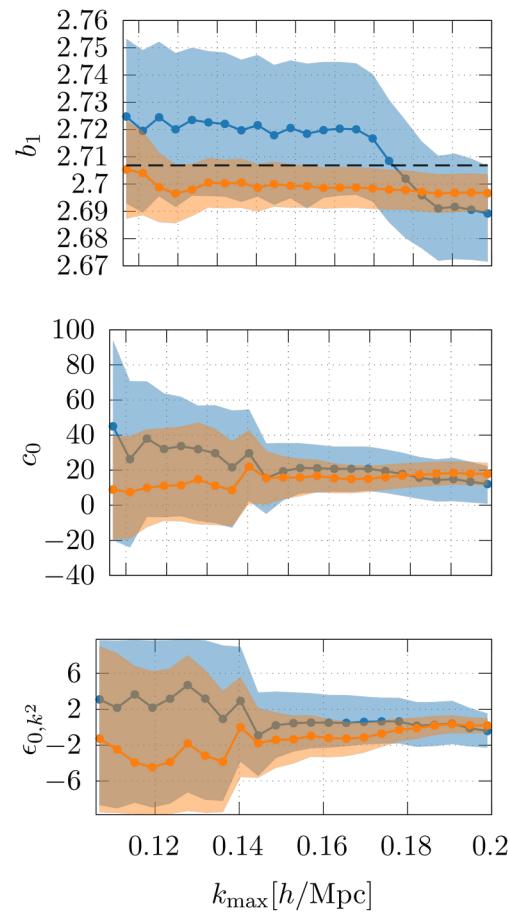
- Higher order multipoles



Bispectrum in redshift space [Rizzo,CM+ 2204.13628]

- Higher order multipoles
- Model selection (selecting minimal model & test bias relations)
- Different scale cuts for the multipoles, covariance approximation, binning effects

P+B in redshift space [CM+ in prep]

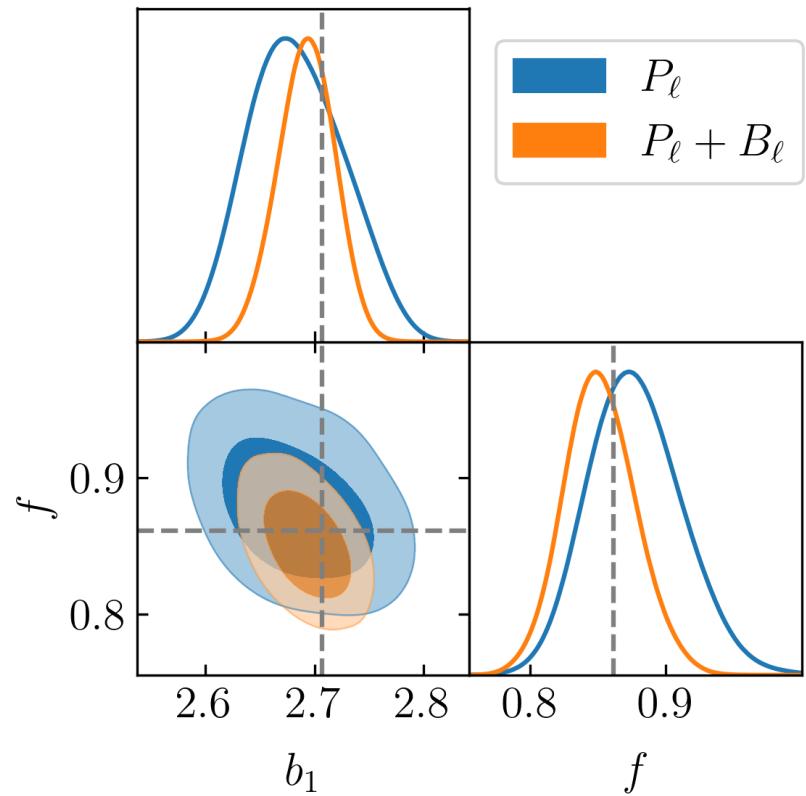


P+B in redshift space [CM+ in prep]

Rescale the covariance to $\sim 10 \text{ Gpc}^3/\text{h}^3$

(match first redshift bin of Euclid)

→ adding B0,**B2** improves constraints on f-bl



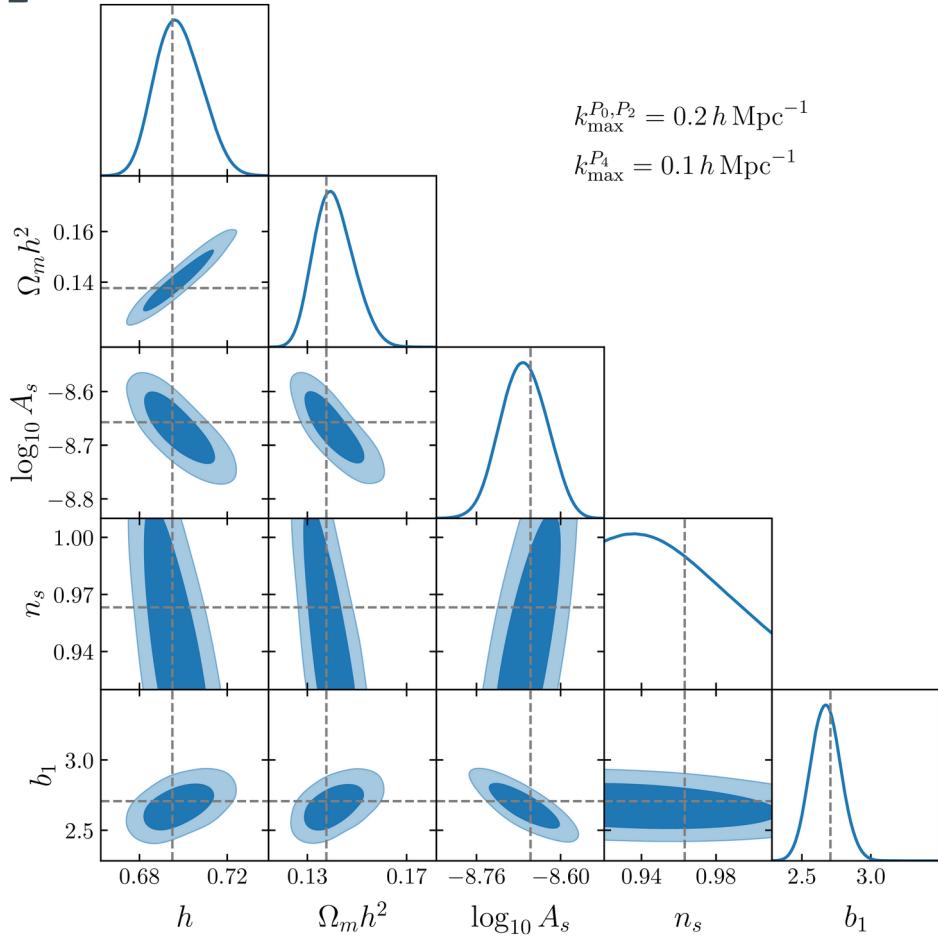
P+B in redshift space [CM+ in prep]

Rescale the covariance to $\sim 10 \text{ Gpc}^3/\text{h}^3$

(match first redshift bin of Euclid)

→ adding B0,B2 improves constraints on f-bl

→ cosmological parameters (covariance rescaled to match Euclid redshift bin):



Beyond LCDM

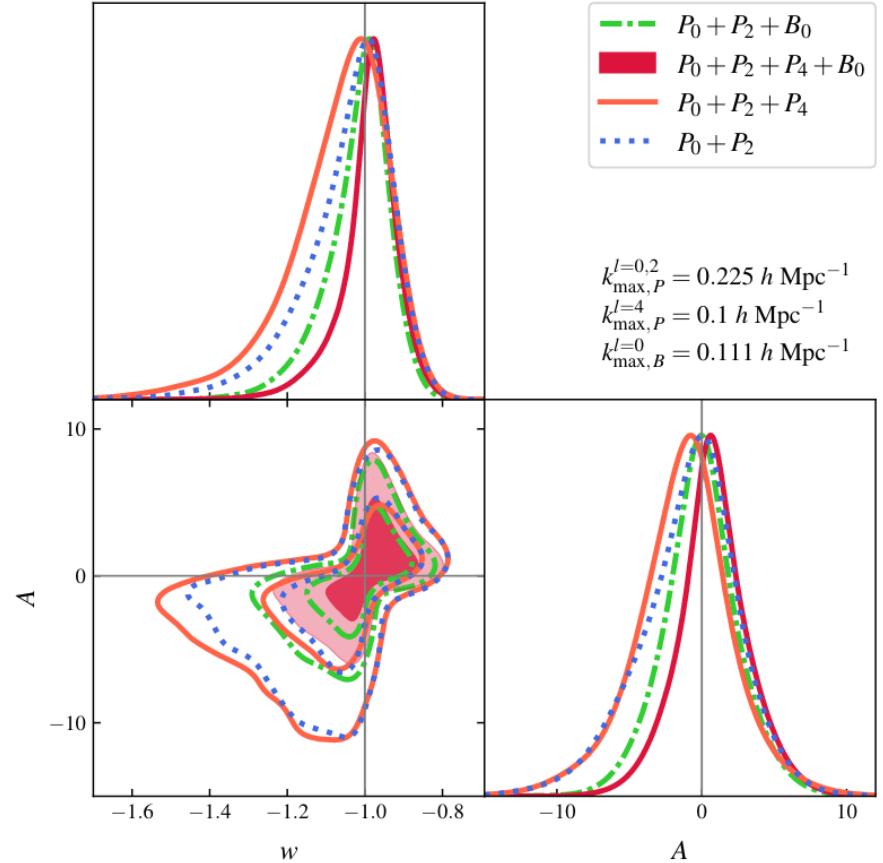
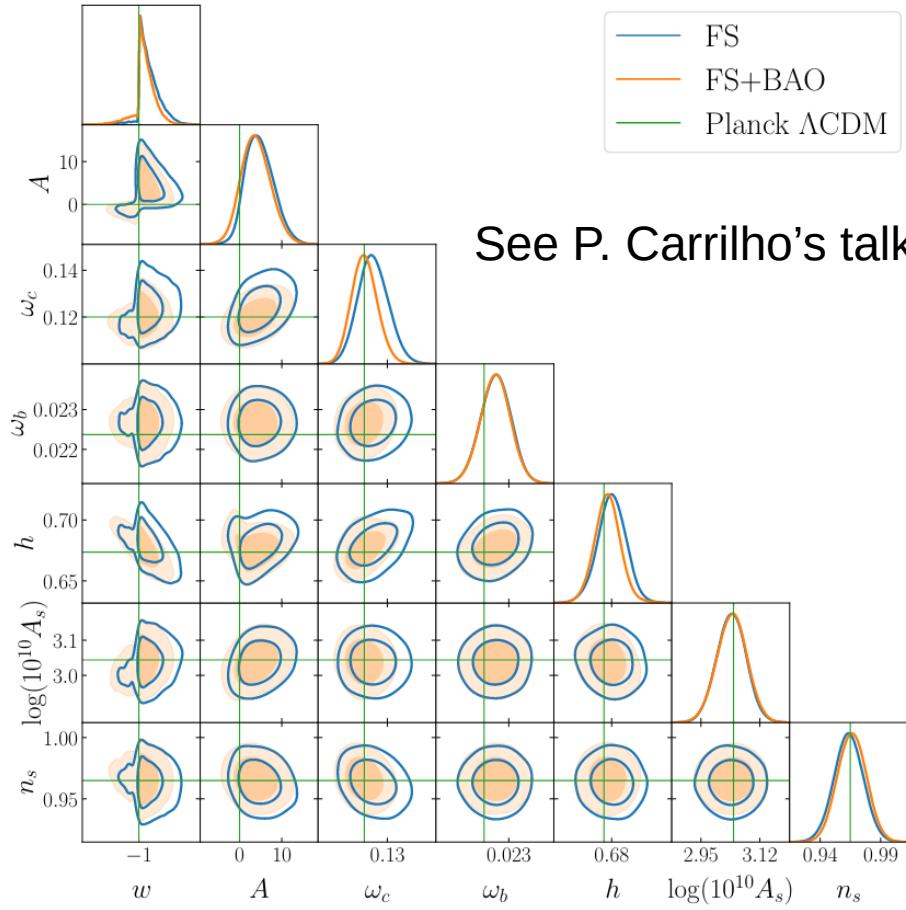
With P. Carrilho, M. Tsedrik, A. Pourtsidou
2207.14784, 2207.13011

- Interacting DE model: elastic scattering with momentum transfer
[Simpson+10, Pourtsidou+13]
- Rescale P_L with appropriate growth factor
- 2 additional parameters: w, A
- Validation on simulations [2106.13163] + BOSS analysis
[Carrilho, CM+] + inclusion of bispectrum [Tsedrik, CM+]

See P. Carrilho's talk

Beyond LCDM

With P. Carrilho, M. Tsedrik, A. Pourtsidou
 2207.14784, 2207.13011

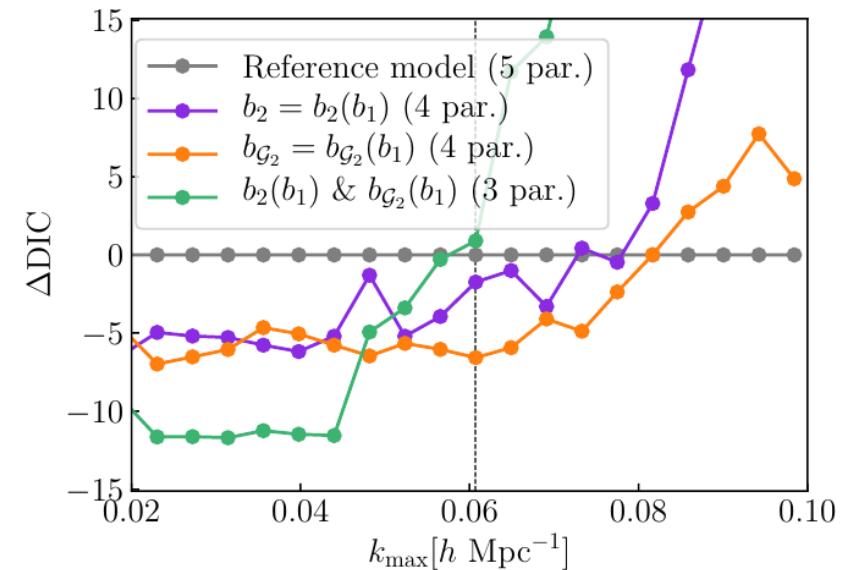
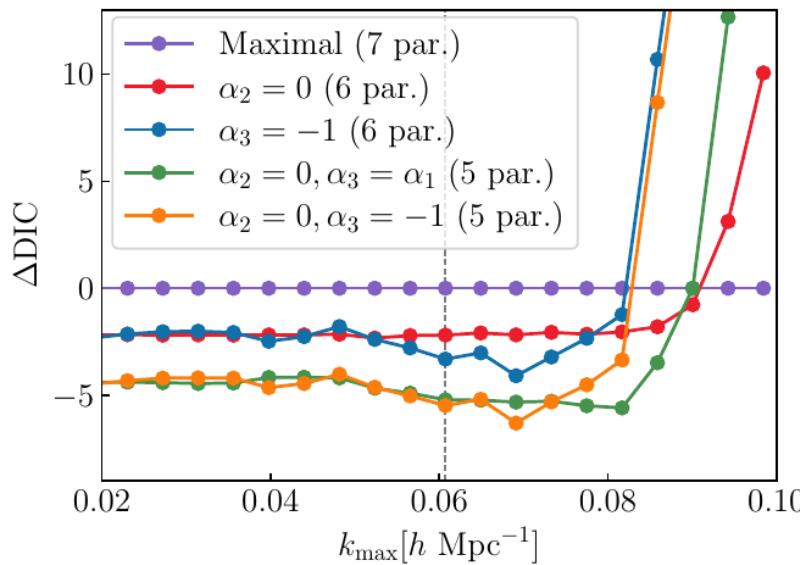


Summary & conclusion

- Pipeline for the power spectrum + bispectrum joint analysis of biased tracers
→ soon to be public!
- 1-loop power spectrum + tree-level bispectrum
- Validation on large set of sim → robust test of theoretical model
[Oddo+20,21; Pardede+22; Rizzo,CM+ 2204.13628; CM+ in prep]
- Validation and forecasts for Euclid + ported to official likelihood code
- Beyond-LCDM: IDE constraints from BOSS [Carrilho,CM+ 2207.14784],
inclusion of bispectrum [Tsedrik,CM+ 2207.13011]

Bispectrum in redshift space [Rizzo,CM 2204.13628]

- Higher order multipoles
- Model selection



P+B in redshift space [CM+ in prep]

