

The Renormalization Group for Large-Scale Structure (RGforLSS)

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With Fabian Schmidt and Charalampos Nikolis

Parma, November 2024

2307.15031,
2404.16929,
2405.21002

Message to take home

We derive the Callan-Symanzik equation for the galaxy bias+stochastic+PNG parameters

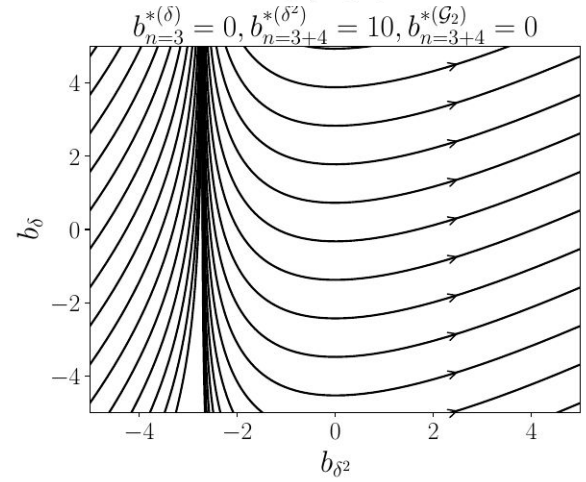
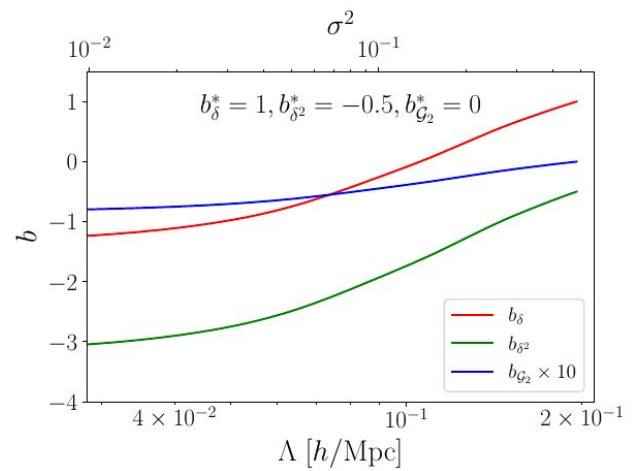
$$\frac{db_\delta}{d\Lambda} = - \left[\frac{68}{21} b_{\delta^2} + 3b_{\delta^3}^* - \frac{4}{3} b_{\mathcal{G}_2\delta}^* \right] \frac{d\sigma_\Lambda^2}{d\Lambda},$$

$$\frac{db_{\delta^2}}{d\Lambda} = - \left[\frac{8126}{2205} b_{\delta^2} + \frac{17}{7} b_{\delta^3}^* - \frac{376}{105} b_{\mathcal{G}_2\delta}^* + b_{n=4}^{*(\delta^2)} \right] \frac{d\sigma_\Lambda^2}{d\Lambda},$$

$$\frac{db_{\mathcal{G}_2}}{d\Lambda} = - \left[\frac{254}{2205} b_{\delta^2} + \frac{116}{105} b_{\mathcal{G}_2\delta}^* + b_{n=4}^{*(\mathcal{G}_2)} \right] \frac{d\sigma_\Lambda^2}{d\Lambda}.$$

Many things to explore:

- Systematic construction of operator basis,
- Systematic renormalization,
- Cross-checks,
- More information from galaxy clustering (to be investigated)



The future

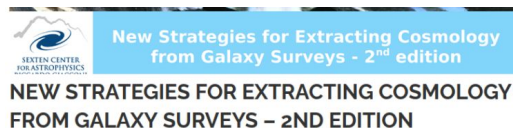
1) Higher n-point functions and higher-loops

- We need: quick loop computations (e.g. COBRA from Bakx, Chisari, Vlah)
- Why? Inflation, neutrinos, DE, mediators, smaller scales, ...

2) Field-level (

- all n-pt function tower
- Difficulties: many degrees of freedom (convergence), hard noise-modelling

3) Other statistics and Multi-tracing



4) Priors

5) Beyond- Λ CDM: new interaction vertex

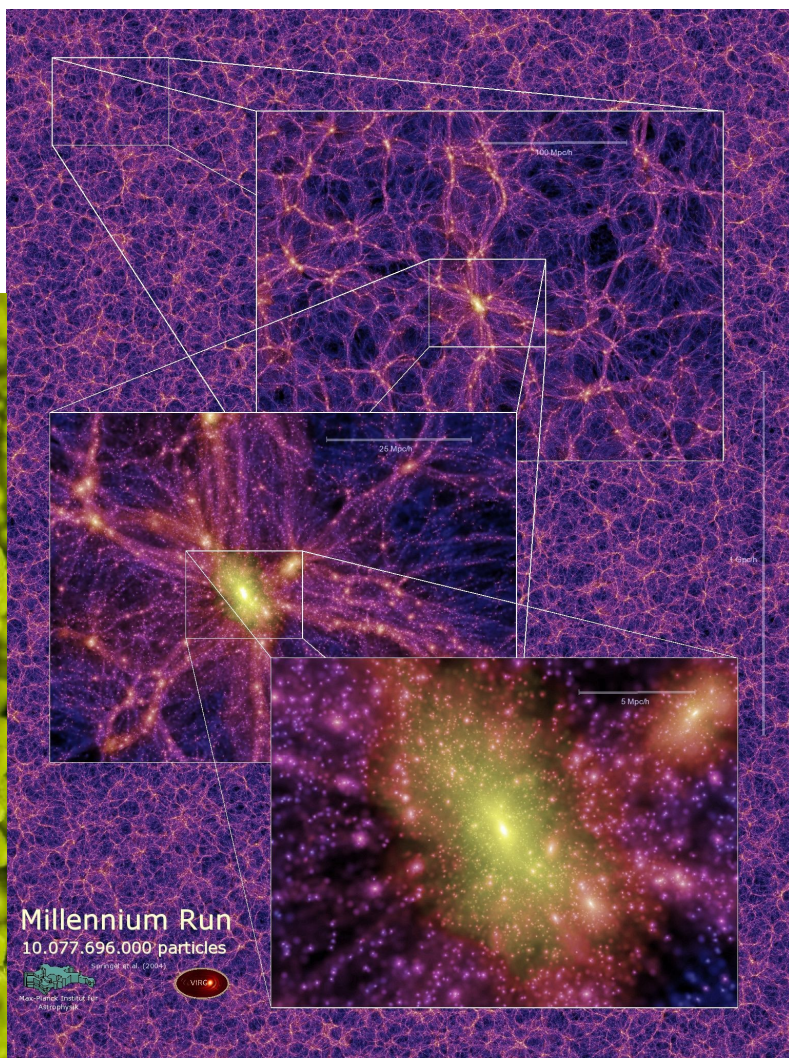


6) Other observables: Lyman-alpha, intensity maps, lensing

7) Theoretical pathway: the **RGforLSS**

Part I - Preamble(s)

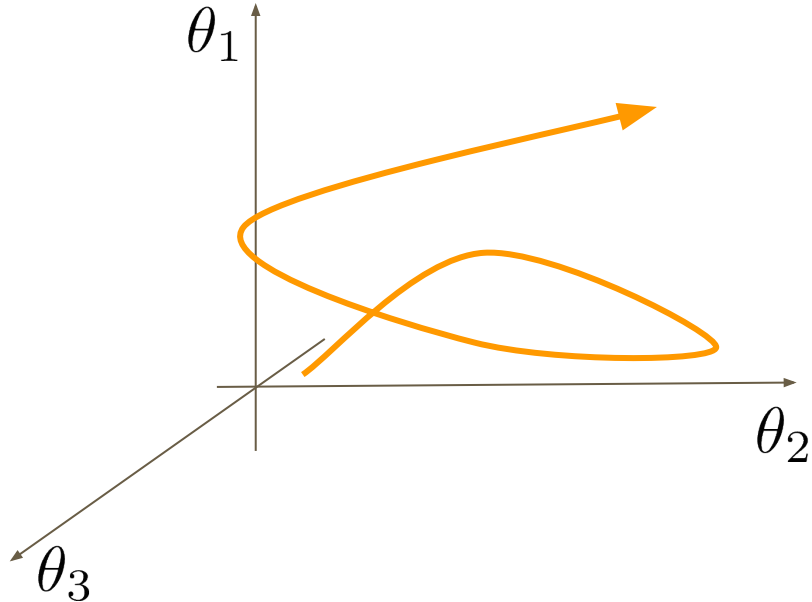
How things change with scale? (from food to galaxies)



QFT101

Renormalization group: coupling constants evolve with the cutoff ("flow").

Observables don't depend on the cutoff!



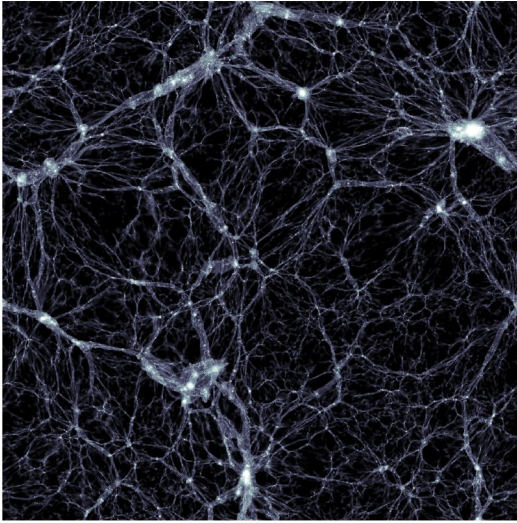
Callan-Symanzik equation:

$$\frac{\partial g}{\partial \ln \mu} = \beta(g)$$

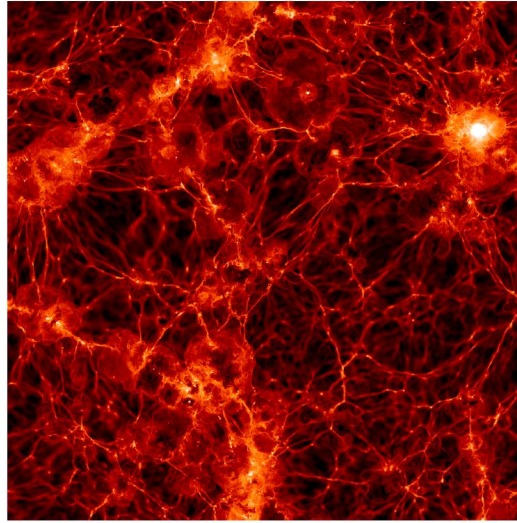
$$\text{QED: } \beta(e) = \frac{e^3}{12\pi^2}$$

$$\text{QCD: } \beta(g) = - \left(11 - \frac{n_s}{6} - \frac{2n_f}{3} \right) \frac{g^3}{16\pi^2}$$

The galaxy bias expansion



(a) dark matter



(b) baryons

From Illustris simulation,
Haiden, Steinhauser, Vogelsberger,
Genel, Springel, Torrey, Hernquist, 15

Stochastic field

$$\delta_g(\mathbf{x}, \tau) \equiv \frac{n_g(\mathbf{x}, \tau)}{\bar{n}_g(\tau)} - 1 = \sum_O \left[b_O(\tau) + c_{\epsilon, O}(\tau) \epsilon(\mathbf{x}, \tau) \right] O(\mathbf{x}, \tau) + \epsilon(\mathbf{x}, \tau)$$

Bias

Part II - Renormalization in LSS

Renormalizing the bias parameters

$$\delta_g(\mathbf{x}, \tau) \equiv \frac{n_g(\mathbf{x}, \tau)}{\bar{n}_g(\tau)} - 1 = \sum_O [b_O(\tau) + c_{\epsilon, O}(\tau)\epsilon(\mathbf{x}, \tau)] O(\mathbf{x}, \tau) + \epsilon(\mathbf{x}, \tau)$$

$$O[\delta](\mathbf{k}) = \int_{\mathbf{p}_1, \dots, \mathbf{p}_n} \delta_D(\mathbf{k} - \mathbf{p}_{1\dots n}) S_O(\mathbf{p}_1, \dots, \mathbf{p}_n) \delta(\mathbf{p}_1) \cdots \delta(\mathbf{p}_n)$$

First order: δ ;

Second order: δ^2, \mathcal{G}_2 ;

Third order: $\delta^3, \delta \mathcal{G}_2, \Gamma_3, \mathcal{G}_3$;

Contribution from arbitrarily small scales!

Renormalizing the bias parameters

$$\delta_g(\mathbf{x}, \tau) \equiv \frac{n_g(\mathbf{x}, \tau)}{\bar{n}_g(\tau)} - 1 = \sum_O [b_O^{\Lambda}(\tau) + c_{\epsilon, O}^{\Lambda}(\tau) \epsilon^{\Lambda}(\mathbf{x}, \tau)] O^{\Lambda}(\mathbf{x}, \tau) + \epsilon^{\Lambda}(\mathbf{x}, \tau) + \text{counter-terms}(\Lambda)$$

$$O[\delta](\mathbf{k}) = \int_{\mathbf{p}_1, \dots, \mathbf{p}_n}^{\Lambda} \delta_{\mathbf{D}}(\mathbf{k} - \mathbf{p}_{1\dots n}) S_O(\mathbf{p}_1, \dots, \mathbf{p}_n) \delta(\mathbf{p}_1) \cdots \delta(\mathbf{p}_n)$$

Notation:

$$[[O]] = O^{\Lambda} + \text{counter-terms}(\Lambda)$$

How to determine the renormalization condition?

First order: δ ;

Second order: δ^2, \mathcal{G}_2 ;

Third order: $\delta^3, \delta \mathcal{G}_2, \Gamma_3, \mathcal{G}_3$;

Contribution from arbitrarily small scales!

Motivation

RENORMALIZATION AND EFFECTIVE LAGRANGIANS

Joseph POLCHINSKI*

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Received 27 April 1983

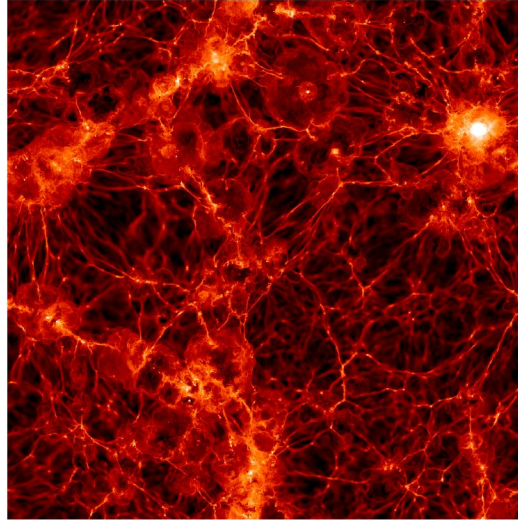
1. Introduction

The understanding of renormalization has advanced greatly in the past two decades. Originally it was just a means of removing infinities from perturbative calculations. The question of why nature should be described by a renormalizable theory was not addressed. These were simply the only theories in which calculations could be done.

A great improvement comes when one takes seriously the idea of a physical cutoff at a very large energy scale Λ . The theory at energies above Λ could be another field

Intuition time

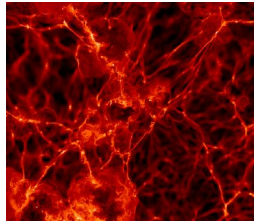
Take a box, smoothed on some scale Λ and measure b_0



*Technically, we are talking about operators constructed from differently the smoothed initial condition. But this picture should work as an intuition

Intuition time

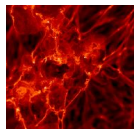
In the same box, smooth on a new scale Λ and measure b_0



Intuition time

You can measure $b_O(\Lambda)$

Extra cross-check:
If the running does not match the
theoretical prediction, something is missing



Does this scaling of the bias parameters carry information?
(Open question, I don't know the answer)

Part III - The Wilson-Polchinski path integral approach

Warning (and apologies in advance):

next 2 slides will be technical, they are just there to trigger interest

The bias partition function (based on Carroll, Leichenauer, Pollack, 13)

$$\mathcal{Z}[J_\Lambda] = \int \mathcal{D}\delta_\Lambda^{(1)} \mathcal{P}[\delta_\Lambda^{(1)}] \exp \left(\int_{\mathcal{O}} J_\Lambda(\mathbf{k}) \left[\sum_{\mathcal{O}} b_{\mathcal{O}}^\Lambda \mathcal{O}[\delta_\Lambda^{(1)}](-\mathbf{k}) \right] + \frac{1}{2} P_\epsilon^\Lambda \int_{\mathbf{k}} J_\Lambda(\mathbf{k}) J_\Lambda(-\mathbf{k}) + \mathcal{O}[J_\Lambda^2 \delta_\Lambda^{(1)}, J_\Lambda^3] \right)$$

Path-integral over linear-smoothed density, normalized
Single-current term
Double-current term captures stochasticity source

N-point correlators evaluated as:

$$\frac{\partial \mathcal{Z}}{\partial J_\Lambda \dots \partial J_\Lambda} \Bigg|_{J_\Lambda=0}$$

The shell expansion (Wilson formalism)

Consider a very thin shell with width: $\Lambda = \Lambda' - \lambda$

$$\delta_{\Lambda'}^{(1)}(\mathbf{k}) = \delta_{\Lambda}^{(1)}(\mathbf{k}) + \delta_{\text{shell}}^{(1)}(\mathbf{k})$$

Idea: Integrate out the shell!

$$\mathcal{Z}[J_{\Lambda}] = \int \mathcal{D}\delta_{\Lambda}^{(1)} \mathcal{P}[\delta_{\Lambda}^{(1)}] \exp \left(\int_{\mathbf{k}} J_{\Lambda}(\mathbf{k}) \left[\sum_{\mathbf{o}} b_{\mathbf{o}}^{\Lambda'} \phi_{[\delta_{\Lambda}^{(1)}]}(-\mathbf{k}) \right] + \frac{1}{2} P_{\epsilon}^{\Lambda'} \int_{\mathbf{k}} J_{\Lambda}(\mathbf{k}) J_{\Lambda}(-\mathbf{k}) + \mathcal{O}[J_{\Lambda}^2 \delta_{\Lambda}^{(1)}, J_{\Lambda}^3] \right)$$

The running of the bias/stochastic operators is done connecting both cutoff

The shell expansion (Wilson formalism)

Consider a very thin shell with width: $\Lambda = \Lambda' - \lambda$

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The running of the bias/stochastic operators is done connecting both cutoff

What appears after integrating out the shell

$$\times \left(1 + \int_{\mathbf{k}} J_{\Lambda}(\mathbf{k}) \left[\sum_{\mathcal{O}} b_{\mathcal{O}}^{\Lambda'} \left(\mathcal{S}_{\mathcal{O}}^1[\delta_{\Lambda}^{(1)}](-\mathbf{k}) + \mathcal{S}_{\mathcal{O}}^2[\delta_{\Lambda}^{(1)}](-\mathbf{k}) + \dots \right) \right] + \frac{1}{2} \int_{\mathbf{k}, \mathbf{k}'} J_{\Lambda}(\mathbf{k}) J_{\Lambda}(\mathbf{k}') \sum_{\mathcal{O}, \mathcal{O}'} b_{\mathcal{O}}^{\Lambda'} b_{\mathcal{O}'}^{\Lambda'} \left[\mathcal{S}_{\mathcal{O}\mathcal{O}'}^{11}[\delta_{\Lambda}^{(1)}](\mathbf{k}, \mathbf{k}') + \dots \right] + \mathcal{O}[J_{\Lambda}^2 \delta_{\Lambda}^{(1)}, J_{\Lambda}^3] \right)$$

The shell expansion (Wilson formalism)

Consider a very thin shell with width: $\Lambda = \Lambda' - \lambda$

$$\delta_{\Lambda'}^{(1)}(\mathbf{k}) = \delta_{\Lambda}^{(1)}(\mathbf{k}) + \delta_{\text{shell}}^{(1)}(\mathbf{k})$$

Idea: Integrate out the shell!

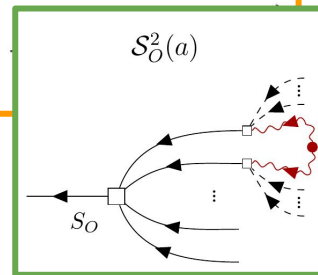
$$\mathcal{Z}[J_{\Lambda}] = \int \mathcal{D}\delta_{\Lambda}^{(1)} \mathcal{P}[\delta_{\Lambda}^{(1)}] \exp \left(\int_{\mathbf{k}} J_{\Lambda}(\mathbf{k}) \left[\sum_{\mathcal{O}} b_{\mathcal{O}}^{\Lambda'} \mathcal{O}[\delta_{\Lambda}^{(1)}](-\mathbf{k}) \right] + \frac{1}{2} P_{\epsilon}^{\Lambda'} \int_{\mathbf{k}} J_{\Lambda}(\mathbf{k}) J_{\Lambda}(-\mathbf{k}) + \mathcal{O}[J_{\Lambda}^2 \delta_{\Lambda}^{(1)}, J_{\Lambda}^3] \right)$$

The running of the bias/stochastic operators is done connecting both cutoff

What appears after integrating out the shell

$$\times \left(1 + \int_{\mathbf{k}} J_{\Lambda}(\mathbf{k}) \left[\sum_{\mathcal{O}} b_{\mathcal{O}}^{\Lambda'} \left(\mathcal{S}_{\mathcal{O}}^1[\delta_{\Lambda}^{(1)}](-\mathbf{k}) + \mathcal{S}_{\mathcal{O}}^2[\delta_{\Lambda}^{(1)}](-\mathbf{k}) + \dots \right) \right] + \frac{1}{2} \int_{\mathbf{k}, \mathbf{k}'} J_{\Lambda}(\mathbf{k}) J_{\Lambda}(\mathbf{k}') \sum_{\mathcal{O}, \mathcal{O}'} b_{\mathcal{O}}^{\Lambda'} b_{\mathcal{O}'}^{\Lambda'} \left[\mathcal{S}_{\mathcal{O}\mathcal{O}'}^{11}[\delta_{\Lambda}^{(1)}](\mathbf{k}, \mathbf{k}') + \dots \right] \right)$$

Bias corrections



The shell expansion (Wilson formalism)

Consider a very thin shell with width: $\Lambda = \Lambda' - \lambda$

$$\delta_{\Lambda'}^{(1)}(\mathbf{k}) = \delta_{\Lambda}^{(1)}(\mathbf{k}) + \delta_{\text{shell}}^{(1)}(\mathbf{k})$$

Idea: Integrate out the shell!

$$\mathcal{Z}[J_{\Lambda}] = \int \mathcal{D}\delta_{\Lambda}^{(1)} \mathcal{P}[\delta_{\Lambda}^{(1)}] \exp \left(\int_{\mathbf{k}} J_{\Lambda}(\mathbf{k}) \left[\sum_{\mathcal{O}} b_{\mathcal{O}}^{\Lambda'} \mathcal{O}[\delta_{\Lambda}^{(1)}](-\mathbf{k}) \right] + \frac{1}{2} P_{\epsilon}^{\Lambda'} \int_{\mathbf{k}} J_{\Lambda}(\mathbf{k}) J_{\Lambda}(-\mathbf{k}) + \mathcal{O}[J_{\Lambda}^2 \delta_{\Lambda}^{(1)}, J_{\Lambda}^3] \right)$$

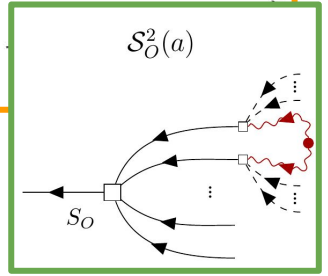
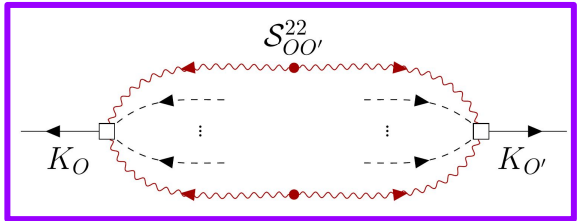
The running of the bias/stochastic operators is done connecting both cutoff

What appears after integrating out the shell

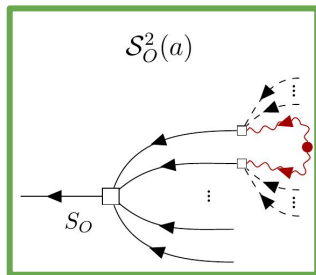
$$\times \left(1 + \int_{\mathbf{k}} J_{\Lambda}(\mathbf{k}) \left[\sum_{\mathcal{O}} b_{\mathcal{O}}^{\Lambda'} \left(\mathcal{S}_{\mathcal{O}}^1[\delta_{\Lambda}^{(1)}](-\mathbf{k}) + \mathcal{S}_{\mathcal{O}}^2[\delta_{\Lambda}^{(1)}](-\mathbf{k}) + \dots \right) \right] + \frac{1}{2} \int_{\mathbf{k}, \mathbf{k}'} J_{\Lambda}(\mathbf{k}) J_{\Lambda}(\mathbf{k}') \sum_{\mathcal{O}, \mathcal{O}'} b_{\mathcal{O}}^{\Lambda'} b_{\mathcal{O}'}^{\Lambda'} \left[\mathcal{S}_{\mathcal{O}\mathcal{O}'}^{11}[\delta_{\Lambda}^{(1)}](\mathbf{k}, \mathbf{k}') + \dots \right] \right)$$

Bias corrections

Stochastic corrections



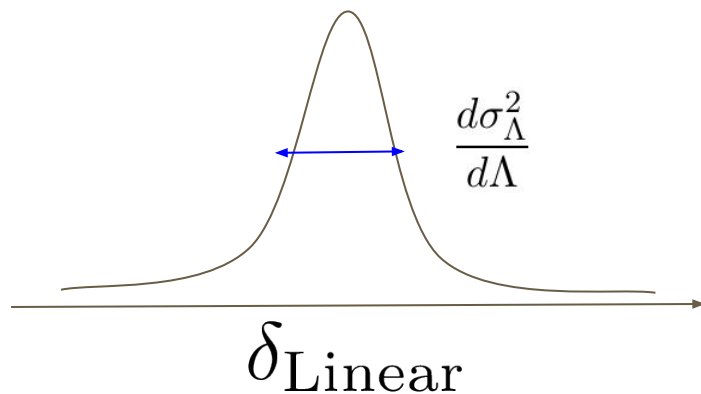
Example...



Correction to those operators!

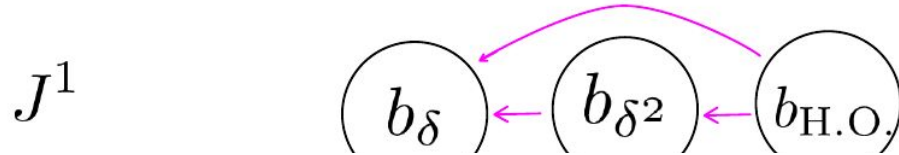
$$\mathcal{S}_{\delta^2}^2[\delta_\Lambda^{(1)}](\mathbf{k}) = \left[\frac{68}{21} \delta^{(1+2)}(\mathbf{k}) + \frac{8126}{2205} \delta^2(\mathbf{k}) \right]^{(2)} + \frac{254}{2205} \mathcal{G}_2^{(2)}(\mathbf{k}) \left[\int \frac{p^2 dp}{2\pi^2} P_{\text{shell}}(p) \right. \\ \left. + \text{higher derivative (h.d.)} + \mathcal{O} \left[\left(\delta_\Lambda^{(1)} \right)^3 \right] \right],$$

$$\int_{\mathbf{p}} P_{\text{shell}}(p) = \int_{\Lambda}^{\Lambda+\lambda} \frac{p^2 dp}{2\pi^2} P_L(p) = \frac{d\sigma_\Lambda^2}{d\Lambda} \Big|_{\Lambda} \lambda + \mathcal{O}(\lambda^2),$$



Results

$$\frac{d}{d\Lambda} b_O(\Lambda) = -\frac{d\sigma_\Lambda^2}{d\Lambda} \sum_{O'} s_{O'}^O b_{O'}(\Lambda),$$



$s_{O'}^O$	δ	δ^2	\mathcal{G}_2	δ^3	\mathcal{G}_3	Γ_3	$\delta\mathcal{G}_2$
$\mathbb{1}$	-	-	-	-	-	-	-
δ	-	$68/21$	-	3	-	-	$-4/3$
δ^2	-	$8126/2205$	-	$68/7$	-	-	$-376/105$
\mathcal{G}_2	-	$254/2205$	-	-	-	-	$116/105$

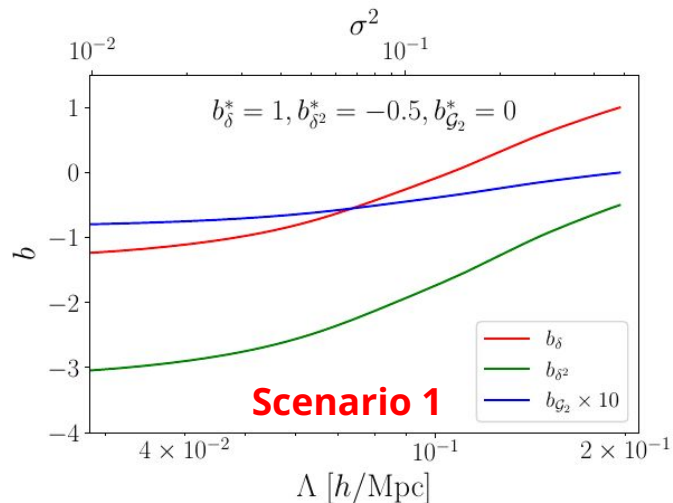
Solutions

Wilson-Polchinski RG-equations

$$\frac{db_\delta}{d\Lambda} = - \left[\frac{68}{21} b_{\delta^2} + 3b_{\delta^3}^* - \frac{4}{3} b_{\mathcal{G}_2\delta}^* \right] \frac{d\sigma_\Lambda^2}{d\Lambda},$$

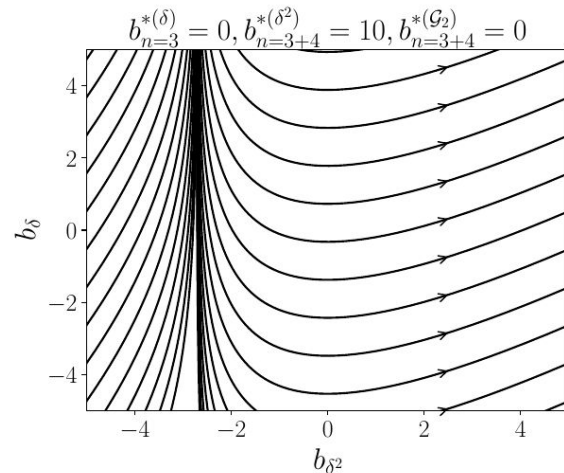
$$\frac{db_{\delta^2}}{d\Lambda} = - \left[\frac{8126}{2205} b_{\delta^2} + \frac{17}{7} b_{\delta^3}^* - \frac{376}{105} b_{\mathcal{G}_2\delta}^* + b_{n=4}^{*(\delta^2)} \right] \frac{d\sigma_\Lambda^2}{d\Lambda},$$

$$\frac{db_{\mathcal{G}_2}}{d\Lambda} = - \left[\frac{254}{2205} b_{\delta^2} + \frac{116}{105} b_{\mathcal{G}_2\delta}^* + b_{n=4}^{*(\mathcal{G}_2)} \right] \frac{d\sigma_\Lambda^2}{d\Lambda}.$$



Notice that:

- Bias parameter that are zero, may be sourced;
- Bias parameters may change sign!



PNGs (2405.21002)

w/ Charalampos Nikolis



PNGs

Spin-0

First order: $\delta, \Psi;$ Second order: $\delta^2, \mathcal{G}_2, \delta\Psi;$ Third order: $\delta^3, \delta^2\Psi, \delta\mathcal{G}_2, \Psi\mathcal{G}_2, \Gamma_3, \mathcal{G}_3$

Spin-2

First order: $\delta;$ Second order: $\delta^2, \mathcal{G}_2, \text{Tr} [\Psi\Pi^{[1]}];$ Third order: $\delta^3, \delta\mathcal{G}_2, \delta \text{Tr} [\Psi\Pi^{[1]}], \Gamma_3, \mathcal{G}_3, \text{Tr} [\Psi\Pi^{[2]}]$

PNGs

Free term

$$\frac{db_\delta}{d\Lambda} = - \left[\frac{68}{21} b_{\delta^2}(\Lambda) + b_{n=3}^{*\{\delta\}_G} \right] \frac{d\sigma_\Lambda^2}{d\Lambda}$$

New interaction

$$- a_0 f_{\text{NL}} \left[-\frac{13}{21} b_\Psi + \frac{13}{21} b_{\Psi\delta} + b_{n=3}^{*\{\delta\}_{\text{NG}}} \right] \left(\frac{H_0}{\Lambda} \right)^2 \frac{3 \Omega_m}{2 T(\Lambda)} \frac{d\sigma_\Lambda^2}{d\Lambda};$$

Now a
coupled set of
ODEs

$$\begin{aligned} \frac{db_\Psi}{d\Lambda} &= -a_0 f_{\text{NL}} b_{n=3}^{*\{\Psi\}_{\text{NG}}} \frac{d\sigma_\Lambda^2}{d\Lambda} - 4a_0 f_{\text{NL}} b_{\delta^2} \frac{d\sigma_\Lambda^2}{d\Lambda}, \\ \frac{db_{\Psi\delta}}{d\Lambda} &= -a_0 f_{\text{NL}} \left[\frac{272}{21} b_{\delta^2} + b_{n=3+4}^{*\{\Psi\delta\}_G} + b_{n=3+4}^{*\{\Psi\delta\}_{\text{NG}}} \right] \frac{d\sigma_\Lambda^2}{d\Lambda}, \end{aligned}$$

Rederivation
of Dalal+ 07
(in an elegant
way)

PNGs

Free term

$$\frac{db_\delta}{d\Lambda} = - \left[\frac{68}{21} b_{\delta^2}(\Lambda) + b_{n=3}^{*\{\delta\}G} \right] \frac{d\sigma_\Lambda^2}{d\Lambda}$$

New interaction

$$- a_0 f_{\text{NL}} \left[-\frac{13}{21} b_\Psi + \frac{13}{21} b_{\Psi\delta} + b_{n=3}^{*\{\delta\}NG} \right] \left(\frac{H_0}{\Lambda} \right)^2 \frac{3 \Omega_m}{2 T(\Lambda)} \frac{d\sigma_\Lambda^2}{d\Lambda};$$

Now a coupled set of ODEs

$$\begin{aligned} \frac{db_\Psi}{d\Lambda} &= -a_0 f_{\text{NL}} b_{n=3}^{*\{\Psi\}NG} \frac{d\sigma_\Lambda^2}{d\Lambda} - 4a_0 f_{\text{NL}} b_{\delta^2} \frac{d\sigma_\Lambda^2}{d\Lambda}, \\ \frac{db_{\Psi\delta}}{d\Lambda} &= -a_0 f_{\text{NL}} \left[\frac{272}{21} b_{\delta^2} + b_{n=3+4}^{*\{\Psi\delta\}G} + b_{n=3+4}^{*\{\Psi\delta\}NG} \right] \frac{d\sigma_\Lambda^2}{d\Lambda}, \end{aligned}$$

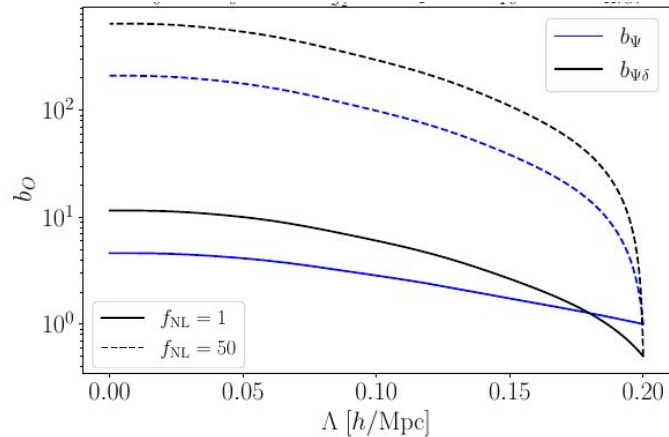
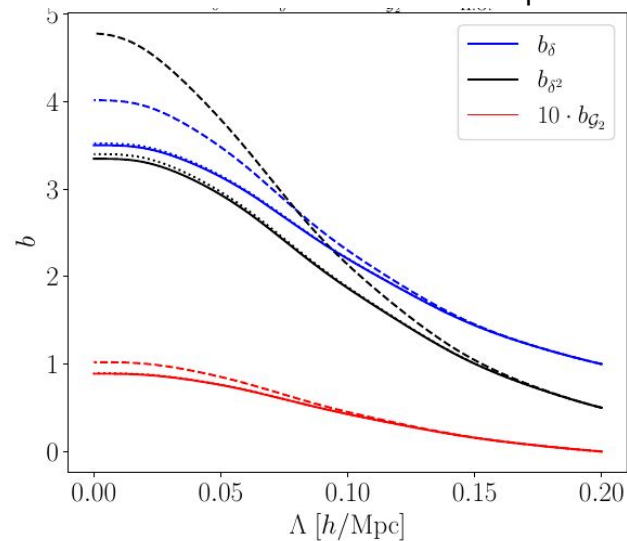
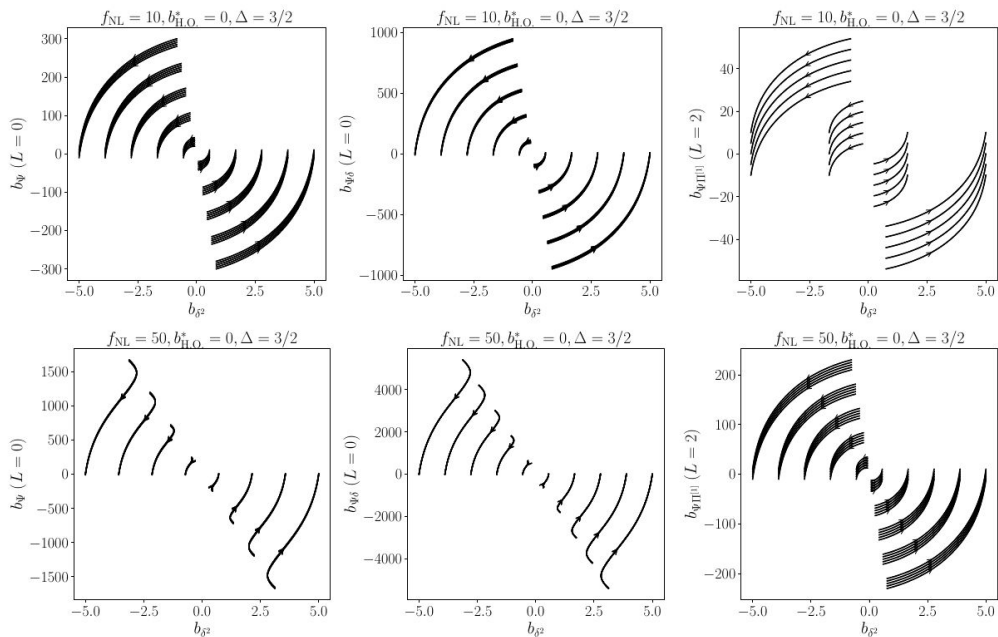
Rederivation of Dalal+ 07 (in an elegant way)

$s_{O'}$	δ^2	δ^3	$\delta\mathcal{G}_2$	Ψ	$\Psi\delta$	$\Psi\delta^2$	$\Psi\mathcal{G}_2$	$\text{Tr } \Psi\Pi^{[1]}$	$\delta \text{Tr } \Psi\Pi^{[1]}$	$\text{Tr } \Psi\Pi^{[2]}$
δ	68/21	3	-4/3	-13/21	13/21	2	-4/3	34/21	1	34/21
δ^2	8126/2205	68/7	-376/105	43/135	478/135	47/21	-31/21	124/315	178/105	14347/6027
\mathcal{G}_2	254/2205	-	116/105	-1699/13230	79/2205	-	-1/21	-661/4410	4/35	-241/735
Ψ	4	-	-	-	-	1	-	-	-	-
$\delta\Psi$	272/21	12	-8/3	-	-	68/21	-	-	-	-
$\text{Tr } \Psi\Pi^{[1]}$	64/105	-	16/15	-	-	-	-	-	8/105	58/305

PNGs

Starting with no PNG in a cutoff, you dynamically generate them!!!
(extending the picture of Assassi, Baumann, Schmidt)

Henrique Rubira



Stochasticity in LSS (2404.16929)

Stochasticity

$$\delta_g(\mathbf{x}, \tau) \equiv \frac{n_g(\mathbf{x}, \tau)}{\bar{n}_g(\tau)} - 1 = \sum_O [b_O(\tau) + c_{\epsilon, O}(\tau) \epsilon(\mathbf{x}, \tau)] O(\mathbf{x}, \tau) + \epsilon(\mathbf{x}, \tau)$$

Noise is a central part in modelling galaxy distribution

Properties of the noise:

How those C coefficients evolve?

$$\langle \epsilon(\mathbf{k}_1) \dots \epsilon(\mathbf{k}_m) O(\mathbf{k}_{m+1}) \rangle = \hat{\delta}_D(\mathbf{k}_{1\dots m}) C_{\epsilon, O}^{(m)} O(\mathbf{k}_{m+1})$$

$$\langle \epsilon(\mathbf{k}_1) O(\mathbf{k}_2) O'(\mathbf{k}_3) \dots \rangle = 0 \quad (\text{linearly does not correlate with O's})$$

Example:

The shot-noise terms

$$\langle \epsilon(\mathbf{k}_1) \dots \epsilon(\mathbf{k}_m) \rangle = \hat{\delta}_D(\mathbf{k}_{1\dots m}) C_{\epsilon, \mathbb{1}}^{(m)}.$$

Stochasticity

Coupled to higher powers of J

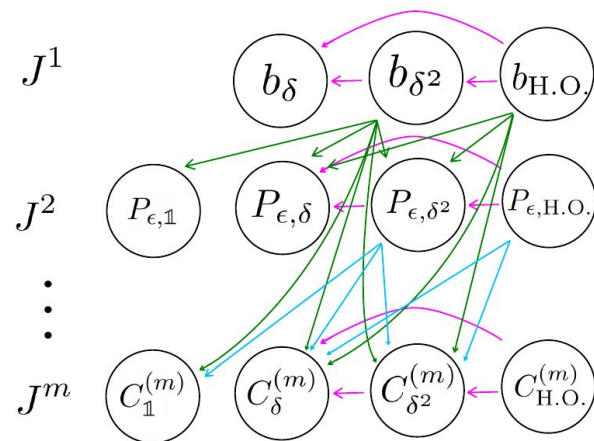
$$\mathcal{Z}[J_\Lambda] = \int \mathcal{D}\delta_\Lambda^{(1)} \mathcal{P}[\delta_\Lambda^{(1)}] \exp \left(\sum_m \left\{ \frac{1}{m!} \int_{\mathbf{x}} \left[(J_\Lambda(\mathbf{x}))^m \sum_O C_O^{(m)}(\Lambda') O[\delta_\Lambda^{(1)}](\mathbf{x}) \right] + \zeta^{(m)}[J_\Lambda, \delta_\Lambda^{(1)}] \right\} \right)$$

Shell corrections

$$\frac{d}{d\Lambda} b_O(\Lambda) = -\frac{d\sigma_\Lambda^2}{d\Lambda} \sum_{O'} s_{O'}^O b_{O'}(\Lambda),$$



$$\frac{d}{d\Lambda} C_O^{(m)}(\Lambda) \propto -[P_L(\Lambda)]^{p-1} \frac{d\sigma_\Lambda^2}{d\Lambda} \sum_{O_1, O_2, \dots, O_m} s_{O_1 O_2 \dots O_m}^O C_{O_1}^{(i_1)}(\Lambda) \dots C_{O_p}^{(i_p)}(\Lambda)$$



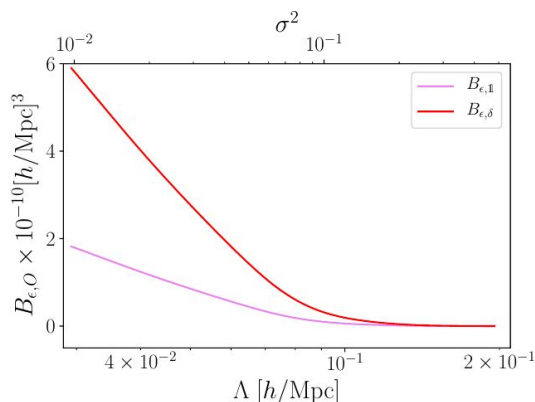
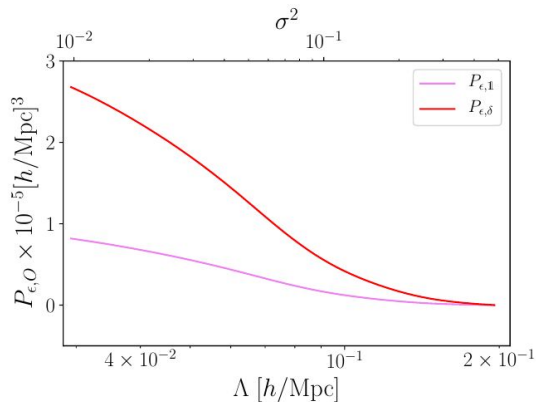
Stochasticity

The δ^2 bias generates the whole tower of stochastic parameters!!!

Examples (shot-noise terms):

$$\frac{dP_{\epsilon,1}}{d\Lambda} = -2[b_{\delta^2}(\Lambda)]^2 [2P_L(\Lambda)] \frac{d\sigma_\Lambda^2}{d\Lambda}$$

$$\frac{dB_{\epsilon,1}}{d\Lambda} = -2P_{\epsilon,\delta}^* b_{\delta^2}(\Lambda) [2P_L(\Lambda)] \frac{d\sigma_\Lambda^2}{d\Lambda} - 8[b_{\delta^2}(\Lambda)]^3 [P_L(\Lambda)]^2 \frac{d\sigma_\Lambda^2}{d\Lambda}$$



Again: starting with no stochasticity, we generate them

Stochasticity

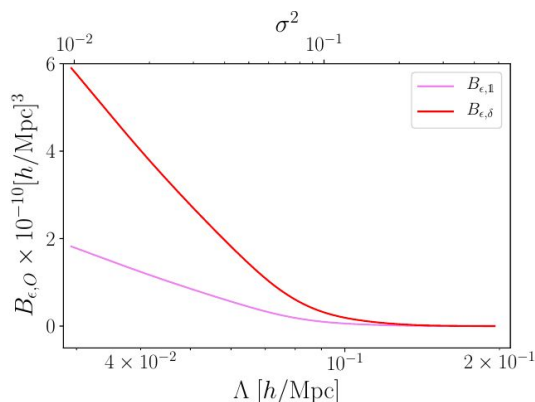
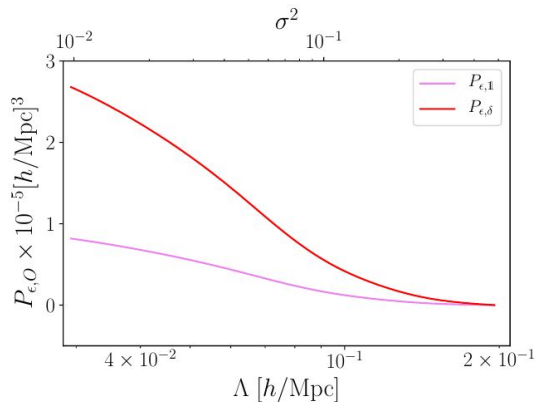
The δ^2 bias generates the whole tower of stochastic parameters!!!

Examples (shot-noise terms):

$$\frac{dP_{\epsilon,1}}{d\Lambda} = -2[b_{\delta^2}(\Lambda)]^2 [2P_L(\Lambda)] \frac{d\sigma_\Lambda^2}{d\Lambda}$$

Chain reaction

$$\frac{dB_{\epsilon,1}}{d\Lambda} = -2P_{\epsilon,\delta}^* b_{\delta^2}(\Lambda) [2P_L(\Lambda)] \frac{d\sigma_\Lambda^2}{d\Lambda} - 8[b_{\delta^2}(\Lambda)]^3 [P_L(\Lambda)]^2 \frac{d\sigma_\Lambda^2}{d\Lambda}$$



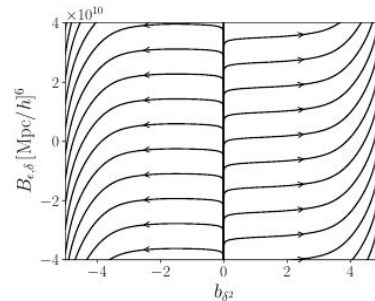
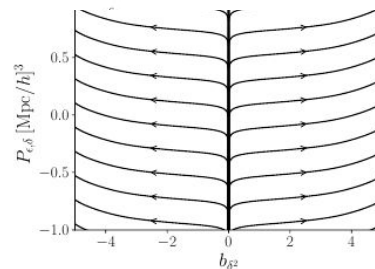
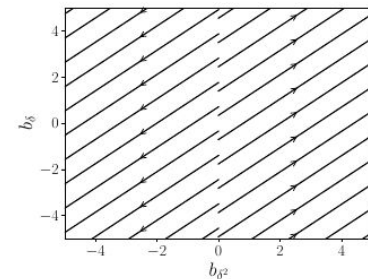
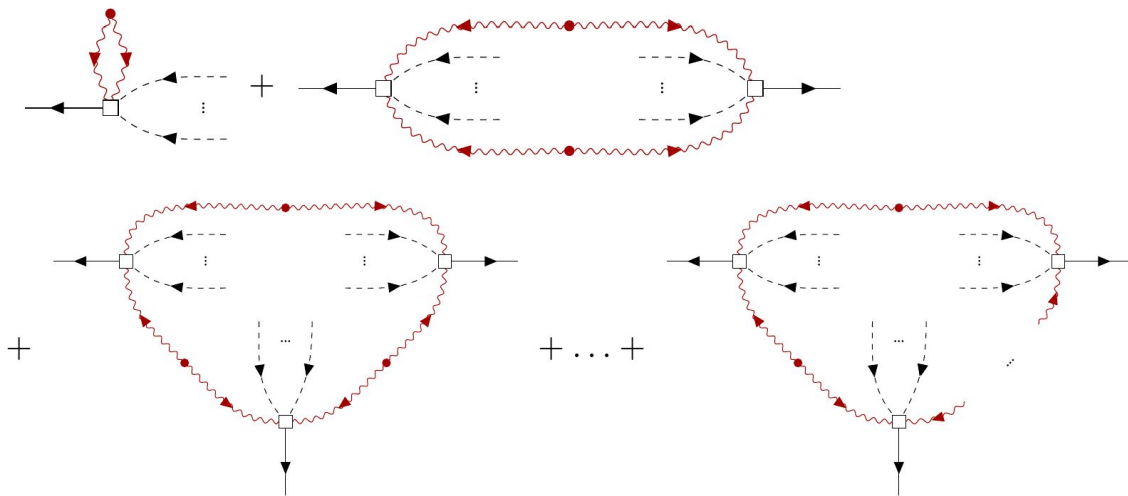
Again: starting with no stochasticity, we generate them

Stochasticity

Conclusion: very simple expression for how general terms in the partition function couple to each other

$$\frac{d}{d\Lambda} C_O^{(m)}(\Lambda) \propto - [P_L(\Lambda)]^{p-1} \frac{d\sigma_\Lambda^2}{d\Lambda} \sum_{O_1, O_2, \dots, O_m} s_{O_1 O_2 \dots O_m}^O C_{O_1}^{(i_1)}(\Lambda) \dots C_{O_p}^{(i_p)}(\Lambda)$$

Simple diagrammatic interpretation



Part IV - Final remarks

How to relate the renormalization schemes?

N-point function renormalized bias
(Assassi, Baumann, Green, Zaldarriaga)

Finite cutoff bias
(This work)

$$\llbracket O' \rrbracket(\mathbf{k}')$$

How to connect both?



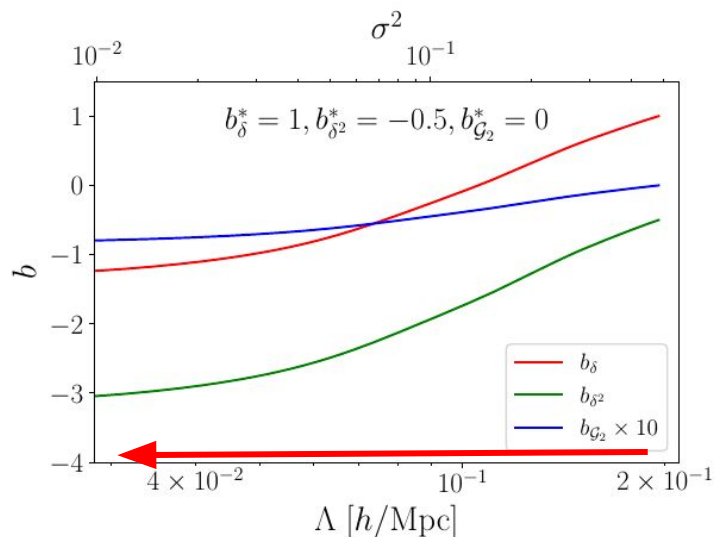
$$O'[\delta_{\Lambda}^{(1)}](\mathbf{k}')$$

How to relate the renormalization schemes?

N-point function renormalized bias
(Assassi, Baumann, Green, Zaldarriaga)

Finite cutoff bias
(This work)

$$\llbracket O' \rrbracket(\mathbf{k}') \quad \xleftrightarrow{\text{How to connect both?}} \quad O'[\delta_{\Lambda}^{(1)}](\mathbf{k}')$$



Solution: Run the bias
towards

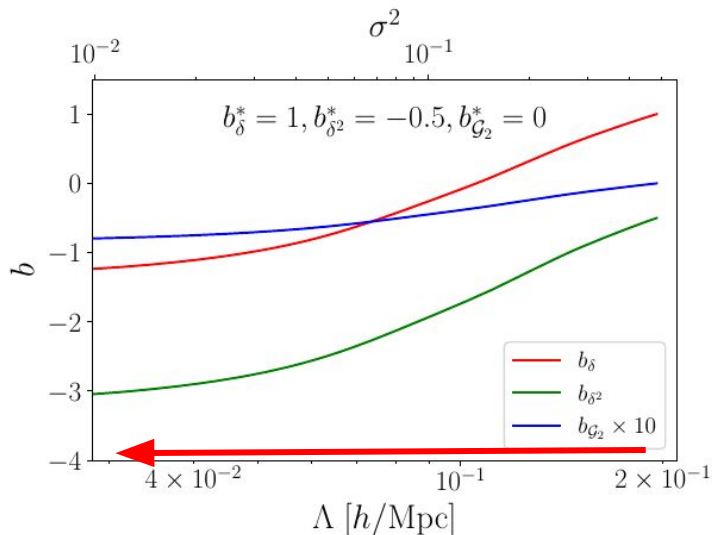
$$\Lambda \rightarrow 0$$

How to relate the renormalization schemes?

N-point function renormalized bias
(Assassi, Baumann, Green, Zaldarriaga)

Finite cutoff bias
(This work)

$$\lim_{\Lambda \rightarrow 0; k/\Lambda \text{ fixed}} \langle O[\delta_{\Lambda}^{(1)}](\mathbf{k}) [O'](\mathbf{k}') \rangle = \lim_{\Lambda \rightarrow 0; k/\Lambda \text{ fixed}} \langle O[\delta_{\Lambda}^{(1)}](\mathbf{k}) O'[\delta_{\Lambda}^{(1)}](\mathbf{k}') \rangle$$



Solution: Run the bias
towards

$$\Lambda \rightarrow 0$$

Conclusion: Why you should care

- Additional cross-check for EFT inference;
- Systematic renormalization of bias/stochastic parameters (including PNG);
- Self-consistent renormalization for $P(k)$, $B(k_1, k_2, k_3)$, ... Also field level
- (Unambiguously) Define Priors for EFT analysis in $\Lambda \rightarrow 0$
- Absorb cutoff dependence in the counter-terms keeping also sub-leading contributions;
- More information? Connection to other fields is manifest, like phase transitions, critical exponents, etc (TBD)



Thanks a lot!

Why just not taking $\Lambda \rightarrow \infty$?

$$\delta_{\Lambda'}^{(1)}(\mathbf{k}) = \delta_{\Lambda}^{(1)}(\mathbf{k}) + \delta_{\text{shell}}^{(1)}(\mathbf{k}) \quad \Lambda = \Lambda' - \lambda$$

$$\mathcal{Z}[J_{\Lambda}] = \int \mathcal{D}\delta_{\Lambda}^{(1)} \mathcal{P}[\delta_{\Lambda}^{(1)}] \exp\left(\int_{\mathbf{k}} J_{\Lambda}(\mathbf{k}) \left[\sum_{\mathbf{o}} b_{\mathbf{o}}^{\Lambda'} \mathcal{O}[\delta_{\Lambda}^{(1)}](-\mathbf{k})\right]\right)$$

Continuum of the theory is determined by taking $\Lambda' \rightarrow \infty$

This determines local terms to be added to the action that will cancel out UV dependence (the counter-terms)

In Wilson-Polchinski we integrate modes up to the cutoff of the theory k_{NL}

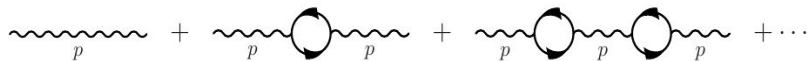
Renormalization scale $\Lambda_* < k_{\text{NL}}$

Logs in QFT

Logs in QFT: Arise when we have a hierarchy of scales

$$\lim_{E \rightarrow \infty} \Gamma(E, m) = E^d \Gamma(1, \frac{m}{E}) \times O \left[\ln \left(\frac{E}{m} \right) \right]$$

Approach 1) Resum diagrams by hand
(if you can)



$$\tilde{V}(p^2) = \frac{e_R^2}{p^2} \left[1 + \frac{e_R^2}{12\pi^2} \ln \frac{p^2}{p_0^2} + \left(\frac{e_R^2}{12\pi^2} \ln \frac{p^2}{p_0^2} \right)^2 + \dots \right] = \frac{1}{p^2} \left[\frac{e_R^2}{1 - \frac{e_R^2}{12\pi^2} \ln \frac{p^2}{p_0^2}} \right]$$

$$e_{\text{eff}}^2(p^2) = \frac{e_R^2}{1 - \frac{e_R^2}{12\pi^2} \ln \frac{p^2}{p_0^2}}$$

Approach 2) Direct from the RGE

$$p_0^2 \frac{d}{dp_0^2} \tilde{V}(p^2) = 0$$

$$p_0^2 \frac{de_{\text{eff}}}{dp_0^2} = \frac{e_{\text{eff}}^3}{24\pi^2}$$

$$e_{\text{eff}}^2(p^2) = \frac{e_R^2}{1 - \frac{e_R^2}{12\pi^2} \ln \frac{p^2}{p_0^2}}$$

The n-point function renormalized bias

(Assassi, Baumann, Green, Zaldarriaga, 2014)

Intuition: Define the bias parameter of order "n" as the large-scale limit of "n+1"-point functions

Example 1: Define the linear bias in the large-scale limit of $P(k)$:

$$b_\delta = \lim_{k \rightarrow 0} \frac{\langle \delta_g \delta \rangle}{\langle \delta \delta \rangle}$$

Example 2: Define the 2nd-order bias parameters in the large-scale limit of $B(k_1, k_2, k_3)$

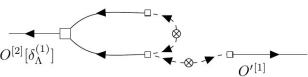

More formally:

$$\langle \delta^{(1)}(\mathbf{k}_1) \dots \delta^{(1)}(\mathbf{k}_m) \llbracket O \rrbracket(\mathbf{k}) \rangle \xrightarrow{k_i \rightarrow 0} \langle \delta^{(1)}(\mathbf{k}_1) \dots \delta^{(1)}(\mathbf{k}_m) O[\delta](\mathbf{k}) \rangle_{\text{LO}}$$

Example:

$$\llbracket \delta^2 \rrbracket = \delta^2 - \sigma_\infty^2 \left(1 + \frac{68}{21} \delta + \frac{8126}{2205} \delta^2 + \frac{254}{2205} \mathcal{G}_2 \right)$$

Differences between renormalization schemes

	N-point renormalization	Finite Λ
In practice, one has to:	Subtract all UV-dep part	Nothing to subtract. Bias params run... and we know how
Potential to:	Error-prone, missing finite contributions	Extra sanity-checks
$\langle (\delta^1)^{(1)} (\delta^2)^{(3)} \rangle$ 	Completely removed by c.t., but missing sub-leading $k^2 P_L(k) \int_{\mathbf{p}} p^{-2} P_L(p)$	Finite: $4b_\delta^\Lambda b_{\delta^2}^\Lambda \int_{\mathbf{p}} F_2(\mathbf{p}, \mathbf{k} - \mathbf{p}) P_L^\Lambda(p) P_L^\Lambda(k)$
$\langle (\delta^2)^{(2)} (\delta^2)^{(2)} \rangle$ 	Subtracted to the stochastic term, but missing sub-leading $k^2 \int_{\mathbf{p}} p^{-2} [P_L(p)]^2$	Finite and contributes to the stochastic running: $2(b_{\delta^2}^\Lambda)^2 \int_{\mathbf{p}} P_L^\Lambda(p) P_L^\Lambda(\mathbf{k} - \mathbf{p})$

Logs in LSS

$$\Delta_{1-loop}^2 = \left(\frac{k}{k_{NL}}\right)^{3+n} + \left(\frac{k}{k_{NL}}\right)^{2(3+n)} \left[\alpha(n) + \tilde{\alpha}(n) \ln\left(\frac{k}{k_{NL}}\right) \right]$$

n	-2	-3/2	-1	-1/2	0	1/2	1	3/2	2	5/2	3
α_{13}	$\frac{5\pi^2}{112}$	$\frac{992\pi}{6,615}$...	$-\frac{416\pi}{8,085}$	$-\frac{\pi^2}{336}$	$-\frac{\pi^2}{168}$
α_{22}	$\frac{75\pi^2}{784}$	-0.232698	$\frac{29\pi^2}{784}$	$\frac{\pi^2}{392}$
$\tilde{\alpha}_{13}$	0	0	$\frac{61}{315}$	0	0	0	$-\frac{4}{105}$	0	0	0	$\frac{20}{1,323}$
$\tilde{\alpha}_{22}$	0	0	0	0	0	$-\frac{9}{98}$	0	$\frac{31}{16,464}$	0	$-\frac{359}{26,880}$	0
α	1.38	.239537	.336	-.0336
$\tilde{\alpha}$	0	0	.194	0	0	-.0918	.0381	-.00188	0	-.0134	.0151

The shell expansion (Wilson formalism)

Consider a very thin shell with width: $\Lambda = \Lambda' - \lambda$

$$\delta_{\Lambda'}^{(1)}(\mathbf{k}) = \delta_{\Lambda}^{(1)}(\mathbf{k}) + \delta_{\text{shell}}^{(1)}(\mathbf{k})$$

Idea: Integrate out the shell!

$$\mathcal{Z}[J_{\Lambda}] = \int \mathcal{D}\delta_{\Lambda}^{(1)} \mathcal{P}[\delta_{\Lambda}^{(1)}] \int \mathcal{D}\delta_{\text{shell}}^{(1)} \mathcal{P}[\delta_{\text{shell}}^{(1)}] \times \exp\left(\int_{\mathbf{k}} J_{\Lambda}(\mathbf{k}) \left[\sum_{\mathcal{O}} b_{\mathcal{O}}^{\Lambda'} \mathcal{O}[\delta_{\Lambda}^{(1)} + \delta_{\text{shell}}^{(1)}](-\mathbf{k}) \right] + \frac{1}{2} P_{\epsilon}^{\Lambda'} \int_{\mathbf{k}} J_{\Lambda}(\mathbf{k}) J_{\Lambda}(-\mathbf{k}) + \mathcal{O}[J_{\Lambda}^2 \delta_{\Lambda}^{(1)}, J_{\Lambda}^3] \right) \quad (2.7)$$

1) Expand the operators in terms of the number of shell fields and integrate those out!

$$\begin{aligned} \mathcal{O}^{(n)}[\delta_{\Lambda}^{(1)} + \delta_{\text{shell}}^{(1)}] &= \mathcal{O}^{(n)}[\delta_{\Lambda}^{(1)}] + \mathcal{O}^{(n),(1)\text{shell}}[\delta_{\Lambda}^{(1)}, \delta_{\text{shell}}^{(1)}] + \mathcal{O}^{(n),(2)\text{shell}}[\delta_{\Lambda}^{(1)}, \delta_{\text{shell}}^{(1)}] \\ &\quad + \dots + \mathcal{O}^{(n),(n-1)\text{shell}}[\delta_{\Lambda}^{(1)}, \delta_{\text{shell}}^{(1)}] + \mathcal{O}^{(n)}[\delta_{\text{shell}}^{(1)}], \end{aligned} \quad (2.8)$$

2) Integrate the shells

$$\begin{aligned} \mathcal{S}_O^2[\delta_{\Lambda}^{(1)}] &= \sum_{n \geq 2} \int \mathcal{D}\delta_{\text{shell}}^{(1)} \mathcal{P}[\delta_{\text{shell}}^{(1)}] \mathcal{O}^{(n),(2)\text{shell}}[\delta_{\Lambda}^{(1)}, \delta_{\text{shell}}^{(1)}](\mathbf{k}) \\ \mathcal{S}_{OO'}^{11}[\delta_{\Lambda}^{(1)}](\mathbf{k}, \mathbf{k}') &= \sum_{n, n' \geq 1} \int \mathcal{D}\delta_{\text{shell}}^{(1)} \mathcal{P}[\delta_{\text{shell}}^{(1)}] \mathcal{O}^{(n),(1)\text{shell}}[\delta_{\Lambda}^{(1)}, \delta_{\text{shell}}^{(1)}](\mathbf{k}) \mathcal{O}'^{(n'),(1)\text{shell}}[\delta_{\Lambda}^{(1)}, \delta_{\text{shell}}^{(1)}](\mathbf{k}') \end{aligned}$$

Solutions

- 1) Neglect fourth-order+ bias;
- 2) Neglect **the running** of fourth-order+ bias;
- 3) Ansatz for higher-order bias running.

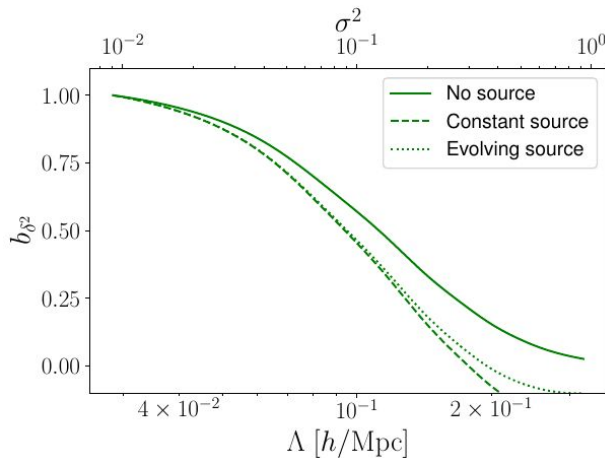
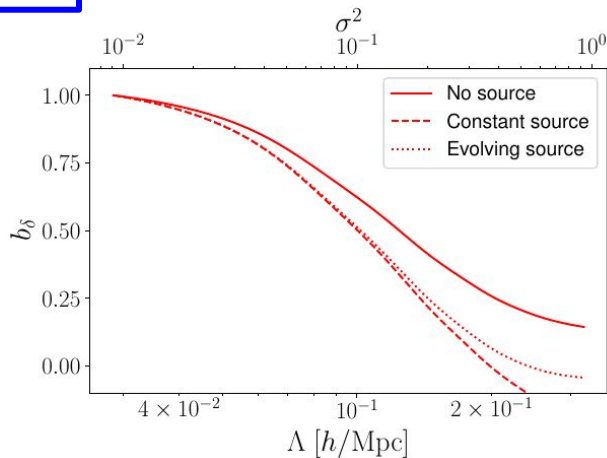
Wilson-Polchinski RG-equations

$$\frac{db_\delta}{d\Lambda} = - \left[\frac{68}{21} b_{\delta^2} + 3b_{\delta^3}^* - \frac{4}{3} b_{\mathcal{G}_2\delta}^* \right] \frac{d\sigma_\Lambda^2}{d\Lambda},$$

$$\frac{db_{\delta^2}}{d\Lambda} = - \left[\frac{8126}{2205} b_{\delta^2} + \frac{17}{7} b_{\delta^3}^* - \frac{376}{105} b_{\mathcal{G}_2\delta}^* + b_{n=4}^{*(\delta^2)} \right] \frac{d\sigma_\Lambda^2}{d\Lambda},$$

$$\frac{db_{\mathcal{G}_2}}{d\Lambda} = - \left[\frac{254}{2205} b_{\delta^2} + \frac{116}{105} b_{\mathcal{G}_2\delta}^* + b_{n=4}^{*(\mathcal{G}_2)} \right] \frac{d\sigma_\Lambda^2}{d\Lambda}.$$

$$b_{n=3+4}^{(O)}(\sigma^2) = b_{n=3+4}^{*(O)} e^{-c^{(O)}(\sigma^2 - \sigma_*^2)}$$



Conclusions:

- 1) **Neglecting source affects result;**
- 2) **Evolving source does not affect the result!**

EFTofLSS via Wilson Polchinski

(based on Carroll,
Leichenauer, Pollack,
13)

$$Z[J] = \int \mathcal{D}\phi_{\text{in}} \exp(S_0[\phi_{\text{in}}] + J_i \phi^i[\phi_{\text{in}}]) \quad \text{with} \quad S_0[\phi_{\text{in}}] = -\frac{1}{2} \phi^i [P(\Lambda)^{-1}]_{ij} \phi^j$$

We use $\left. \frac{\partial \mathcal{Z}}{\partial J_\Lambda \dots \partial J_\Lambda} \right|_{J_\Lambda=0}$ Since $\langle \phi^{i_1} \dots \phi^{i_n} \rangle = \int \mathcal{D}\phi_{\text{in}} \phi^{i_1}[\phi_{\text{in}}] \dots \phi^{i_n}[\phi_{\text{in}}] e^{S_0[\phi_{\text{in}}]}$

$$\phi_{\text{SPT}}^i \equiv K_{\text{SPT}j}^i \phi_{\text{in}}^j + \frac{1}{2} K_{\text{SPT}jk}^i \phi_{\text{in}}^j \phi_{\text{in}}^k + \dots$$

Advantages:

- Path integral formulation
- Systematic generation EFT structure (coefficients are closed under RG flow)
- Keeps small (yet-perturbative) modes in the theory

$$\frac{d}{d\Lambda} K_{i_1 \dots i_m}^{j_1 \dots j_n} = -\frac{1}{2} \left(\frac{dP^{ij}}{d\Lambda} K_{ij i_1 \dots i_m}^{j_1 \dots j_n} + \frac{dP^{ij}}{d\Lambda} \sum_{k=0}^m \sum_{l=0}^n \binom{m}{k} \binom{n}{l} K_{i i_1 \dots i_k}^{j_1 \dots j_l} K_{j_{l+1} \dots j_n}^{i_{k+1} \dots i_m} \right)$$

Historical overview and frameworks

- **Dim Reg, scale transformations and applications to QED:** Stueckelberg, Petermann, Gell-Mann, Low ~1953
- **RG in condensed matter:** Kadanoff, 1966
- **RG in the continuum, derivation of RG equations and critical phenomena:** Callan and Symanzik 1970, Kenneth Wilson, 1970/71 (Nobel Prize 1982)
- **RG via path integrals:** Polchinski, 1984

Framework 1 (a la Wilson/Polchinski):

$$\Lambda \frac{d}{d\Lambda} Z[J] = 0$$

Sliding cutoff,

integrate out modes between cutoffs

$$\Lambda \rightarrow \Lambda'$$

Framework 2:

Sliding renormalization conditions (e.g. Dim Reg), no UV regulator

$$\frac{\partial g}{\partial \ln \mu} = \beta(g)$$

- More practical for computations