

Field level bias modeling, assembly bias, and primordial non-Gaussianity

Theoretical Modeling of Large-Scale Structure of the Universe
June 5 2024

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(based on work w/ Stephen Chen, Uroš Seljak)



Simplest Local Primordial Non-Gaussianity

Initial gravitational potential largely Gaussian

But primordial physics can add non-Gaussianity (PNG)

(Oliver's talk)

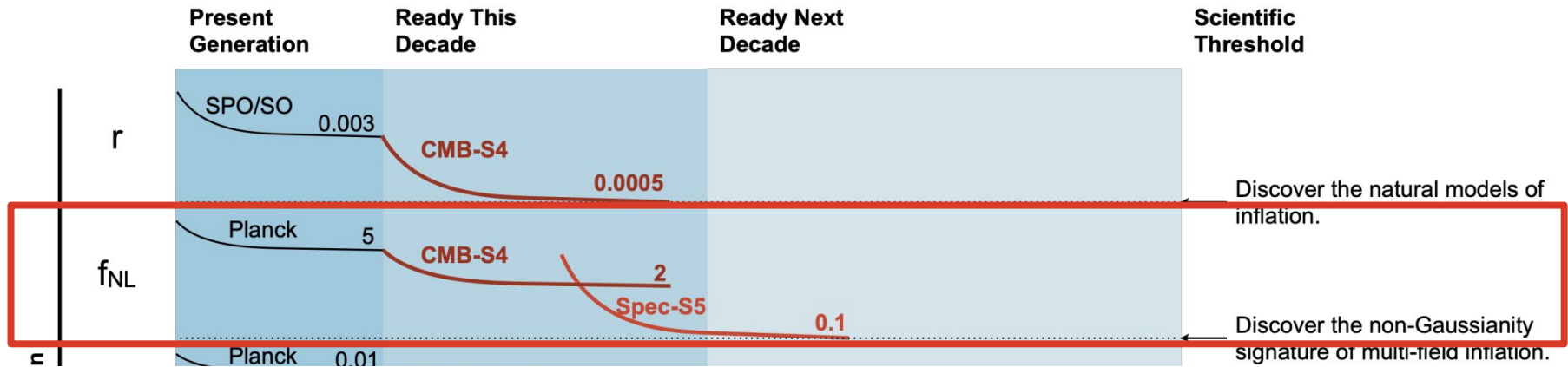
E.g. multi-field inflation produces *local* PNG:

$$\boxed{\phi} = \phi_G + \boxed{f_{NL}^{\text{loc}}} [\phi_G^2 - \langle \phi_G^2 \rangle]$$

Seed for structure

Counts multiple fields

f_{NL} a Prime Target of Future Galaxy Surveys

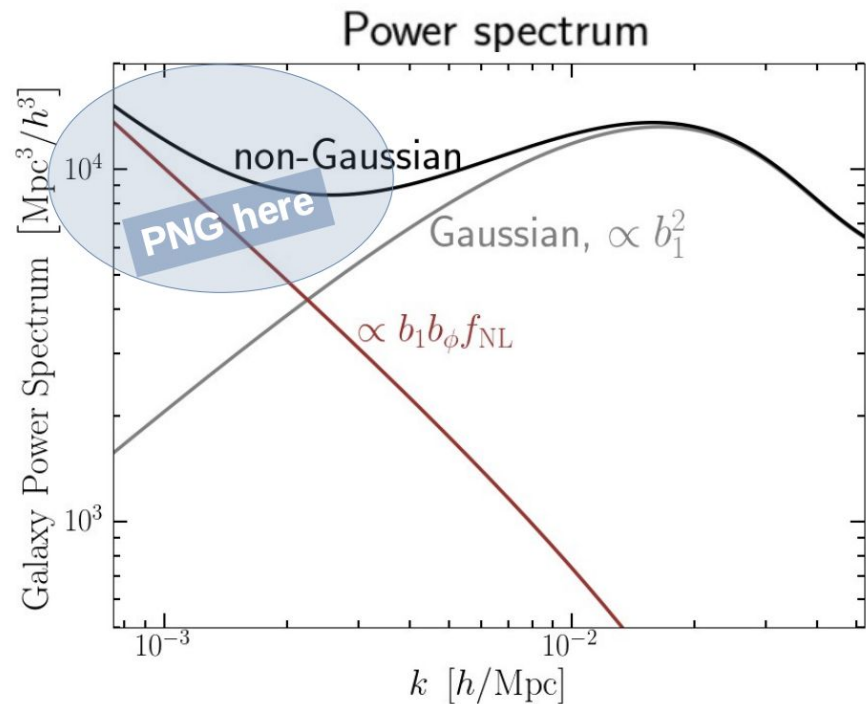


Spec-S5 (Martin's talk):

- DESI, Euclid, SPHEREx, PFS...
- Also SO x Rubin-LSST, CMB-S4

Measuring LPNG in Galaxy Surveys

Measure power spectrum



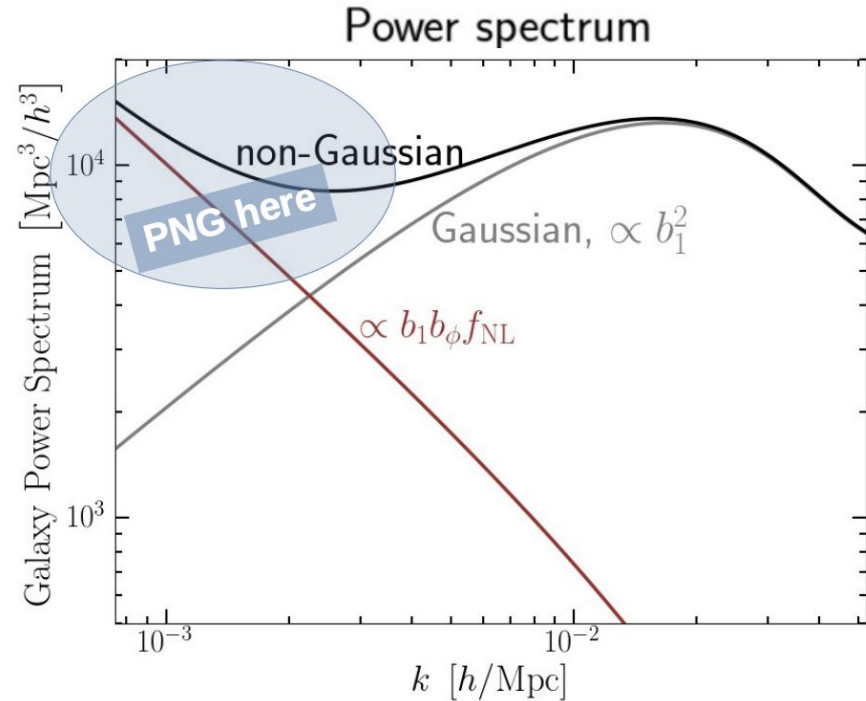
Measuring LPNG in Galaxy Surveys

Measure power spectrum

To model it need a galaxy bias model:

$$\boxed{\delta_g} = b \boxed{\delta} + \dots$$

Galaxy **Matter**
overdensity **overdensity**



Measuring LPNG in Galaxy Surveys

Measure power spectrum

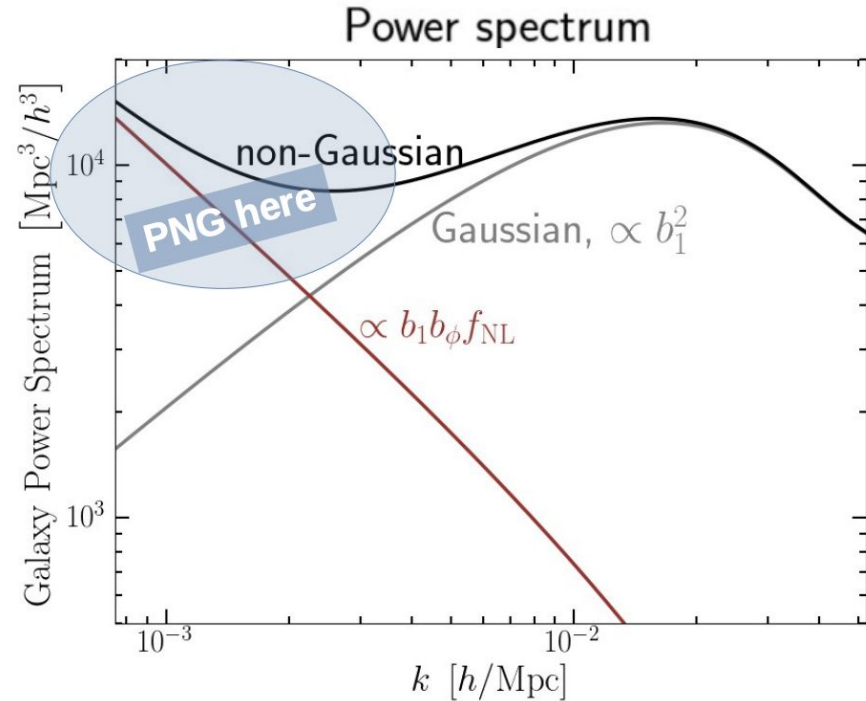
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Galaxy
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15 years ago, it was realized there is an **extra** bias signal

$$\delta_g = b \delta + \boxed{b_\phi f_{\text{NL}}^{(\text{loc})} \phi} + \dots$$



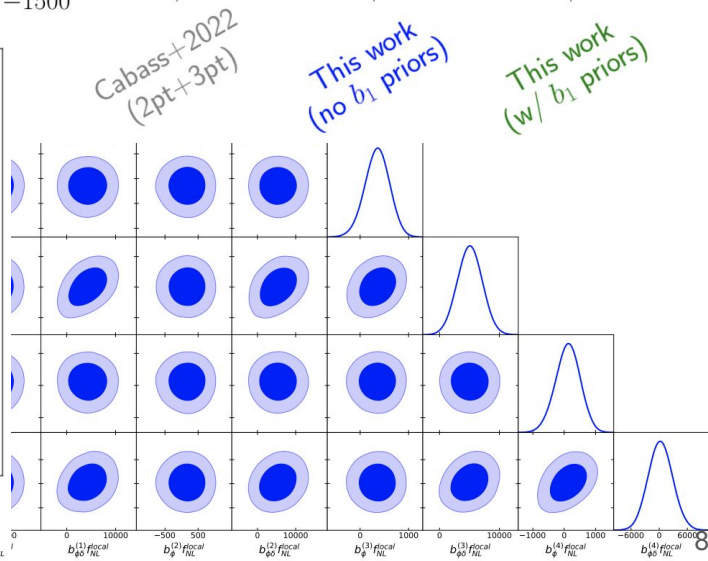
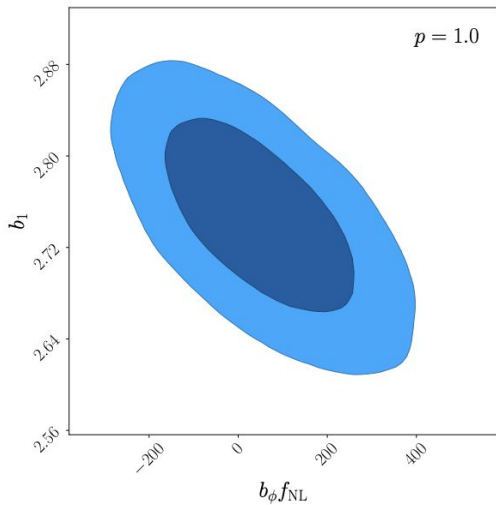
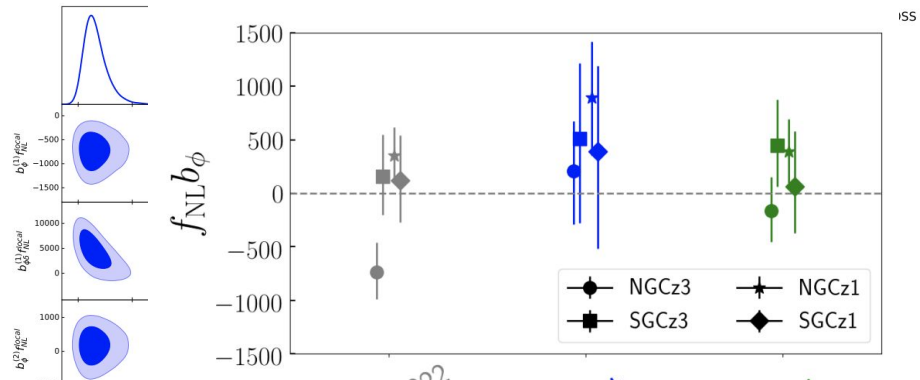
Constraining f_{NL}

Option 1 - LPNG amplitude

Assumption free!

Option 2 - Constrain f_{NL}

Requires knowing the value of b_ϕ
(Also plan surveys this way)



Constraining f_{NL}

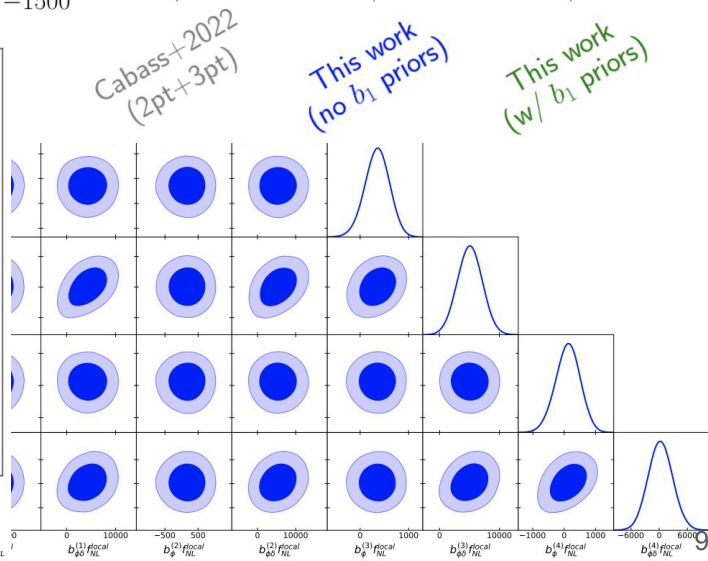
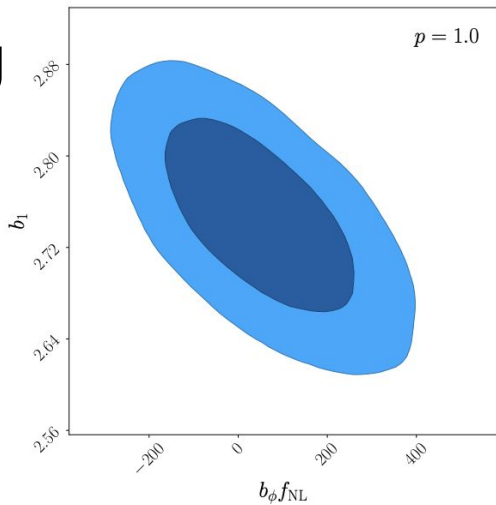
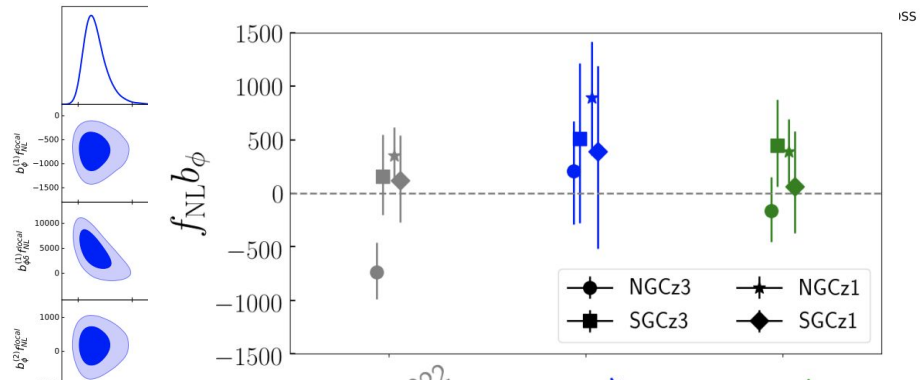
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Option 2 - Constrain f_{NL}

Requires knowing
the value of b_ϕ

Assume we are
ambitious...



Assuming a b_ϕ - Universal Mass Function

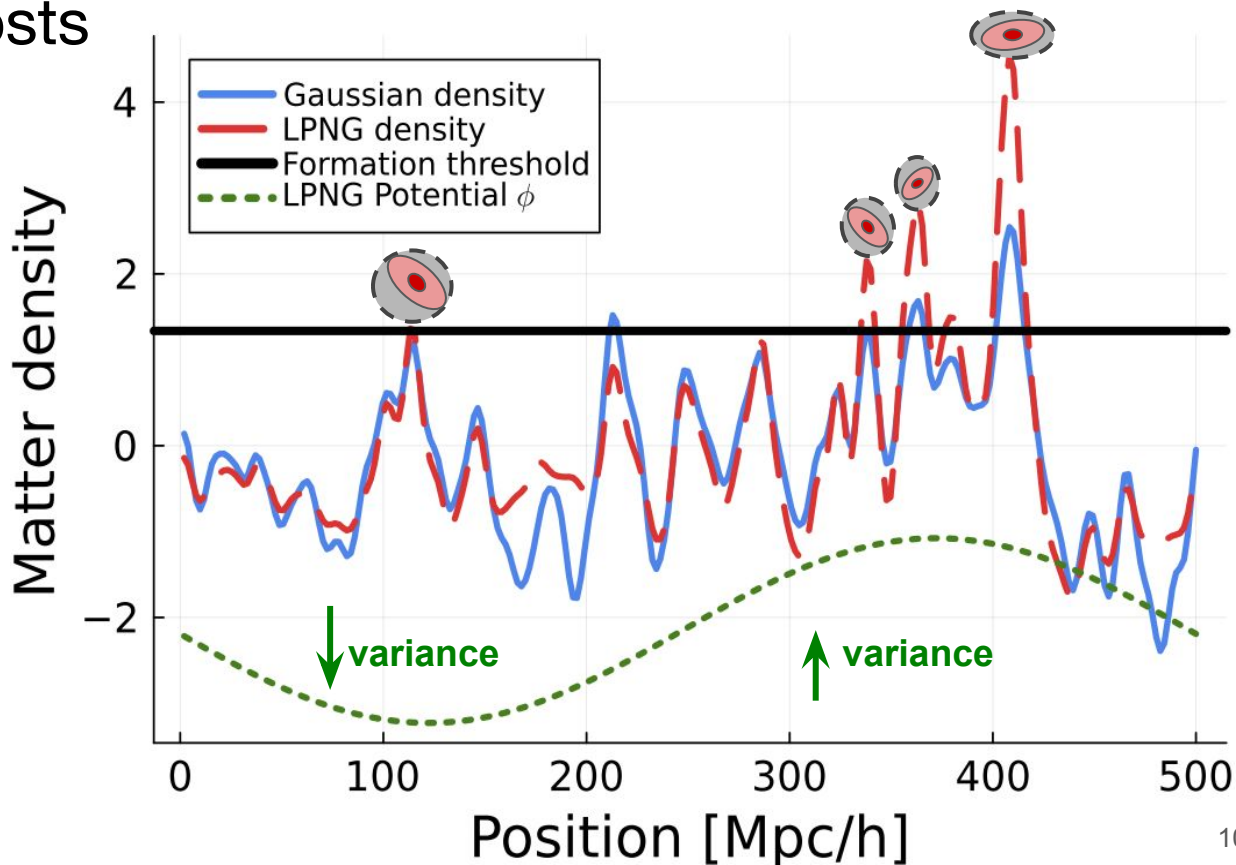
Cartoon: LPNG “boosts local variance”

Halos form after crossing threshold

Crossing affected by LPNG

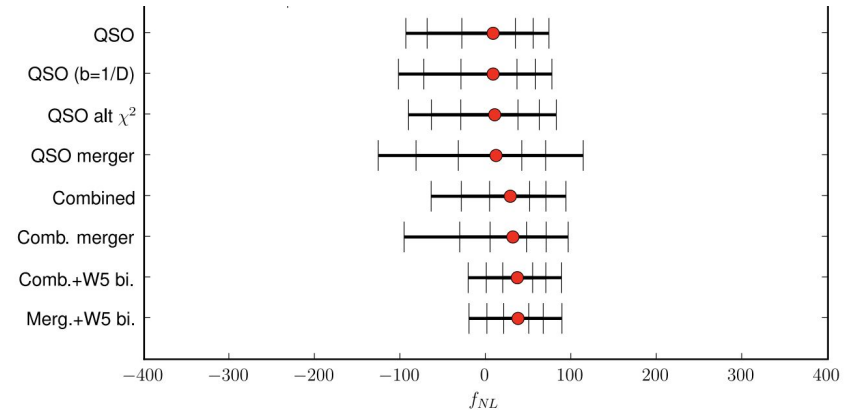
Assume UMF form:

$$b_\phi(b, p) \propto b - p$$



Galaxy survey f_{NL} - SDSS quasars

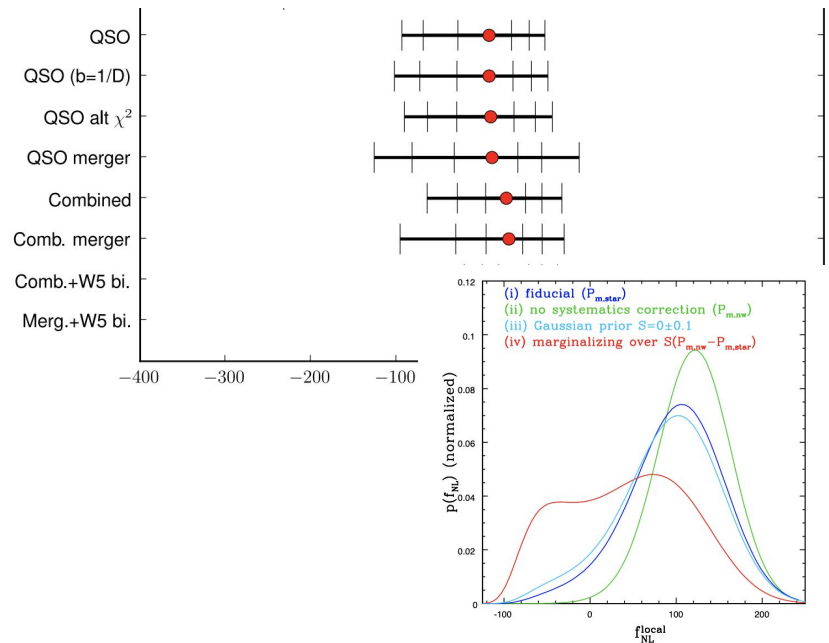
Slosar++08



Galaxy survey f_{NL} - SDSS quasars

Slosar++08

Ross++12

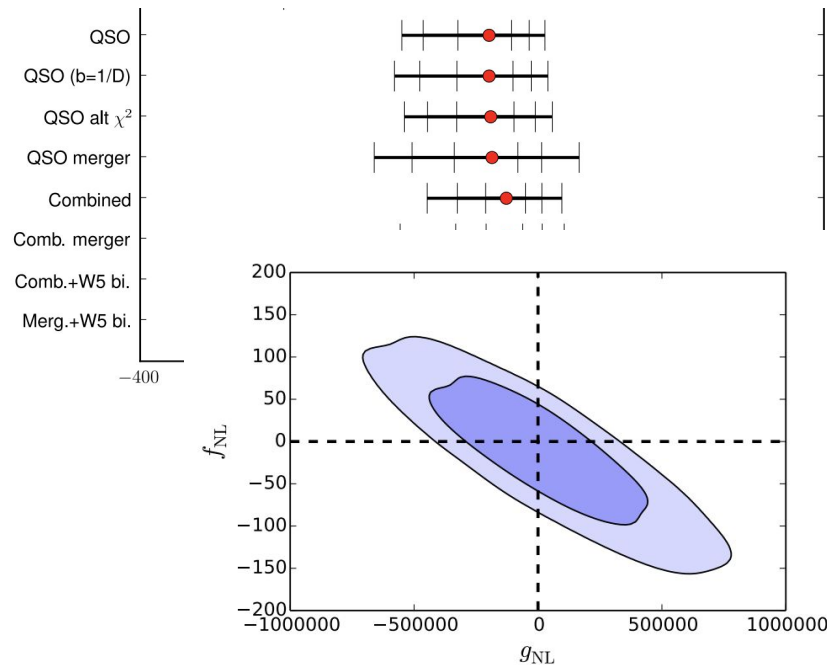


Galaxy survey f_{NL} - SDSS quasars

Slosar++08

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Leistedt++14



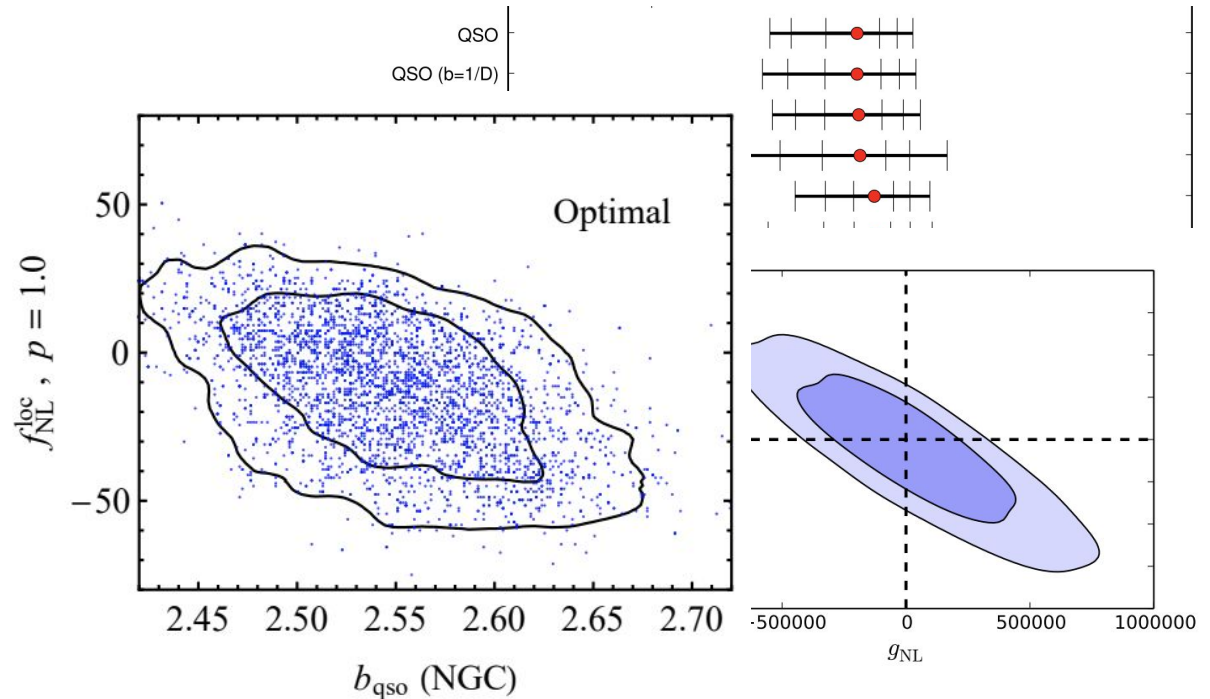
Galaxy survey f_{NL} - eBOSS quasars

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Ross++12

Leistedt++14

Castorina++19



Galaxy survey f_{NL} - eBOSS quasars

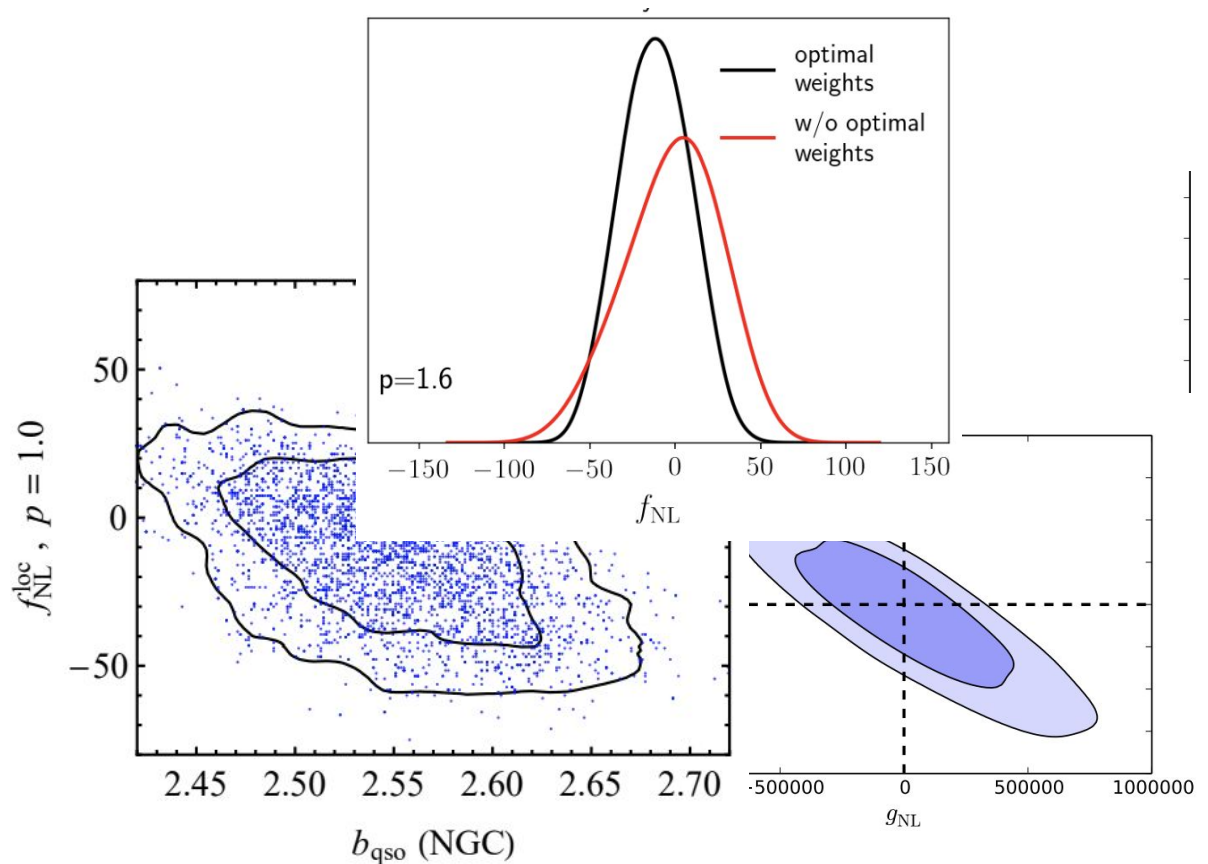
Slosar++08

Ross++12

Leistedt++14

Castorina++19

Mueller++21



Galaxy survey f_{NL} - BOSS LRGs

Slosar++08

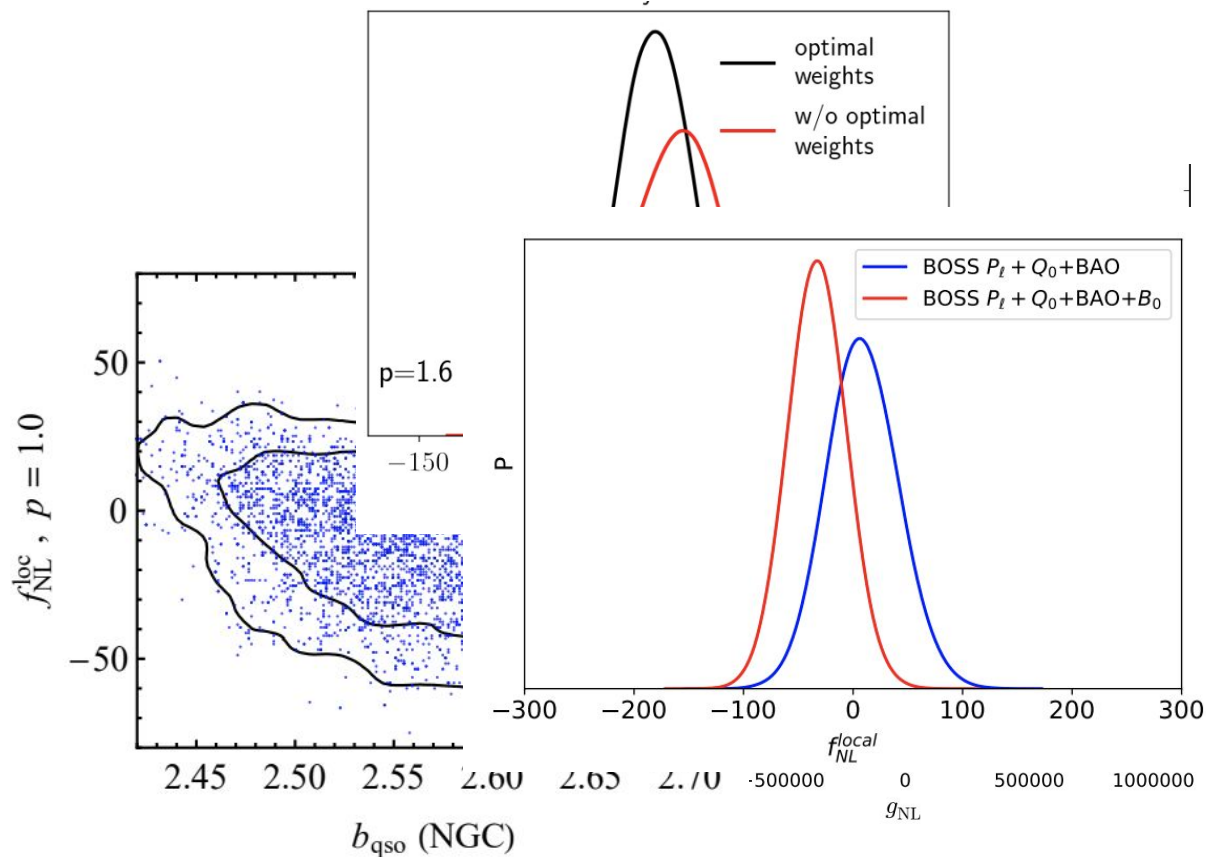
Ross++12

Leistedt++14

Castorina++19

Mueller++21

D'Amico++22



Galaxy survey f_{NL} - BOSS LRGs

Slosar++08

Ross++12

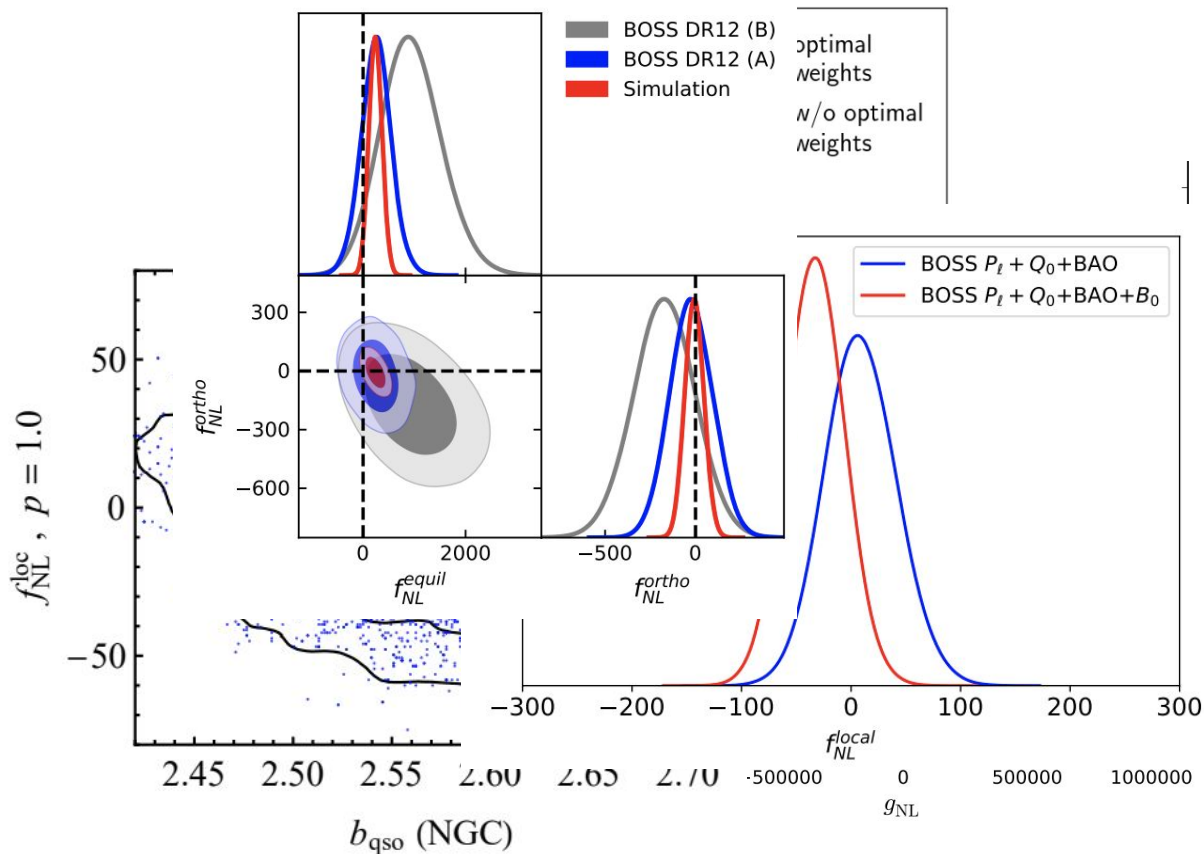
Leistedt++14

Castorina++19

Mueller++21

D'Amico++22

Cabass++22a



Galaxy survey f_{NL} - eBOSS/DESI

Slosar++08

Ross++12

Leistedt++14

Castorina++19

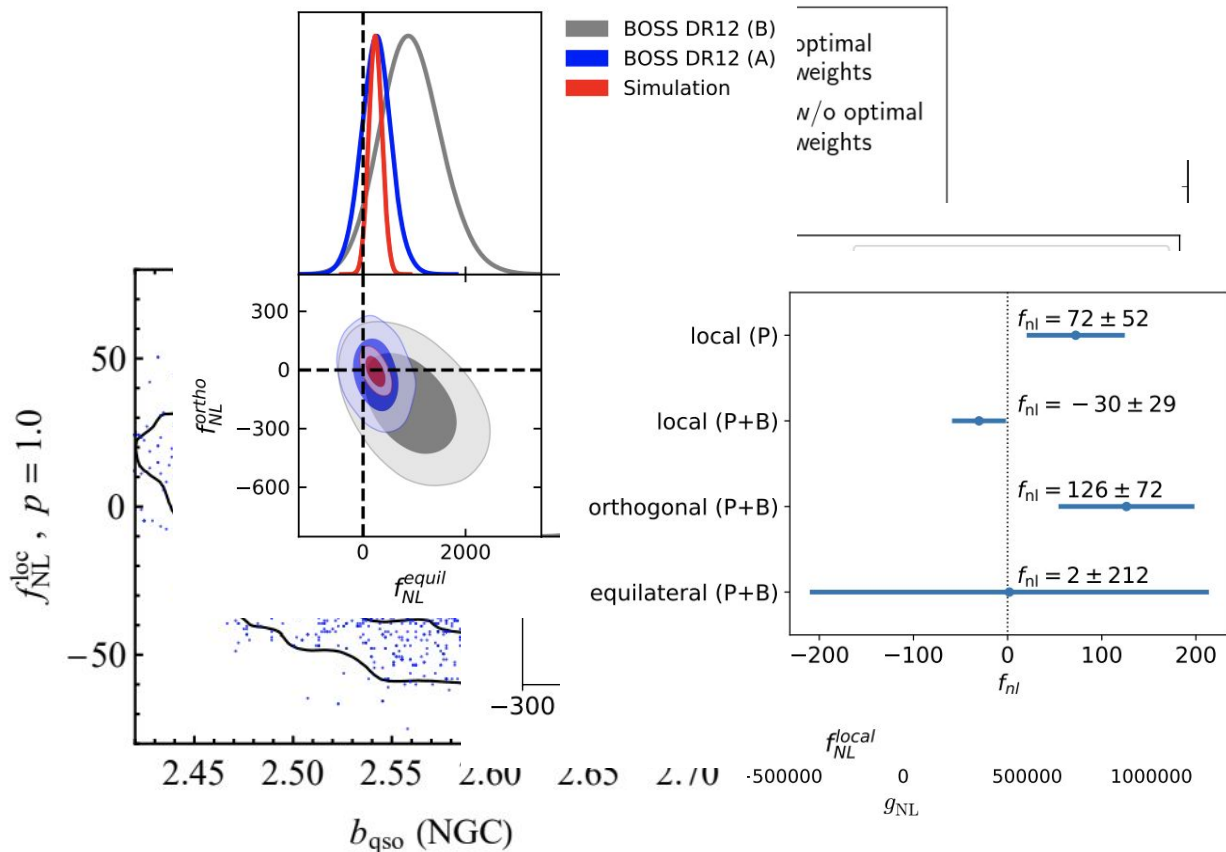
Mueller++21

D'Amico++22

Cabass++22a

Rezaie++23

Cagliari++23



Understanding b_ϕ

How worried should we be about standard assumption?

Can we break the degeneracy between b_ϕ and f_{NL} ?

Attacking b_ϕ from **3 angles**:

1. Test b_ϕ in simulations at field level
2. *Data-driven* b_ϕ prior?
3. Can we *model* deviations from standard assumption?
(Assembly bias, see also 2303.08901)

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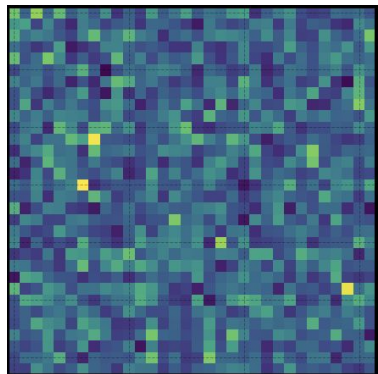
Field-level Bias Model

Idea: test of PNG bias at the field level (quadratic Lag model)

Field-level likelihood - simple regression - NL displacements

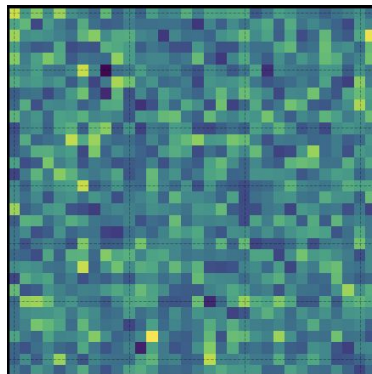
See Francisco's, Matteo's, Marcos' talks. (+Juila's and Kazu's talks for PT flavor)

Halos



$= b_1$

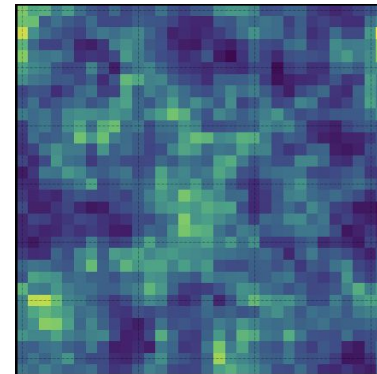
Matter



$+$

$b_{\phi} f_{NL}$

Potential



Field-level Bias Model

At 2nd order in bias, **2 Local PNG terms**

Neglect position-dependent variance

$$\mathcal{P}[\delta_t|\delta] = \prod_{\mathbf{x}} (2\pi\sigma_0^2)^{-\frac{1}{2}} \exp\left(-\frac{|\delta_t(\mathbf{x}) - \delta_{t,\text{fwd}}(\mathbf{x}, \delta)|^2}{2\sigma_0^2}\right)$$

$$\begin{aligned}\tilde{\delta}_{t,\text{fwd}}(\mathbf{x}, \delta) = & \delta(\mathbf{x}) + b_\delta D(z) \delta^{\text{adv}}(\mathbf{q}) \\ & + b_{\delta^2} D^2(z) \delta^{2,\text{adv}}(\mathbf{q}) + b_{K^2} D^2(z) K_{ij}^{2,\text{adv}}(\mathbf{q}) \\ & + c_{\nabla^2\delta} D(z) (\nabla^2\delta)^{\text{adv}}(\mathbf{q}) \\ & + \epsilon_t(\mathbf{x}),\end{aligned}$$

$$\begin{aligned}\delta_{t,\text{fwd}}(\mathbf{x}) = & \tilde{\delta}_{t,\text{fwd}}(\mathbf{x}, \delta) + b_\phi f_{NL}^{\text{loc}} \phi^{\text{adv}}(\mathbf{q}) \\ & + b_{\phi\delta} f_{NL}^{\text{loc}} [\phi\delta]^{\text{adv}}(\mathbf{q})\end{aligned}$$

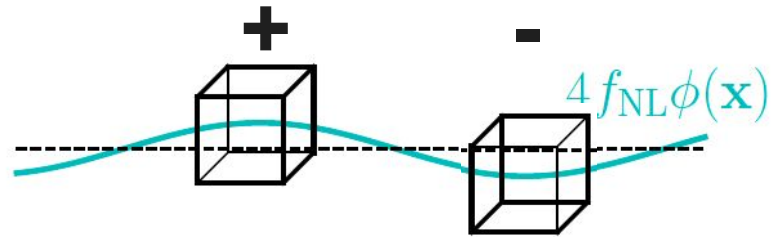
Separate Universe (response) b_ϕ

Separate Universe (-> peak-background split)

Finite-difference 2 sims

Uses infinite-wavelength
limit

Separate Universe \mathcal{A}_s



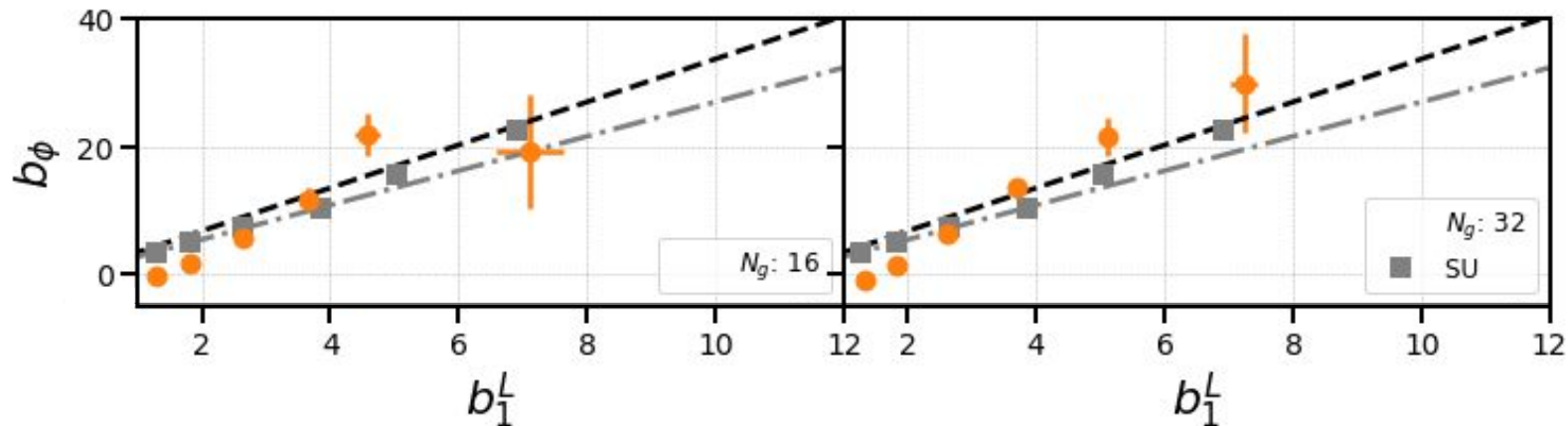
$$\mathcal{A}_s \rightarrow \mathcal{A}_s [1 + \delta\mathcal{A}_s] , \delta\mathcal{A}_s = 4f_{\text{NL}}\phi_L$$

$$b_\phi(M, t) = 4d\ln n_g(M, t)/d\delta\mathcal{A}_s$$

UMF Prediction

Roughly agree with UMF and SU on large scales

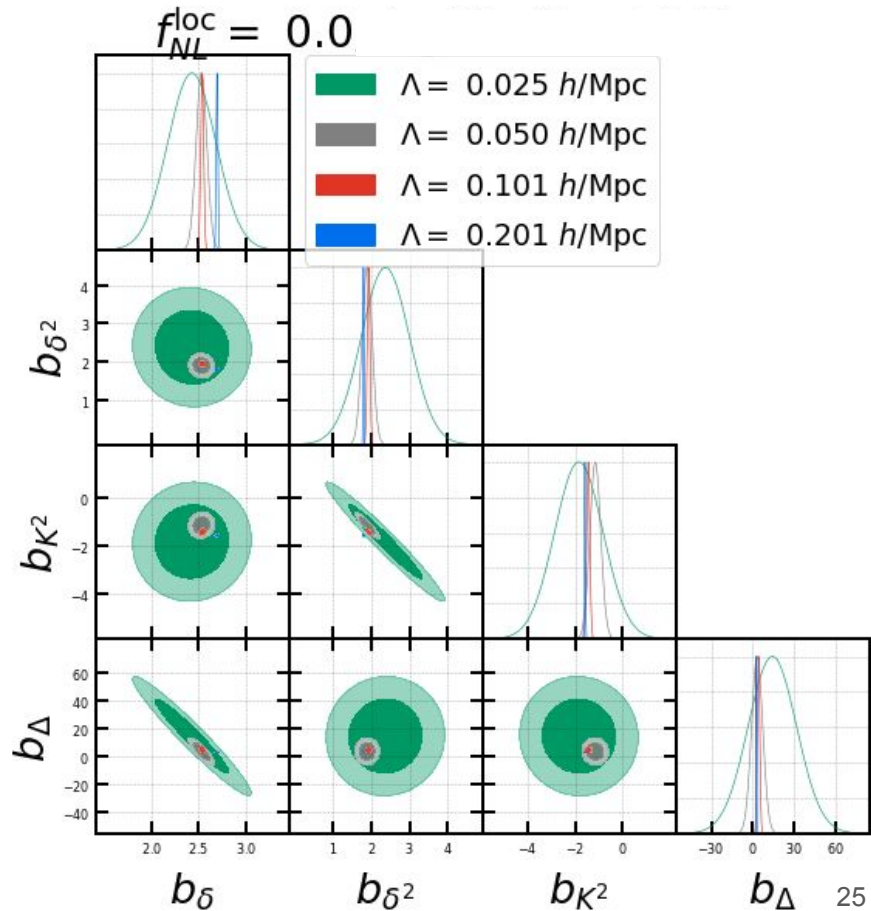
Somewhat resolution dependent...



Quadratic Bias Parameters

How are we doing with
the cutoff Λ ?

Looks good for Gaussian
up to **red scale**



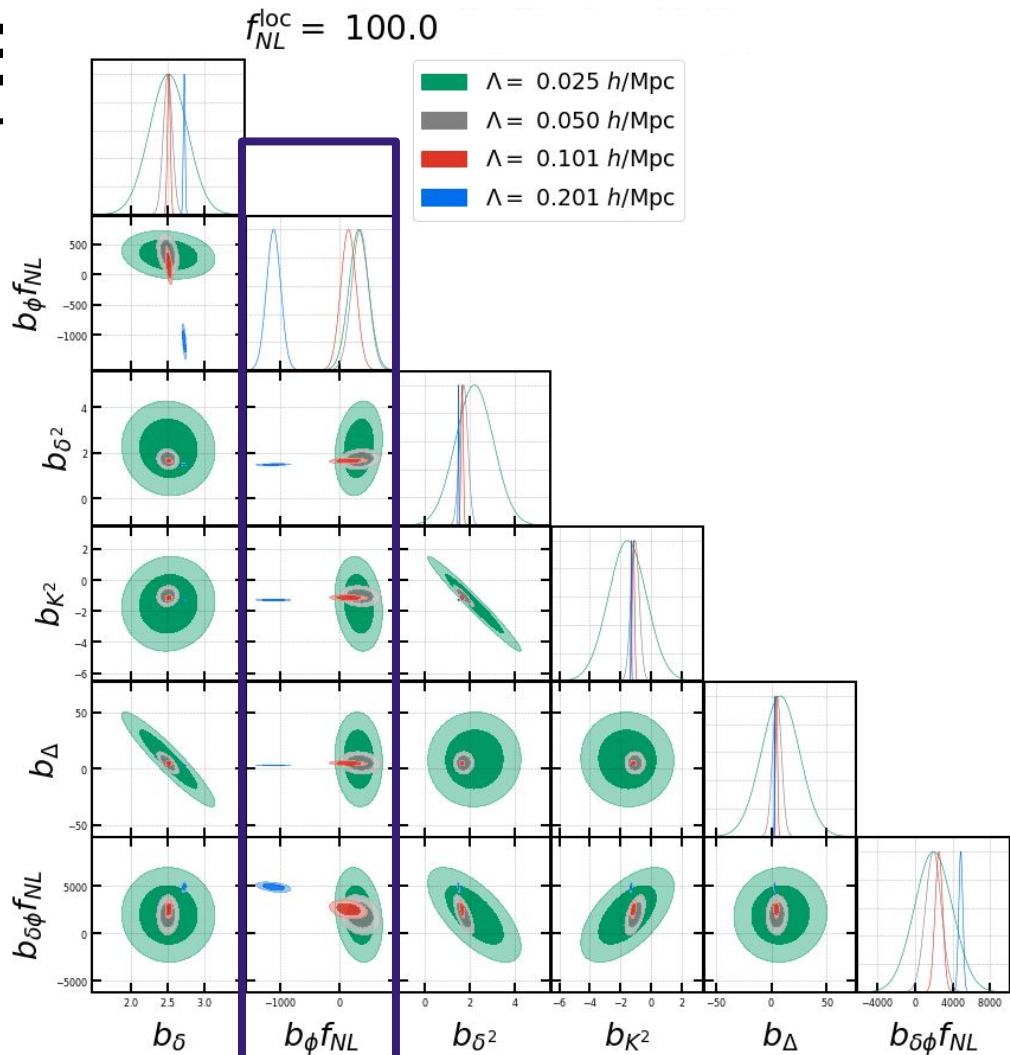
Quadratic I

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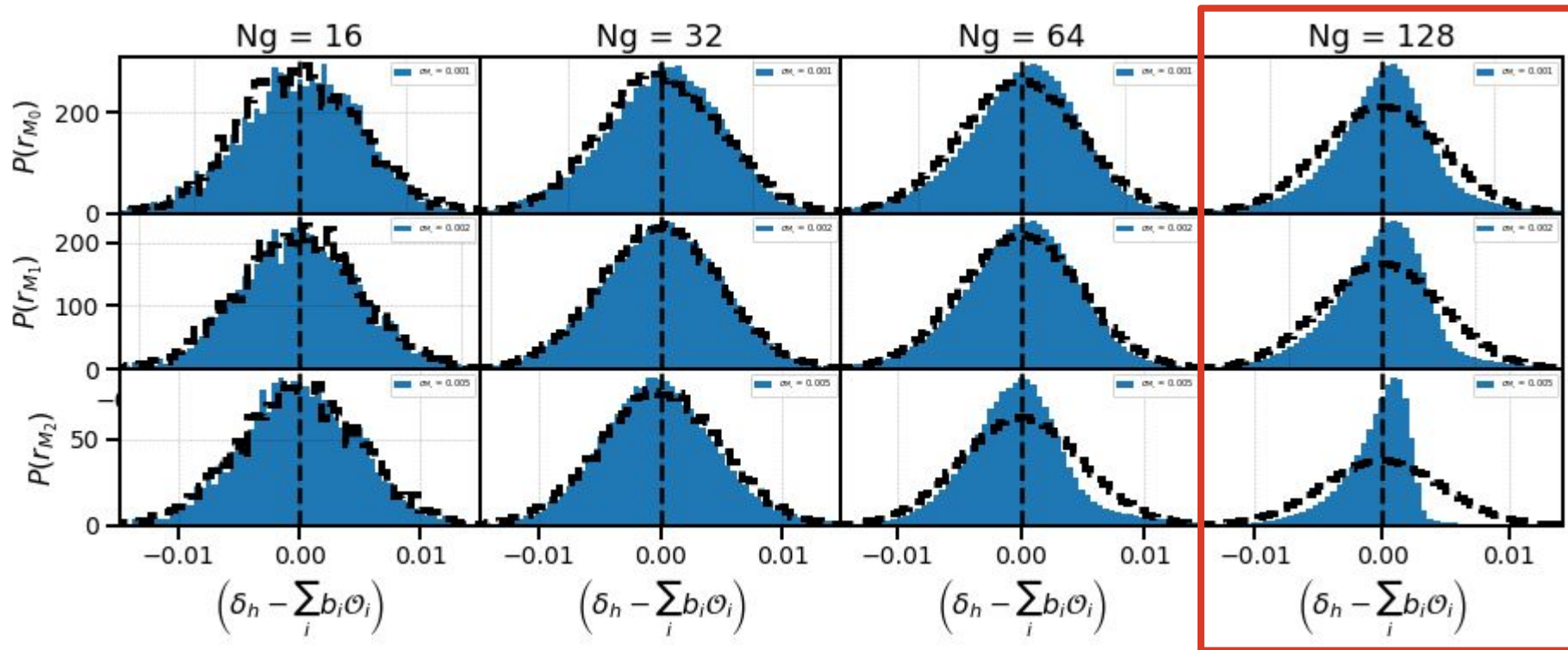
Adding PNG, much the same*

(*w/ renormalized operators)



Check with PDFs

Does this breakdown make sense? -> yes, **small-scale failure**



Inferring Local PNG

Easy mode - fixed phases

Inferring f_{NL} , marginalizing over from PT mocks?

Yes

Inferring f_{NL} from halos?

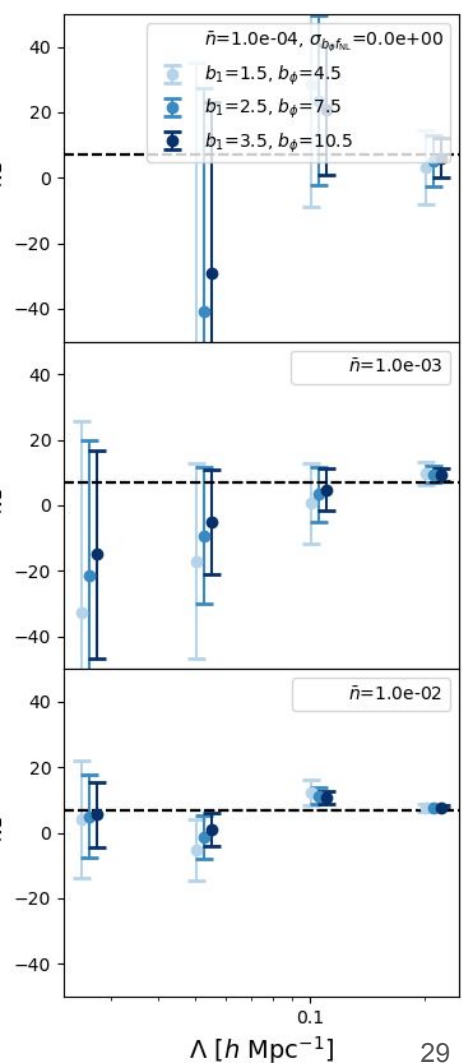
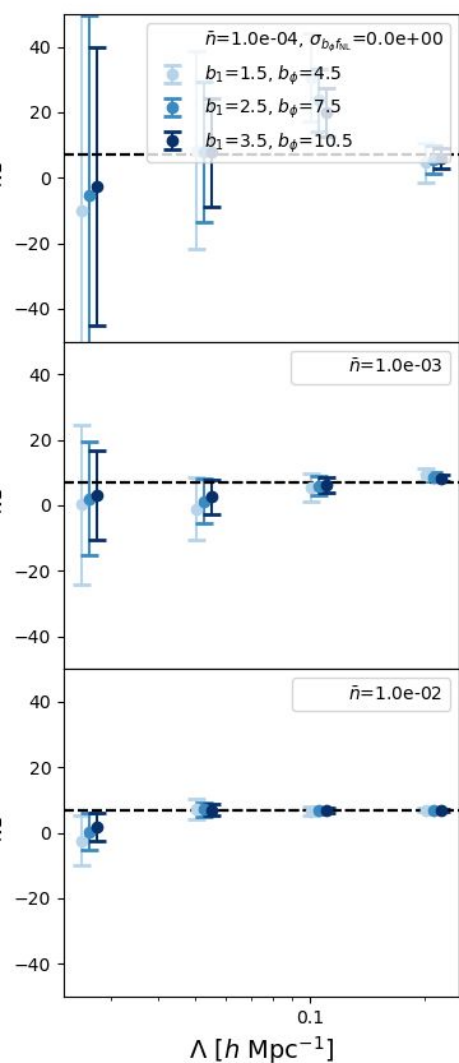
No...*

Inferring Local PNG

PT mock test

Bias-marginalized profile likelihood

2-3x degradation on $\sigma(f_{NL})$ w/
quadratic bias



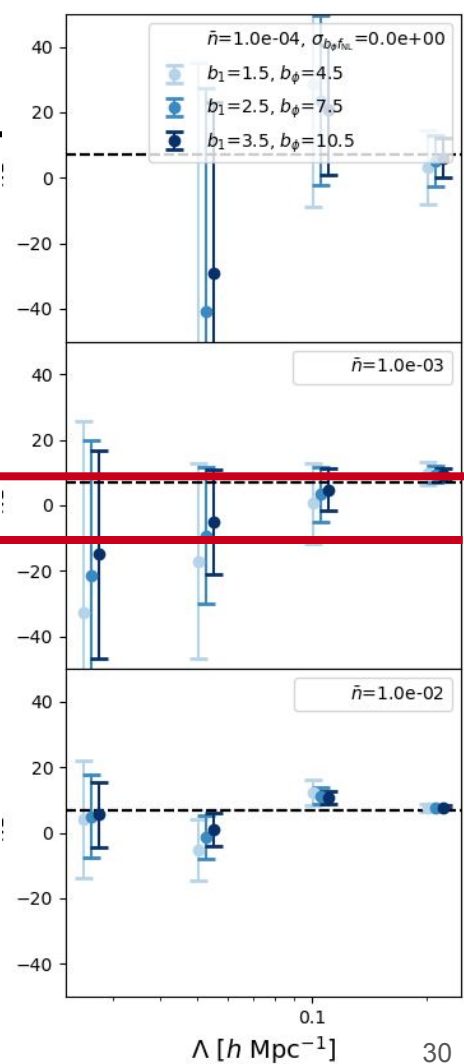
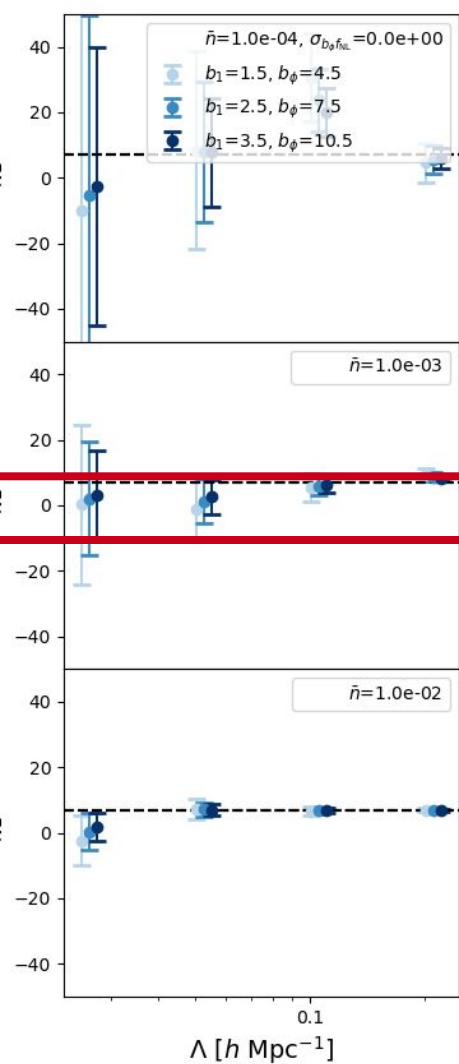
Inferring Local PNG

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Linear



Inferring Local PNG

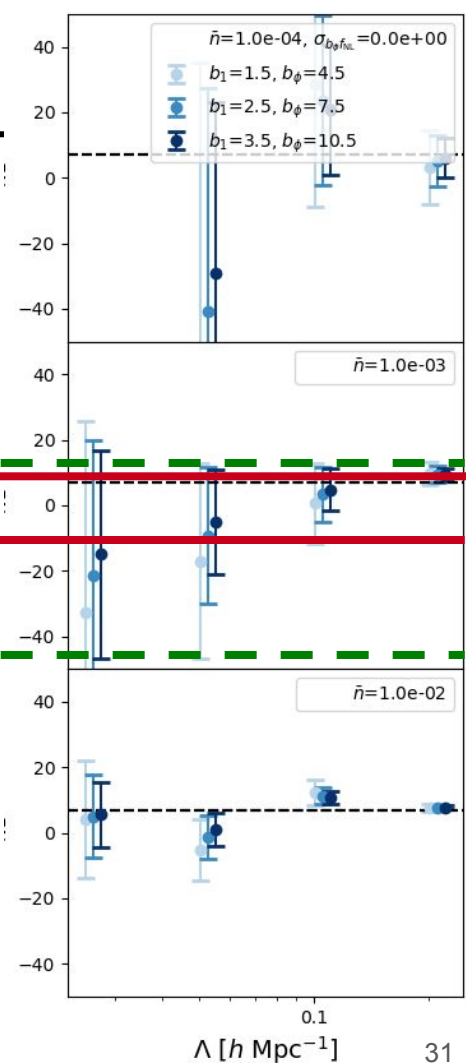
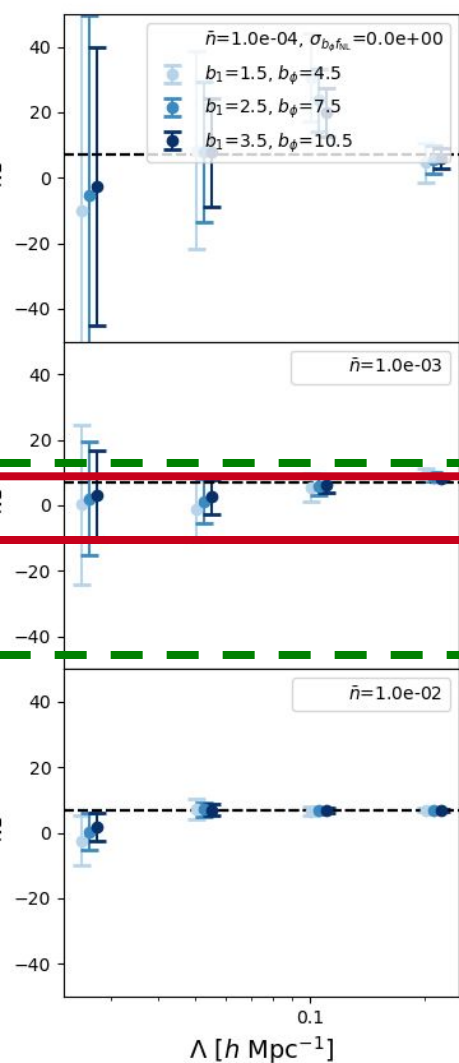
PT mock test

Bias-marginalized profile likelihood

~~2-3x degradation on $\sigma(f_{NL})$ w/ quadratic bias~~

Linear

Quadratic



Inferring Local PNG

PT mock test

Bias-marginalized profile likelihood

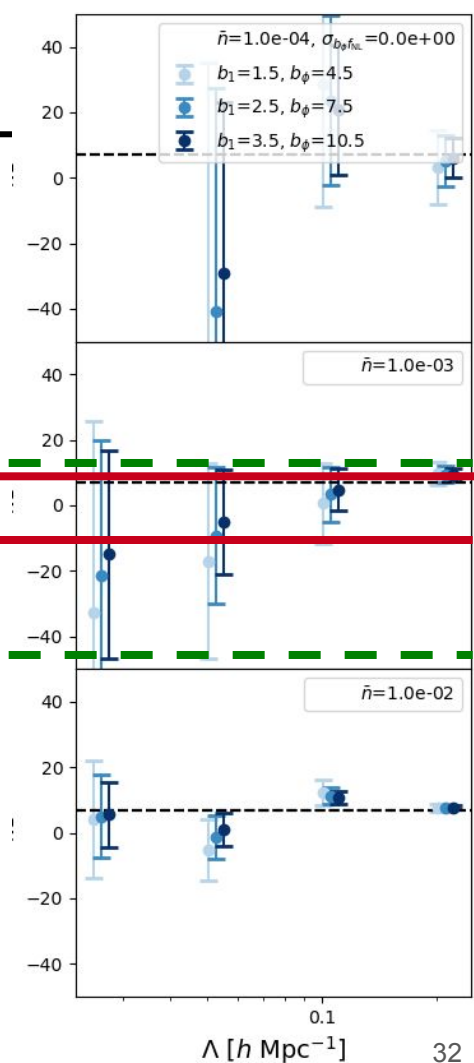
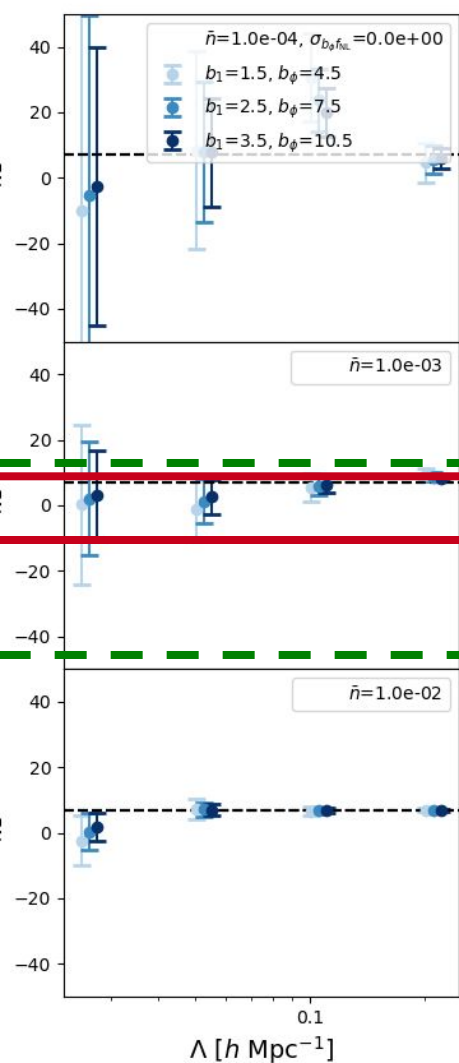
~~2-3x degradation on $\sigma(f_{NL})$ w/~~

quadratic bias

Linear

Quadratic

Even easy-mode pessimistic...



Bias from Time Evolution - Idea

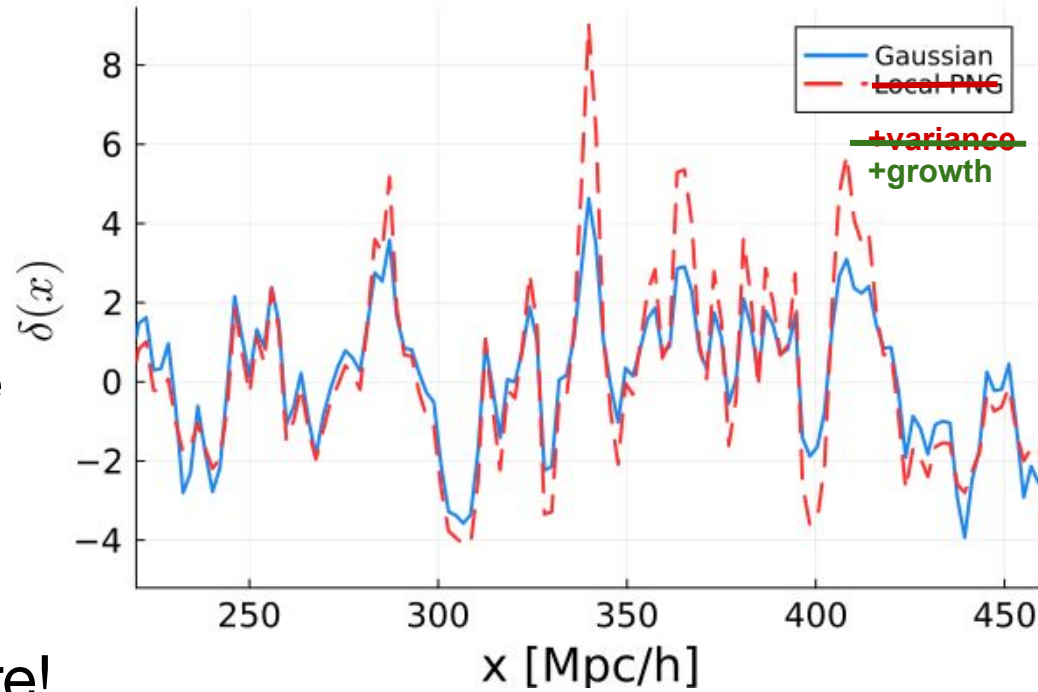
LPNG is “like” boosting
underlying variance

Bias from Time Evolution - Idea

LPNG is “like” boosting
underlying variance

Can measure LPNG
bias by running 2
simulations w/ diff variance

But boosting variance is
~equivalent to
boosting growth of structure!



Separate universe with 1 universe? - UMF

Universality of mass function a decent first approximation

Peak-background split relates bias to peak height response

$$\sigma^2(M, z) = \frac{1}{2\pi^2} \int k^2 dk |W(R(M, z)k)|^2 D^2(z) P_L(k)$$

$$n(M) = n(M, \nu) = M^{-2} \nu f(\nu) \frac{d \ln \nu}{d \ln M} \quad \nu = \delta_c^2 / \sigma^2(M)$$

Growth and change in variance perfectly degenerate via variance

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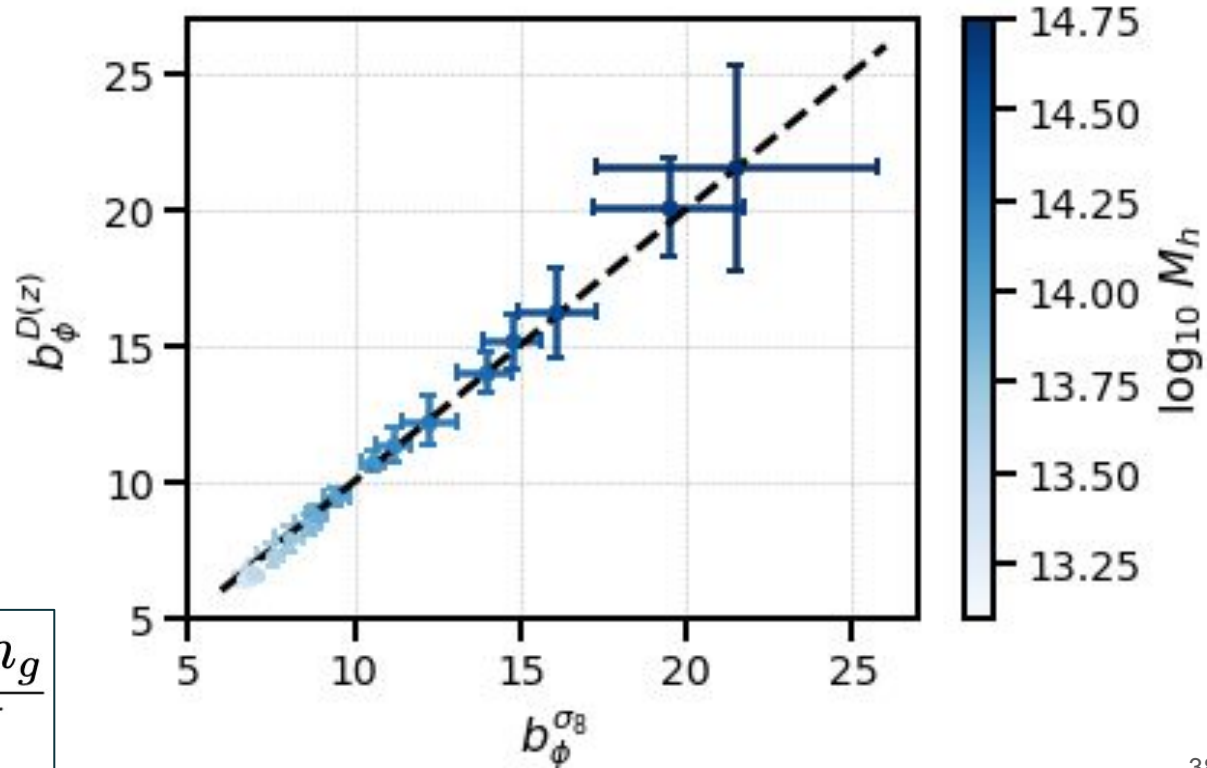
Bias from Time Evolution - Simulated Halos

N-body halos at $z = 1$

Evaluate bias via
finite difference
response to:

1. variance (σ_8)
2. growth

$$b_{\phi}^{X=\{\sigma_8, D(z)\}} = 2 \frac{d \log n_g}{d \delta X}$$



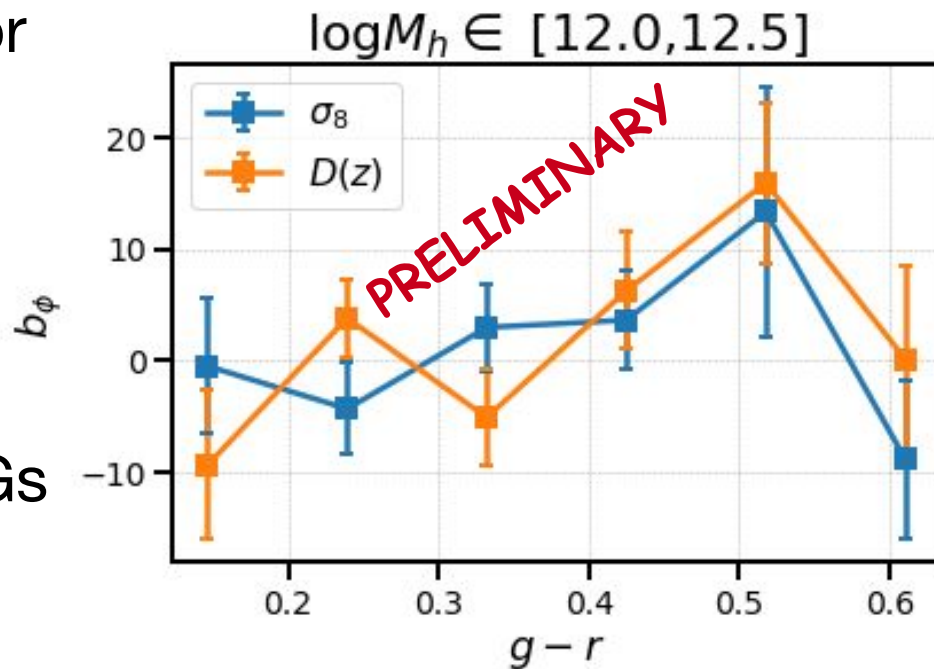
Bias from Time Evolution - Hydro

Can do the same w/ LPNG
assembly bias - here w/ color

Holds roughly across mass

Hydro sims out there
are limited

Now-> looking at BOSS LRGs
(selection function)



LPNG & Assembly Bias

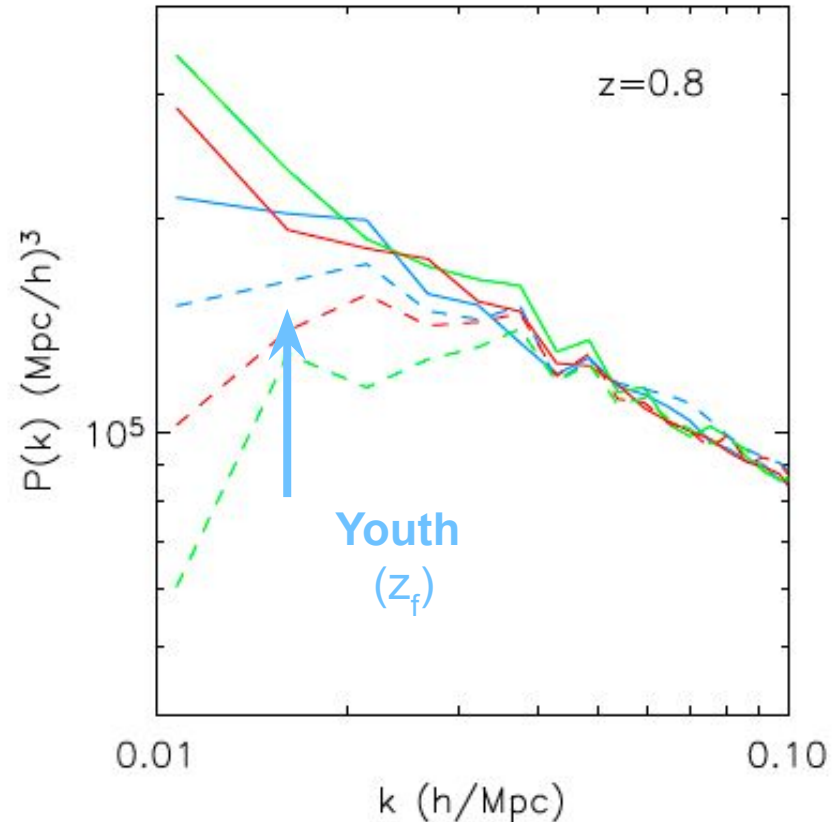
Halo assembly bias:

Slosar 08 -> “merger”

Reid 10 -> formation time

...

Lazeyras 22 -> concentration
(also spin, axis ratio)

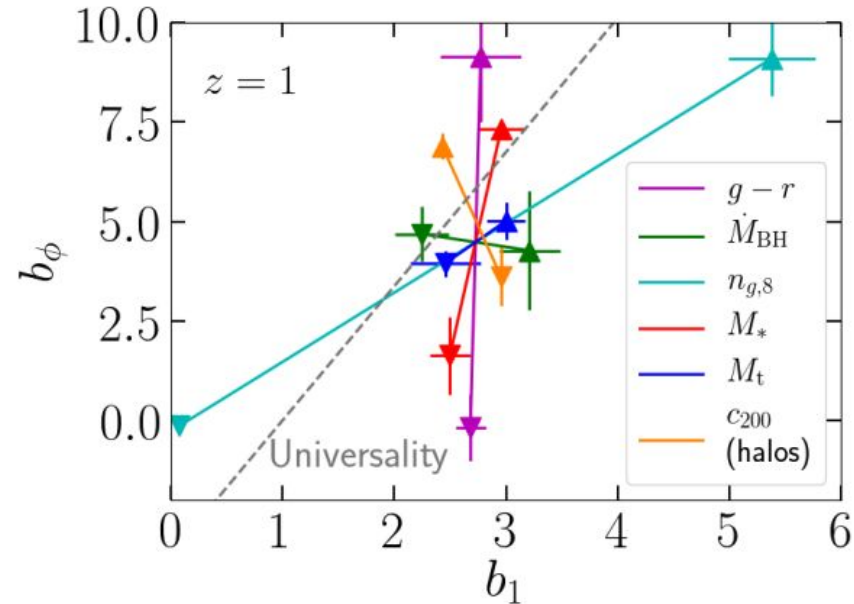
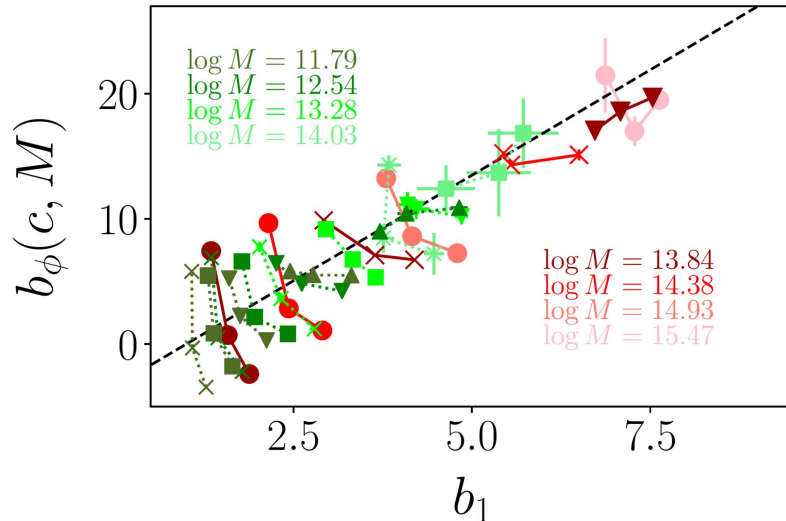


LPNG & Assembly Bias

Halo concentration c has a large effect (see 2303.08901)

Especially at low mass, enormous variation

Galaxy color is even stronger! **Why?!**



Final Thoughts

Unanswered questions:

- Information content of field-level vs n -point for PNG generally?
- What more to learn about b_{ϕ} assembly bias?
 - > Physical understanding even for halos
- Is time-evolution bias applicable to other samples?
(Working on LRGs now)
 - > Deeper understanding of galaxy populations

Final Thoughts

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- Information content of field-level vs n -point for PNG generally?
- **V** **Lunch!**
 - > Physical understanding even for halos
- Is time-evolution bias applicable to other samples?
(Working on LRGs now)
 - > Deeper understanding of galaxy populations