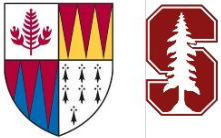
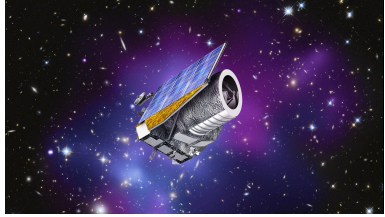


# Comparing Galaxy Formation Models using the Bias Expansion

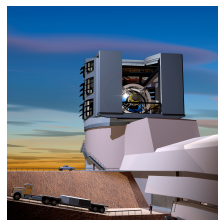
Mahlet Shiferaw

Advisors: Risa Wechsler, Nickolas Kokron

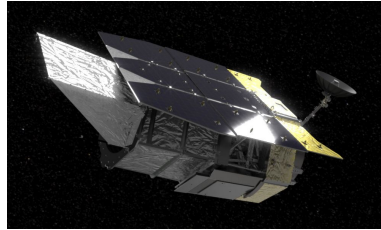




Euclid



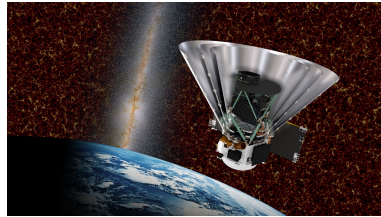
Spherex



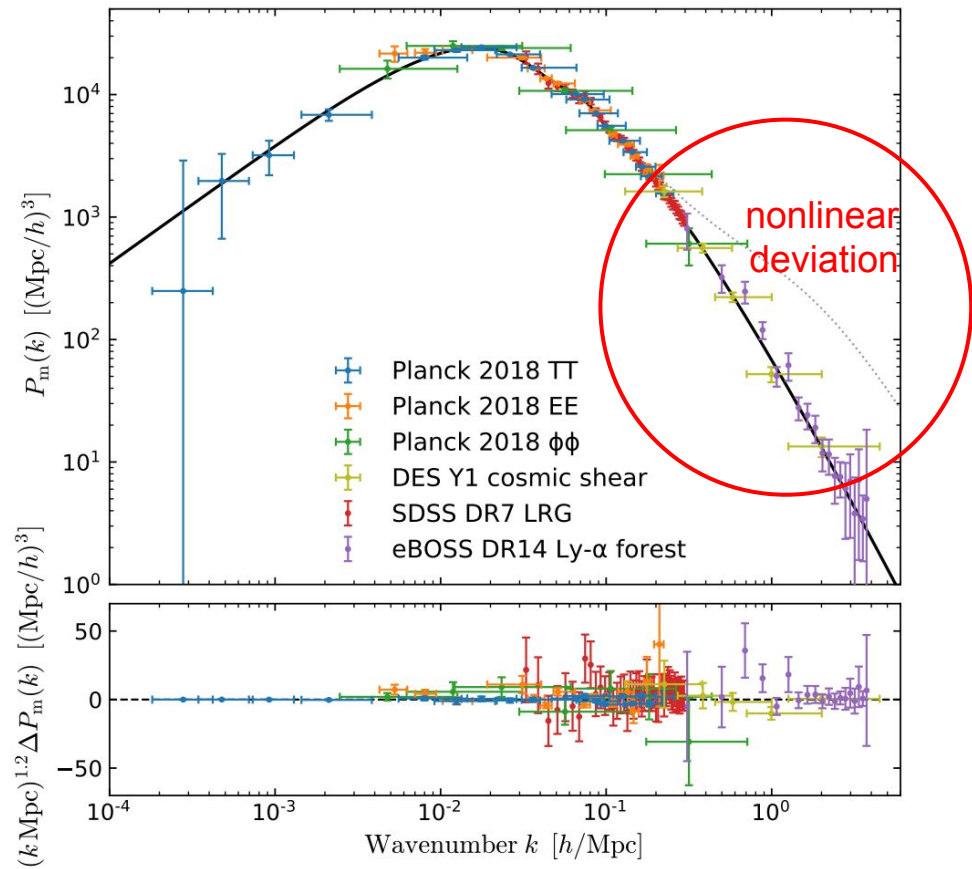
Roman Space Telescope



DESI



Rubin/LSST



We will have a lot of data! (especially at small scales)

# Galaxy Bias

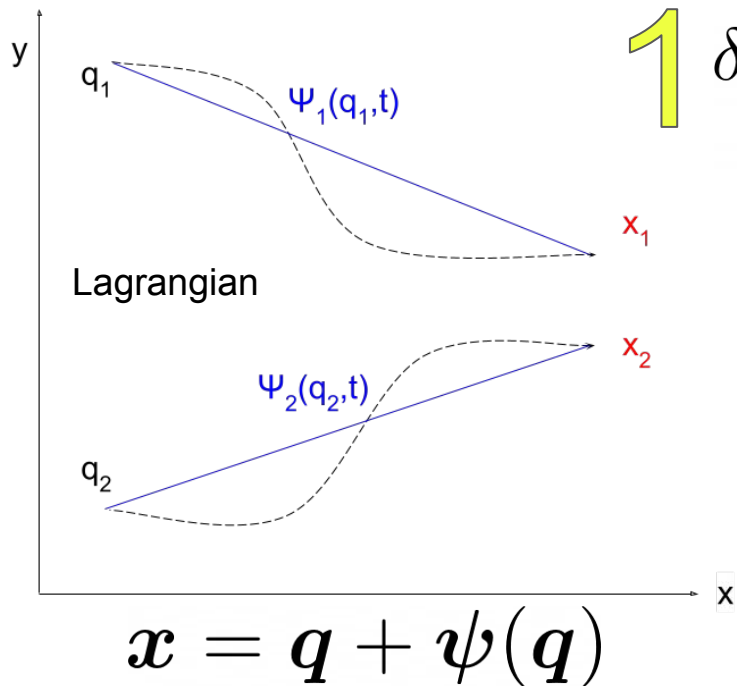
$$1 \quad \delta_g(\mathbf{x}, \tau) = \sum_O \underline{b_O(\tau)} O(\mathbf{x}, \tau)$$

Perturbation Theory

$$1 \quad \delta_g(\mathbf{x}, \tau) = \sum_O \underline{b_O(\tau)} O(\mathbf{x}, \tau)$$

Perturbation Theory

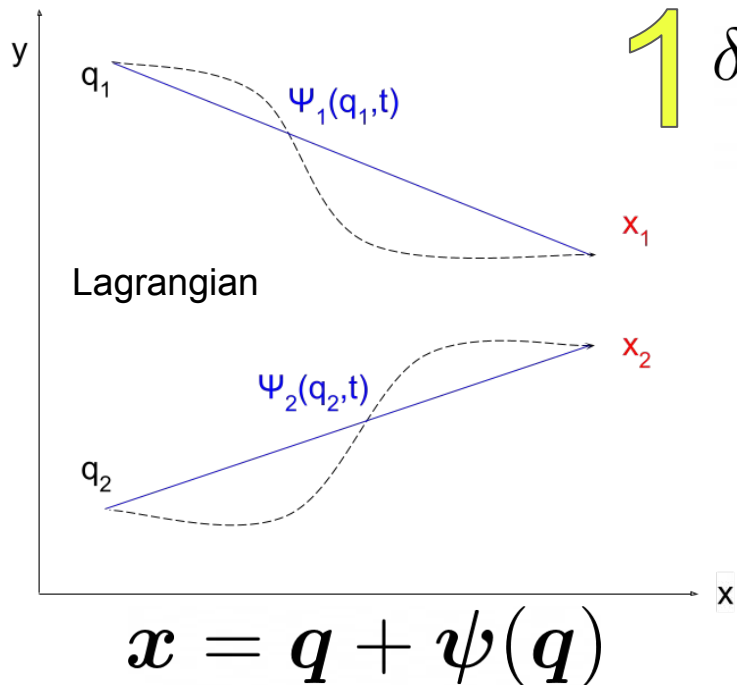
2



$$1 \quad \delta_g(\mathbf{x}, \tau) = \sum_O \underline{b_O(\tau)} O(\mathbf{x}, \tau)$$

Perturbation Theory

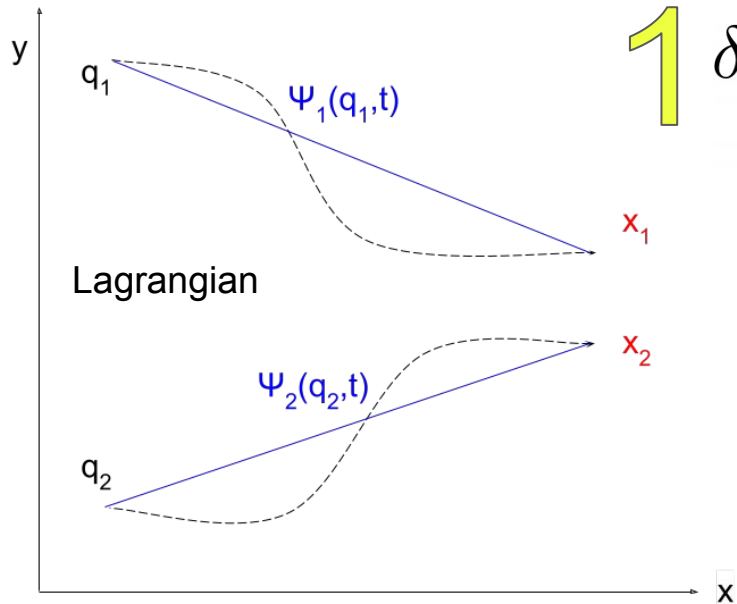
2



Hybrid Effective Field Theory (HEFT)

3

$$\delta_g(\mathbf{x}) = \delta_m(\mathbf{x}) + \underline{b_1} \mathcal{O}_\delta(\mathbf{x}) + \underline{b_{\nabla^2}} \mathcal{O}_{\nabla^2 \delta}(\mathbf{x}) + \underline{b_2} \mathcal{O}_{\delta^2}(\mathbf{x}) + \underline{b_{s^2}} \mathcal{O}_{s^2}(\mathbf{x}) + \epsilon(\mathbf{x})$$



$$\mathbf{x} = \mathbf{q} + \boldsymbol{\psi}(\mathbf{q})$$

$$1 \quad \delta_g(\mathbf{x}, \tau) = \sum_O \underline{b_O(\tau)} O(\mathbf{x}, \tau)$$

Perturbation Theory

$$4 \quad S = \langle [\epsilon(\mathbf{x})]_{k_{\max}}^2 \rangle$$

Loss function

Hybrid Effective Field Theory (HEFT)

3

$$\delta_g(\mathbf{x}) = \delta_m(\mathbf{x}) + \underline{b_1} \mathcal{O}_\delta(\mathbf{x}) + \underline{b_{\nabla^2}} \mathcal{O}_{\nabla^2} \delta(\mathbf{x}) + \underline{b_2} \mathcal{O}_{\delta^2}(\mathbf{x}) + \underline{b_{s^2}} \mathcal{O}_{s^2}(\mathbf{x}) + \epsilon(\mathbf{x})$$

### Approaches to modeling the galaxy-halo connection

← physical models		empirical models →		
Hydrodynamical Simulations	Semi-analytic Models	Empirical Forward Modeling	Subhalo Abundance Modeling	Halo Occupation Models
Simulate halos & gas; Star formation & feedback recipes	Evolution of density peaks plus recipes for gas cooling, star formation, feedback	Evolution of density peaks plus parameterized star formation rates	Density peaks (halos & subhalos) plus assumptions about galaxy—(sub)halo connection	Collapsed objects (halos) plus model for distribution of galaxy number given host halo properties



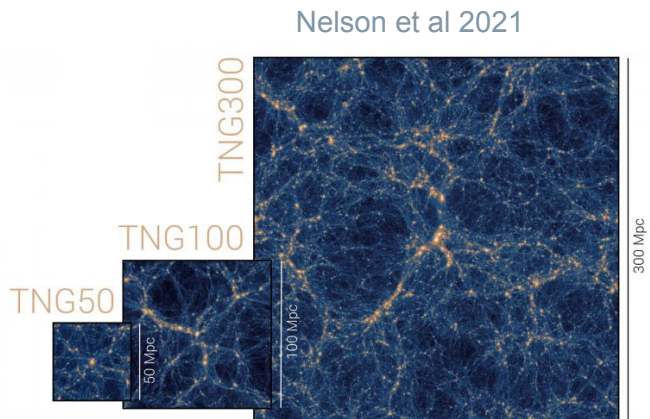
### Approaches to modeling the galaxy-halo connection

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**Our Approach:** We examine quenched and star-forming galaxy samples from models with disparate approaches using a perturbative bias expansion.

1. Enables a direct comparison between distinct modeling approaches
2. See how uncertainties in galaxy-halo connection impact bias parameters
3. Allows us to inform priors on these bias parameters for future surveys

**Initial Conditions:**  $2500^3$  dark matter particles,  $L_{\text{box}} = 205 \text{ Mpc}/h$   
 $z=0$  (for now)



**IllustrisTNG Project**

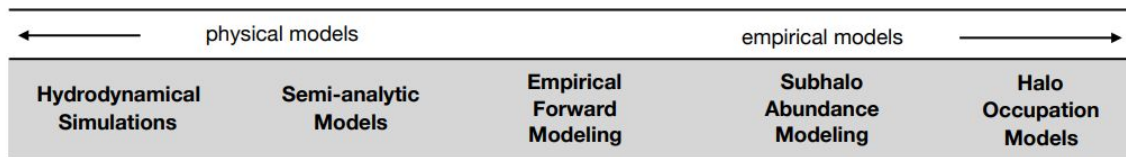
Full, hydrodynamical (more physical) model



**UniverseMachine (UM)**

Semi-analytic (more empirical) model

**Models**



# Preliminary Results



# Bias as a function of $k_{\text{max}}$ :

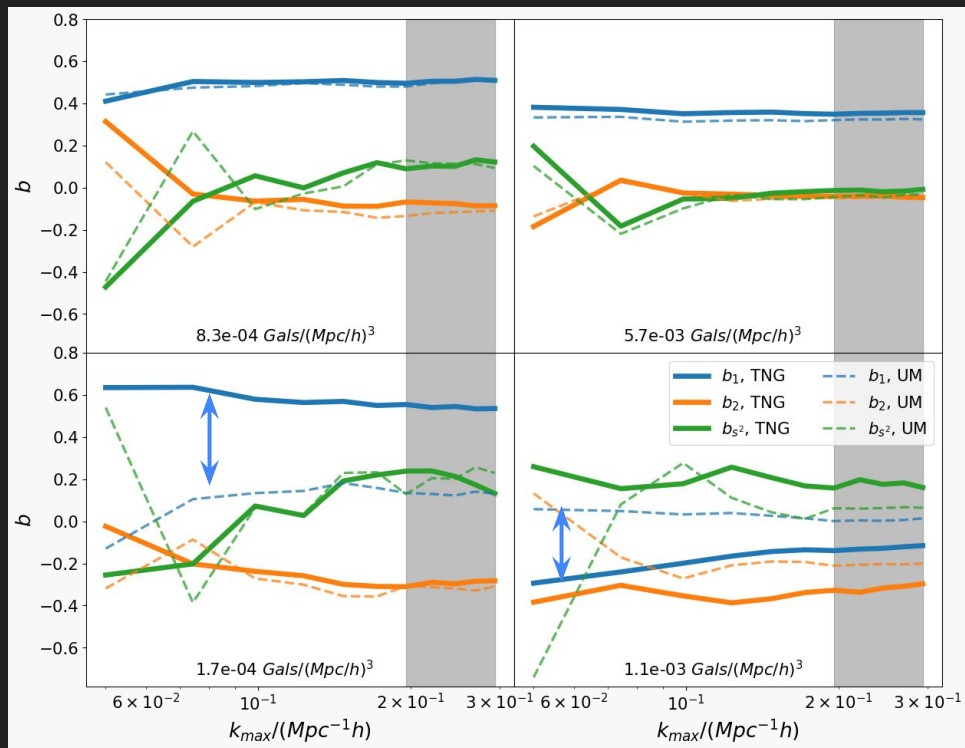
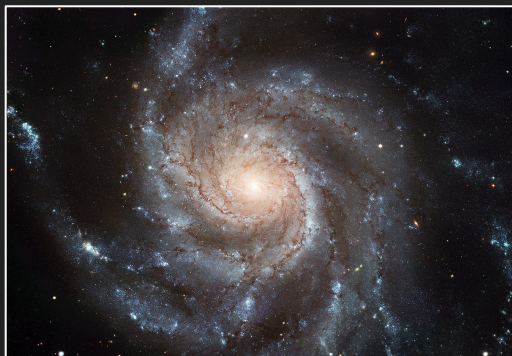
Low density,  
High stellar mass

High density,  
Low stellar mass

Quenched



Star-forming



Bias parameters have significant disagreement between UM & TNG in star-forming samples!

# Conclusions

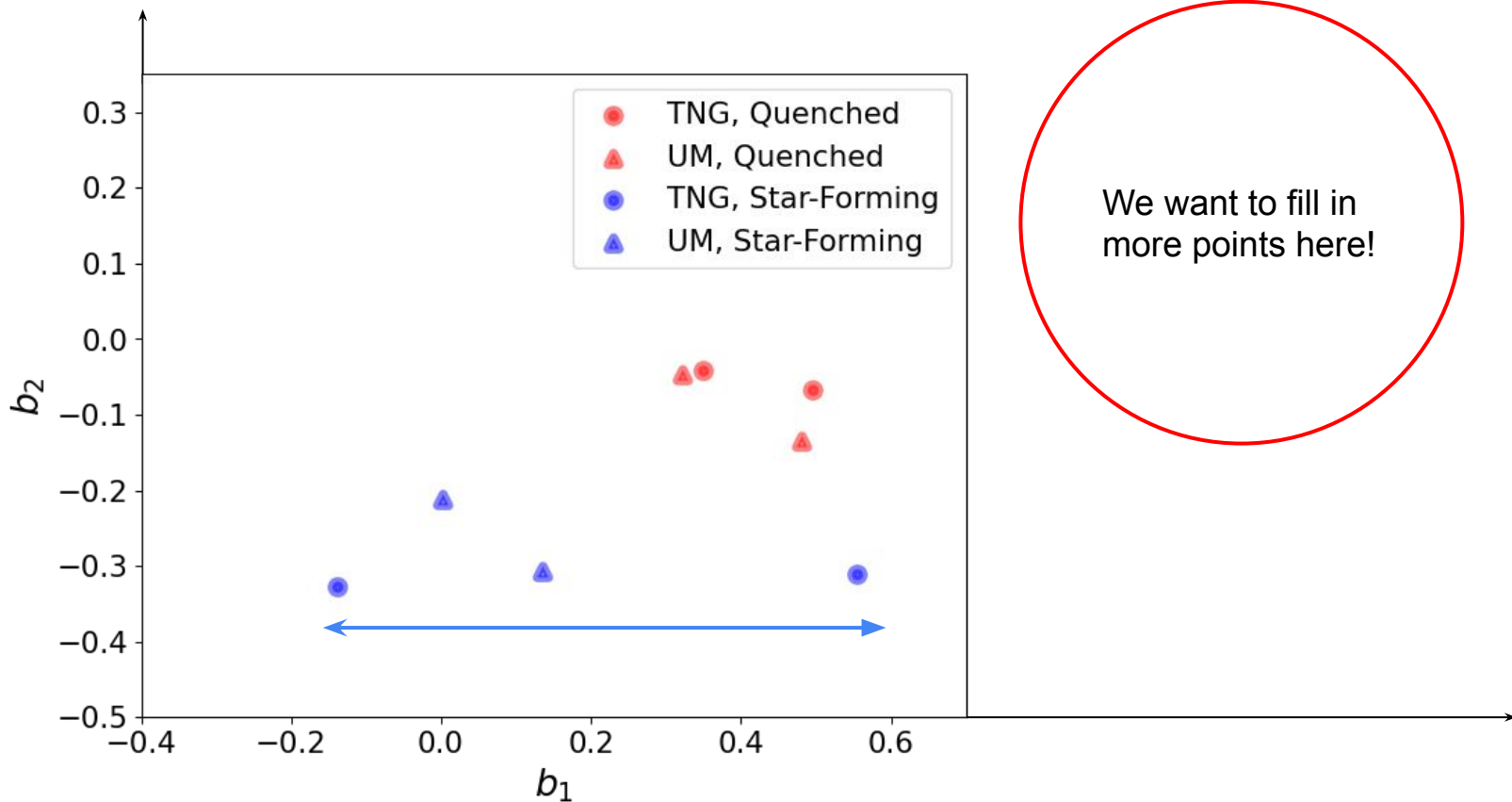
1. Using the language of bias, we have successfully created a framework for a **1:1 comparison between models** with disparate approaches
2. The linear bias relation between the galaxy and matter distributions **agree for for UM and TNG in quenched samples**, but **differ in star-forming samples**
3. This suggests that we need a **wider prior for star-forming galaxies**, but can put **tighter constraints on quenched galaxies**

Next steps:

- Fill in parameter space at higher  $z$  (0.5-1.5)
- Fill in priors with intermediate number densities



# A closer look at $k_{\text{max}} = 0.2 \text{ h/Mpc}$





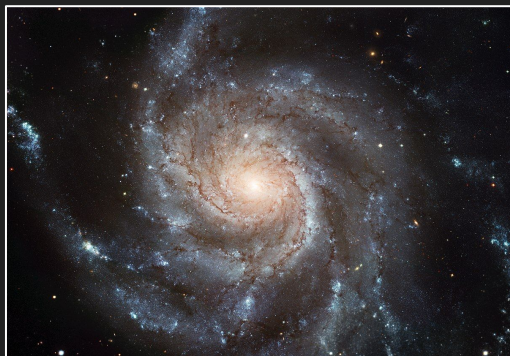
# Galaxy Samples

**Quenched**



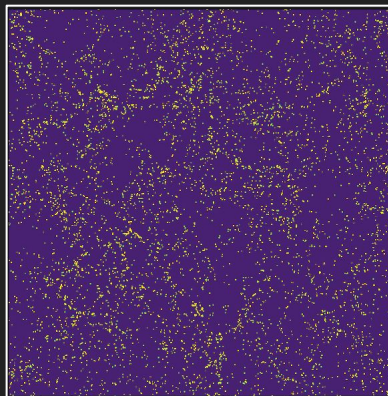
NASA, ESA, R.M. Crockett et al

**Star-forming**

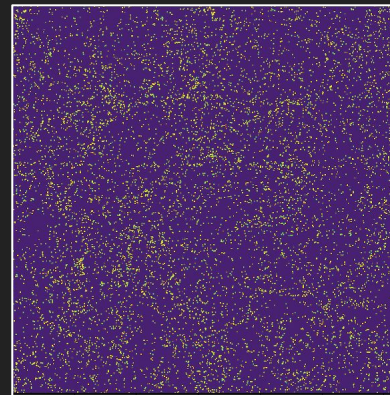
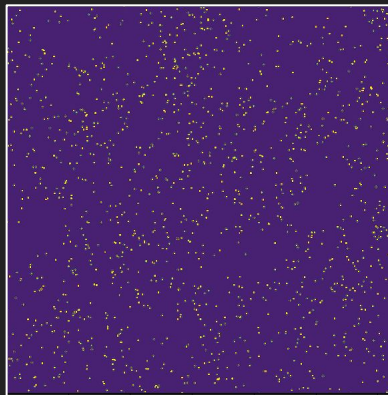
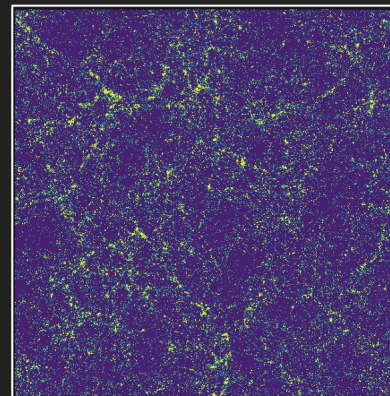


Credit: ESA/Hubble

**Low density,  
High stellar mass**



**High density,  
Low stellar mass**



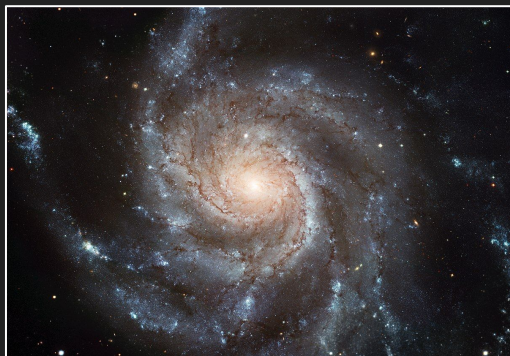
# Galaxy Samples

**Quenched**



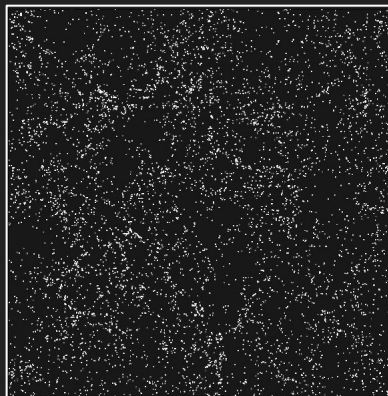
NASA, ESA, R.M. Crockett et al

**Star-forming**

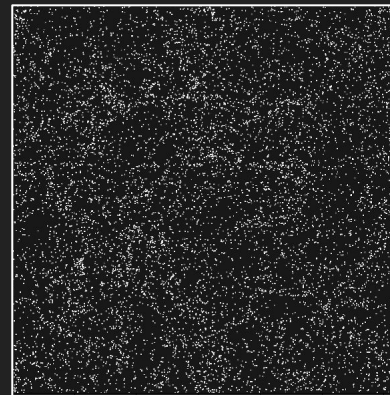
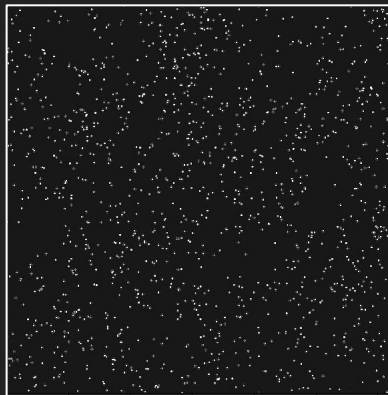
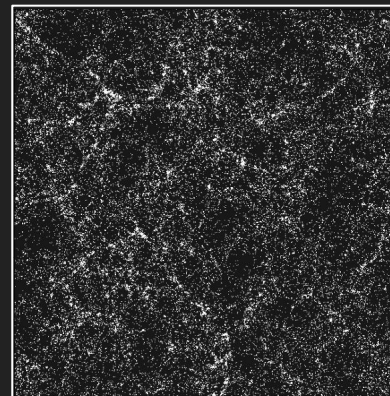


Credit: ESA/Hubble

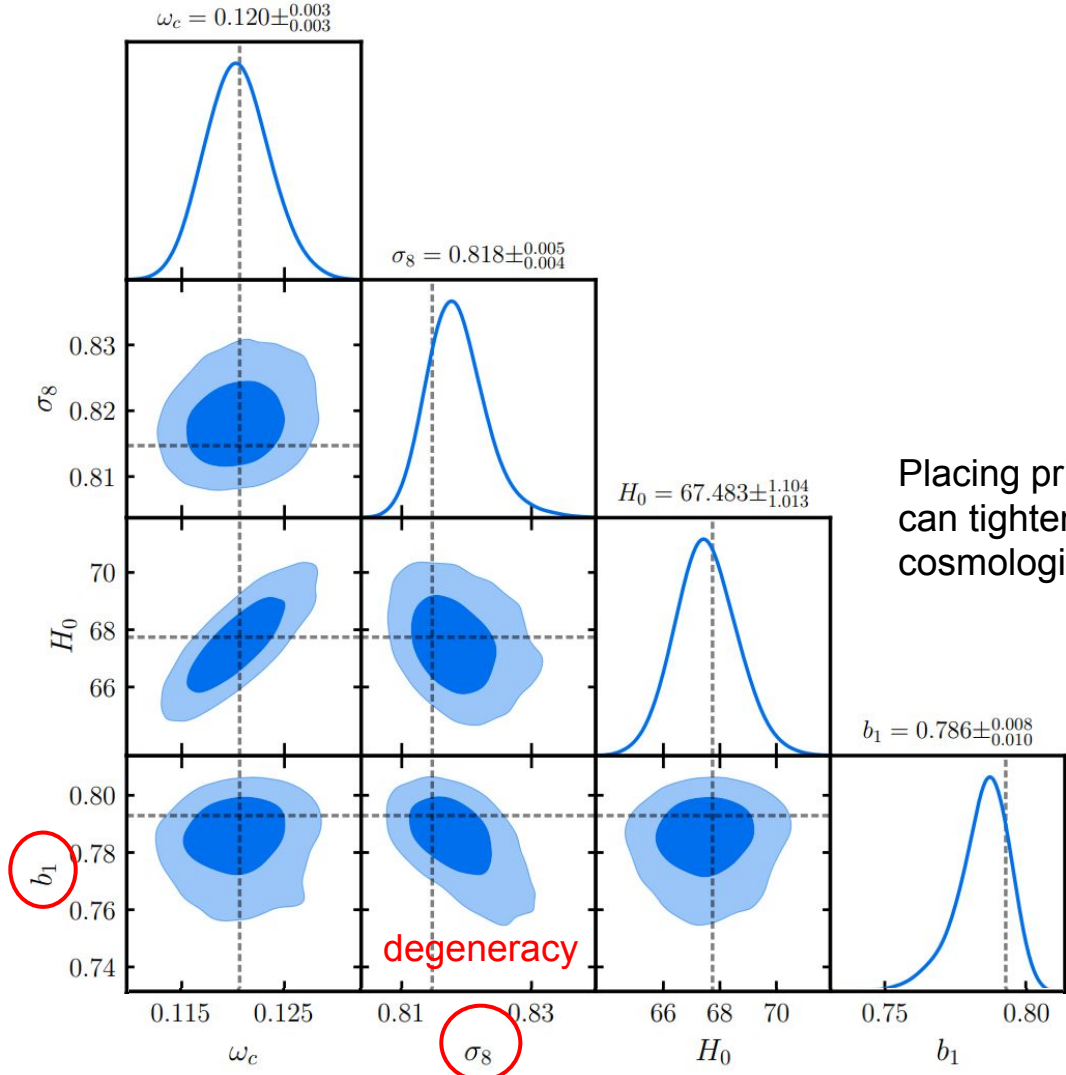
**Low density,  
High stellar mass**



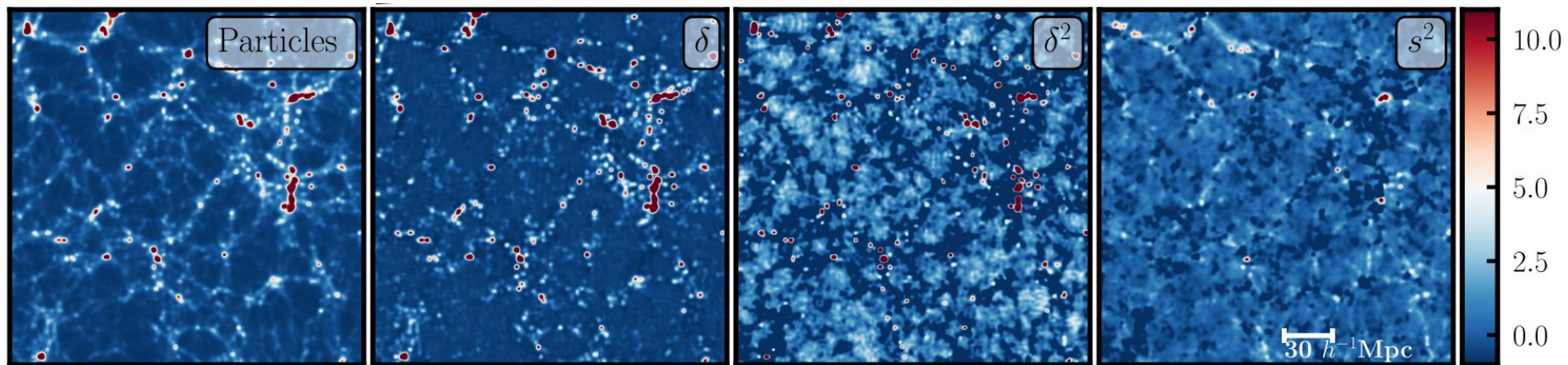
**High density,  
Low stellar mass**

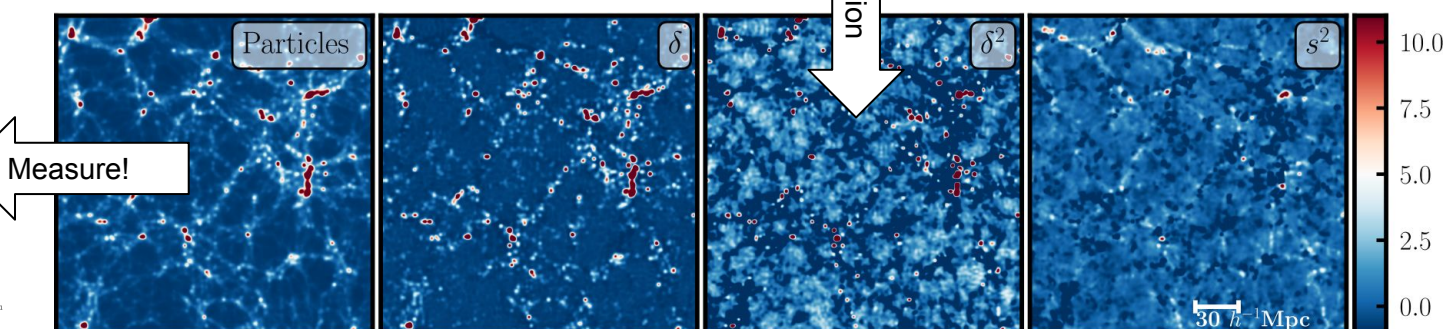
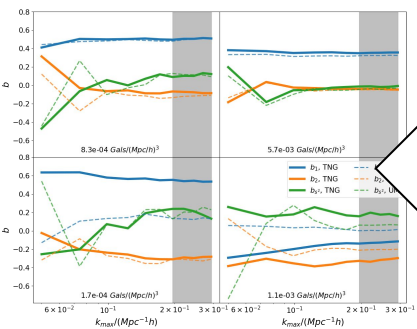
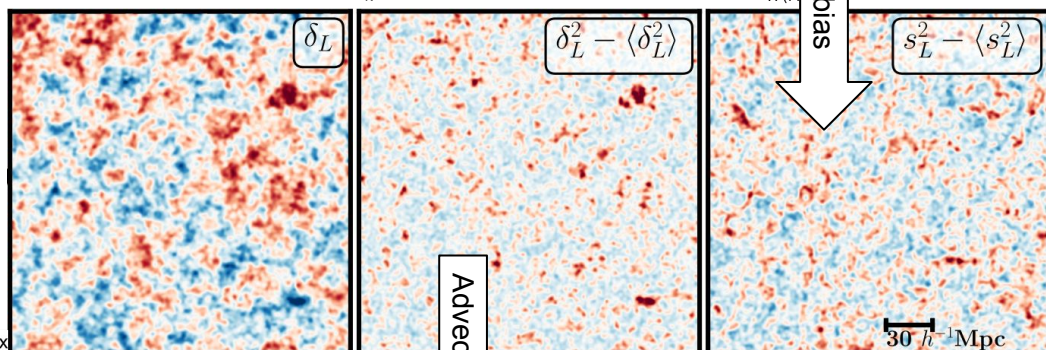
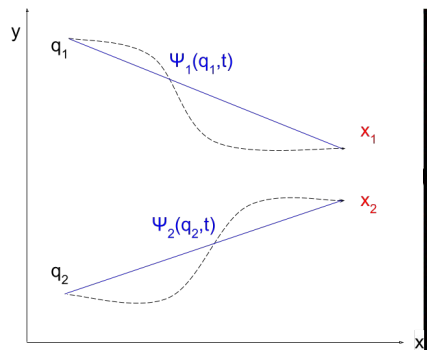
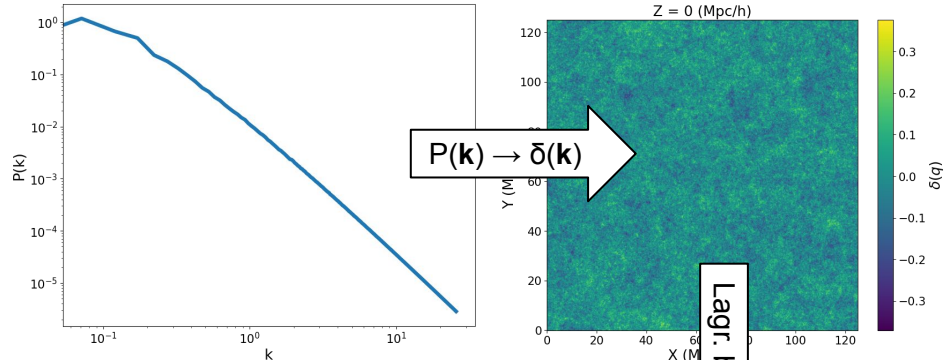




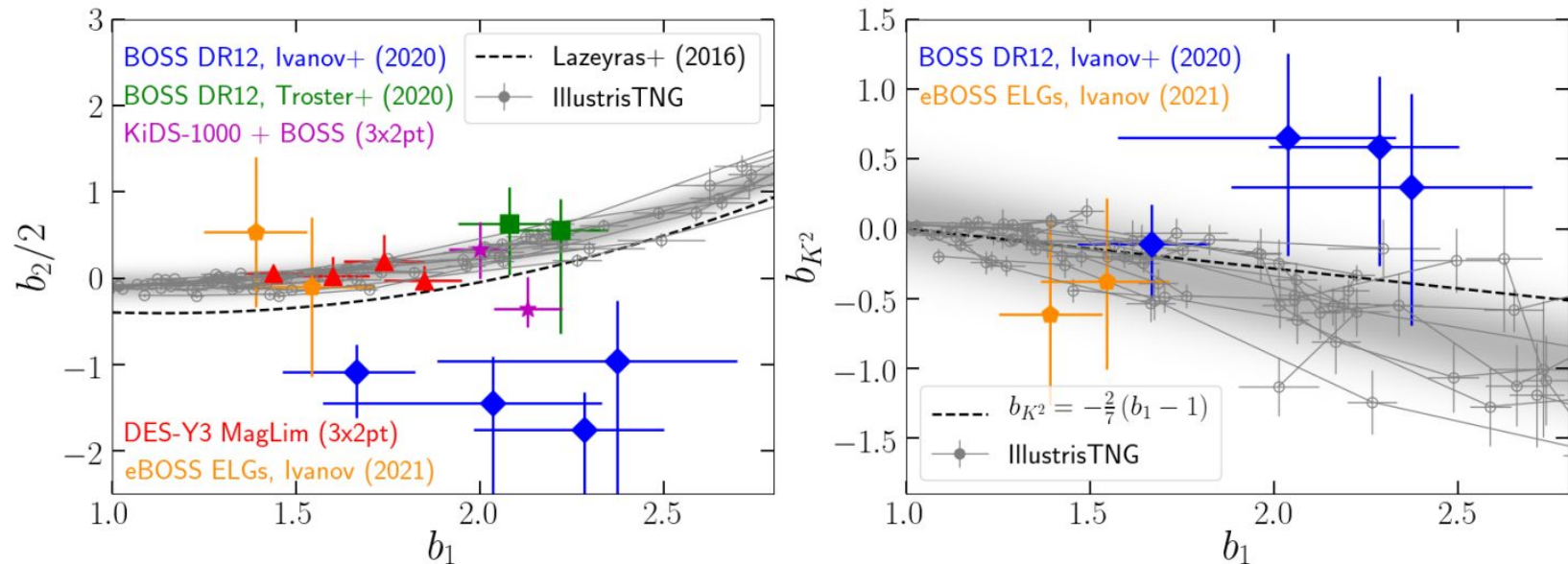


Placing priors on bias parameters can tighten constraints on cosmological parameters!









**Figure 7.** Comparison between the  $b_2(b_1)$  and  $b_{K^2}(b_1)$  relations obtained in this paper for IllustrisTNG galaxies and recent estimates from real galaxy data analyses. The grey data points show the result for all of the simulated galaxy samples in Figs. 4 and 5 put together, without distinguishing by redshift or selection criterion. The grey shaded map shows a Gaussian prior based on the IllustrisTNG data points (cf. Eqs. (3.8) and (3.9)). The colored data points show the relations inferred from recent cosmological inference analyses of observed galaxies: blue diamonds from Ref. [86], green squares from Ref. [87], magenta stars from Ref. [88], red triangles from Ref. [90], and orange pentagons from Ref. [89]. All error bars are 68% confidence limits.

## Expand perturbatively...

1) To first order:  $\delta_g(\mathbf{x}) = \underline{b_1} \delta_m(\mathbf{x})$

2) Taylor expand:  $\delta_g(\mathbf{x}, \tau) = \sum_O \underline{b_O(\tau)} O(\mathbf{x}, \tau)$

3) Hybrid expansion:  $\mathbf{x} = \mathbf{q} + \boldsymbol{\psi}(\mathbf{q})$

$$\delta_g(\mathbf{x}) = \delta_m(\mathbf{x}) + \underline{b_1} \mathcal{O}_\delta(\mathbf{x}) + \underline{b_{\nabla^2}} \mathcal{O}_{\nabla^2} \delta(\mathbf{x}) + \underline{b_2} \mathcal{O}_{\delta^2}(\mathbf{x}) + \underline{b_{s^2}} \mathcal{O}_{s^2}(\mathbf{x}) + \epsilon(\mathbf{x})$$

# Measuring the bias parameters via maximum likelihood

1) Define the stochasticity field:

$$\epsilon(\mathbf{x}) = \delta_g(\mathbf{x}) - \left( \delta_m(\mathbf{x}) + \sum_O \underline{b_O}(\tau) O(\mathbf{x}, \tau) \right)$$

2) Take the Fourier Transform and define its mean/variance:

$$\begin{aligned} \langle \epsilon(\mathbf{k}) \rangle &= 0 \\ \langle \epsilon(\mathbf{k}) \epsilon(\mathbf{k}') \rangle &\equiv P_{\text{err}}(|\mathbf{k}|) \end{aligned}$$

3) Find the bias parameters that **minimize this loss function** (up to some  $k_{\text{max}}$ ):

$$S = \left\langle [\epsilon(\mathbf{x})]_{k_{\text{max}}}^2 \right\rangle$$

$$\begin{aligned}
P\left(\delta_h(\mathbf{k}) - \delta_{h,\text{det}}(\mathbf{k})\right) &= \int d\varepsilon_m(\mathbf{k}) P\left(\delta_h(\mathbf{k}) - \delta_{h,\text{det}}(\mathbf{k}) - b_1\varepsilon_m(\mathbf{k}), \varepsilon_m(\mathbf{k})\right) \\
&= (2\pi)^{-1/2} \left| P_{hh}^\varepsilon(k) + 2b_1 P_{hm}^\varepsilon(k) + b_1^2 P_{mm}^\varepsilon \right|^{-1/2} \\
&\quad \times \exp \left[ -\frac{1}{2} \frac{|\delta_h(\mathbf{k}) - \delta_{h,\text{det}}(\mathbf{k})|^2}{P_{hh}^\varepsilon(k) + 2b_1 P_{hm}^\varepsilon(k) + b_1^2 P_{mm}^\varepsilon} \right]. \tag{3.4}
\end{aligned}$$

$$\epsilon(\mathbf{x}) = \delta_g(\mathbf{x}) - (\delta_m(\mathbf{x}) + \sum_O b_O(\tau) O(\mathbf{x}, \tau))$$

$$S = \langle [\epsilon(\mathbf{x})]_{k_{\max}}^2 \rangle^O$$

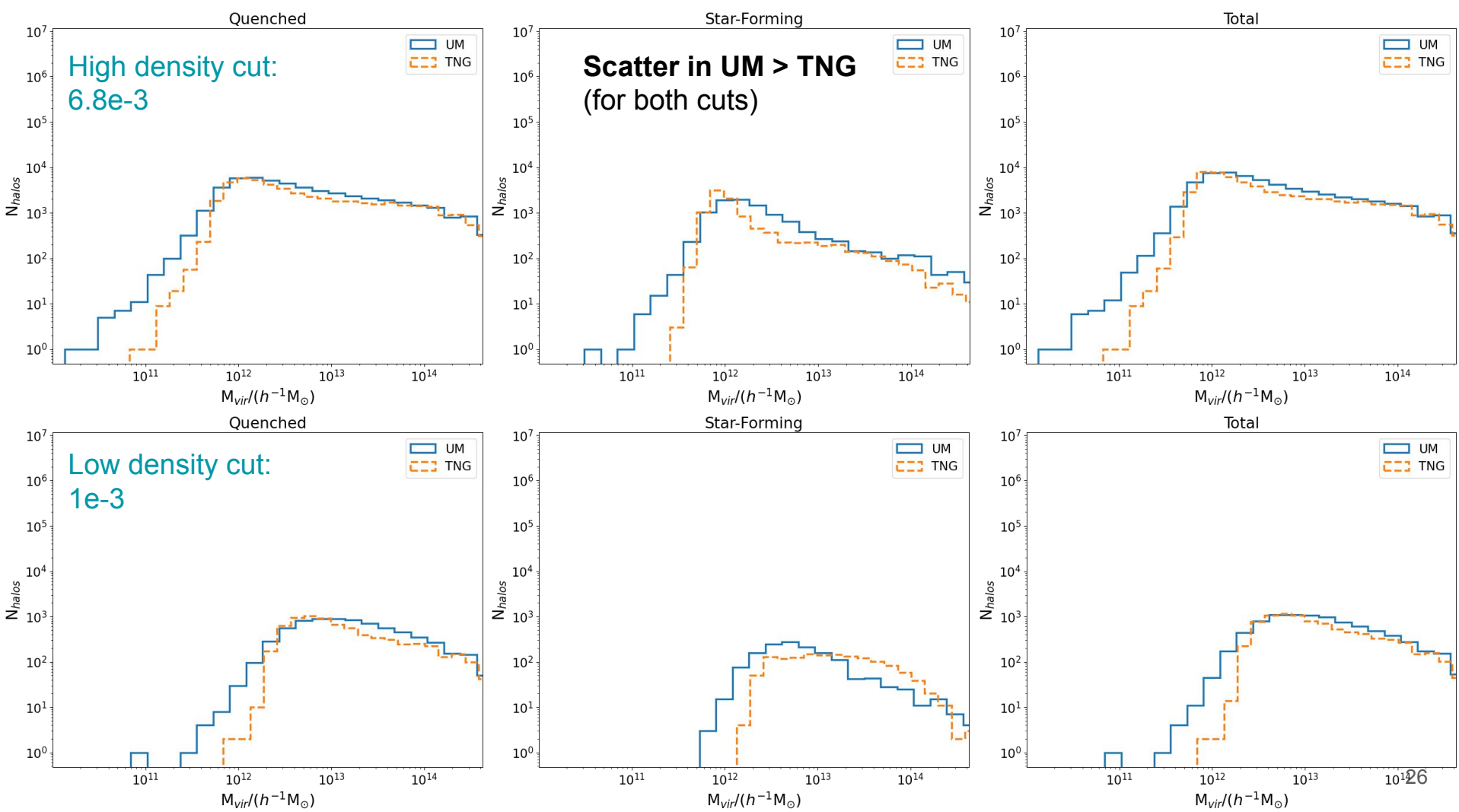
$$S \approx \int_{|\mathbf{k}| < k_{\max}} \frac{d^3 k}{(2\pi)^3} \left\| \delta_h(\mathbf{k}) - \delta_m(\mathbf{k}) - \sum_i b_i \mathcal{O}_i(\mathbf{k}) \right\|^2. \quad (14)$$



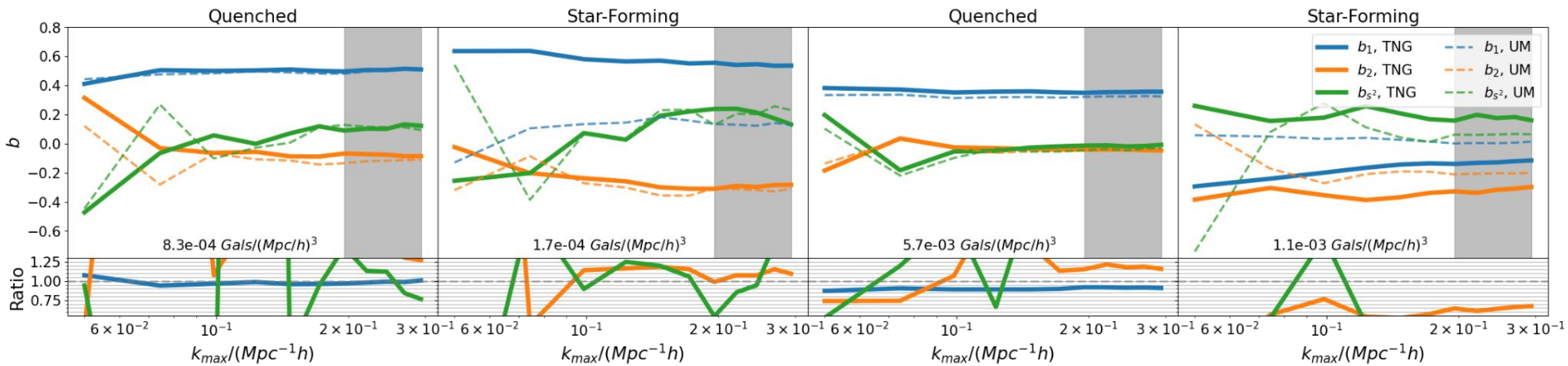
# Galaxy Samples

1. Set the total number density with a **stellar mass cut**
2. Split the galaxies into **quenched** and **star-forming** samples with a **specific star formation rate cut**

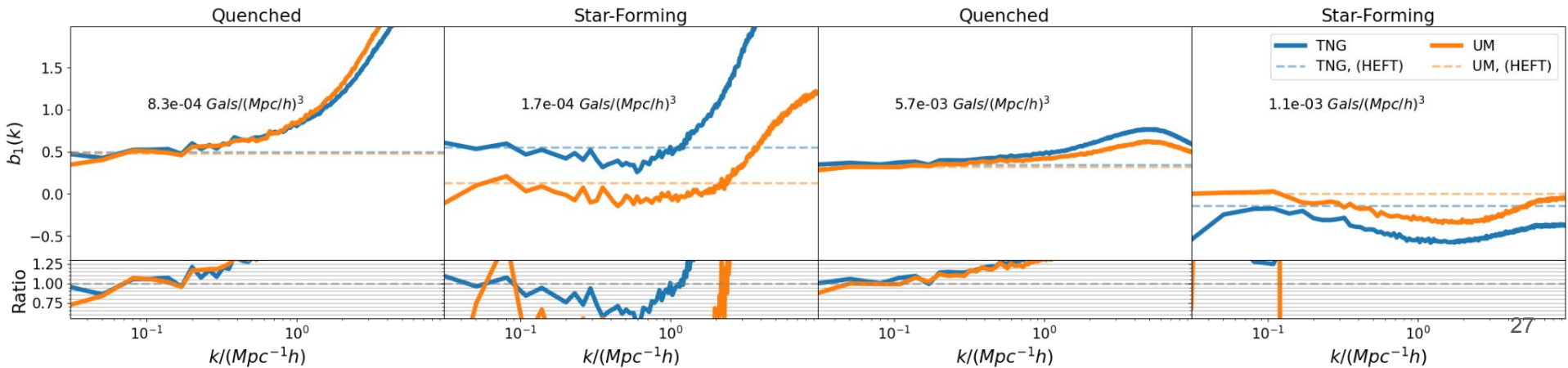
	Low density (Gals/(Mpc/h) <sup>3</sup> )	High density (Gals/(Mpc/h) <sup>3</sup> )
Quenched	8.3e-4	5.7e-3
Star-forming	1.7e-4	1.1e-3
Total	1e-3	6.8e-3



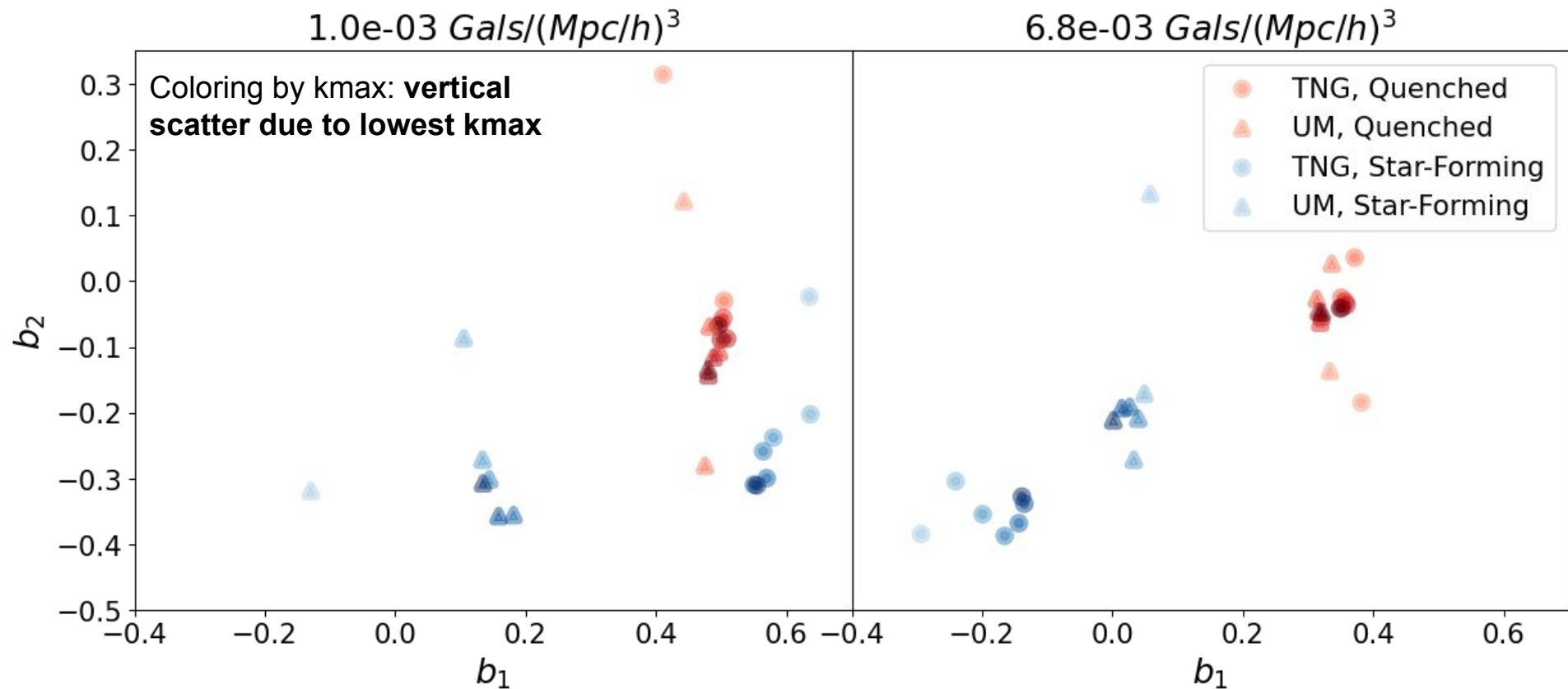
# Bias parameters as a function of $k_{\max}$ :



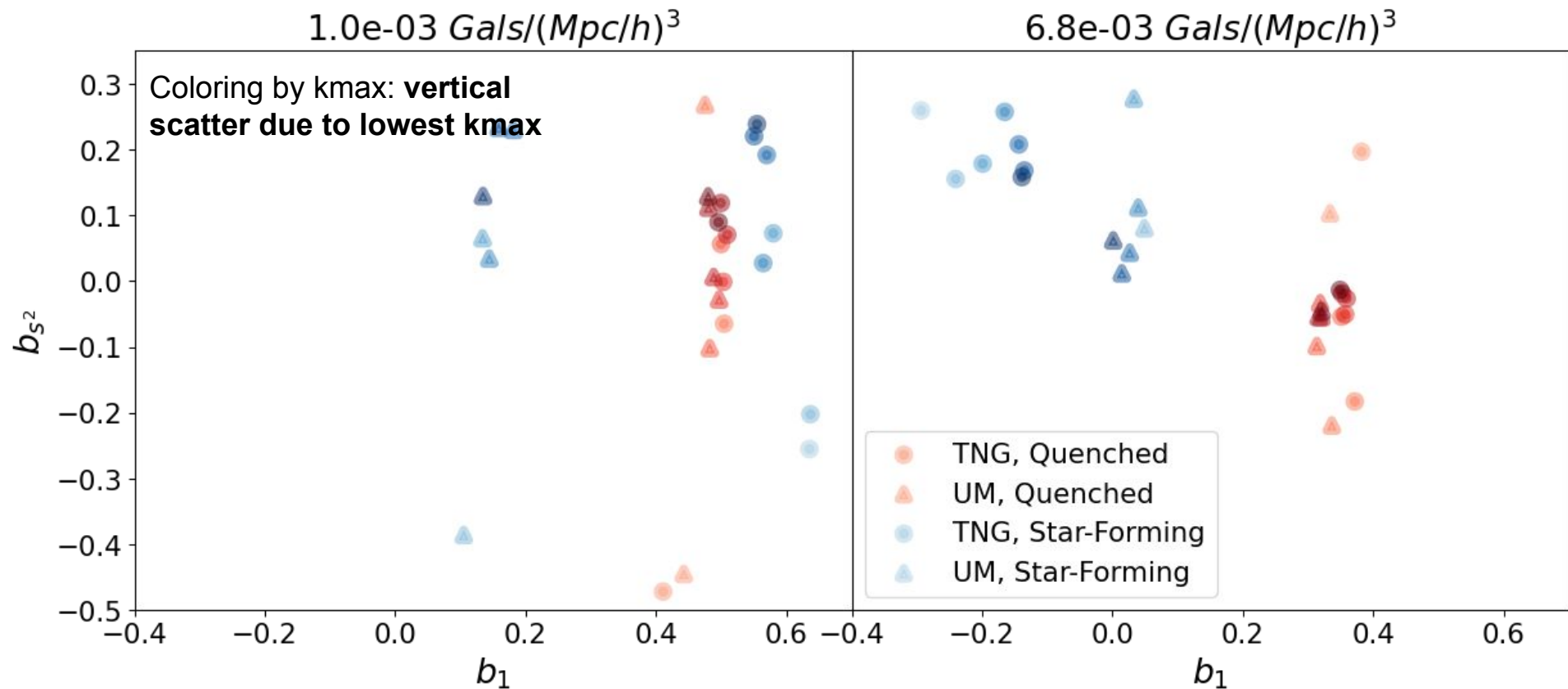
Sanity check with linear theory at large scales:  $\langle \delta_g(\mathbf{k})\delta_m(\mathbf{k}') \rangle = b_1 \langle \delta_m(\mathbf{k})\delta_m(\mathbf{k}') \rangle$



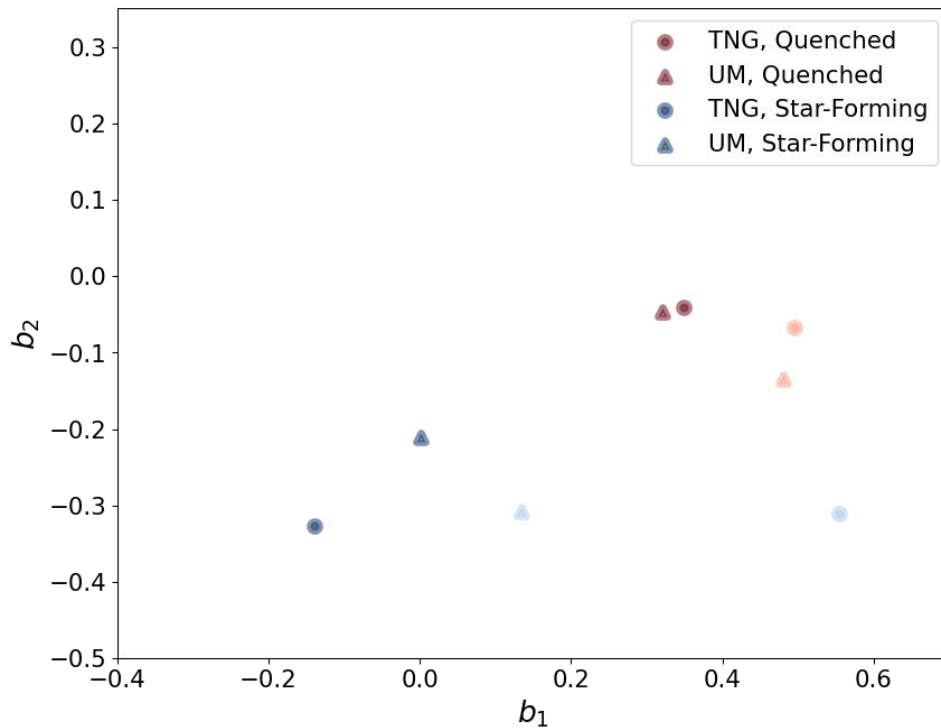
# b1 vs b2 in large scale regime ( $k_{\text{max}} < 0.2 \text{ h/Mpc}$ )



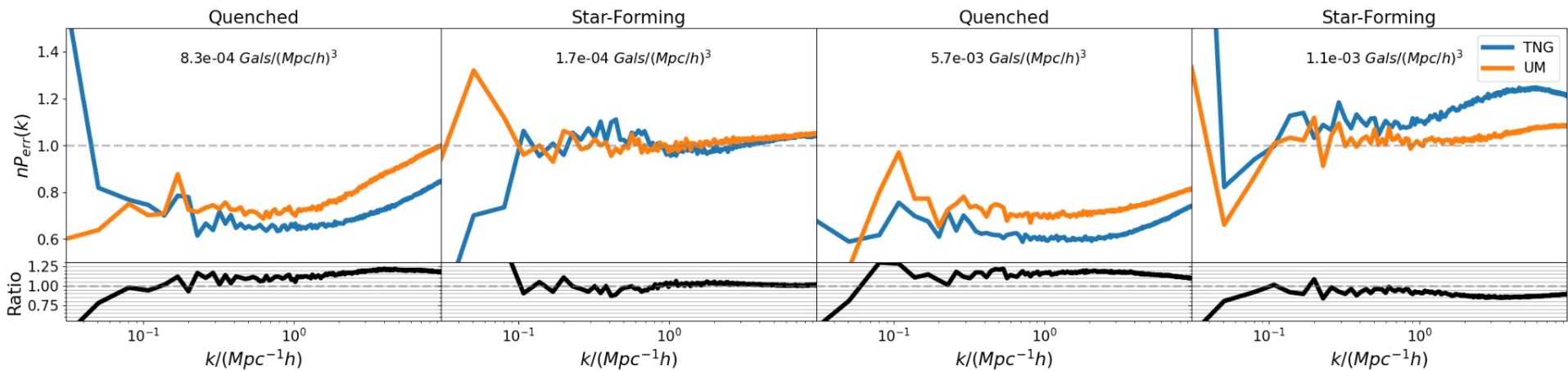
# $b_1$ vs $b_{s^2}$ in large scale regime ( $k_{\max} < 0.2 \text{ h/Mpc}$ )



# A closer look at $k_{\max} = 0.2 \text{ h/Mpc}$



## Comparing stochasticity power spectrum at $z=0$ :



## Comparing HEFT/galaxy power spectrum at $z=0$ :

