

Hybrid Lagrangian Bias Expansion for Covariance Matrices

Francisco Germano Maion
Donostia International Physics Center



European Research Council
Established by the European Commission

Fitting a Model to Data

Model: $x_i(\theta)$

Data: \hat{x}_i

Quality of the Fit $\longrightarrow \chi^2 = \sum_{i,j} (\hat{x}_i - x_i(\theta)) C_{ij}^{-1} (\hat{x}_k - x_j(\theta))$

Tightness of the Constraints $\longrightarrow \mathcal{L}(\theta) = \frac{1}{(2\pi)^{n/2} \sqrt{\det C}} \exp -\frac{1}{2} \sum_{i,j} (\hat{x}_i - x_i(\theta)) C_{ij}^{-1} (\hat{x}_k - x_j(\theta))$

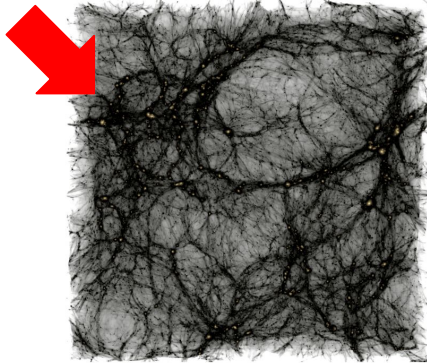
Maximum
Likelihood Point

Assuming the wrong covariance can lead to:

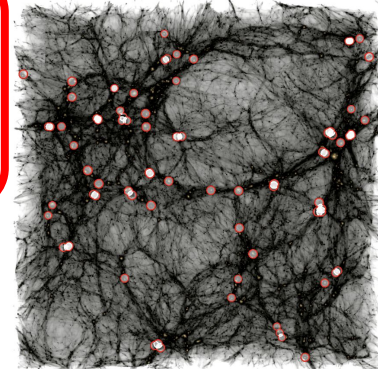
- Poor evaluation on quality of fit
- Distorted confidence regions
- Biased constraints on parameters

Covariances in Cosmology

Assume cosmology



Preferred galaxy formation
model (HOD, SHAM,
semi-analytics)

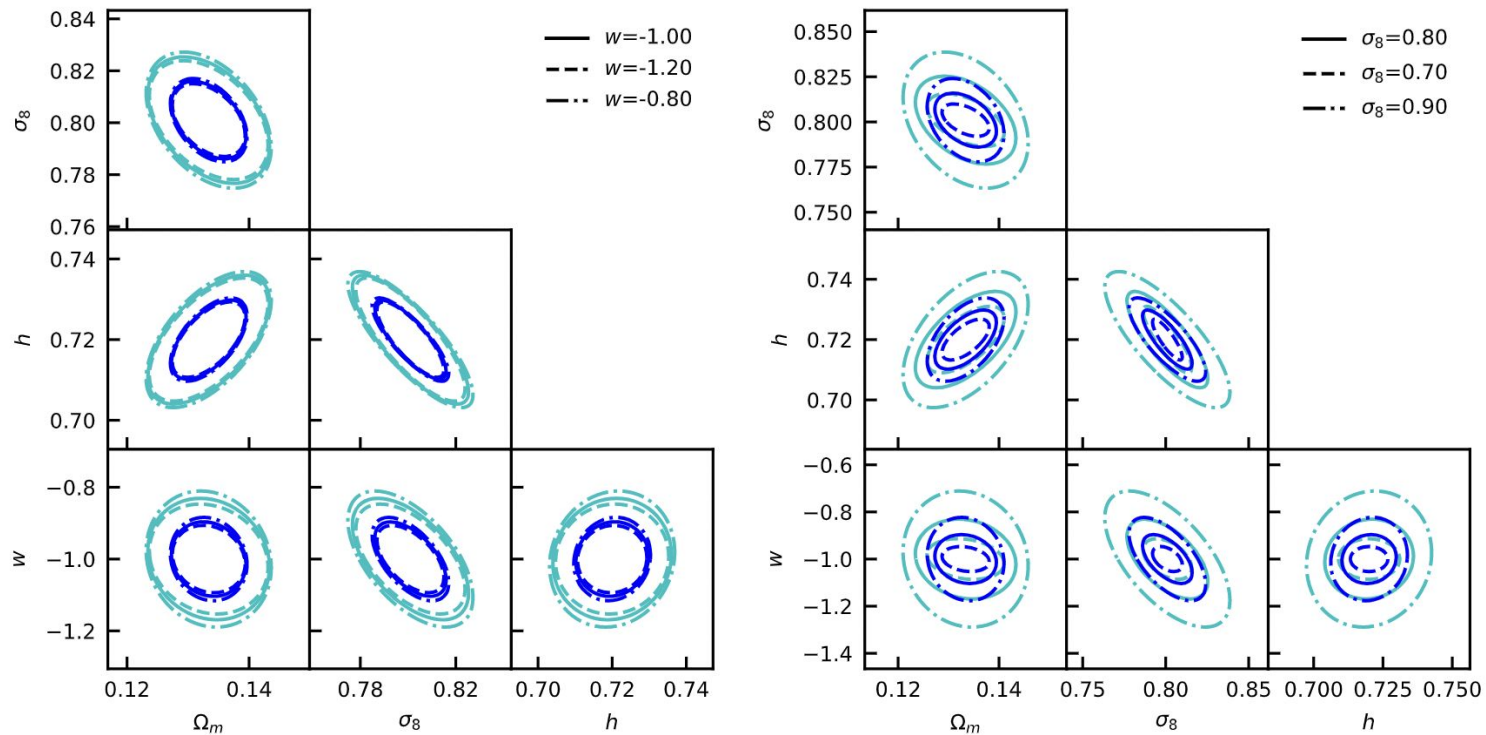


Repeat $\mathcal{O}(10^3)$ times

Estimate covariance from them

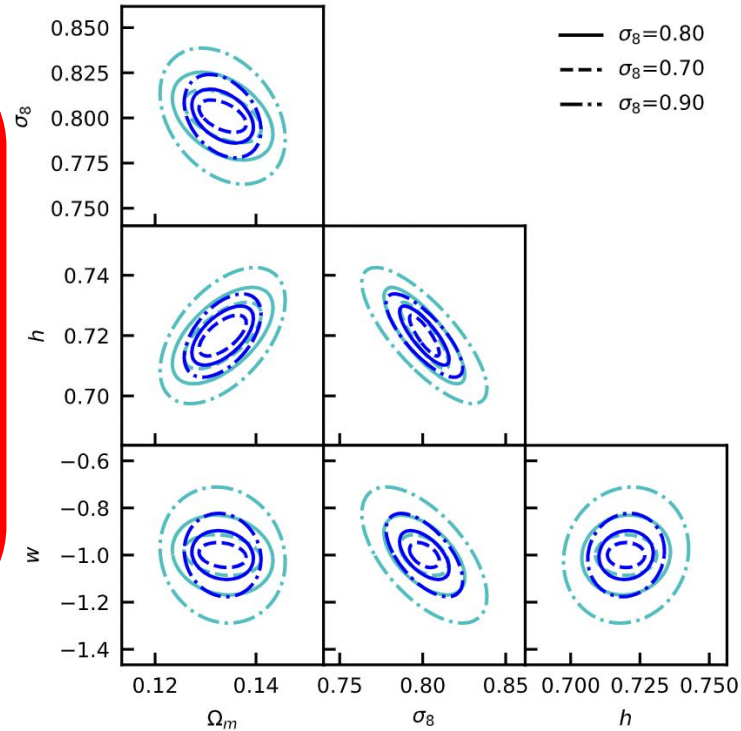
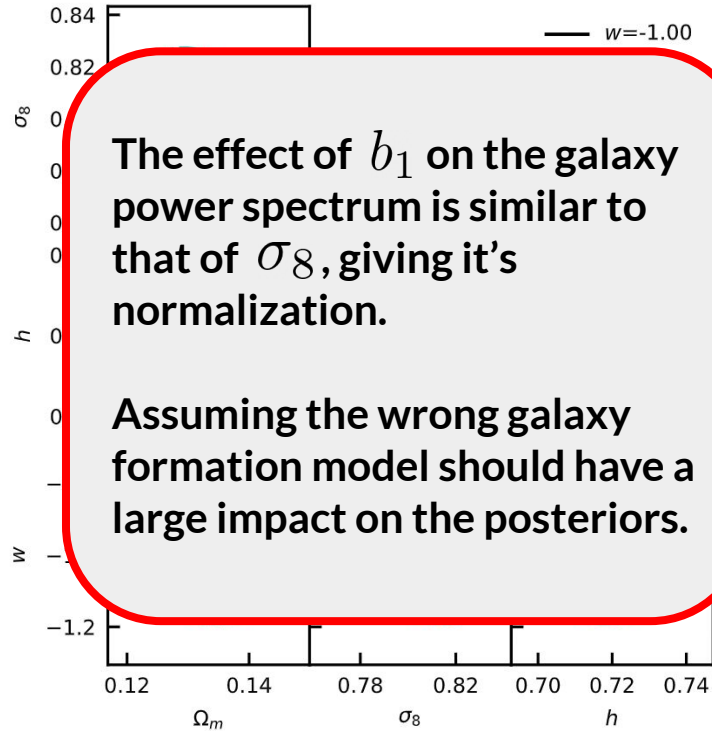
$$C(k, k') = \frac{1}{N-1} \sum_{i=0}^N (P_i(k) - \langle P(k) \rangle) (P_i(k') - \langle P(k') \rangle)$$

Choosing Wrong Fiducial Cosmology



Extracted from Blot et. al (2021)

Choosing Wrong Fiducial Cosmology



Bias Expansion for Covariance Matrices

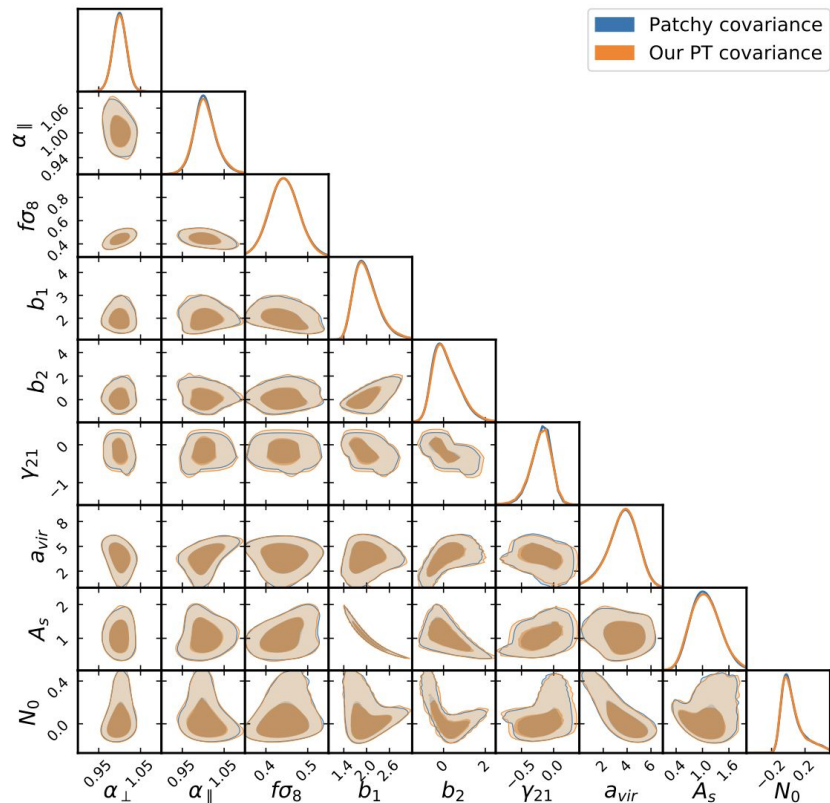
Bias Expansion for Covariances

$$C_{gg} = \sum_{ij,lm} b_i b_j b_l b_m C_{ij,lm}$$

+

Perturbative calculation of the
DM statistics

Works very well until scales of $k \approx 0.3 [h/Mpc]$

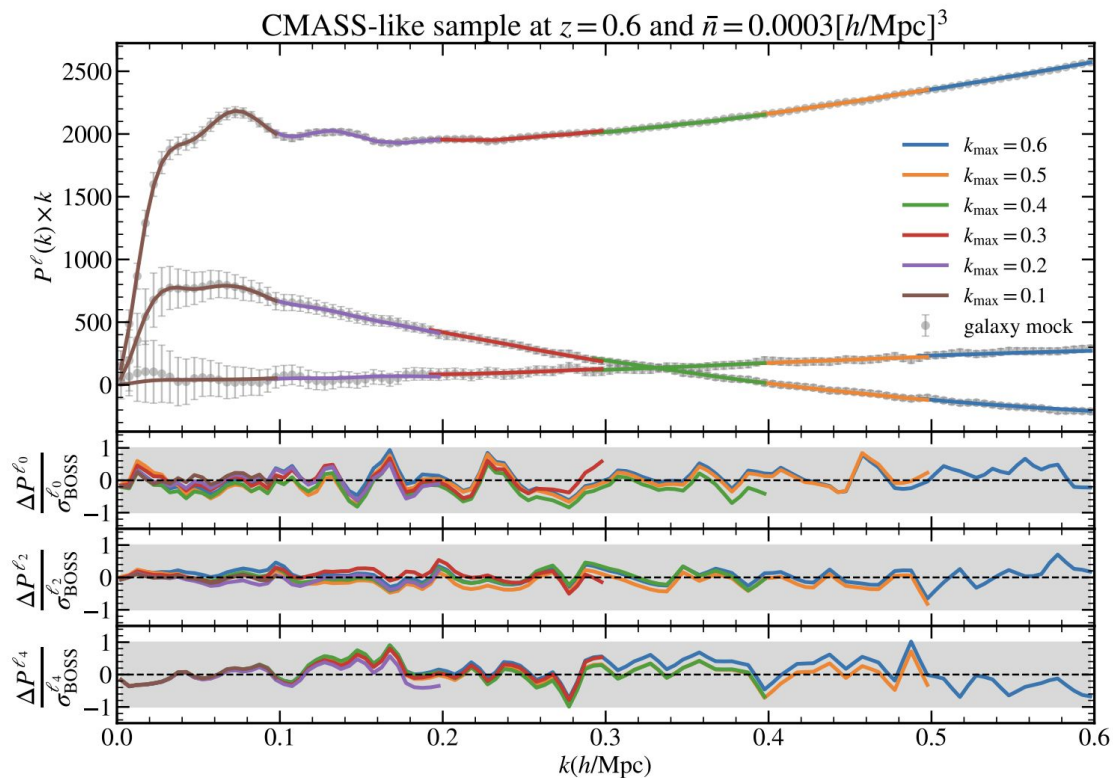


Taken from Wadekar & Scoccimarro (2019)

Hybrid Lagrangian Bias Model

We have developed a model which is accurate until very small scales.

Thus, we need a covariance matrix that is also accurate in this regime.



Hybrid Lagrangian Bias Model

Second-order Lagrangian bias expansion plus non-local term

$$1 + \delta_g(\mathbf{q}) = 1 + b_1 \delta_L + b_2 \delta_L^2 + b_{s^2} s^2 + b_\nabla \nabla^2 \delta_L + \mathcal{O}(\delta_L^3)$$

use displacements of particles computed from N-Body simulations to move tracers from Lagrangian to Eulerian space.

$$\rho_E(\mathbf{x}) = \int d^3 \mathbf{q} \rho_L(\mathbf{q}) \delta^D(\mathbf{x} - \mathbf{q} - \psi(\mathbf{q}))$$

The expression for the galaxy density in Eulerian space then becomes:

$$\delta_g(\mathbf{x}) \approx \delta + b_1 \delta_\delta + b_2 \delta_{\delta^2} + b_{s^2} s^2 + b_\nabla \delta_{\nabla^2}$$

and computing the galaxy power spectrum, one can write it as a linear combination of 15 auto and cross-spectra between the basis fields

$$P_{gg}(k) = \sum_{i,j \in [1, \delta, \delta^2, s^2, \nabla^2 \delta]} b_i b_j P_{ij}(k) + \frac{1}{\bar{n}}$$

Choosing Wrong Fiducial Cosmology

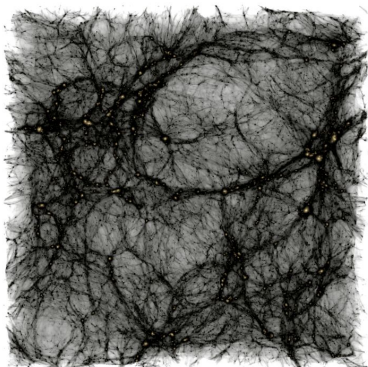
$$P_{gg}(k) = \sum_{i,j \in [1, \delta, \delta^2, s^2, \nabla^2 \delta]} b_i b_j P_{ij}(k) + \frac{1}{\bar{n}} \quad \rightarrow \quad C_{gg}(k, k') = \langle (P_{gg}(k) - \langle P_{gg}(k) \rangle) (P_{gg}(k') - \langle P_{gg}(k') \rangle) \rangle$$

$$C_{gg} = \sum_{ij, lm} b_i b_j b_l b_m C_{ij, lm} + \text{shot-noise}$$

$$C_{ij, lm}(k, k') = \frac{1}{N-1} \sum_{n=1}^N (P_{ij}^{(n)}(k) - \langle P_{ij}(k) \rangle) (P_{lm}^{(n)}(k') - \langle P_{lm}(k') \rangle)$$

Choosing Wrong Fiducial Cosmology

θ \longrightarrow Fiducial Cosmology



credit: Wechsler & Tinker



P_{ij}

$i, j \in [1, \delta, \delta^2, s^2, \nabla^2 \delta]$



$$C_{ij,lm}(k, k') = \frac{1}{N-1} \sum_{n=1}^N (P_{ij}^{(n)}(k) - \langle P_{ij}(k) \rangle) (P_{lm}^{(n)}(k') - \langle P_{lm}(k') \rangle)$$

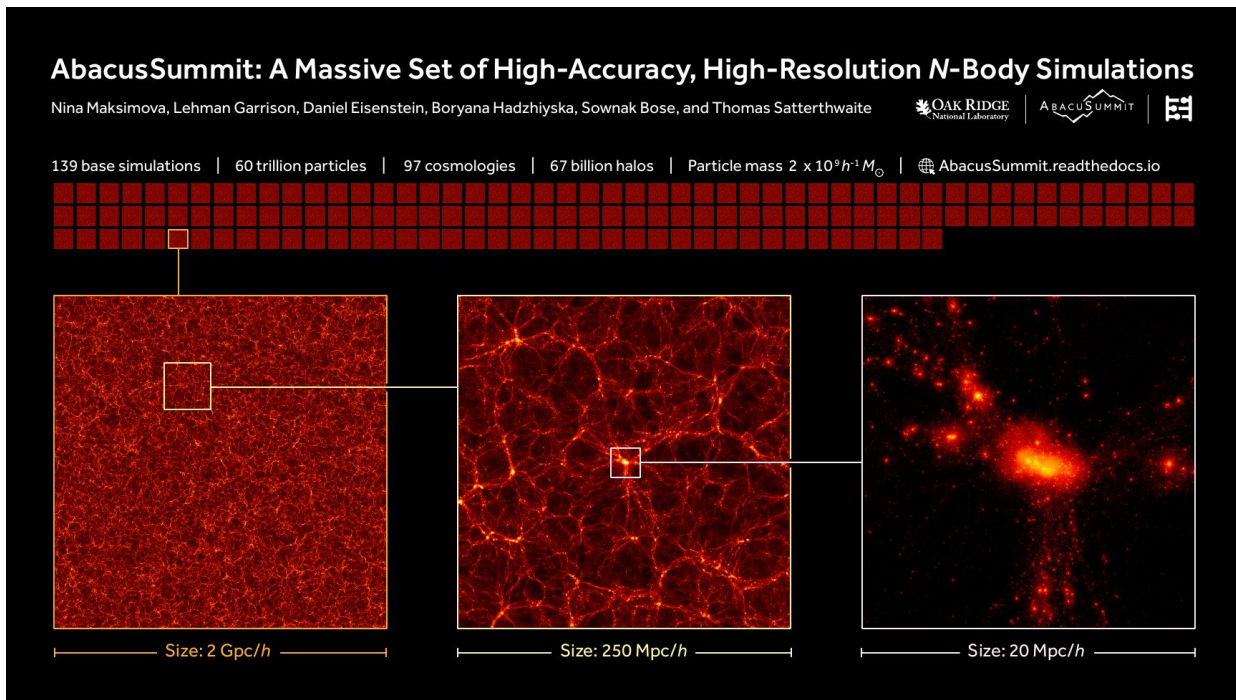
$$C_{gg} = \sum_{ij,lm} b_i b_j b_l b_m C_{ij,lm} + \text{shot-noise}$$

Repeat $\mathcal{O}(10^3)$ times

Set of 1800 simulations run
with Planck 2018 cosmology

$$L = 500[h/\text{Mpc}]$$

$$N_p = 536^3$$

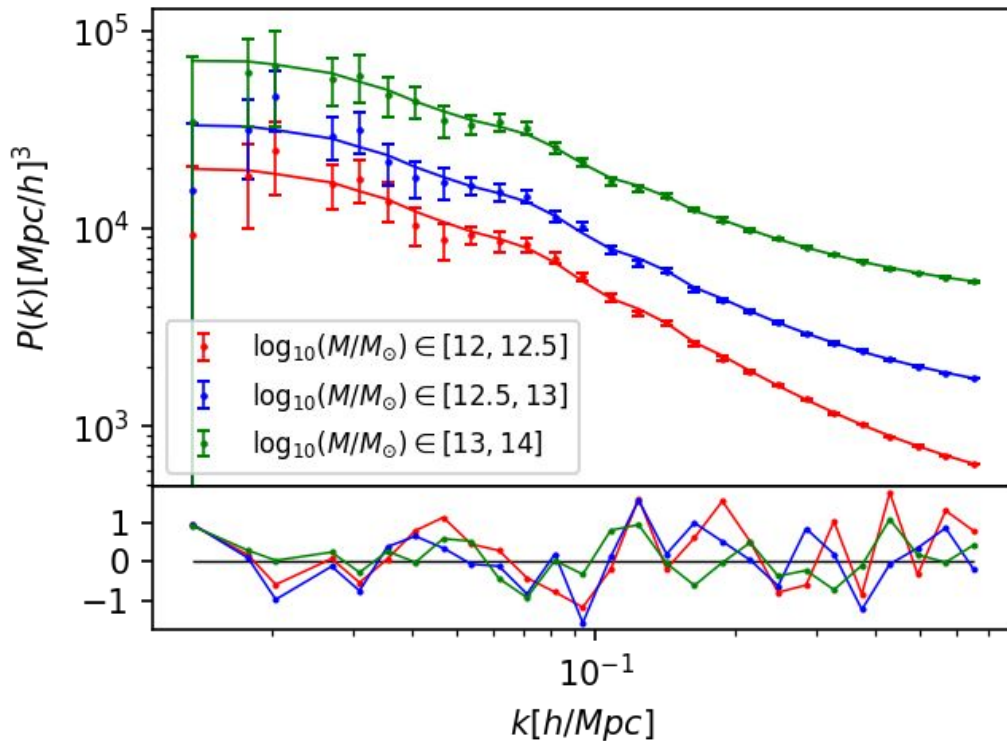


Bias Expansion for Covariances: Halos

Halos from the **AbacusSummit** suite of simulations divided into three mass bins.

We fit their power spectra using the coevolution relations from Zennaro et. al. (2021)

$$P_{hh}(k) = \sum_{i,j} b_i b_j P_{ij} + \frac{A_{SN}}{\bar{n}}$$



Bias Expansion for Covariances: Halos

Haloes, $k_d = 0.75 h \text{ Mpc}^{-1}$, $k_{\text{max}} = 0.7 h \text{ Mpc}^{-1}$

$$b_2^L(b_1^L) = -0.09143(b_1^L)^3 + 0.7093(b_1^L)^2 - 0.2607b_1^L - 0.3469$$

$$b_{s_2}^L(b_1^L) = 0.02278(b_1^L)^3 - 0.005503(b_1^L)^2 - 0.5904b_1^L - 0.1174$$

$$b_{\nabla^2 \delta}^L(b_1^L) = -0.6971(b_1^L)^3 + 0.7892(b_1^L)^2 + 0.5882b_1^L - 0.1072$$

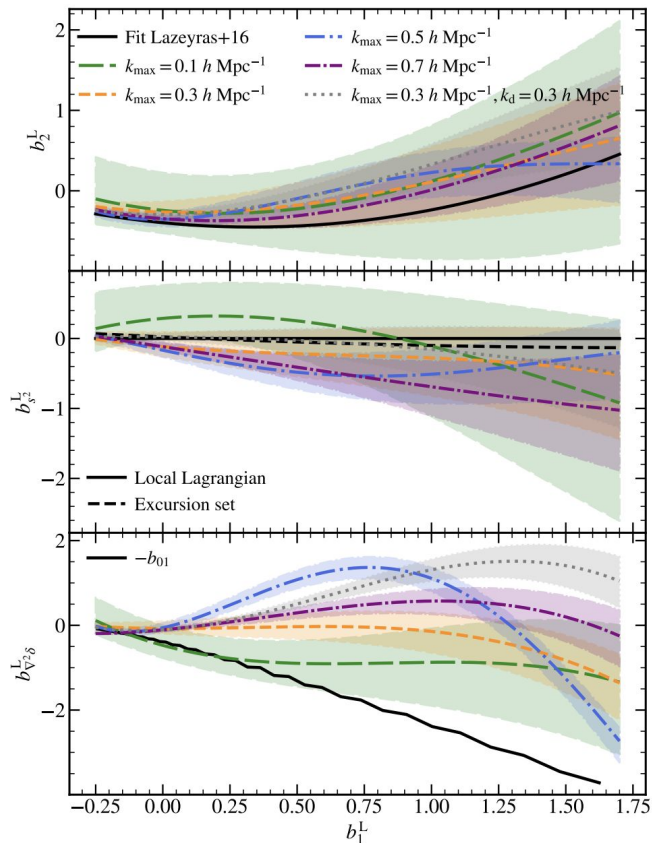
Galaxies, $k_d = 0.75 h \text{ Mpc}^{-1}$, $k_{\text{max}} = 0.7 h \text{ Mpc}^{-1}$

$$b_2^L(b_1^L) = 0.01677(b_1^L)^3 - 0.005116(b_1^L)^2 + 0.4279b_1^L - 0.1635$$

$$b_{s_2}^L(b_1^L) = -0.3605(b_1^L)^3 + 0.5649(b_1^L)^2 - 0.1412b_1^L - 0.01318$$

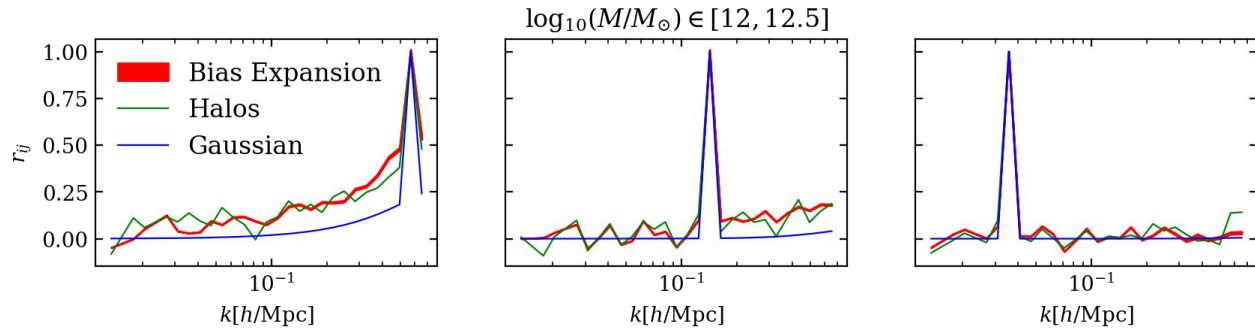
$$b_{\nabla^2 \delta}^L(b_1^L) = 0.2298(b_1^L)^3 - 2.096(b_1^L)^2 + 0.7816b_1^L - 0.1545$$

Zennaro, Angulo, Contreras, Pellejero-Ibanez, **FM (2022)**



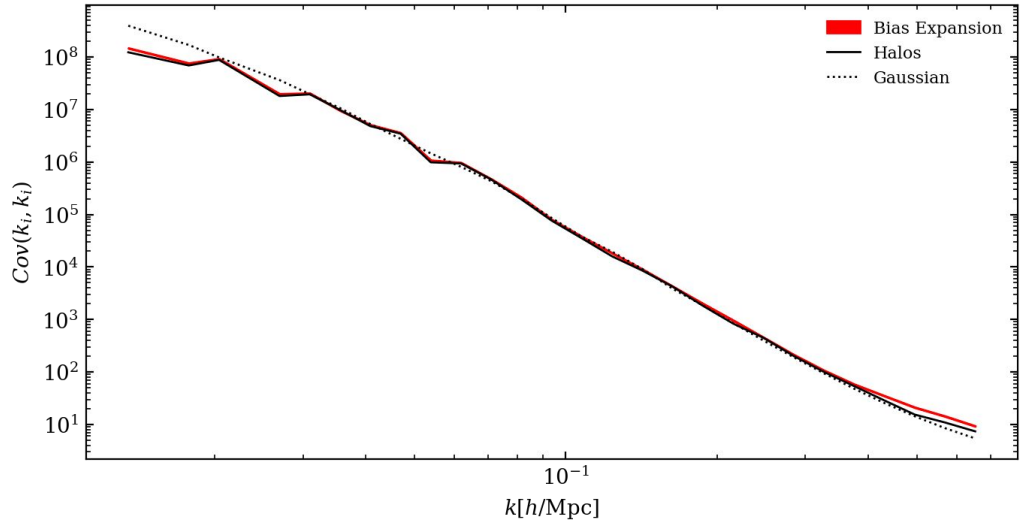
Results

$$r_{ij} = \frac{C(k_i, k_j)}{\sqrt{C(k_i, k_i)C(k_j, k_j)}}$$



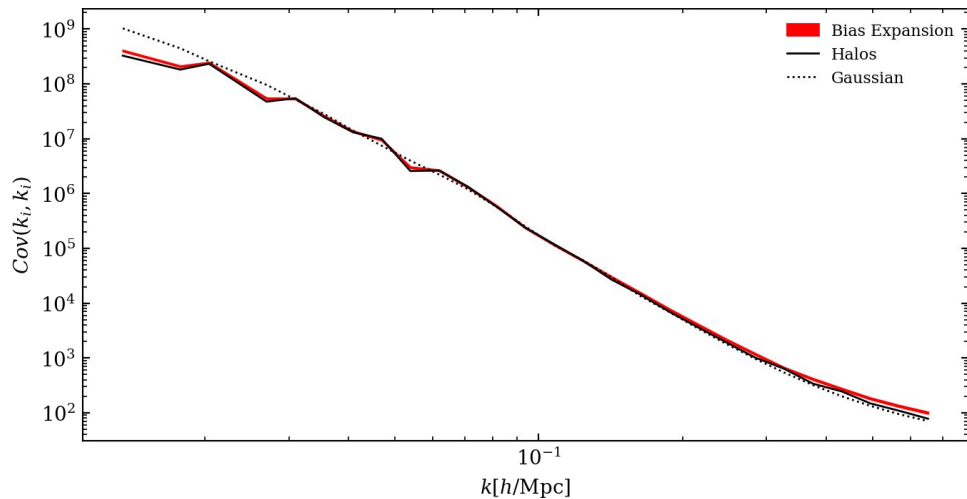
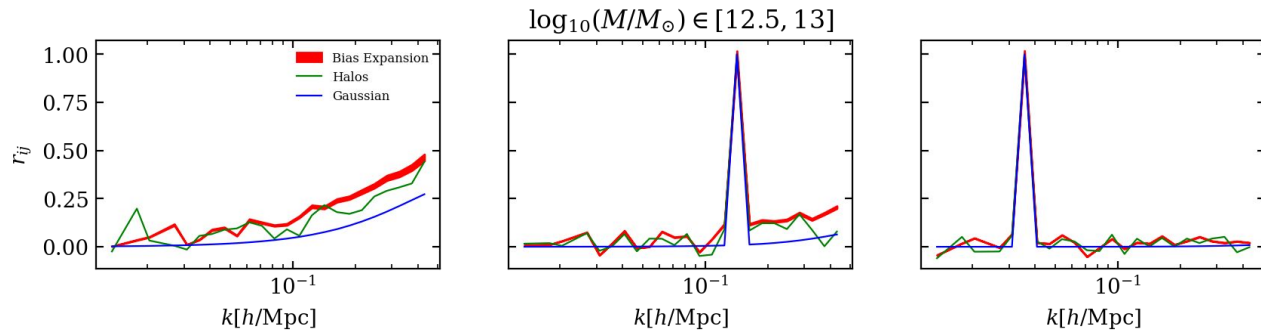
Good description of the covariance matrix
for scales such that

$$P(k) > \frac{1.5}{\bar{n}}$$



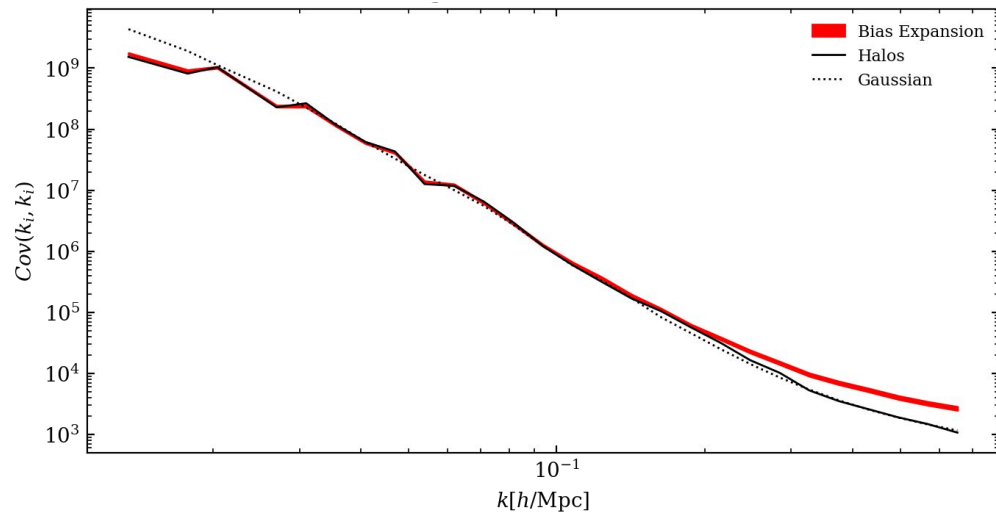
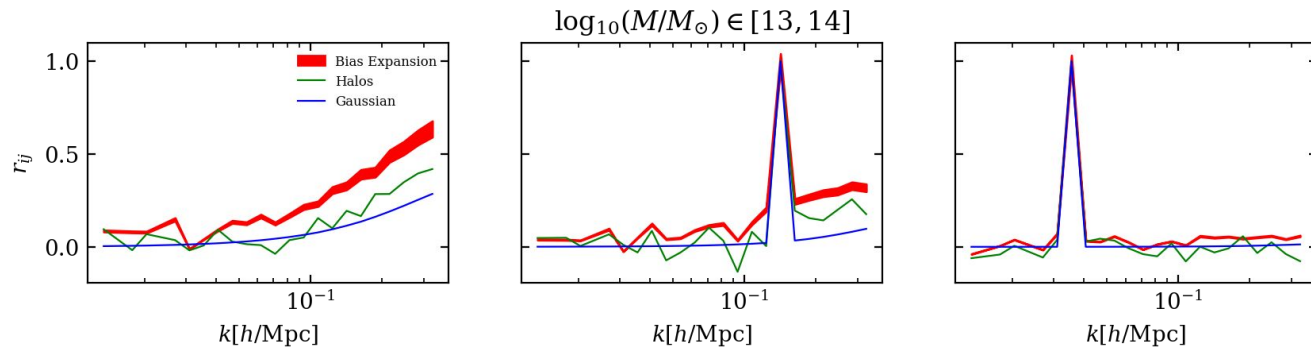
Results

$$r_{ij} = \frac{C(k_i, k_j)}{\sqrt{C(k_i, k_i)C(k_j, k_j)}}$$



Results

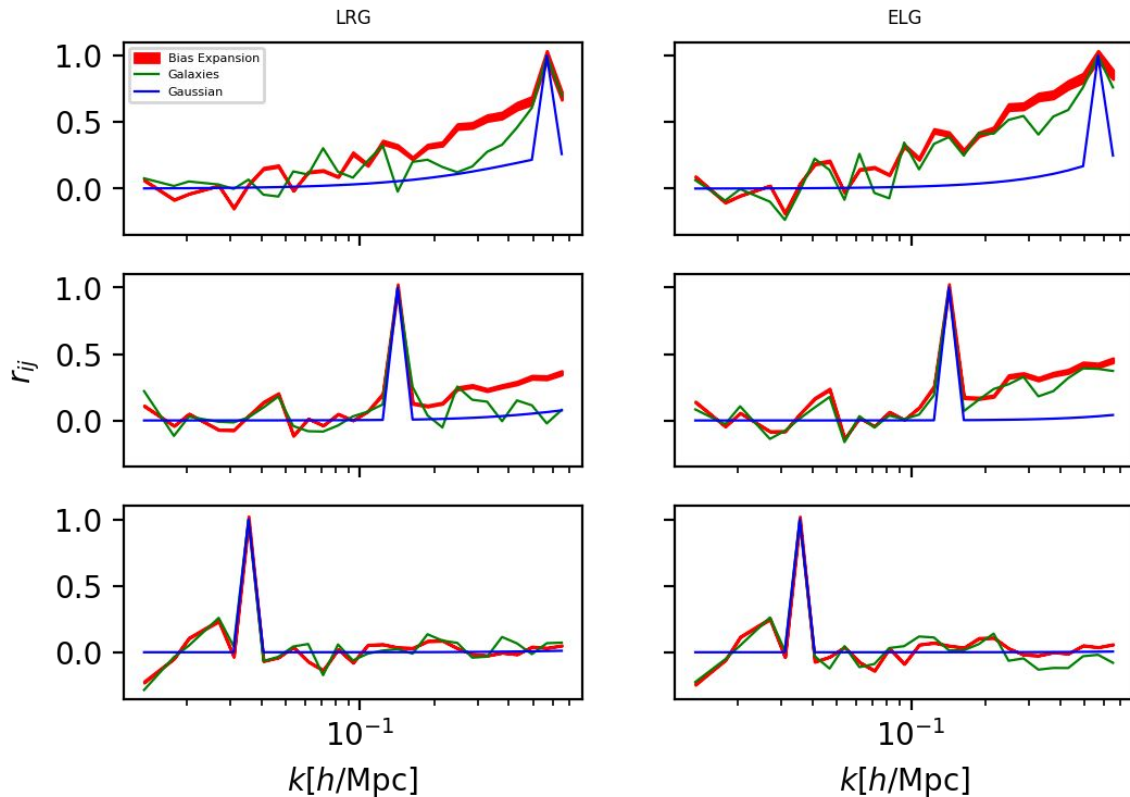
$$r_{ij} = \frac{C(k_i, k_j)}{\sqrt{C(k_i, k_i)C(k_j, k_j)}}$$



Results

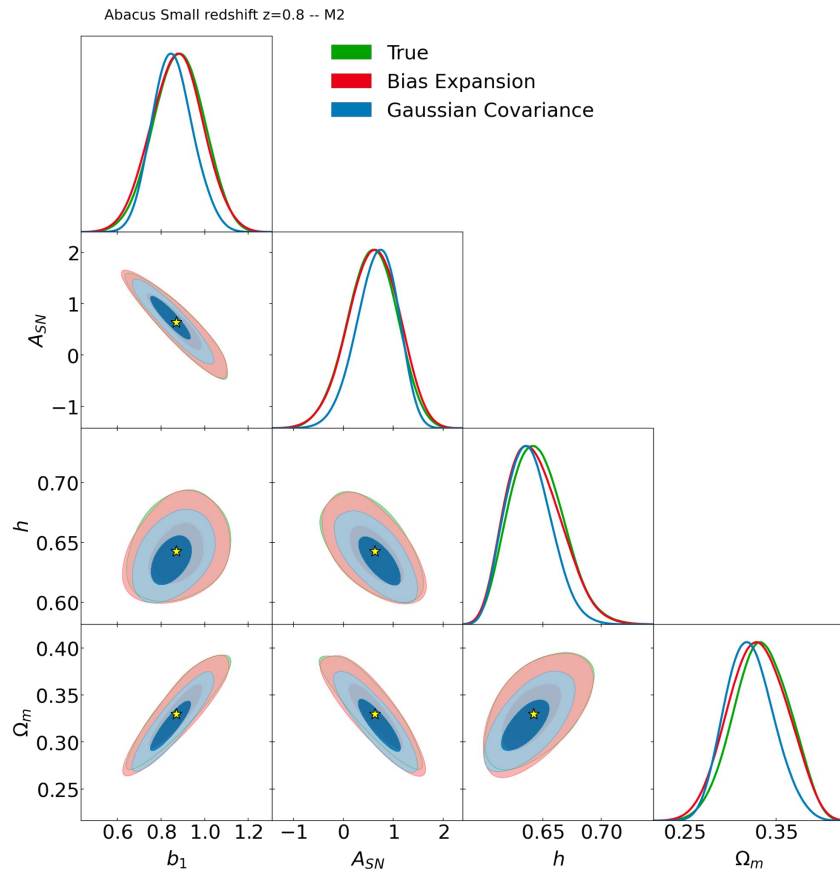
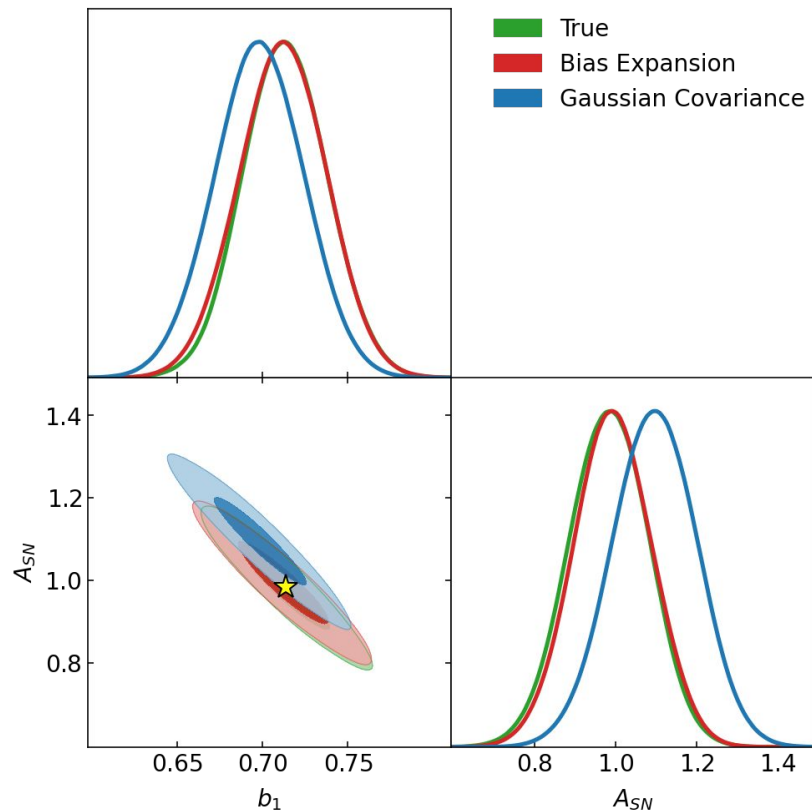
Using the Multi-Trace HOD implemented for AbacusUtils, we can populate the simulations with galaxies. This HOD is based in Alam et al. (2021).

$$r_{ij} = \frac{C(k_i, k_j)}{\sqrt{C(k_i, k_i)C(k_j, k_j)}}$$

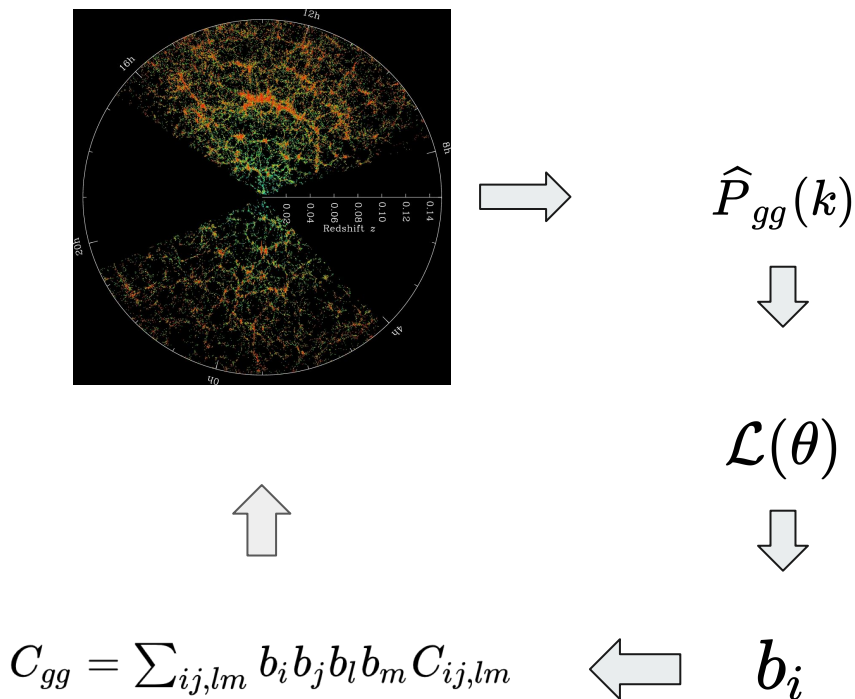


Bias Expansion for Covariances: Halos

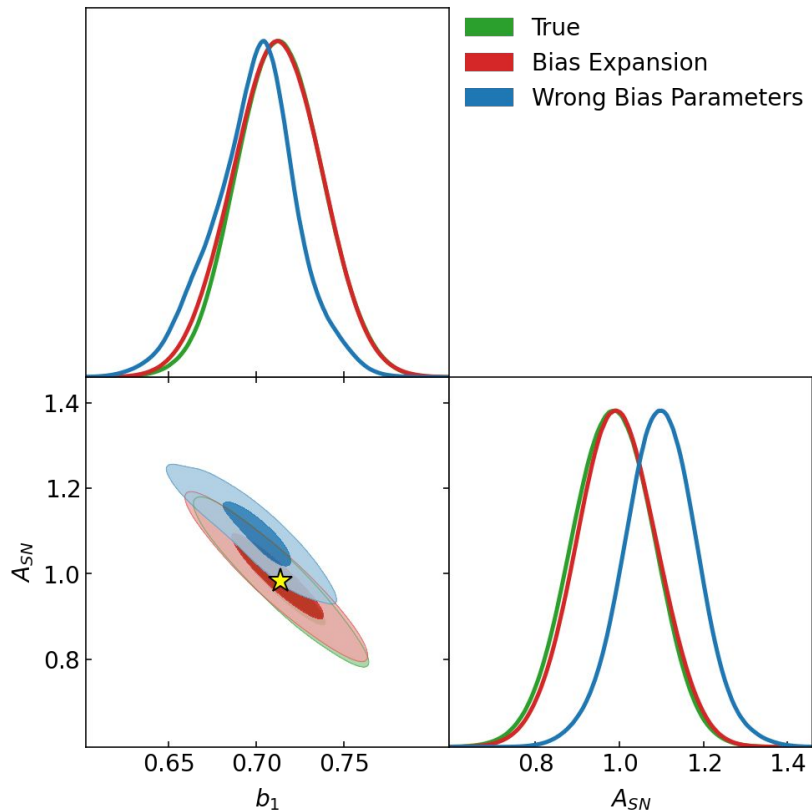
Abacus Small redshift $z=0.8$ -- M2



Bias Expansion for Covariances: Halos



Abacus Small redshift $z=0.8$ -- M2



Conclusions

Upshots

- Hybrid Lagrangian Bias Expansion describes well the covariance matrix of simulated halos
- Correct posterior regions are recovered with good precision
- Allows for great flexibility in the galaxy-formation model
- Works well even in mildly non-linear scales.

Future Developments

- Include third-order operators in the bias expansion
- Test the covariances for realistic analyses
- Further quantify the deviations caused by assuming wrong galaxy formation model

Bias Expansion for Covariances: Halos

Counting noise for a distribution of halos with Poisson statistics:

$$\begin{aligned} C_{SN} = & \frac{1}{\bar{n}} [B(\mathbf{k}, -\mathbf{k}) + B(\mathbf{k}, \mathbf{k}') + B(\mathbf{k}, -\mathbf{k}') \\ & + B(-\mathbf{k}, \mathbf{k}') + B(-\mathbf{k}, -\mathbf{k}') + B(\mathbf{k}', -\mathbf{k}')] \\ & + \frac{1}{\bar{n}^2} [P(\mathbf{k} + \mathbf{k}') + P(0) + P(\mathbf{k}' - \mathbf{k}) + P(\mathbf{k}) + P(-\mathbf{k}) + P(\mathbf{k}') + P(-\mathbf{k}')] \\ & + \frac{1}{\bar{n}^3} \end{aligned}$$