

Panel Discussion on  
Limits of  $pt$ , setting external priors, setting  $k_{NL}$ ,  
how to squeeze info from LSS on small scales

Maria Tsedrik, Jamie Sullivan, Guido D'Amico

# Why do we need priors?

$$\chi_m^2(\Omega | d) \approx \chi^2(\Omega, n_{\text{best-fit}} | d) + \log [\det F_2(\Omega)]$$

$\uparrow$   
 in linear nuisance

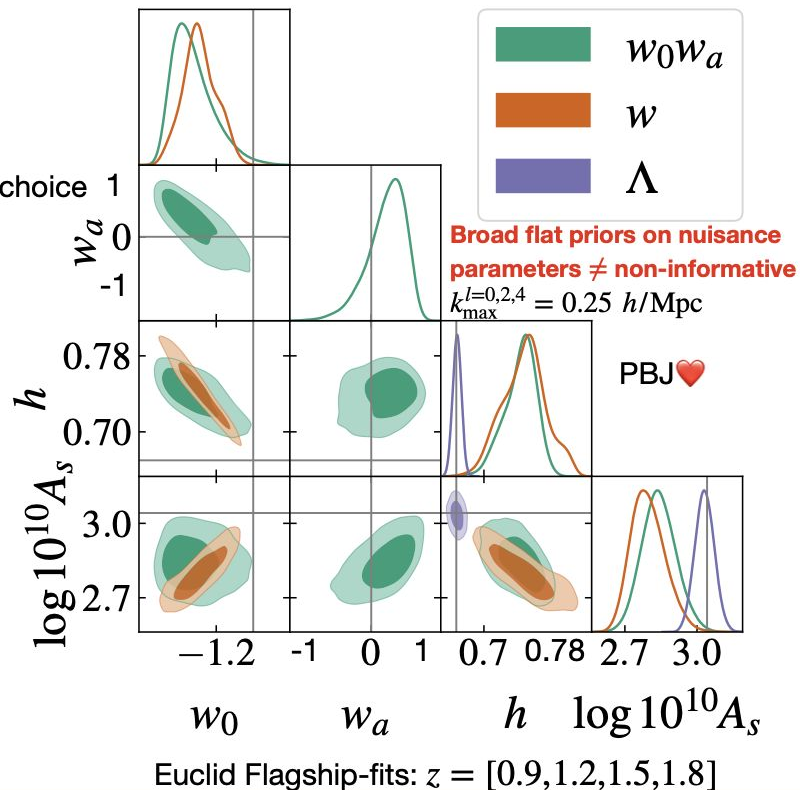
Profile likelihood      Laplace term

$F_{2,ij} = P_{i,l}^m \text{Cov}_{l'l'}^{-1} P_{j,l'}^m + C_{n,ij}^{-1}$

sensitive to parameterisation      sensitive to prior-choice

## Potential “solutions”

- adding more data (bispectrum/BAO/CMB)
- statistical solutions (re-parametrisation of parameters, profiling)
- get more info on nuisance parameters



Copilot, show me a typical telecon on priors and projection effects



# Priors...

- In Bayesian analysis, data update our belief on the model and its parameters. Must start from a probability measure on parameters.  
**There is no “uninformative prior”.**
- And what if data are not precise enough?

The status of LSS EFT analysis: two manifestly equivalent implementations of the manifestly correct theory on manifestly the same data produce manifestly different results. [#cosmology](#)

This seems a little strong. Ignoring bispectrum loops (which are quite preliminary), the main power spectrum results from the three groups are similar. This is true with / without windows (and with broad priors on eft parameters). But we should check for residual differences!

Priors on counterterms? The encarnation of the devil on Earth.



- Just wait. Sooner or later data will just not care about the priors
- Perturbativity prior

$$\mathcal{P}_P = \frac{1}{2N_{\text{bins}}^P} \sum_{i \in \text{bins}_P} \left( \frac{P_{1\text{-loop}}^h(k_i)}{\sigma_P^{\text{P.P.}}(k_i)} \right)^2 \quad \mathcal{P}_B = \frac{1}{2N_{\text{bins}}^B} \sum_{i \in \text{bins}_B} \left( \frac{B_{1\text{-loop}}^h(k_1^i, k_2^i, k_3^i)}{\sigma_B^{\text{P.P.}}(k_1^i, k_2^i, k_3^i)} \right)^2$$

$$\sigma_P^{\text{P.P.}}(k) \sim S^P(k) P_{1\text{-loop}}^{k_{\text{max}}}, \quad \sigma_B^{\text{P.P.}}(k_1, k_2, k_3) \sim S^B(k_1, k_2, k_3) B_{1\text{-loop}}^{k_{\text{max}}}$$

$$S_{1\text{-loop}}^P(k) \sim b_1^2 P_{11}(k) \left( \frac{k}{k_{\text{NL}}} \right)^{3+n(k)} \quad S_{1\text{-loop}}^B(k_1, k_2, k_3) \sim B_{211}^h(k_1, k_2, k_3) \sum_{i=1}^3 \left( \frac{k_i}{k_{\text{NL}}} \right)^{3+n(k_i)}$$

- Get some controlled UV information.  
We used some old fitting formulas from simulations.  
Now, the idea is to do a dedicated search. As a first step, fitting HOD. But have to marginalise over them.

# Small-scale Information?

How small is “small”? 10 Mpc/h? 2 Mpc/h? 0.5 Mpc/h?

Could you ever trust near or below the halo scale?

And at high- $z$ ?

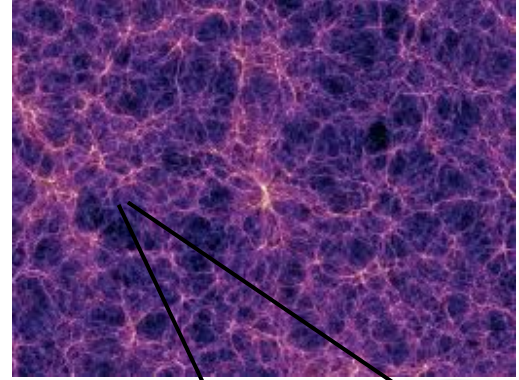
Roll up our sleeves and do astrophysics? (Priors)

Simulations *and* data? Redshift evolution?

Or maybe some statistics?

Abandon the straightforward bias parameterization?

(Baryons, satellites, FoG, mass profiles.....)



# Summary of Questions

- How do we get more information on the nuisance parameters of PT models?
- Can we determine theoretically what the scales of nonlinearities are? How do they depend on cosmology?
- Will we still need simulations as a check of PT and  $k_{\text{max}}$ ?  
We could use self-consistent NNLO methods
- What physical priors can we place on nuisance parameters?
- Can higher-loop PT extract more information from smaller scales? Is it limited by the large number of extra nuisance parameters?