# PBJ: Bayesian analysis of the galaxy power spectrum and bispectrum

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#### **Motivation: Stage-IV surveys**

Mapping the Universe over unprecedented volumes  $\rightarrow$  high precision measurements

Dark energy? Dark matter? Massive neutrinos?

- Need **fast** and **accurate** tools
- Robust validation on simulations/synthetic datavectors
- Modelling **systematics** is crucial



### Galaxy clustering

Homogeneous distribution of overdensities

- ightarrow gravity
- $\rightarrow$  clustered distribution of galaxies

## Galaxy distribution $\leftrightarrow$ cosmological model



#### **Galaxy clustering**

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*Joint likelihood pipeline for galaxy power spectrum + bispectrum* 

- EFT model for power spectrum ightarrow **ported to Euclid likelihood CLOE** 
  - FastPT for fast loop corrections
  - $\circ\,$  Emulators for  $P_L$  (or Boltzmann solver)
  - wiggle-nowiggle decomposition for infra-red resummation
- Tree level bispectrum
  - expansion for fast Alcock-Paczinsky distortions
- Gaussian likelihood + corrections for noise in numerical covariance matrix

- Fully in python
- Extremely fast:  $P_{gg}$  evaluation in  $\sim 0.04$ s,  $B_{ggg}$  in  $\sim 0.1$ s  $\circ$  Euclid-like datavector: convergence in  $\mathcal{O}(50)$  cpu hours
- Analytic marginalisation for nuisance parameters
- Option to run in fast mode when cosmology is fixed
- Several samplers: Metropolis-Hastings, affine invariant (emcee), nested (ultranest), ML powered (pocomc, nautilus)

Modular structure:

- **theory module**: implements computation of  $P_L$ ,  $P_{\rm NL}$ ,  $B_T$ , as well as several cosmological quantities (background evolution, growth functions, distance measurements)
- **likelihood module**: implements likelihoods functions and samplers
- **binning module**: implements binning strategies for observables
- **PBJ main module**: class to initialise all required quantities starting from the parameter file

Only needs three input files to run:

- parameter file (yaml file with path to data and run specifications)
- prior file (yaml file with prior distributions for parameters)
- minimal python script to run mcmc:

```
import PBJ
from tools.param_handler import read_file
init_dict = read_file("paramfile.yaml")
pbj = PBJ.pbj(init_dict)
pbj.initialise_full()
pbj.run_sampler(NmaxP=33.5, nsteps=20000, nwalker=200)
```

#### The EFT model

Anisotropic (redshift space) galaxy power spectrum:

# Validation on simulations

Minerva dataset: 298 N-body simulations+10k Pinocchio mocks for covariance

 $\Rightarrow$   $\sim$   $1000~({
m Gpc}/h)^3$  volume, extremely tight errorbars

- validate bias relations to reduce parameter space
   model selection tools
- inclusion of bispectrum improves constraints



#### BOSS analysis: growth index and massive neutrinos

#### CM+23, 2306.09275

- constraints on  $\gamma$  +  $M_{
  u}$  from full shape
- forecasts for Stage IV surveys
- profile likelihood to mitigate projections



#### Euclid

Currently used in several preparatory projects:

- modelling challenge
   Pezzotta+23 milestone 7
- Beyond  $\Lambda$ CDM models Bose+23 milestone 7
- IST:NL pre-launch KPs milestone 8?
- forecasts for SPV3 for P and P+B with  $w_0, w_a$



#### Work in progress: future KPIs

- Window convolution (with Jacopo Salvalaggio) milestone 8?
- Inclusion of post-reconstruction BAO (with Elena Sarpa) milestone 8? 🗸
- Public code release milestone 8? 🔴
- Interlopers (with Matilde Barberi Squarotti) milestone 9? 🔴
- Massive neutrinos (with Emilio Bellini) milestone 9? —