

Perturbation theory challenge for cosmological parameters estimation

PTchat@Kyoto

April 8

Yukawa Institute of Theoretical Physics

Ken Osato

Institut d'Astrophysique de Paris

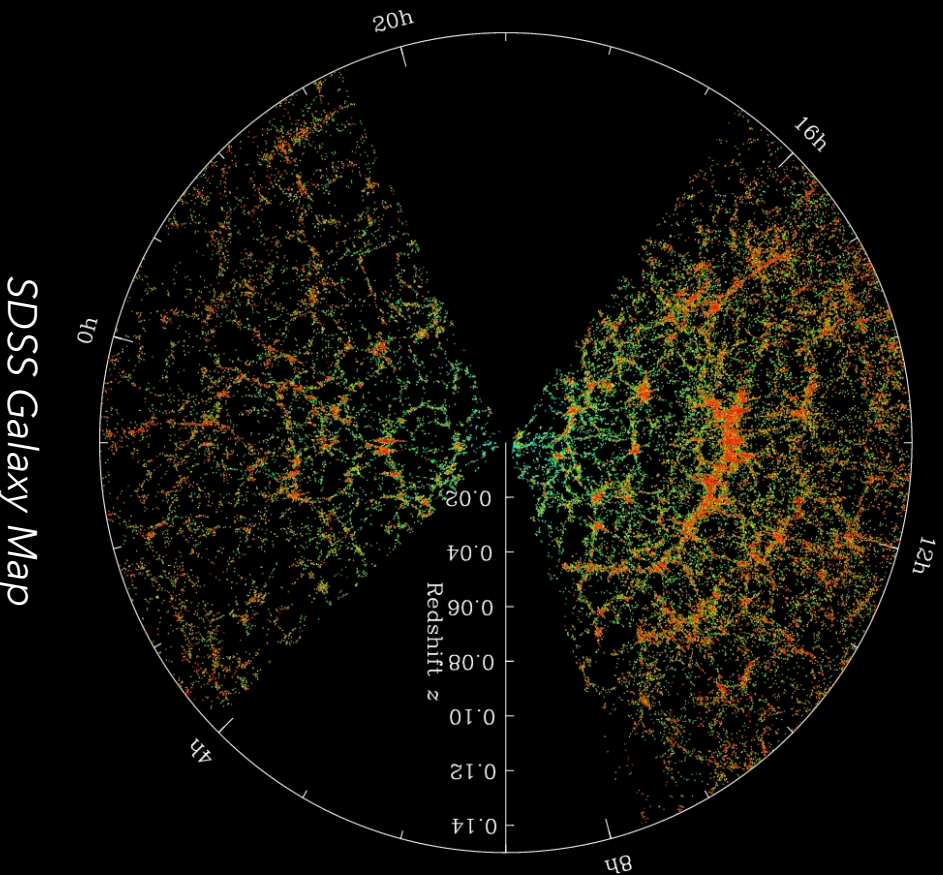
In collaboration with

Takahiro Nishimichi (YITP), Francis Bernardreau (LAP/IPhT),
and Atushi Taruya (YITP)

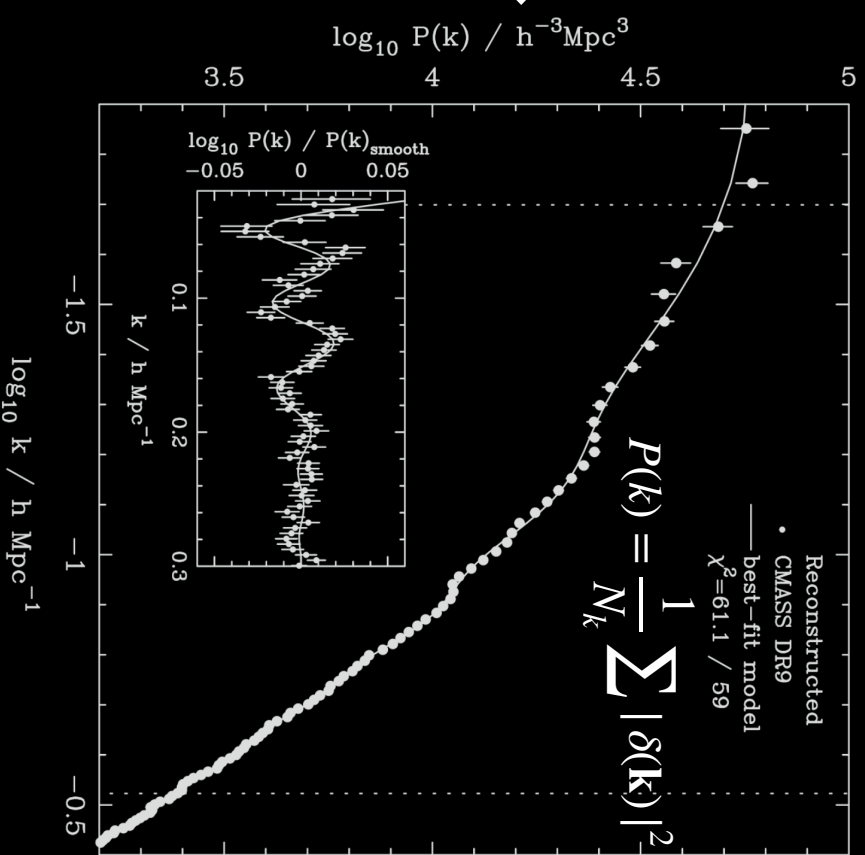
Based on PRD 99, 063530 (2019)

Cosmology with Galaxy Clustering

- ◆ The matter distribution, **large-scale structures**, reflects underlying physics (**modified gravity**, **massive neutrinos**...).
- ➔ This information is imprinted on **power spectrum** and it can be measured from spectroscopic surveys of galaxy distribution.

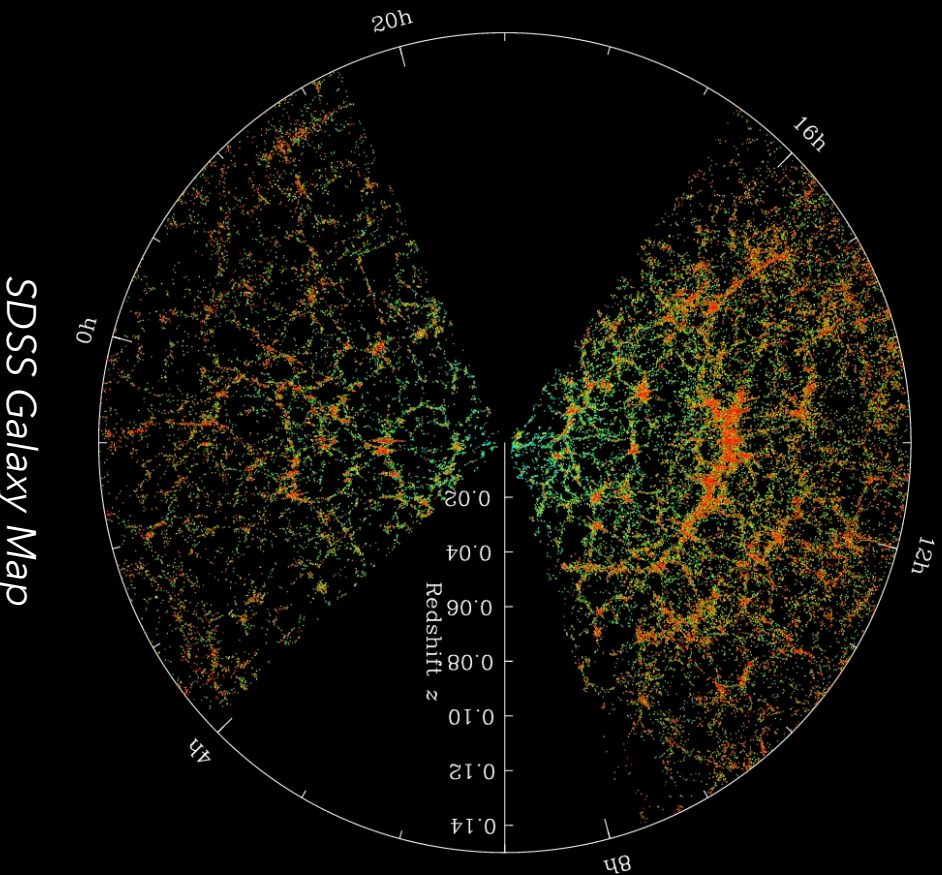


Statistics

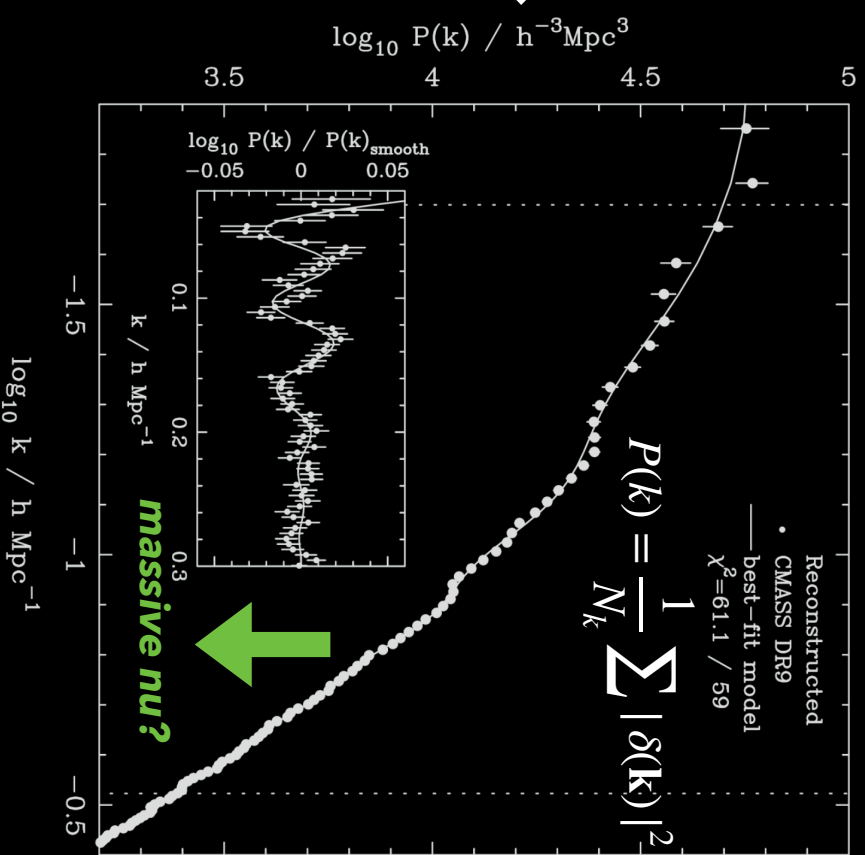


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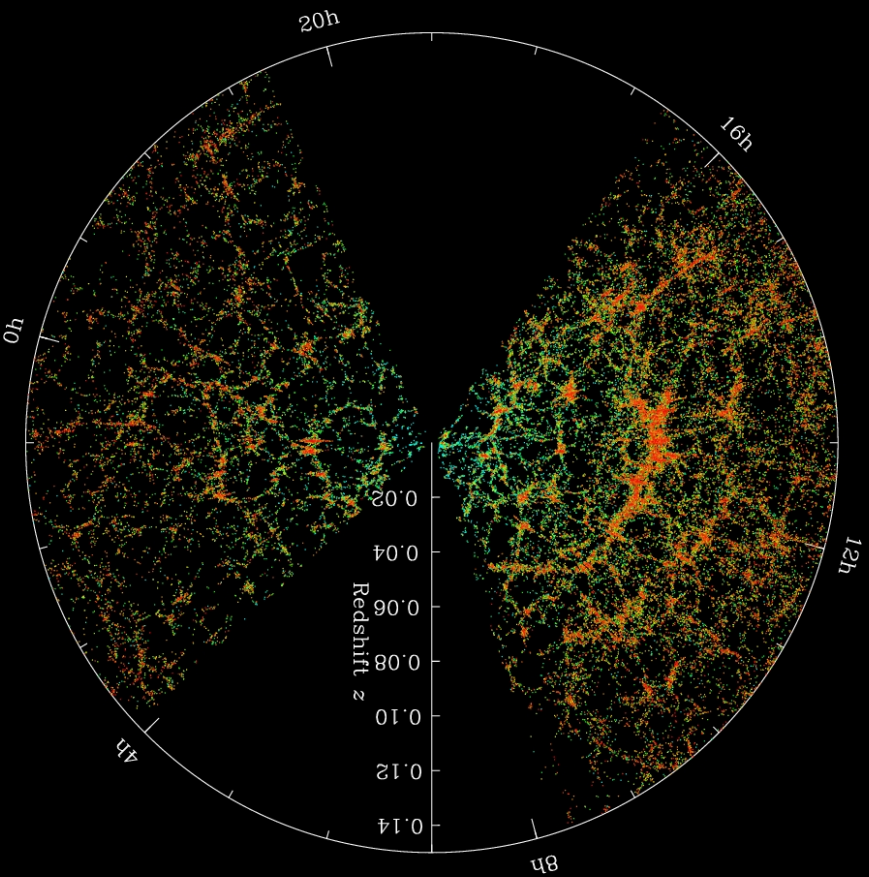
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Cosmology with Galaxy Clustering

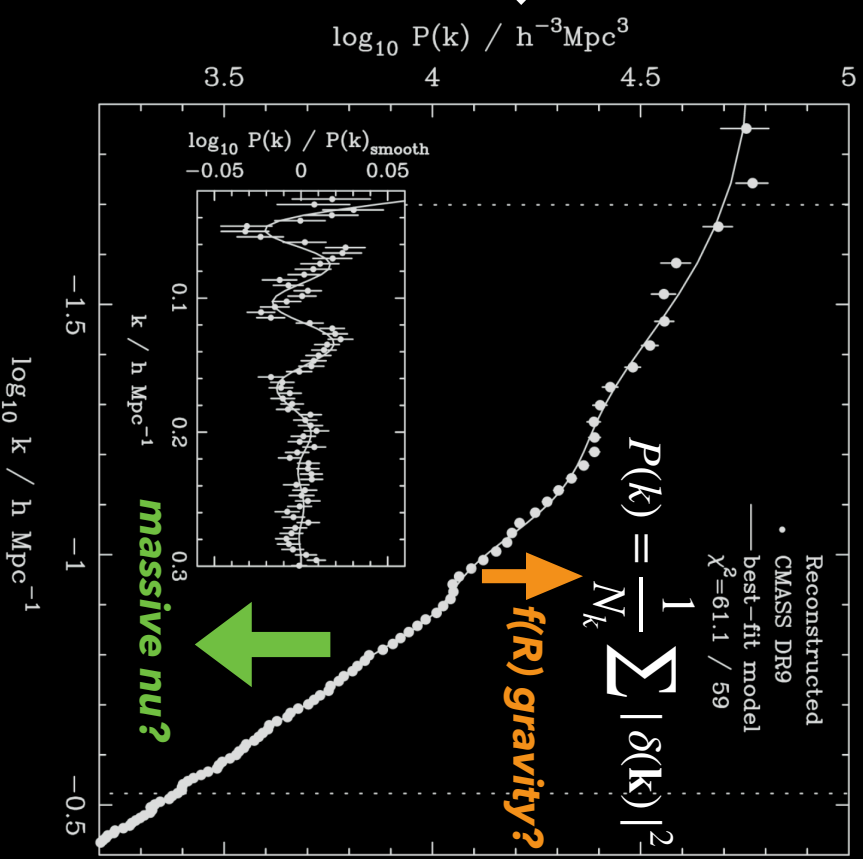
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SDSS Galaxy Map



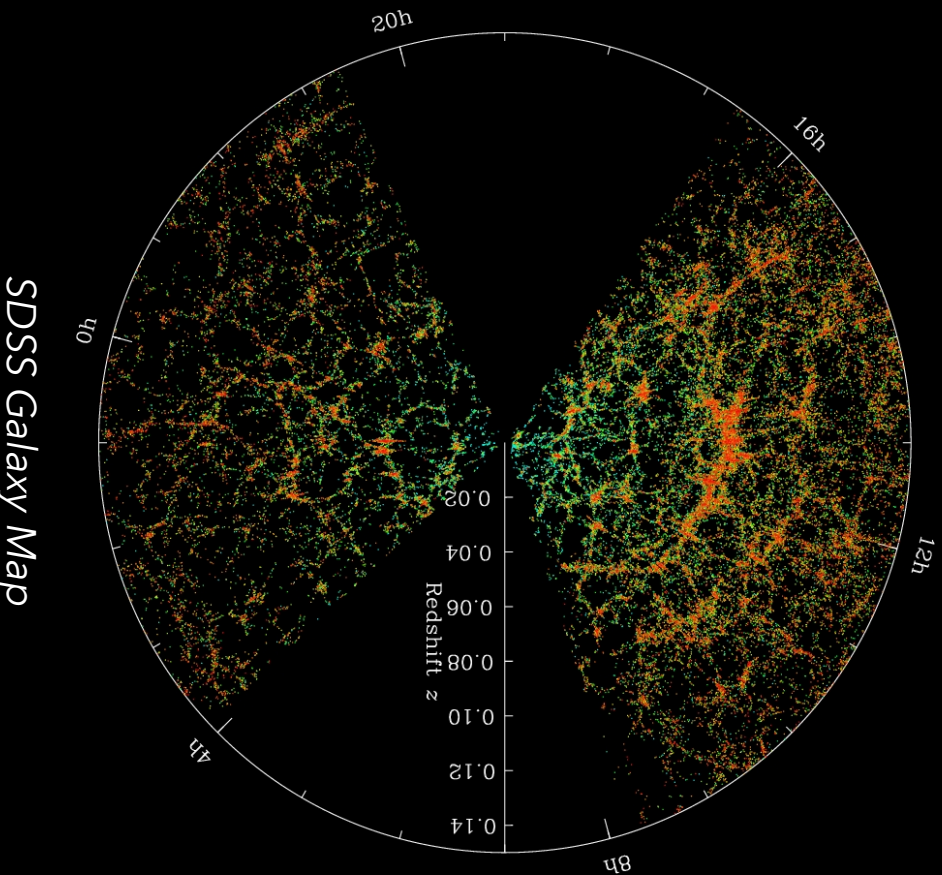
Statistics

Power spectrum

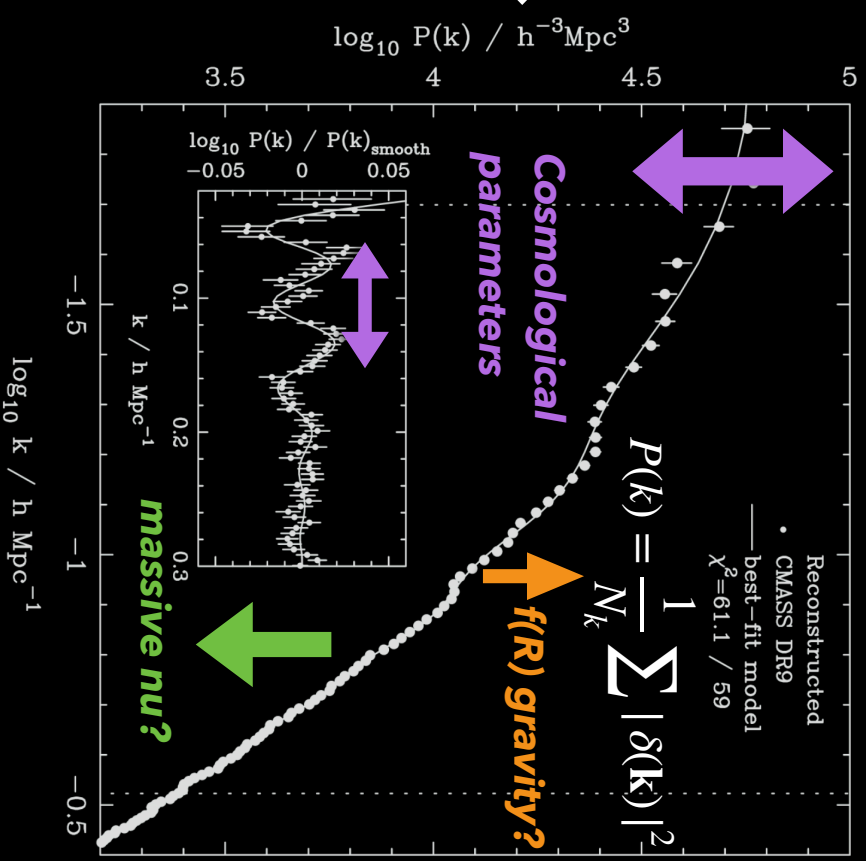


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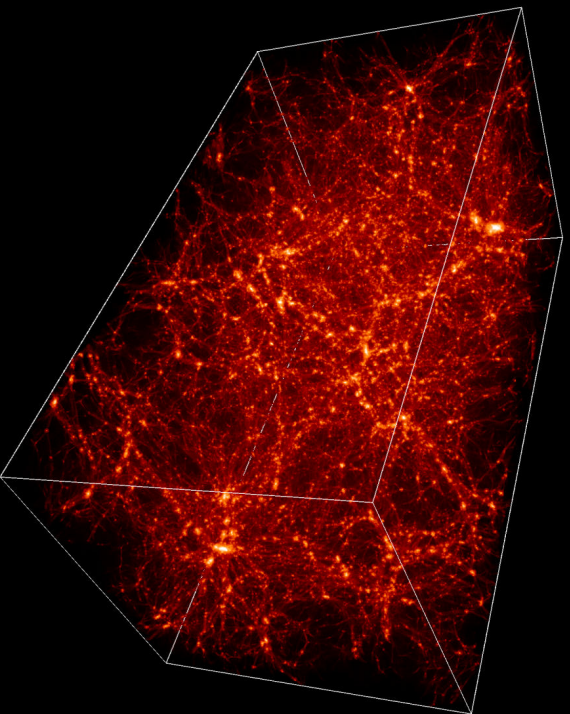
Modeling of LSS at nonlinear regimes

N-body simulations

Matter distribution is discretized as particles. The evolution is governed by Newtonian gravity.

Pros: Accurate down to resolution

Cons: Computationally expensive
Not suitable for MCMC



Perturbation theory (PT)

Based on single-stream approx., fluid equations are expanded with respect to density contrast.

Pros: Analytical and fast

Cons: The applicable range is limited to mildly nonlinear regime.

$$\delta(\mathbf{k}) = \sum_{n=1} D_+^n \delta^{(n)}(\mathbf{k})$$
$$\delta^{(n)}(\mathbf{k}) = \int \frac{d^3 q_1 \dots d^3 q_n}{(2\pi)^{3(n-1)}} \delta_D(\mathbf{k} - \mathbf{q}_1 - \dots - \mathbf{q}_n)$$
$$\times F_{\text{sym}}^{(n)}(\mathbf{q}_1, \dots, \mathbf{q}_n) \delta_0(\mathbf{q}_1) \dots \delta_0(\mathbf{q}_n)$$

PT Models

Standard Perturbation Theory (SPT) : Widely used standard way to expand fluid eqs. 2-loop level expansion gives reasonable predictions.

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Regularized Perturbation Theory (RegPT) : Bernardeau+ (2008), Tarrya+ (2012)

Reorganizing SPT expansion with **propagator** (Γ -expansion)

At high- k regime, the power spectrum damps as $e^{-\sigma_d^2 k^2/2}$.

The r.m.s. displacement can be a free parameter.



RegPT+

(IR-resummed) Effective Field Theory (EFT) : Baumann+ (2012), Carrasco+ (2012)

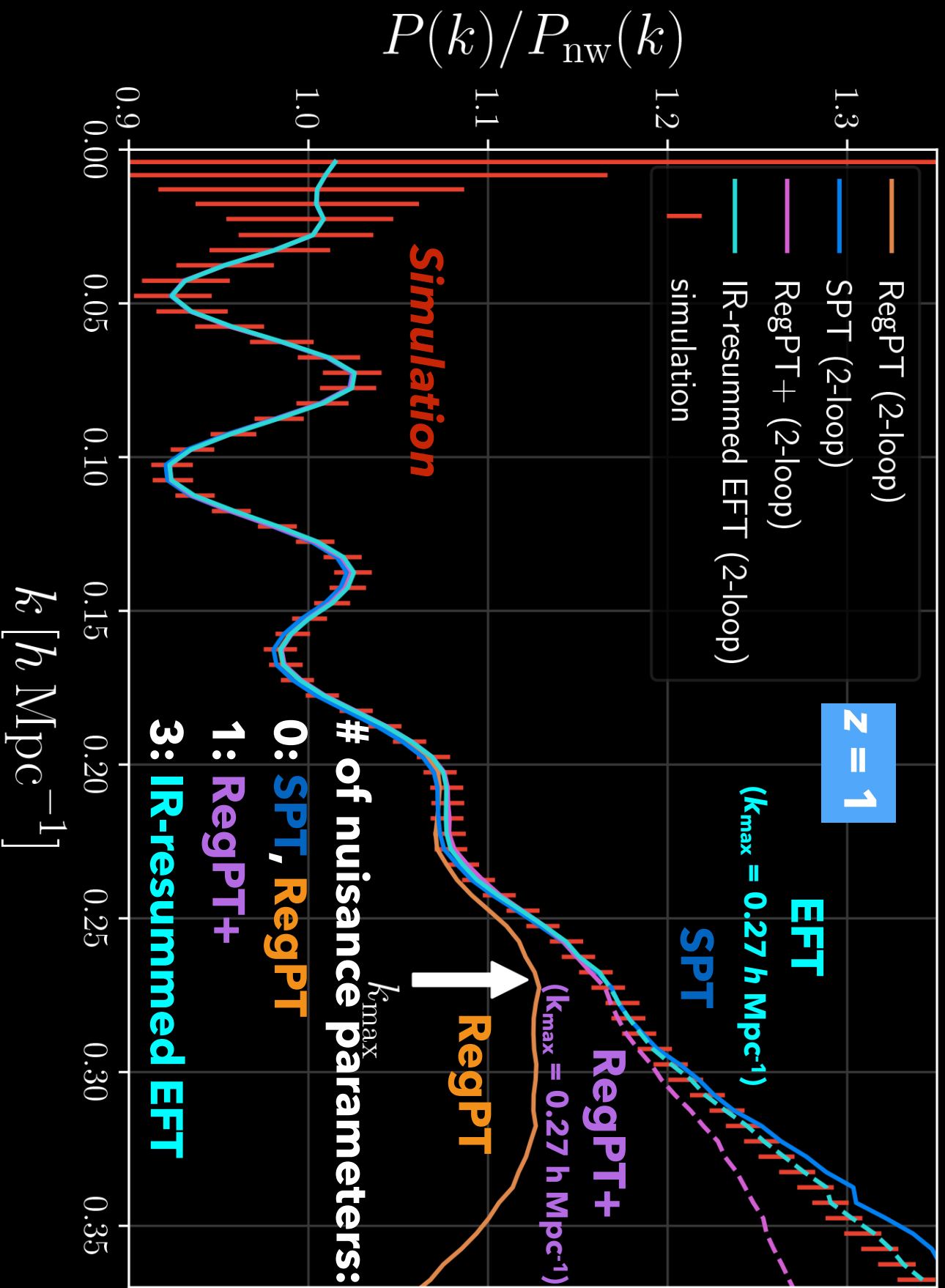
Introducing counter terms to describe small-scale behavior characterized by effective sound speed.

The sound speed is treated as free parameter.

$$-2(2\pi)^2 c_s^{(1)2} \left(\frac{k}{k_{\text{NL}}} \right)^2 P_L(k)$$

LPT, closure, Time-RG, RPT, and more...

Predictions of Power Spectra



Motivation

There are many PT schemes proposed so far
but details are different for each method.

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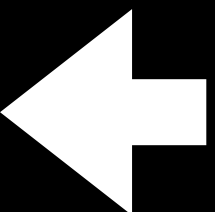
- **Is it applicable to smaller scales?**
- **How accurate in cosmological parameter estimation?**
- **Some models contain nuisance parameters but is it fair to compare just the smallest applicable scales?**

Motivation

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Key question: which PT scheme performs better?

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This study: Systematic tests of various 2-loop level PT schemes with N -body simulations in the context of cosmological parameter estimation with $P(k)$

Strategy

◆ *End-to-end test in cosmological parameter inference*

1. **Generate initial condition**

h Hubble parameter

Ω_m Matter density

A_s Amplitude of scalar perturbation

Strategy

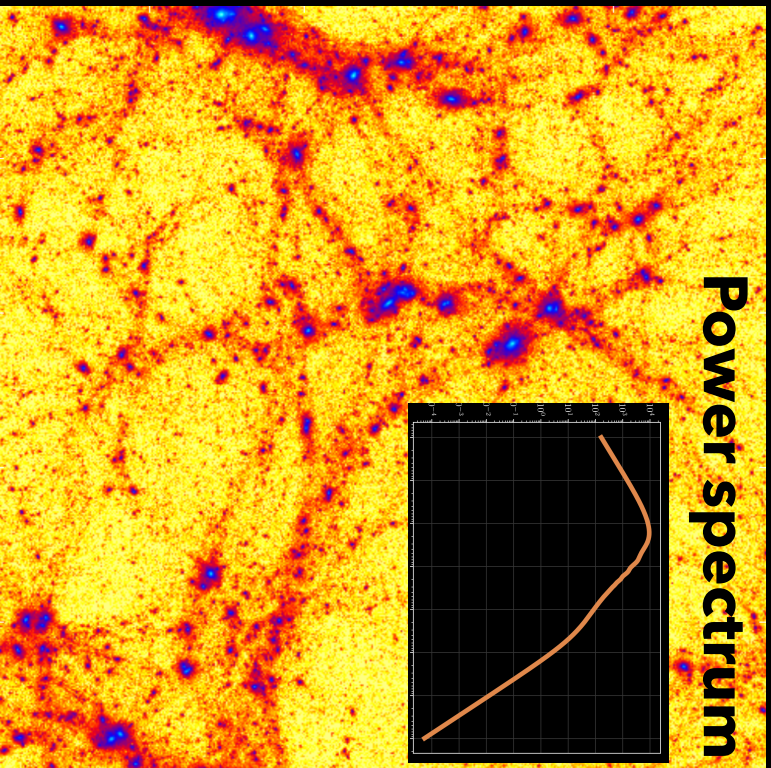
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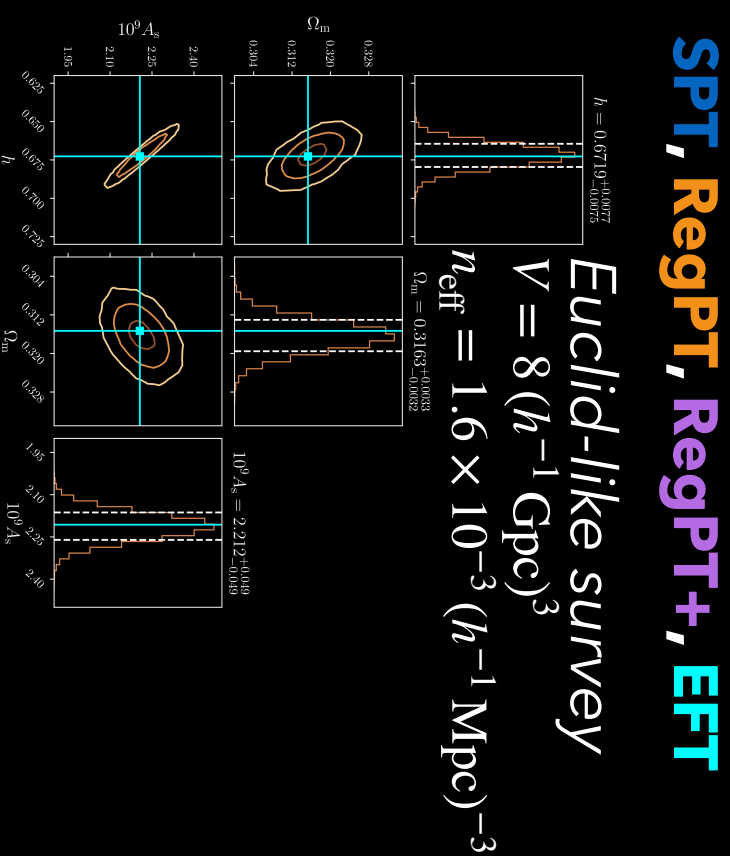
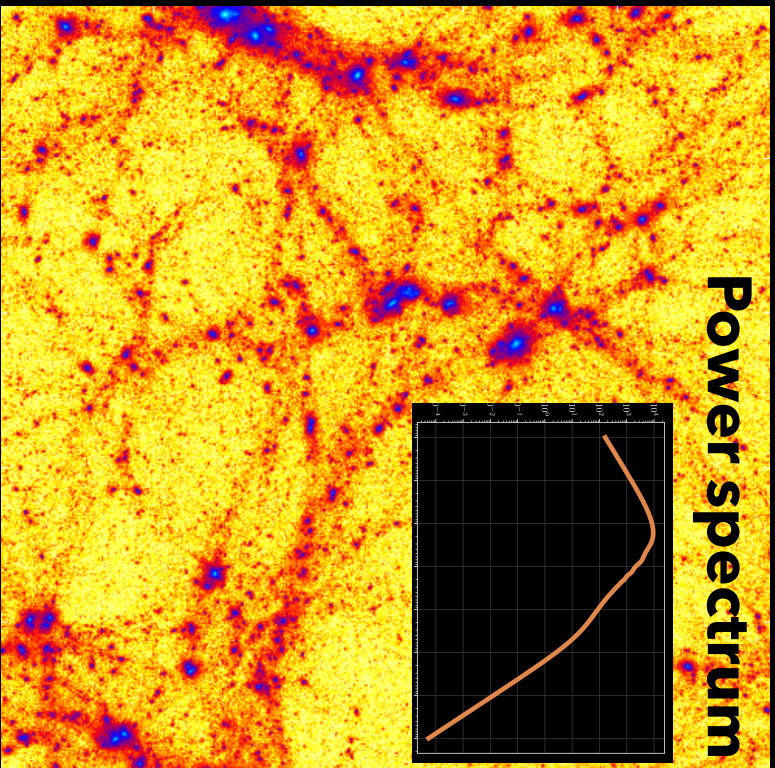


2. Run *N*-body sim. & measure
power spectrum @ $z = 1$



Strategy

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- 1. **Generate initial condition**
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- 2. **Run N -body sim. & measure power spectrum @ $z = 1$**
- 3. **Infer cosmological params. with MCMC**



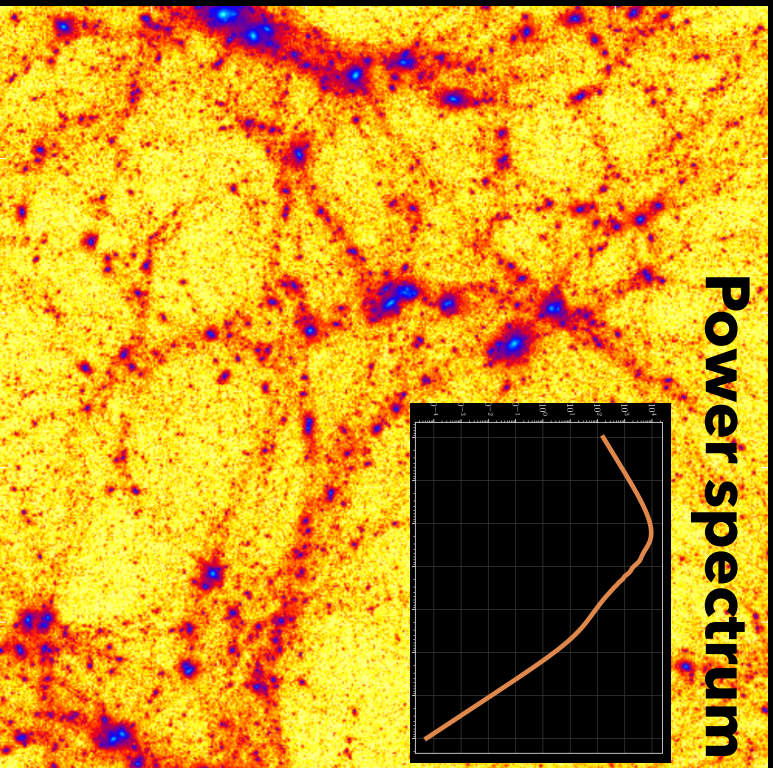
Strategy

◆ **End-to-end test in cosmological parameter inference**

1. Generate initial condition

Can PT reproduce?

2. Run N -body sim. & measure power spectrum @ $z = 1$

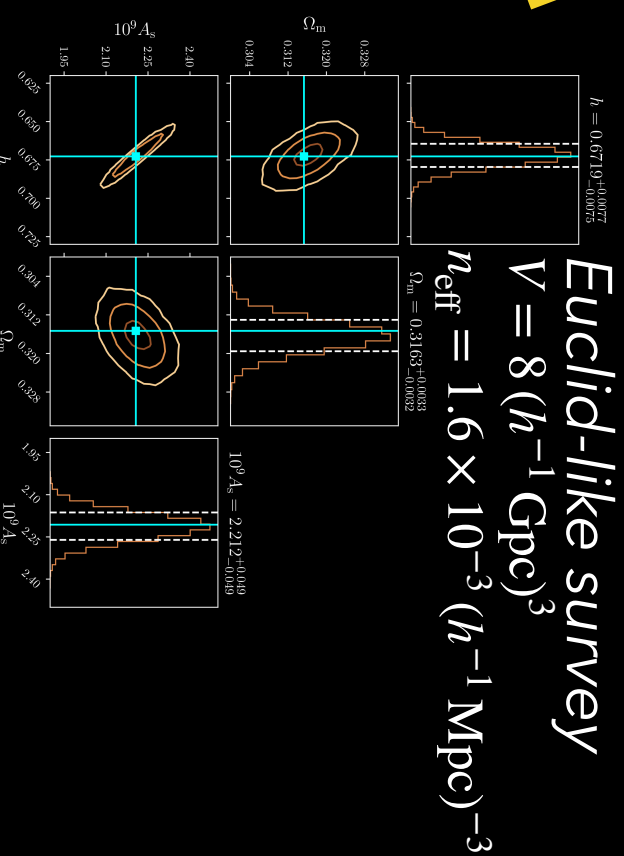


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2-loop level PT schemes of

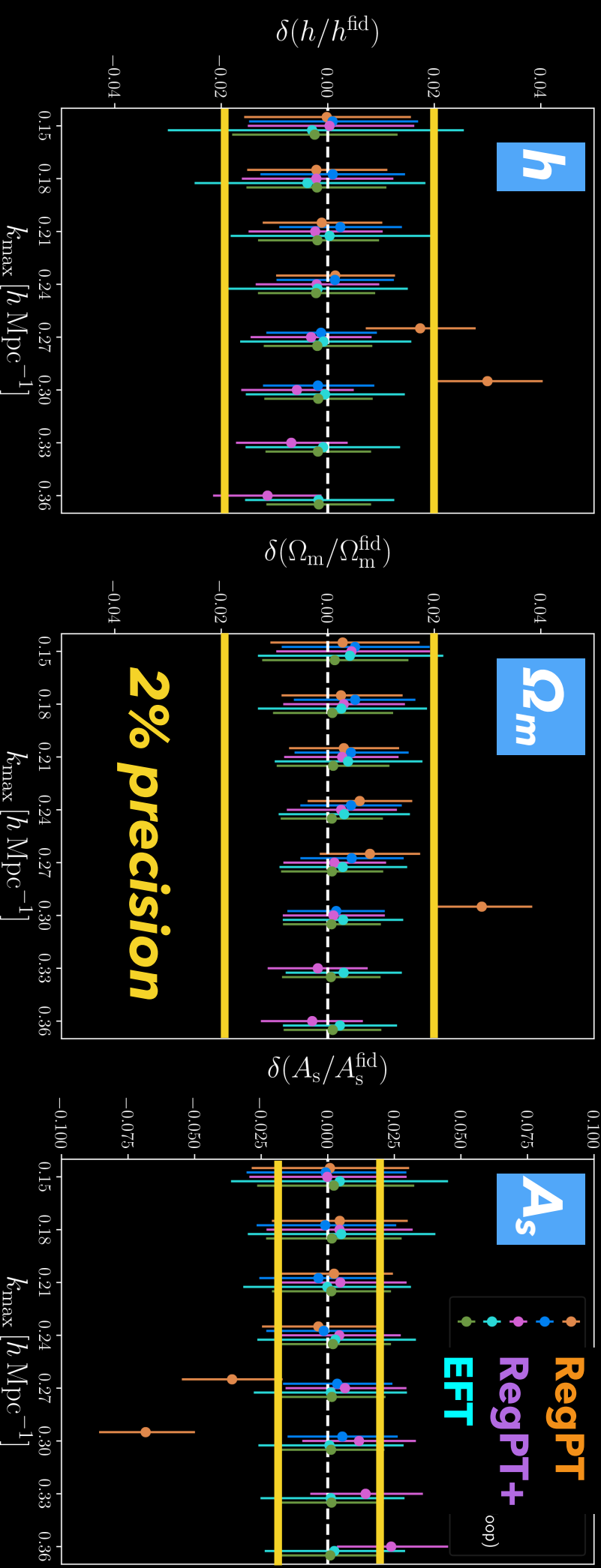
SPT, RegPT, RegPT+, EFT



Results: Estimated Parameters

- ◆ Inferred parameters normalized by fiducial values for three cosmological parameters as a function of k_{\max} .

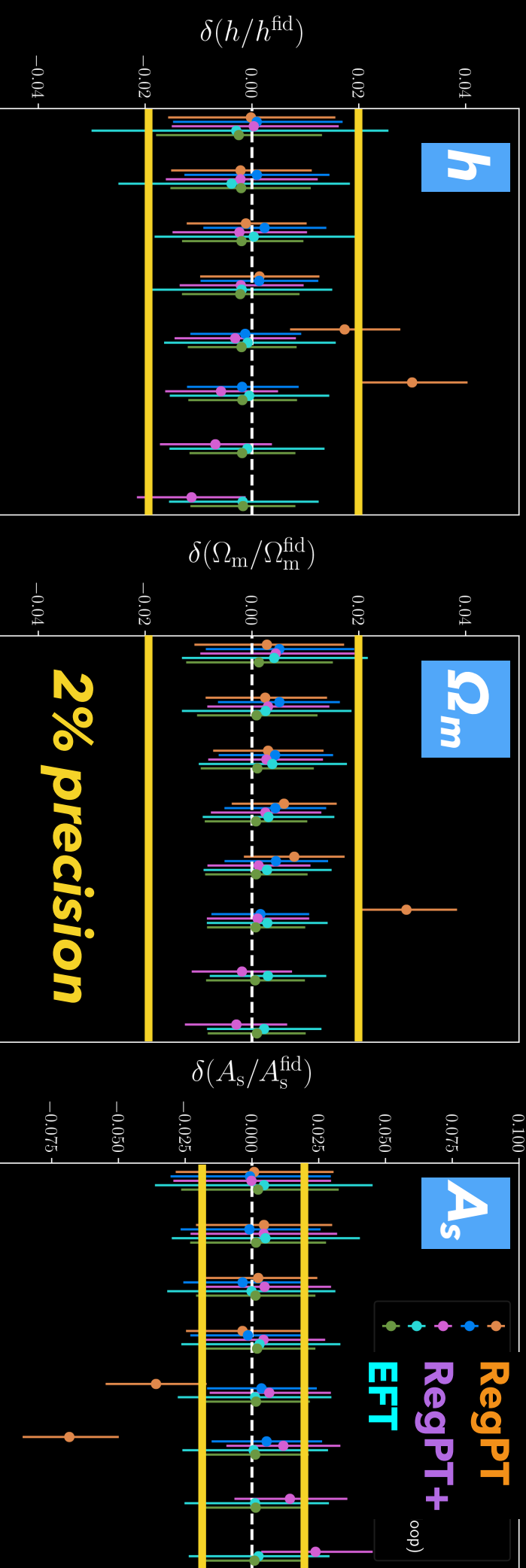
$$\log L(\theta | \hat{P}) = -\frac{1}{2} \sum_{k_i, k_j < k_{\max}} (\hat{P}(k_i) - P(k_i; \theta))(C^{-1})_{ij} (\hat{P}(k_j) - P(k_j; \theta))$$



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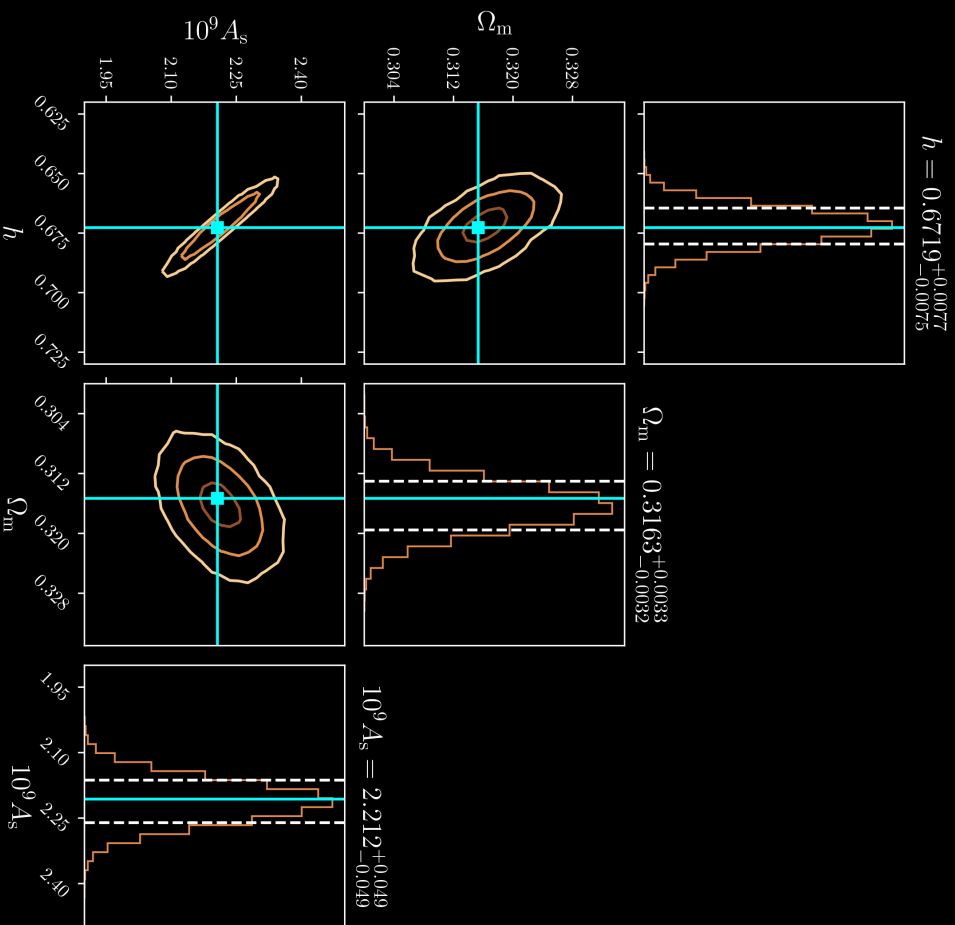
2% precision

RegPT: small error bars (= high precision) but biased
EFT: unbiased even for high k_{\max} but large error bars

Results: 2D Contours

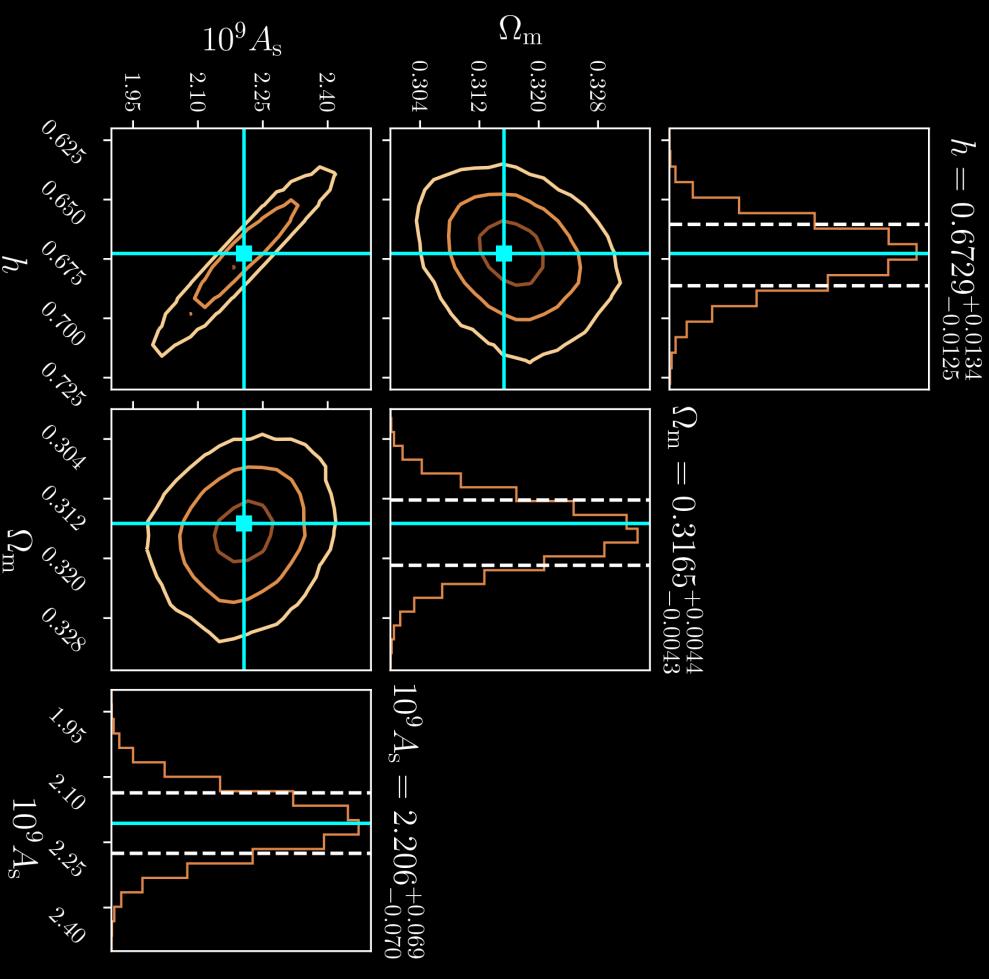
RegPT

(No nuisance param.)



IR-resummed EFT

(3 nuisance params.)

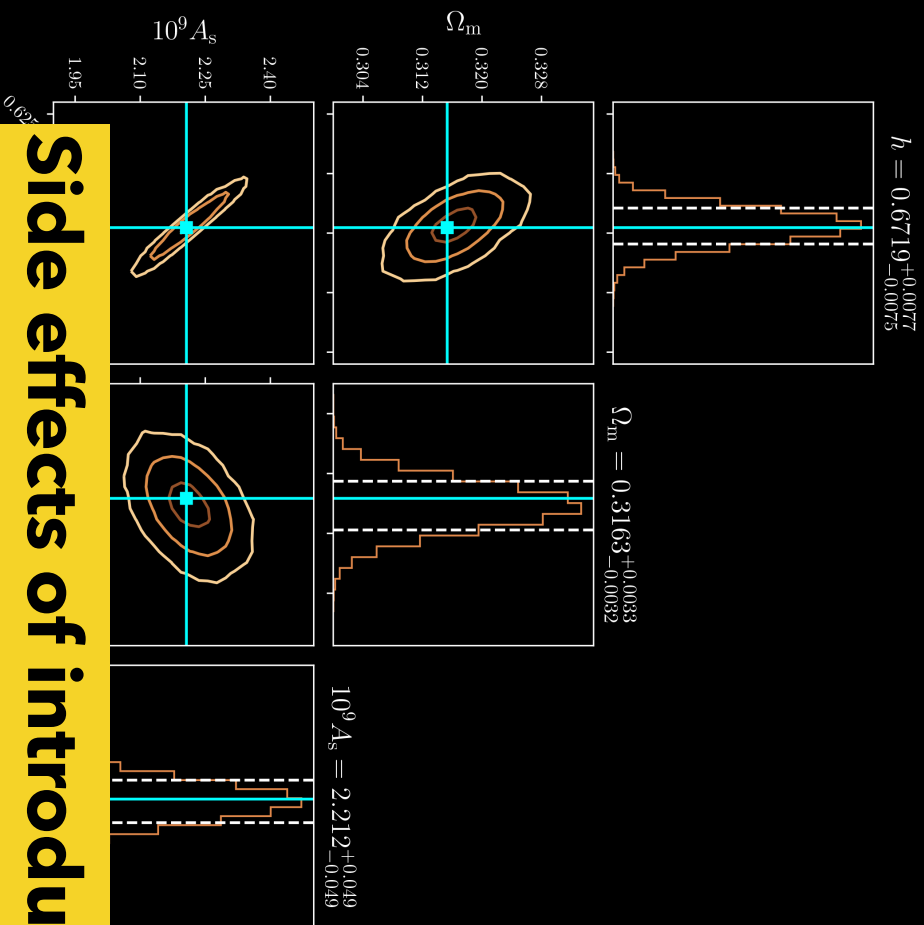


$z = 1.0, k_{\max} = 0.21 h \text{ Mpc}^{-1}$

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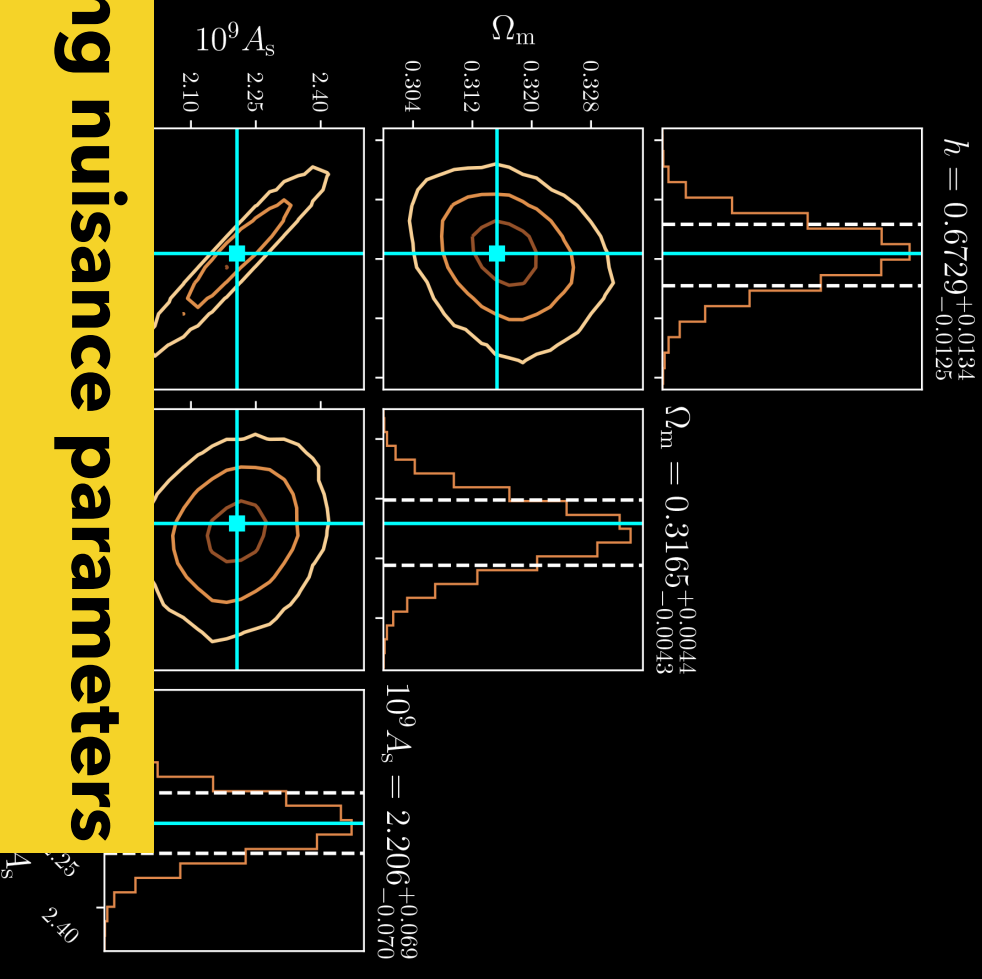
RegPT

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IR-resummed EFT

(3 nuisance params.)



Side effects of introducing nuisance parameters

- 1. Degrades constraints**
- 2. Distorts parameter degeneracy**

FOM & FOB

Introduce goodness of fit and parameter bias as summary quantities

Correlation matrix: estimated from MCMC chains

$$S_{\alpha\beta} = \frac{1}{N-1} \sum_k^N (\theta_\alpha^k - \bar{\theta}_\alpha)(\theta_\beta^k - \bar{\theta}_\beta)$$

Parameter vector
 $\theta = (h, \Omega_m, A_s, \dots)$

Figure of Merit (FoM): goodness of fit

$$\text{FoM} = [\det \tilde{S}]^{-\frac{1}{2}} \sim (\text{Volume of } 1\text{-}\sigma \text{ C.L.})^{-1}$$

marginalized for nuisance params.

Figure of Bias (FoB): parameter bias

$$\text{FoB} = \left[(\bar{\theta}_\alpha - \theta_\alpha^{\text{fid.}}) \tilde{S}_{\alpha\beta} (\bar{\theta}_\beta - \theta_\beta^{\text{fid.}}) \right]^{-\frac{1}{2}}$$

Distance between true and estimated params. normalized by variances.

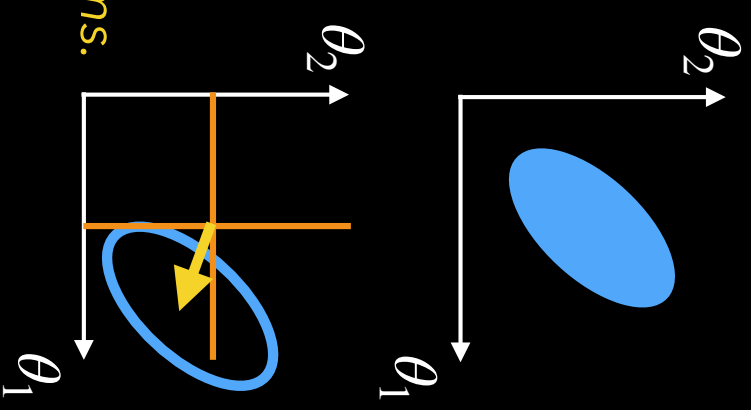


Figure of Bias

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Applicable scale without biased

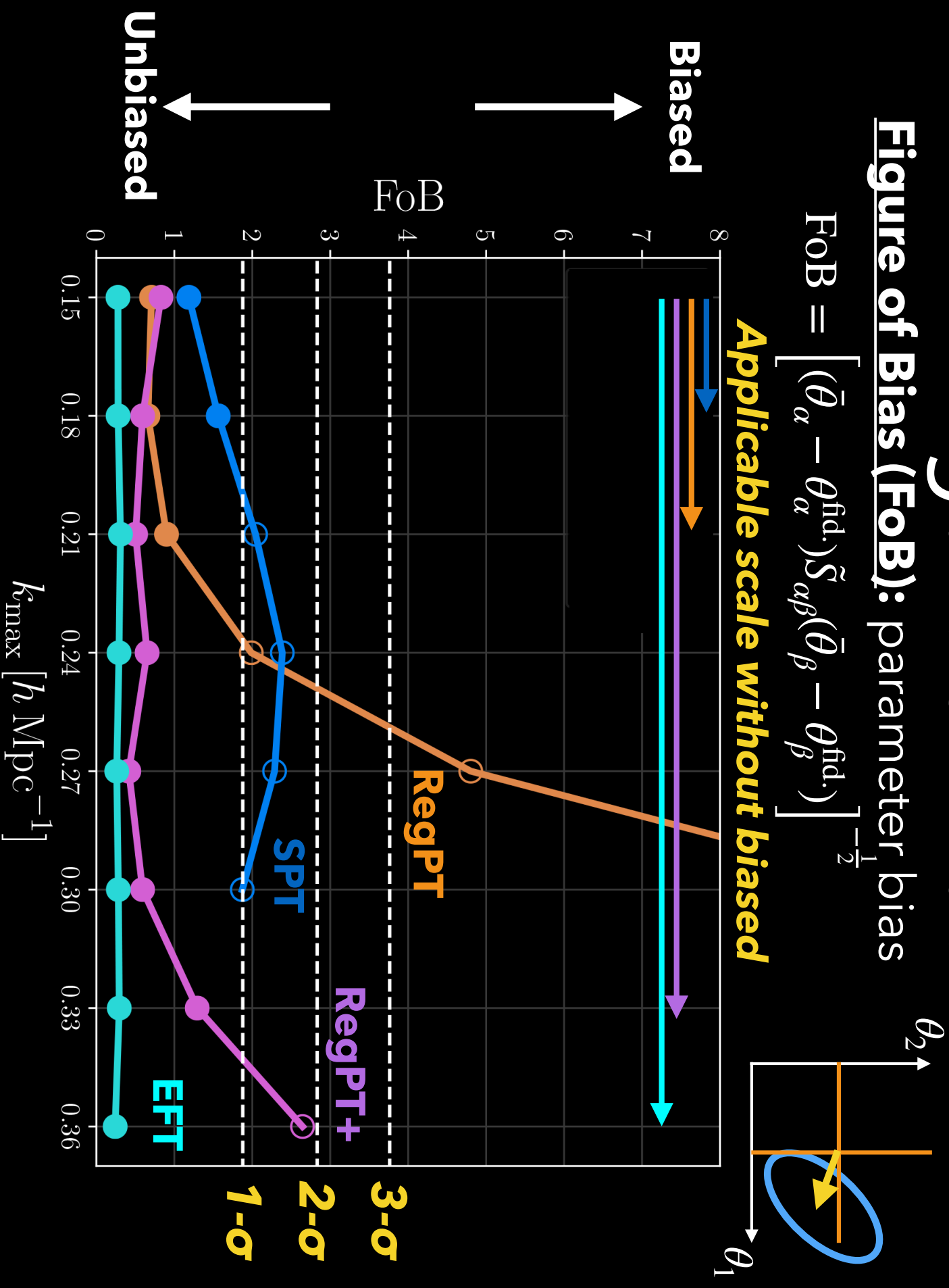
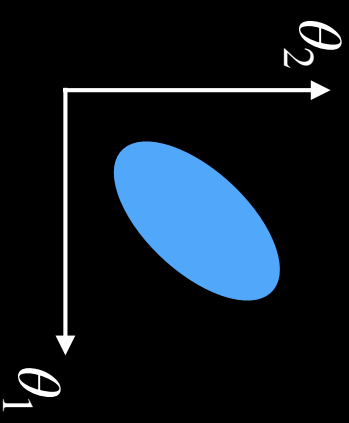
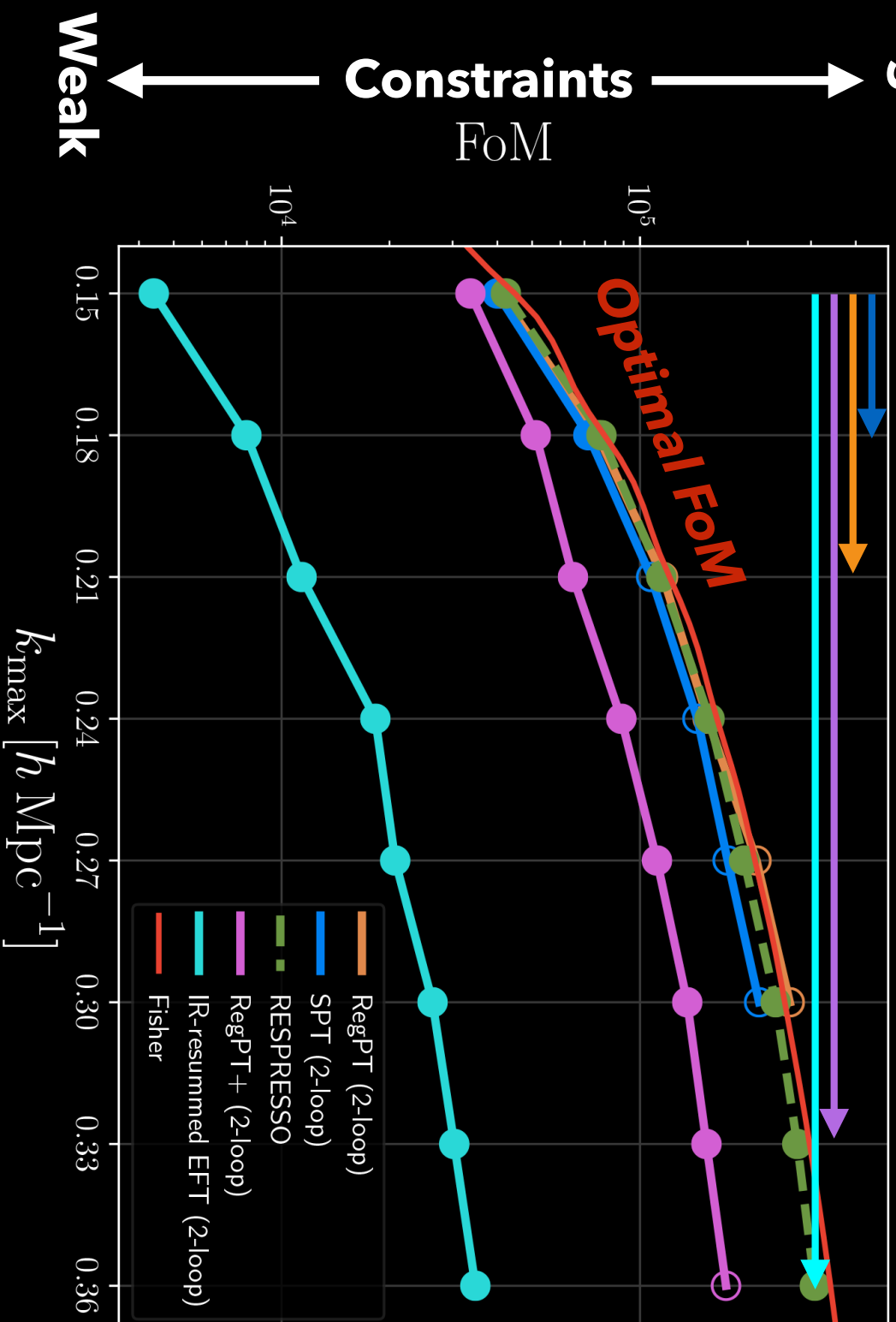


Figure of Merit

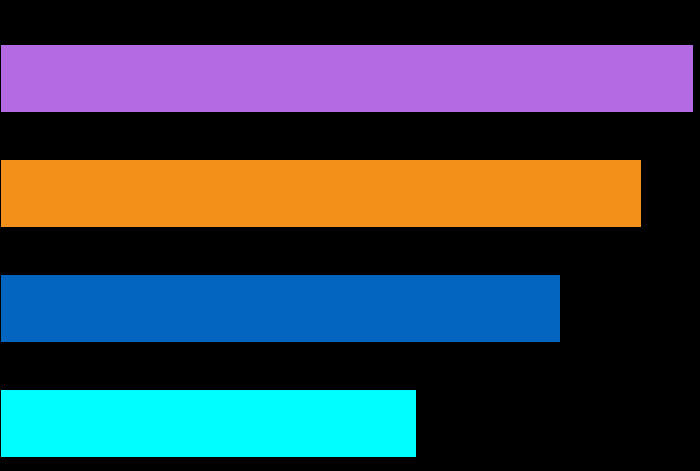
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Applicable scale without biased



Maximum FoM without biased



RegPT+

RegPT

SPT

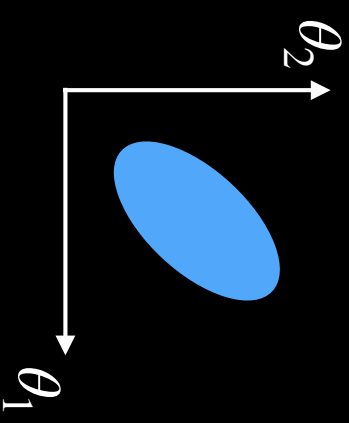
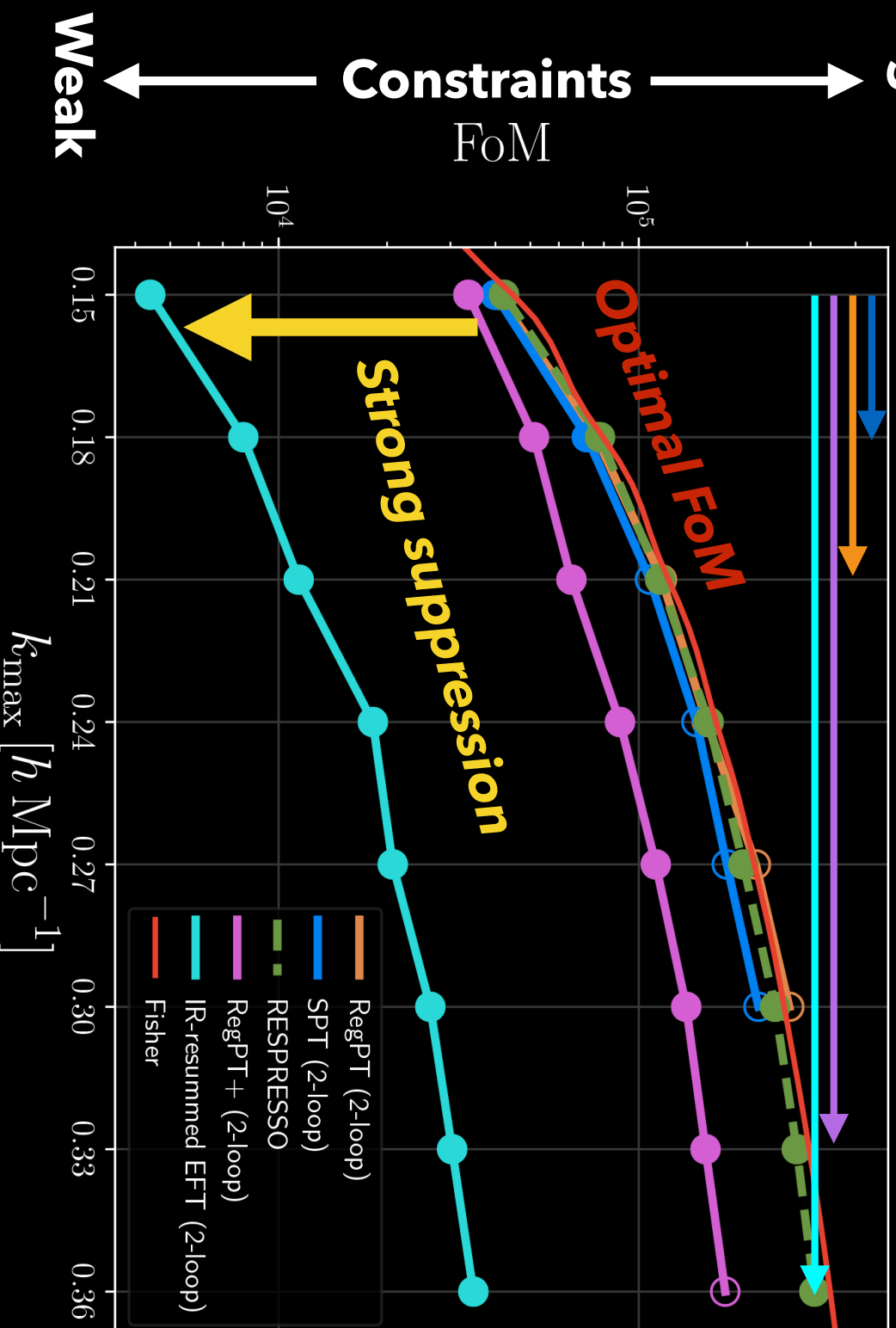
EFT

Figure of Merit

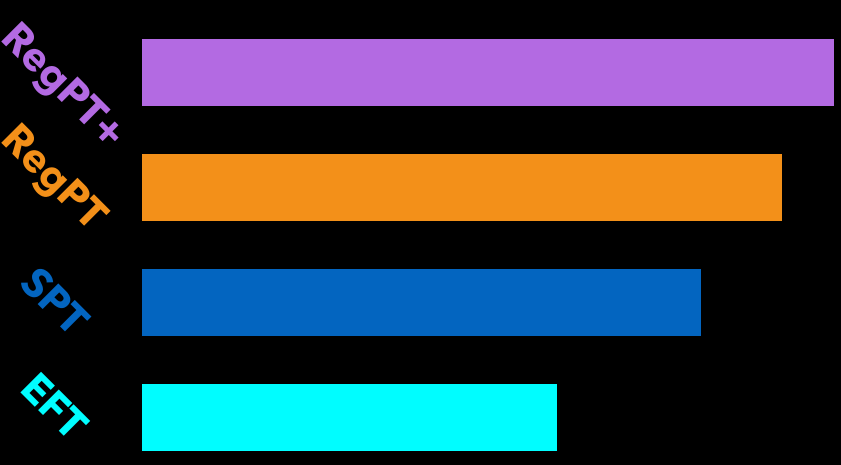
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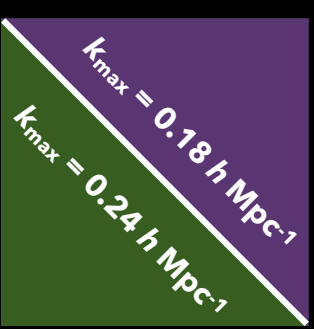
Maximum FoM without biased



Correlation Coefficient

Correlation coefficient: $R_{\alpha\beta} = \frac{S_{\alpha\beta}}{\sqrt{S_{\alpha\alpha}S_{\beta\beta}}}$

$R_{\alpha\beta} = \pm 1$ **Positive/Negative correlation**
 $R_{\alpha\beta} = 0$ **No correlation**



RegPT+

IR-resummed EFT

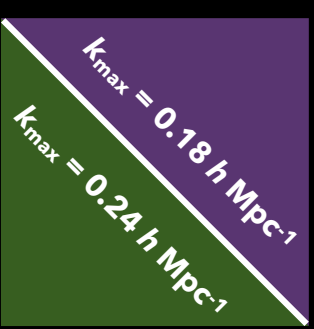
σ_d	h	Σ_m	A_s	h
σ_d	0.31	0.14	-0.30	
A_s	-0.99	0.39		
Σ_m	-0.49		0.34	
h		-0.48	-0.98	0.25

Σ	α_1	A_s	Σ_m	h	α_1	α_2	Σ
Σ	-0.26	-0.09	0.21	0.36	-0.52		
α_2	0.55	0.46	-0.39	-0.91			-0.28
α_1	-0.71	-0.63	0.55				-0.94
A_s	-0.97	-0.15		0.60	-0.48	0.12	
Σ_m	0.24		-0.12	-0.54	0.40	-0.04	
h		0.12	-0.98	-0.71	0.59	-0.14	

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RegPT+

IR-resummed EFT

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0.31	0.14	-0.26	-0.09	0.21	0.36	-0.52	-0.28
-0.99	0.39	0.55	0.46	-0.39	-0.91	-0.94	0.19
-0.49	-0.27	-0.71	-0.63	0.55	-0.91	0.19	0.12
	0.34	-0.97	-0.15	0.60	-0.48	0.12	0.12
	0.35	0.24	-0.12	-0.54	0.40	-0.04	-0.04

Strong correlation with cosmological parameters degrades constraints!

Summary

End-to-end test of PT schemes (**SPT**, **RegPT**, **RegPT+**, **EFT**) in the analysis of cosmological parameter inference with real-space matter power spectrum.

- **Best FoB model** : **IR-resummed EFT** (3 nuisance params.)
but weak constraining power
- **Best FoM model** : **SPT** / **RegPT** (no nuisance param.)
but biased parameter estimation
- **Reasonable choice** : **RegPT+** (1 nuisance param.)
works well up to even $k_{\max} \sim 0.33 h \text{ Mpc}^{-1}$

Prospects: Incorporating RSD and galaxy bias for more realistic modeling of power spectrum and fast calculation suitable for MCMC analysis