# The Effective Field Theory of Large-Scale Structure and Multi-tracer

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# Main Result



#### Multi-tracer block diagonalize the correlation matrix of parameters!

### (Gaussian) Fisher Forecast

For two tracers, the fisher matrix is diagolanized by the three degrees of freedom (L. Raul Abramo, *et al.*, 2021):

$$\begin{split} \mathcal{Q}_1 &= \mathcal{P}\left[1 + \frac{1}{2} \mathrm{log}\left(\frac{\mathcal{P}_{11}^2 + \mathcal{P}_{22}^2 + 2\mathcal{P}_{12}^2}{\mathcal{P}^2}\right)\right] \,,\\ \mathcal{Q}_2 &= \mathrm{log}\left(\frac{\mathcal{P}_{11}^2 + \mathcal{P}_{12}^2}{\mathcal{P}_{22}^2 + \mathcal{P}_{12}^2}\right) \,,\\ \mathcal{Q}_3 &= \frac{\mathcal{P}_{12}^2 - \mathcal{P}_{11}\mathcal{P}_{22}}{\mathcal{P}^2} \,. \end{split}$$

In the linear limit where  $\mathcal{P}_{12}^2 = \mathcal{P}_{11}\mathcal{P}_{22}$  these reduces to:

$$(\mathcal{Q}_1, \mathcal{Q}_2, \mathcal{Q}_3) = (\mathcal{P}, \log (\mathcal{P}_{11}/\mathcal{P}_{22}), 0).$$

## (Gaussian) Fisher Forecast

The information in the cross-spectrum is of the same order of the information in the auto-spectra.



L. Raul Abramo, et al., 2021.

#### Real Space EFT

The power spectra, for multi-tracer, is given by [computed with CLASS-PT (Anton Chudaykin, Mikhail M. Ivanov, Oliver H.E. Philcox and Marko Simonović, 2020)]:

$$\begin{split} P^{AB}(k) &= b_{1}^{A}b_{1}^{B}\left[P_{0}(k) + P_{1}(k)\right] + \frac{1}{2}\left(b_{1}^{A}b_{2}^{B} + b_{1}^{B}b_{2}^{A}\right)\mathcal{I}_{\delta^{2}}(k) + \left(b_{1}^{A}b_{\mathcal{G}_{2}}^{B} + b_{1}^{B}b_{\mathcal{G}_{2}}^{A}\right)\mathcal{I}_{\mathcal{G}_{2}}(k) \\ &+ \left[\left(b_{1}^{A}b_{\mathcal{G}_{2}}^{B} + b_{1}^{B}b_{\mathcal{G}_{2}}^{A}\right) + \frac{2}{5}\left(b_{1}^{A}b_{\Gamma_{3}}^{B} + b_{1}^{B}b_{\Gamma_{3}}^{A}\right)\right]\mathcal{F}_{\mathcal{G}_{2}}(k) \\ &+ \frac{1}{4}b_{2}^{A}b_{2}^{B}\mathcal{I}_{\delta^{2}\delta^{2}}(k) + b_{\mathcal{G}_{2}}^{B}b_{\mathcal{G}_{2}}^{A}\mathcal{I}_{\mathcal{G}_{2}\mathcal{G}_{2}}(k) + \frac{1}{2}(b_{2}^{A}b_{\mathcal{G}_{2}}^{B} + b_{2}^{B}b_{\mathcal{G}_{2}}^{A})\mathcal{I}_{\delta^{2}\mathcal{G}_{2}}(k) \\ &+ P_{\nabla^{2}\delta}^{AB}(k) + P_{\varepsilon^{A}\varepsilon^{B}}(k)\,, \end{split}$$

where

$$\begin{split} \mathcal{P}^{AB}_{\nabla^{2}\delta} &= -k^{2} \mathcal{P}_{0} \left[ 2 \frac{c_{s}^{2} b_{1}^{A} b_{1}^{B}}{k_{\mathrm{NL}^{2}}} + b_{1}^{A} \left( \mathcal{R}^{B}_{*} \right)^{2} + b_{1}^{B} \left( \mathcal{R}^{A}_{*} \right)^{2} \right] \\ &= - \left( b^{A}_{\nabla^{2}\delta} b^{B}_{1} + b^{B}_{\nabla^{2}\delta} b^{A}_{1} \right) k^{2} \mathcal{P}_{0} \,, \end{split}$$

and

$$P_{\varepsilon^A\varepsilon^B}(z,k) = \frac{1}{\sqrt{\bar{n}_A\bar{n}_B}} \left[ c_0^{AB} + c_2^{AB} \frac{k^2}{k_{\rm norm}^2} + \mathcal{O}(k^4) \right] \,.$$

 $\{b_{1}^{A}, b_{2}^{A}, b_{\mathcal{G}_{2}}^{A}, b_{\Gamma_{3}}^{A}, b_{\nabla^{2}\delta}^{A}, b_{1}^{B}, b_{2}^{B}, b_{\mathcal{G}_{2}}^{B}, b_{\Gamma_{3}}^{B}, b_{\nabla^{2}\delta}^{B}, c_{0}^{AA}, c_{2}^{AA}, c_{0}^{BB}, c_{2}^{BB}, c_{0}^{AB}, c_{2}^{AB}\} - 16 \, \text{parameters}$ 

#### Simulation

We have used the halos of the (huge) MultiDark simulation (A. Klypin, *et al.*, 2014).

Halo Data set	Mass range [log $M_{\odot}/h$ ]	$\bar{n} \left[ (\mathrm{Mpc}/h)^{-3} \right]$	number count
Halo A	[13.2, 13.5]	$1.44 imes10^{-4}$	$9.21 imes10^{6}$
Halo B	[13.5, 15.7]	$1.23 imes10^{-4}$	$7.90 imes10^{6}$
Halo $A + B$	[13.2, 15.7]	$2.67 imes10^{-4}$	$1.71  imes 10^7$



Thiago Mergulhão, Henrique Rubira, RV, L. Raul Abramo, 2021.

#### Priors

We used flat prior for the cosmological parameters

	Prior	
$\omega_{cdm}$	Flat [0.095, 0.14]	
h	Flat [0.6, 0.75]	
A	Flat [1.49, 2.8]	

And both flat and Guassian priors for (some) of the bias and stochastic parameters

	Prior <i>Flat</i>	Prior $G_0(\sigma)$	
$b_1$	Flat [1.0, 2.2]		
<i>b</i> <sub>2</sub>	Flat [-5.0, 5.0]	Gauss.(0, $\sigma$ )	
$b_{\mathcal{G}_2}$	Flat [-5.0, 5.0]	$Gauss(0,\sigma)$	
b <sub>F3</sub>	Flat [-10.0, 10.0]	$Gauss(0,2\sigma)$	
$b_{ abla^2\delta}$	Flat [-5.0, 5.0]		
<i>c</i> 0	Flat [-5.0, 5.0]		
<i>c</i> <sub>2</sub>	Flat [-5.0, 5.0]		

# Cosmological Constraints

#### Unbiased and stronger constraints!



# Model Complexity

# The constraining power is weakly affected by the number of parameters!



Thiago Mergulhão, Henrique Rubira, RV, L. Raul Abramo, 2021.

#### Stochastic cross-term

The cross stochastic term seems to not be very important.



# HOD Galaxies

More constraint for galaxies with populations with the similar  $b_1$ !



### Redshift Space EFT

The power spectra, for multi-tracer in redshift space, is given by [computed with CLASS-PT (Anton Chudaykin, Mikhail M. Ivanov, Oliver H.E. Philcox and Marko Simonović, 2020)]:

 $P^{AB}(k) =$  Usual thing with symmetrized bias parameters where

$$\begin{split} P_{\rm ct}^{AB}(k,\mu) &= \frac{k^2}{k_{\rm norm}^2} P_{\rm lin}(k) \left[ Z_1^A \left( c_{\rm ct,20}^B + c_{\rm ct,22}^B \mu^2 + c_{\rm ct,24}^B \mu^4 + c_{\rm ct,44}^B \frac{k^2}{k_{\rm norm}^2} \mu^4 + c_{\rm ct,26}^B \mu^6 + c_{\rm ct,46}^B \frac{k^2}{k_{\rm norm}^2} \mu^6 \right) + A \leftrightarrow B \right] \,, \end{split}$$

and

$$P_{\varepsilon^A \varepsilon^B}(k,\mu) = \frac{1}{\sqrt{\bar{n}_A \bar{n}_B}} \left[ c^{AB}_{\mathrm{st},00} + c^{AB}_{\mathrm{st},20} \frac{k^2}{k^2_{\mathrm{norm}}} + c^{AB}_{\mathrm{st},22} \frac{k^2}{k^2_{\mathrm{norm}}} f \mu^2 \right] \,.$$

$$\begin{cases} b_1^A, b_2^A, b_{G_2}^A, b_{\Gamma_3}^A, c_{ct,20}^A, c_{ct,22}^A, c_{ct,24}^A, c_{ct,26}^A, c_{ct,44}^A, c_{ct,46}^A, c_{st,00}^{AA}, c_{st,02}^{AA}, c_{st,22}^A, \\ b_1^B, b_2^B, b_{G_2}^B, b_{G_3}^B, c_{ct,20}^B, c_{ct,22}^B, c_{ct,24}^B, c_{ct,26}^B, c_{ct,44}^B, c_{st,46}^B, c_{st,00}^{BB}, c_{st,02}^{BB}, c_{st,22}^B, \\ c_{st,00}^{AB}, c_{st,02}^{AB}, c_{st,22}^B \end{cases} \} = 29 \, \text{parameters}$$

# Galaxy Catalogue

The galaxy catalogues were generated from BACCO using the SHAMe method (S. Contreras, *et al.*, 2020).

Galaxy set	SFR [ <i>M</i> <sub>☉</sub> /yr]	$\bar{n} [(Mpc/h)^{-3}]$
A	$\gtrsim 10^{-4}$	0.0015
В	$\lesssim 10^{-4}$	0.0015
A + B		0.0030
Galaxy set	SFR [ <i>M</i> <sub>☉</sub> /yr]	$\bar{n} [(Mpc/h)^{-3}]$
Galaxy set	$egin{array}{c c c c c c c c } SFR [M_{\odot}/yr] \ \gtrsim 10^{-1} \end{array}$	$\bar{n} \left[ (Mpc/h)^{-3} \right]$ 0.00015
Galaxy set A B	$egin{array}{c} {\sf SFR} \; [M_{\odot}/{ m yr}] \ \gtrsim 10^{-1} \ \lesssim 10^{-1} \end{array}$	$\frac{\bar{n}[(Mpc/h)^{-3}]}{0.00015}$ 0.00015

We assumed a Gaussian covariance (thanks to Enea di dio)

$$\begin{aligned} \operatorname{Cov}\left[P_{\ell}^{AB}(\boldsymbol{k}), P_{\ell'}^{CD}(\boldsymbol{k}')\right] &= \frac{\left(2\ell+1\right)\left(2\ell'+1\right)}{2V_{s}}\frac{\delta_{D}\left(k-k'\right)}{k^{2}}\\ &\sum_{\ell_{1}\ell_{2}\ell_{3}}\left(-1\right)^{\ell_{2}}\left(2\ell_{3}+1\right)\left(\begin{array}{cc}\ell_{1} & \ell_{2} & \ell_{3}\\ 0 & 0 & 0\end{array}\right)^{2}\left(\begin{array}{cc}\ell & \ell' & \ell_{3}\\ 0 & 0 & 0\end{array}\right)^{2}\\ &\left[P_{\ell_{1}}^{AC}\left(k\right)P_{\ell_{2}}^{BD}\left(k\right)+\left(-1\right)^{\ell'}P_{\ell_{1}}^{AD}\left(k\right)P_{\ell_{2}}^{BC}\left(k\right)\right].\end{aligned}$$

### Galaxy Catalogue



Thiago Mergulhão, Henrique Rubira, RV, 2023.

## Comparison in Redshift Space



Thiago Mergulhão, Henrique Rubira, RV, 2023.

## Comparison in Redshift Space



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# Model Complexity

# Higher order $\mu$ -terms are important for multi-tracer but the cross stochastic terms are not.



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#### Shot Noise Impact

We can not go to deeply on non-linear scales if the shot noise is to small.



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## Conclusion



#### Multi-tracer block diagonalize the correlation matrix of parameters!

Taylor



#### Co-evolution Relations



# HOD



Thiago Mergulhão, Henrique Rubira, RV, L. Raul Abramo, 2021.

#### eBOSS Footprint



Ruiyang Zhao, et al., 2023.