

Random Walks in the Sky (remembering your steps!)

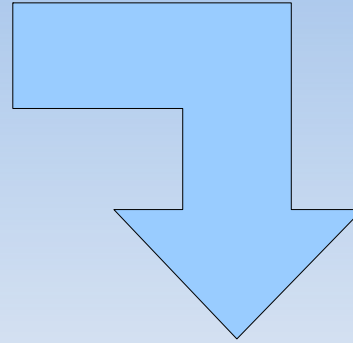
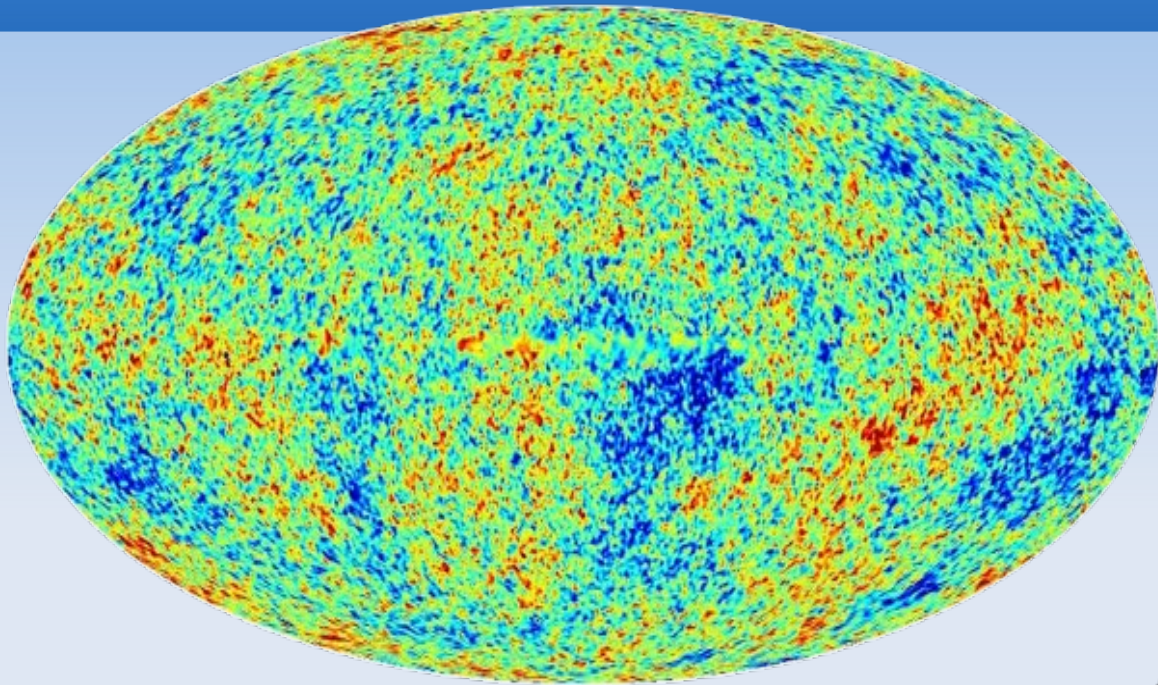
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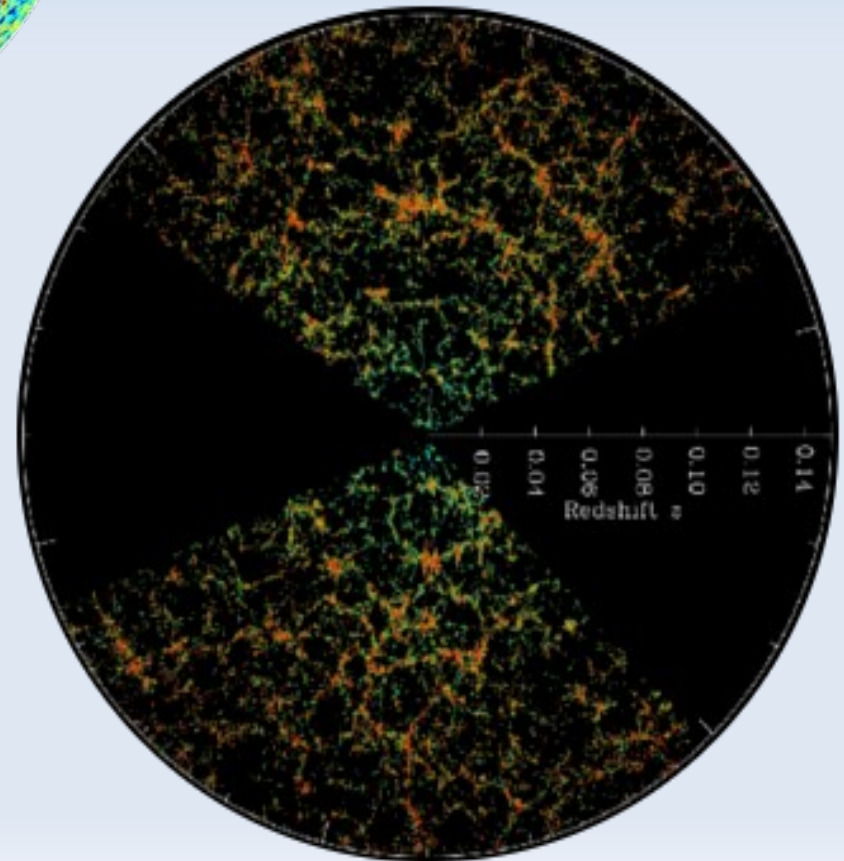
in collaboration with A. Paranjape and R. Sheth
(arXiv: 1201.3876, 1205.3401, 1305.0724, 1306.0551, 1401.????)



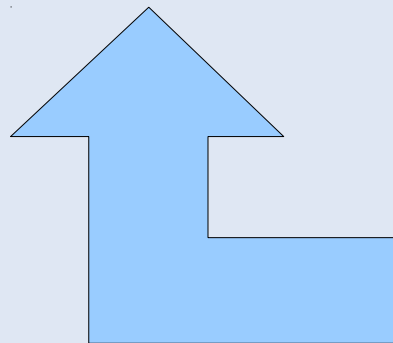
Formation of structures



Halos evolve
from critical
overdensities



Can use halos
to reconstruct
the primordial
distribution



Q. What is a halo?

- A1. “The largest gravitationally bound nearly spherical structure”
- A2. “The largest virialized structure”
- A3. “A cluster of galaxies”
- A4. “A lump of dark matter surrounding a cluster”
- A5. “A lump of dark matter surrounding a galaxy”
- A6. “A lump of dark matter in a simulation found by an algorithm”
- A7. “A lump of dark matter in a simulation found by ANOTHER algorithm”
- A8. “An dark matter overdensity larger than...”
- ...

We don't know what they are... but we do know how to compute them!

Halo Mass Function and Bias

- How many halos of given mass: mass function $n(M, z)$
- Correlation of halo counts with underlying matter field: halo bias

GOALS:

- Analytical halo statistics reproducing (heavy!) N-body simulations
- Information on initial conditions and matter/energy balance from data:

At fixed z :

Properties of
Early Universe

As z changes:

Energy / matter content
 Λ , Quintessence, ModGrav?

EUCLID will try to measure both!

Excursion set theory

- Halos from “dense enough” patches in the initial matter distribution δ_{in}
- Mean density $\delta_R \equiv [\text{average of } \delta_{in} \text{ over volume } R^3] \geq \text{threshold } b$

$$\delta_R(\mathbf{x}) \equiv \frac{1}{V_R} \int d^3y W_R(\mathbf{y} - \mathbf{x}) \delta_{in}(\mathbf{y}) \geq \frac{b}{D(z)}$$

- Find the distribution $f(R)$ of R of first crossing given that of δ_{in}
- Halo mass M proportional to the volume $V \sim R^3$ of the patch.

$$M = \bar{\rho} V_R \equiv \bar{\rho} \int d^3y W_R(\mathbf{y})$$

- Result depends on matter power spectrum $P(k)$, choice of W and b
- If b from spherical collapse in Einstein-dS, then $b = \delta_c = 1.7$. But $D(z)$ is sensitive to Ω_m , Ω_Λ , w , baryons, ν 's, modifications of gravity... Plus scale dependence from ellipsoidal collapse. All the physics is here!

Excursion set theory

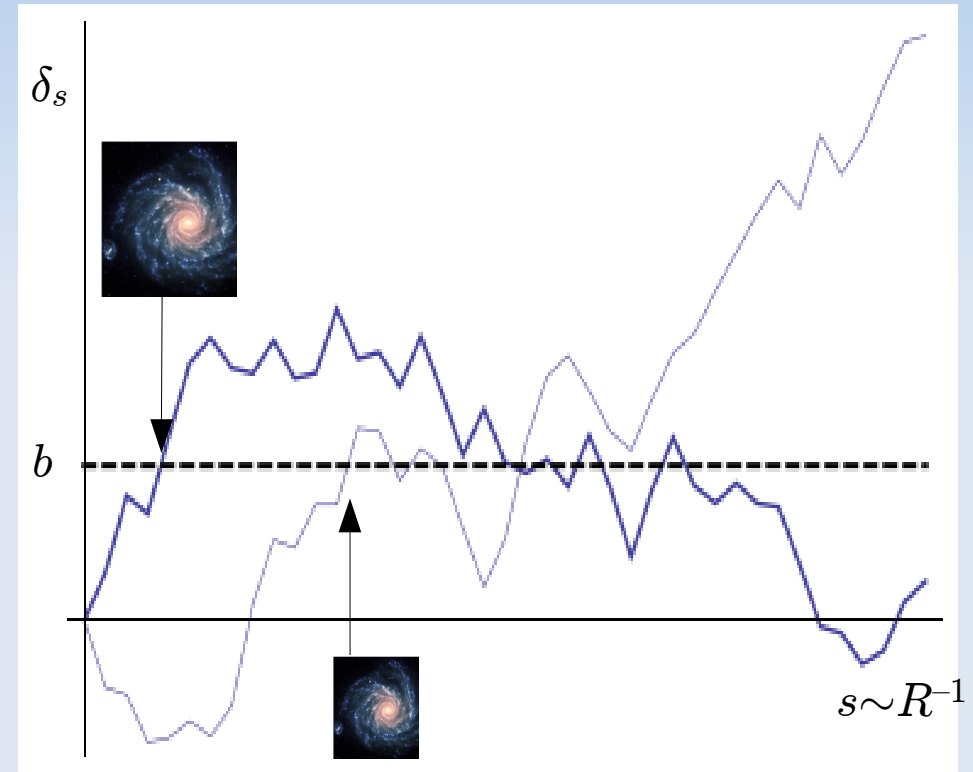
- Different locations realize different random walks: $s(M) \equiv \langle \delta_R^2(x) \rangle$

FIRST PASSAGE PROBLEM!

All the physics is in b

CORRELATED STEPS!

No known solution
NEED BETTER MATHS



- Abundance $n(M) \leftrightarrow$ first crossing probability $f(s)$ at scale $s(M)$

First crossing distribution

- Probability of ANY crossing at s :

$$f(s) = \frac{d}{ds} \langle \vartheta(\delta - b(s)) \rangle = \frac{d}{ds} \int_{b(s)}^{+\infty} d\delta p(\delta; s)$$

Press & Schechter (1974)

- Not any, but **FIRST** crossing (cloud-in-cloud problem); solution only for Gaussian uncorrelated steps with constant or linear barrier

$$f(s) = \frac{\langle \vartheta(b_1 - \delta_1) \dots \vartheta(b_{N-1} - \delta_{N-1}) \vartheta(\delta_N - b_N) \rangle}{\Delta s} \quad \text{Bond et al. (1991)}$$

- Can perturb around that

Maggiore & Riotto (2010)
Corasaniti & Achitouv (2011)

First crossing distribution

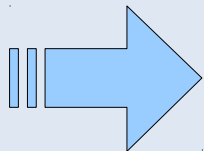
- However: strongly correlated walks are less affected (less zig-zags)

Paranjape, Lam & Sheth (2011)

- Can relax FIRST into simply **UPWARDS**: $\delta = B$; $\delta' \geq B'$

$$f(s) = \left\langle \left[\frac{d}{ds} \vartheta(\delta_s - B) \right] \vartheta(\delta'_s - B') \right\rangle = \int_{B'}^{\infty} dv (v - B') p(B, v; s)$$

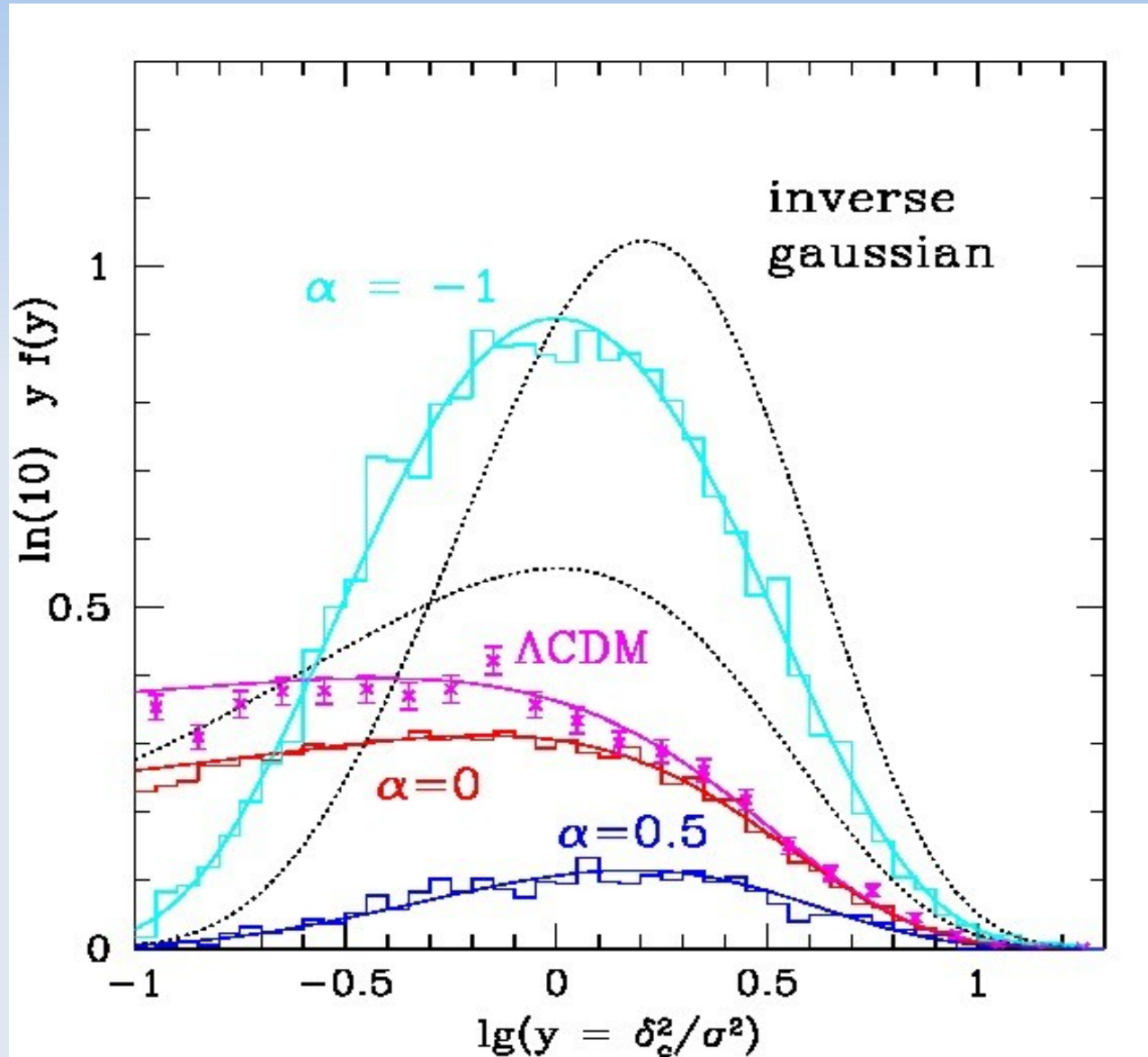
MM & Sheth (2012)



$$f(s) = - \left(\frac{B}{\sqrt{s}} \right)' \frac{e^{-B^2/2s}}{\sqrt{2\pi}} \left[\frac{1 + \operatorname{erf}(X/\sqrt{2})}{2} + \frac{e^{-X^2/2}}{2X\sqrt{2\pi}} \right]$$

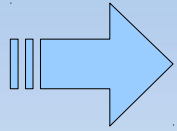
$$X = \frac{\text{mean of } p(v|B) - B'}{\sqrt{\text{var of } p(v|B)}}$$

First crossing distribution

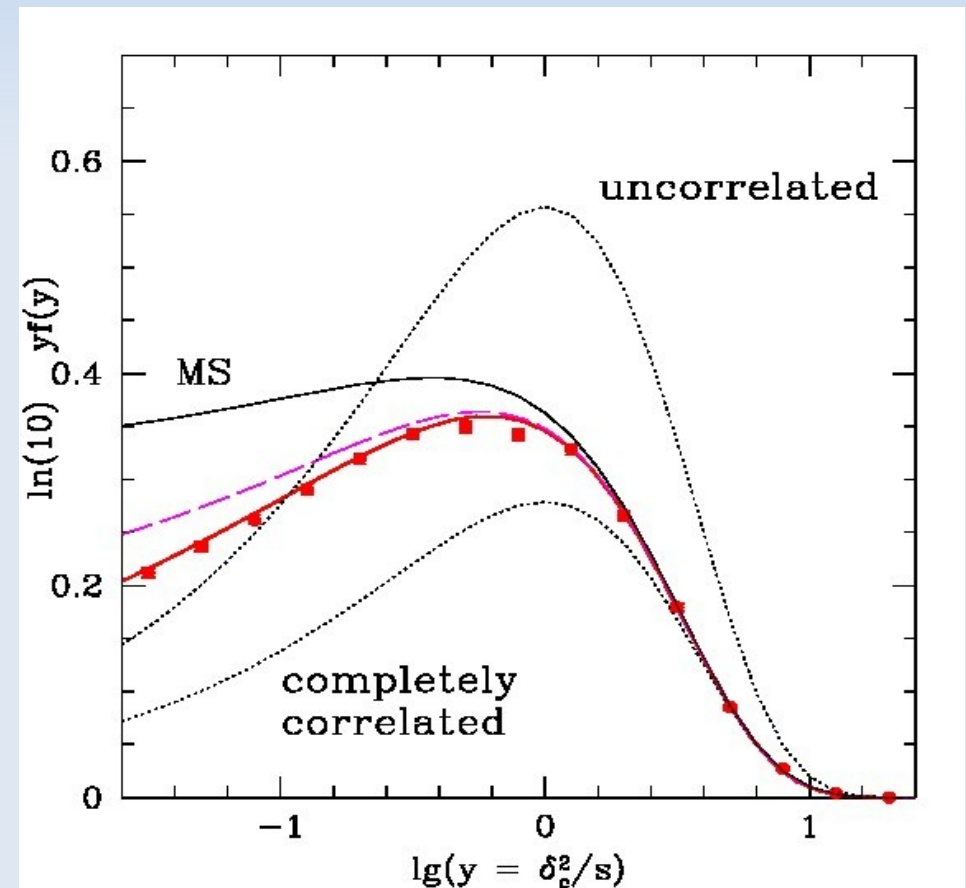
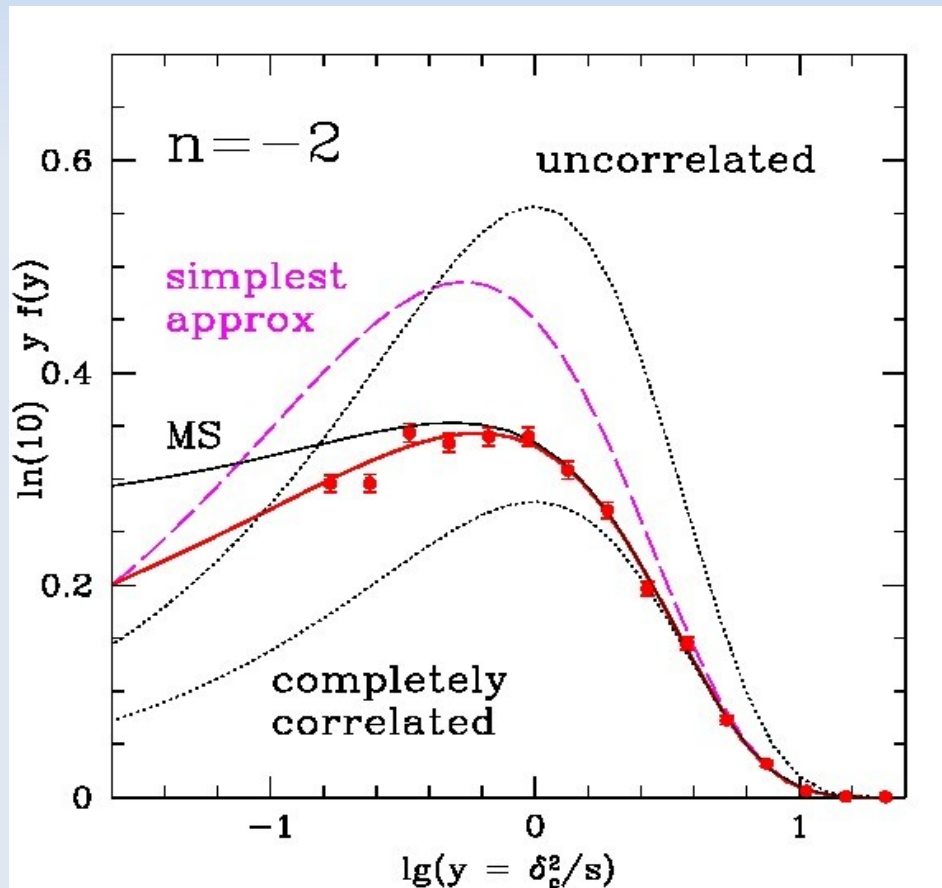


MM & Sheth (2012)

Upward mobility, back-substitution

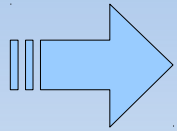


$$p(\delta \geq B, s) = \int_0^s dS f(S) p(\delta \geq B, s | \text{up}, S)$$



MM & Sheth (2013)

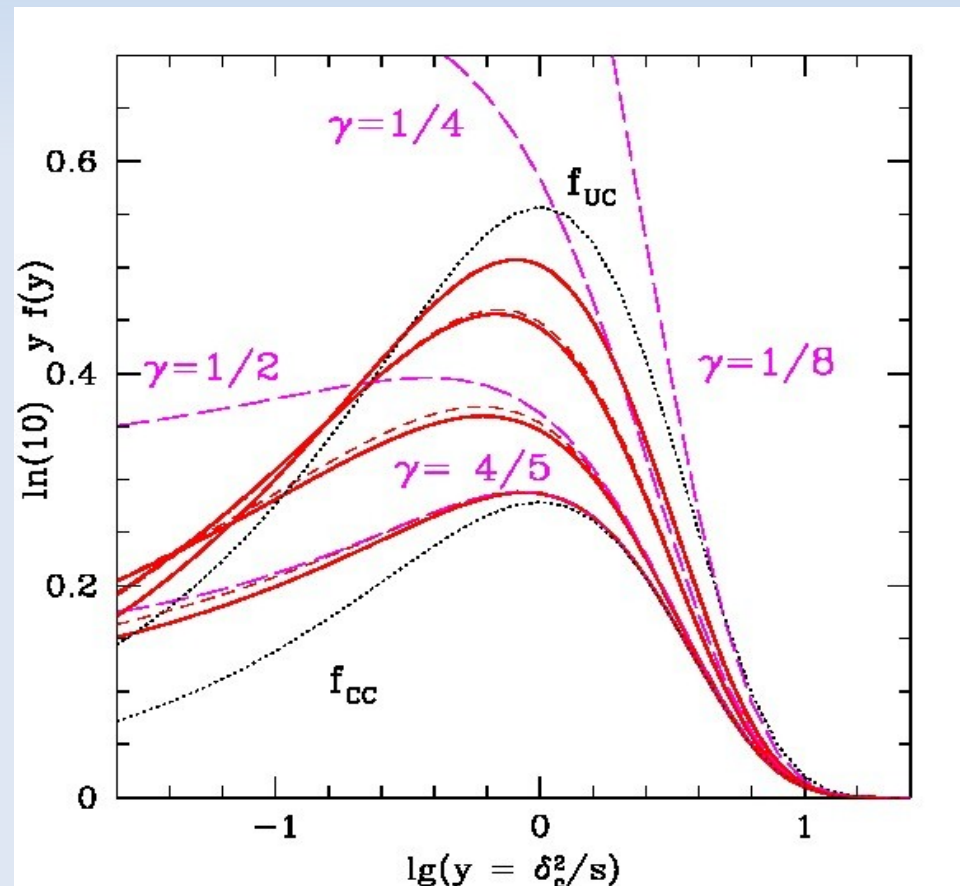
Upward mobility, back-substitution



$$p(\delta \geq B, s) = \int_0^s dS f(S) p(\delta \geq B, s | \text{up}, S)$$

$$\gamma^2 = \frac{\langle \delta v \rangle^2}{\langle \delta^2 \rangle \langle v^2 \rangle} = \frac{1}{4s \langle v^2 \rangle}$$

- It fully captures $f(s)$
- Interpolates between correlated and uncorrelated steps
- Only same-scale statistics matter



MM & Sheth (2014)

Upward mobility, back-substitution

$$p(\delta \geq B, s) = \int_0^s dS f(S) p(\delta \geq B, s | \text{up}, S)$$

- Full understanding of the correlated random walk problem, with any power spectrum and barrier!



MM & Sheth (2013)

- Does this $f(s)$ reproduce N-body mass function? Not really...



- Not surprising, space correlations are neglected. HOWEVER...

Upward mobility, back-substitution

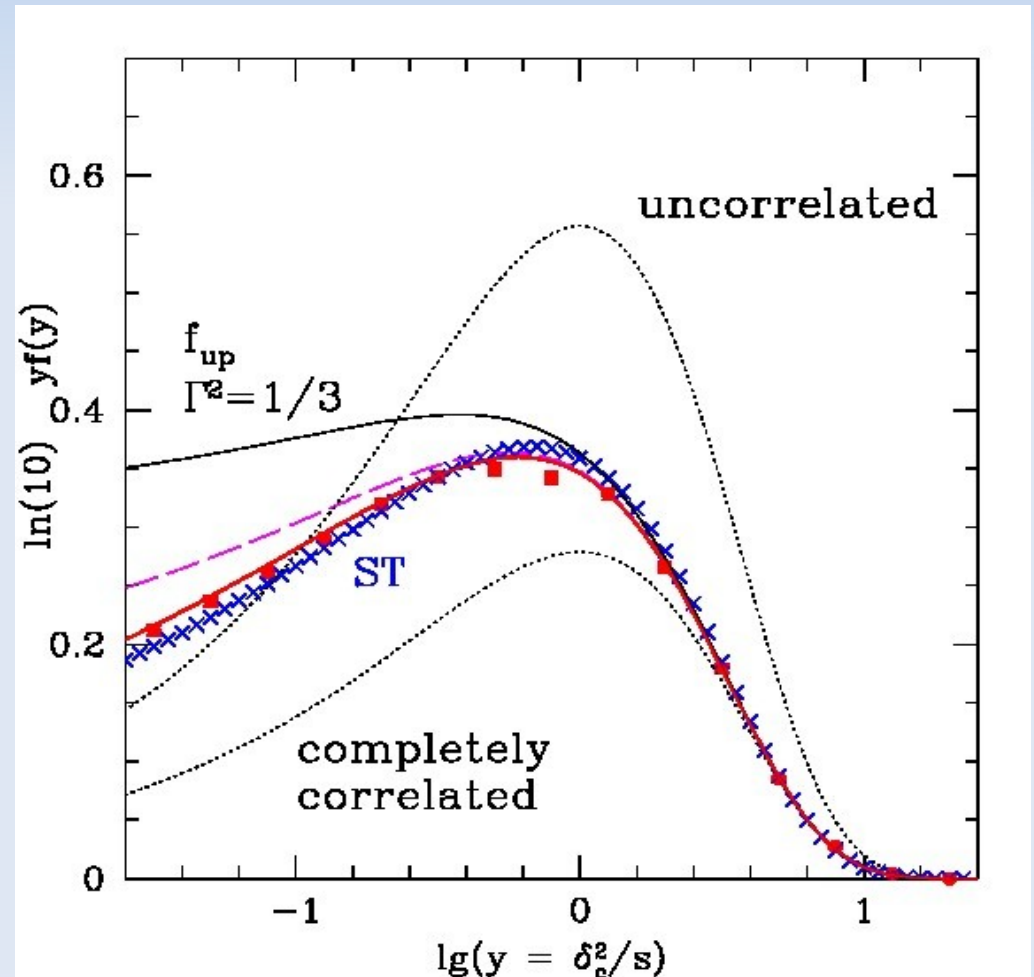
$$p(\delta \geq B, s) = \int_0^s dS f(S) p(\delta \geq B, s | \text{up}, S)$$

HOWEVER:

- Simple model with Markovian velocities (not heights!)
- Rescaled constant barrier:

$$B = \delta_c \rightarrow \sqrt{0.7} \delta_c$$

IT WORKS!!



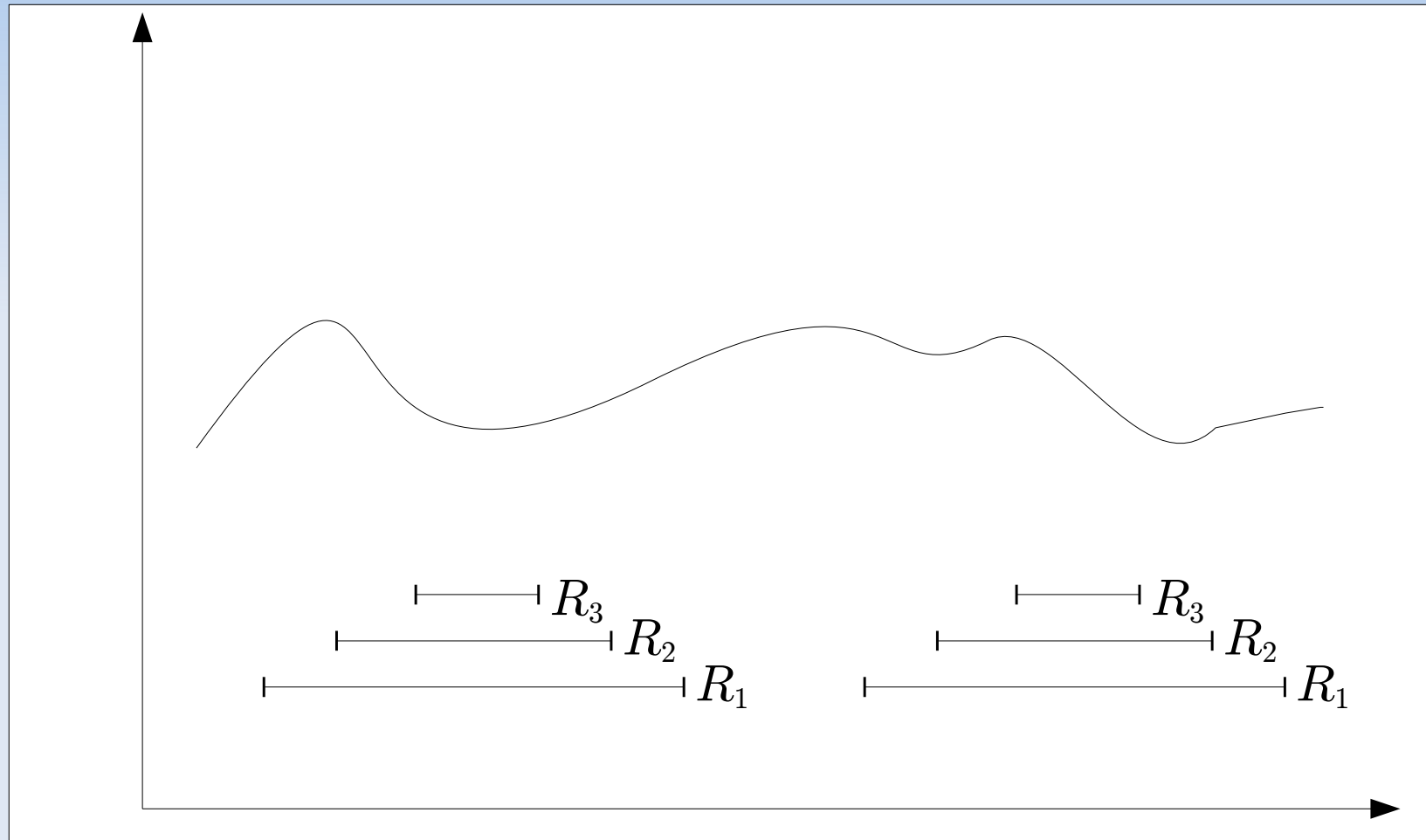
MM & Sheth (2013)

Conclusions

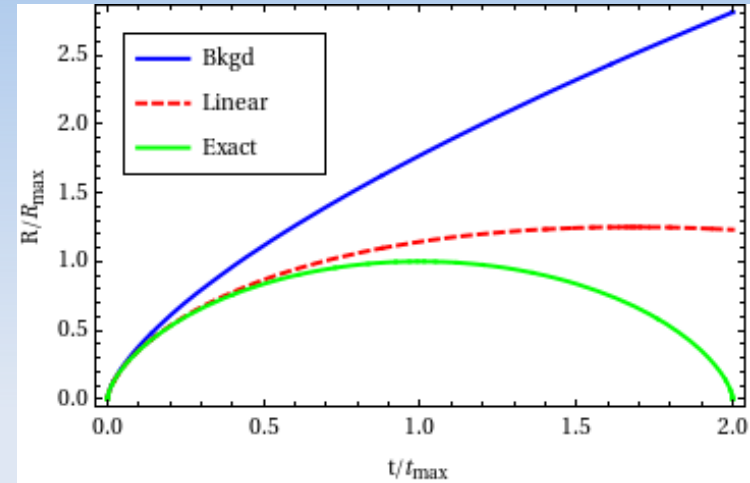
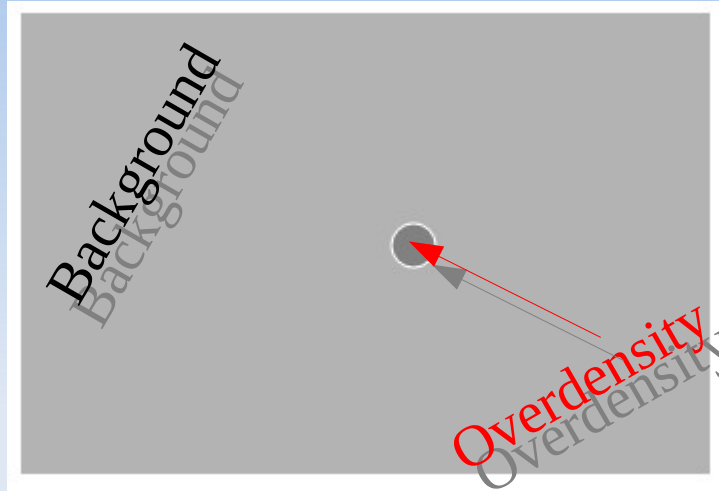
- Accurate solution of first passage of correlated random walks
- Full understanding of the excursion set approach
- Rescaling of the spherical collapse barrier fits correctly Gaussian mass function from simulations
- Straightforward non-perturbative inclusion of NG (primordial, Eulerian field, effects of shear on collapse...)
- Self-consistent predictions of bias functions and coefficients
- Lots of consistency checks to be done!

Thanks!!

Excursion set theory



Spherical collapse



Study the LINEARIZED density contrast: $\delta_{lin}(t) = \frac{3}{20} \left(\frac{6\pi t}{t_{max}} \right)^{3/2}$

At virialization:

$$\delta_{lin} \simeq 1.68 \equiv \delta_c$$

Useless estimator for δ

Good indicator of clusters

(Mild dependence on cosmology)

Adding non-Gaussianity: MF

Same formalism for non-Gaussian initial conditions:

$$f(s) = \underbrace{\left[\frac{d}{ds} \int_{B(s)}^{\infty} d\delta p(\delta; s) \right]}_{\text{old PS result}} \underbrace{\left[\frac{1 + \text{erf}(X/\sqrt{2})}{2} + \frac{e^{-X^2/2}}{2X\sqrt{2\pi}} + \dots \right]}_{\text{Gaussian structure}}$$

Small!
(computed)

$$X = \frac{\text{mean of } p(v|B) - B'}{\sqrt{\text{var of } p(v|B)}}$$

full NG pdf

- Non-perturbative in NG parameters: $p(\delta; s)$ is the exact pdf!

$$p(B; s) = \frac{1}{\sqrt{2\pi}} \exp \left[-\frac{B^2}{2s} + \mu \frac{B^3}{3!} + \dots \right] \quad \left[\mu = \frac{\langle \delta^3 \rangle}{s^3} \sim f_{\text{NL}} \right]$$

- Residual NG corrections are small: OK as perturbations

MM & Sheth (2013)

Adding non-Gaussianity: MF

Same formalism for non-Gaussian initial conditions:

$$f(s) = \underbrace{\left[\frac{d}{ds} \int_{B(s)}^{\infty} d\delta p(\delta; s) \right]}_{\text{old PS result}} \underbrace{\left[\frac{1 + \text{erf}(X/\sqrt{2})}{2} + \frac{e^{-X^2/2}}{2X\sqrt{2\pi}} + \dots \right]}_{\text{Gaussian structure}}$$

Small!
(computed)

$$X = \frac{\text{mean of } p(v|B) - B'}{\sqrt{\text{var of } p(v|B)}}$$

full NG pdf

- The conditional $p(v | B)$ may still be Gaussian. E.g.:

$$\delta = F(\delta_G, s) \quad \longrightarrow \quad v = \frac{\partial F}{\partial \delta_G} v_G + \frac{\partial F}{\partial s}$$

- In this case the result is exact!

Light vs Mass: Halo Bias

- Relation between halo abundance δ_h and underlying DM density
- Usually, computed from $p(\delta_h; s | \delta_0; s_0)$. Is there an easier way?
- Yes, there is. Expanding halo correlation functions in terms of DM correlation functions

Light vs Mass: Halo Bias

- Take the most generic dependence on the matter field:

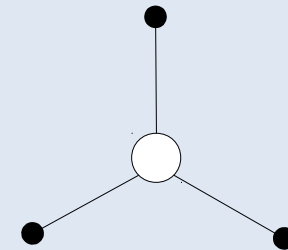
$$\delta_h(\mathbf{x}) = \sum_{k=0}^{\infty} \frac{1}{k!} \int d^3 y_1 \dots d^3 y_k b_k(\mathbf{x} - \mathbf{y}_1, \dots, \mathbf{x} - \mathbf{y}_k) \delta(\mathbf{y}_1) \dots \delta(\mathbf{y}_k)$$

e.g. Matsubara (2011)

- Compute connected correlation function of δ_h and δ :

$$\begin{aligned} & \langle \delta_h(\mathbf{x}) \delta(\mathbf{z}_1) \dots \delta(\mathbf{z}_n) \rangle_c \\ &= \int d^3 \mathbf{x}_1 \dots d^3 \mathbf{x}_n \underbrace{\left\langle \frac{\delta^n \delta_h(\mathbf{x})}{\delta \delta(\mathbf{x}_1) \dots \delta \delta(\mathbf{x}_n)} \right\rangle}_{\text{Bias functions}} \prod_{j=1}^n \langle \delta(\mathbf{x}_j) \delta(\mathbf{z}_j) \rangle \end{aligned}$$

- δ_h acts as an effective vertex for δ :



Halo Bias from Excursion Sets

- Need a prediction for δ_h . Can get it from excursion sets:

$$1 + \delta_h(m) = \frac{\vartheta(B_1 - \delta_1) \dots \vartheta(B_{N-1} - \delta_{N-1}) \vartheta(\delta_N - B_N)}{\langle \vartheta(B_1 - \delta_1) \dots \vartheta(B_{N-1} - \delta_{N-1}) \vartheta(\delta_N - B_N) \rangle}$$

$$c_n(\mathbf{x} - \mathbf{y}_1, \dots, \mathbf{x} - \mathbf{y}_n) = \sum_{i_1, \dots, i_n}^N \left\langle \frac{\partial^n \delta_h(m)}{\partial \delta_{i_1} \dots \partial \delta_{i_n}} \right\rangle W_{i_1}(\mathbf{x} - \mathbf{y}_1) \dots W_{i_n}(\mathbf{x} - \mathbf{y}_n)$$

$$\left\langle \frac{\partial^n \delta_h}{\partial \delta_{i_1} \dots \partial \delta_{i_n}} \right\rangle = \frac{(-1)^n}{f(s)} \frac{\partial^n f(s)}{\partial B_{i_1} \dots \partial B_{i_n}}$$

- Still a bit complicated...

$$\langle \delta_h \delta_0 \rangle = \sum_{i=1}^N \left\langle \frac{\partial \delta_h}{\partial \delta_i} \right\rangle \langle \delta_i \delta_0 \rangle \quad \langle \delta_h \delta_0^2 \rangle = \sum_{i,j=1}^N \left\langle \frac{\partial^2 \delta_h}{\partial \delta_i \partial \delta_j} \right\rangle \langle \delta_i \delta_0 \rangle \langle \delta_j \delta_0 \rangle$$

Halo Bias from Excursion Sets

- Use the **UPWARDS** approximation. Only two variables!

$$1 + \delta_h = \frac{1}{f(s)} \left[\frac{d}{ds} \vartheta(\delta_s - B) \right] \vartheta(\delta'_s - B')$$

- The real space bias functions become easy:

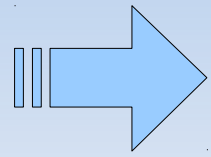
$$c_1(\mathbf{x} - \mathbf{y}) = -\frac{1}{f(s)} \left[W_R(\mathbf{x} - \mathbf{y}) \frac{\partial}{\partial B} + W'_R(\mathbf{x} - \mathbf{y}) \frac{\partial}{\partial B'} \right] f(s)$$

$$c_n(\mathbf{x} - \mathbf{y}_1, \dots, \mathbf{x} - \mathbf{y}_n) = \frac{(-1)^n}{f(s)} \prod_{i=1}^n \left[W_R(\mathbf{x} - \mathbf{y}_i) \frac{\partial}{\partial B} + W'_R(\mathbf{x} - \mathbf{y}_i) \frac{\partial}{\partial B'} \right] f(s)$$

MM, Paranjape & Sheth (to appear)

Halo Bias from Excursion Sets

- What should one measure?



$$\langle \delta_h \delta_0 \rangle = c_{10} \langle \delta \delta_0 \rangle + c_{11} \langle \delta' \delta_0 \rangle$$

$$\langle \delta_h \delta_0^2 \rangle_c = b_{20} \langle \delta \delta_0 \rangle^2 + 2b_{21} \langle \delta \delta_0 \rangle \langle \delta' \delta_0 \rangle + b_{22} \langle \delta' \delta_0 \rangle^2$$

MM, Paranjape & Sheth (2012)

- Easy measurement in real space!
- The coefficients are straightforward:

$$c_{nk} = \frac{(-1)^n}{f(s)} \frac{\partial^{n-k}}{\partial B^{n-k}} \frac{\partial^k}{\partial B'^k} f(s) \quad \text{with} \quad f(s) = \int_{B'}^{\infty} dv (v - B') p(B, v)$$

MM, Paranjape & Sheth (to appear)

- Completely generic structure! (not specific to excursion sets)

Halo Bias from Excursion Sets

- Linear bias in Fourier space:

$$c_1(k) = c_{10} \underbrace{W(kR)}_{\sim 1} + c_{11} \underbrace{2sW'(kR)}_{\sim k^2 R^2}$$

- Quadratic bias:

$$c_2(k_1, k_2) \simeq c_{20} + c_{21}(k_1^2 + k_2^2)R^2 + c_{22}k_1^2 k_2^2 R^4$$

- Numerical predictions for the coefficients $b_{nj}(m)$ from $f(s)$; b_{n0} is the same as in peak-background split
- k -dependence!

MM, Paranjape & Sheth (2012)

Adding non-Gaussianity: Bias

- Expand halo-matter n -point functions in matter polyspectra

$$\langle \delta_h(\mathbf{x}) \delta(\mathbf{z}_1) \rangle_c = \text{[Diagram: circle]---\bullet + \text{[Diagram: overlapping circles]---\bullet + \dots}$$

$$\langle \delta_h(\mathbf{x}) \delta(\mathbf{z}_1) \delta(\mathbf{z}_2) \rangle_c = \text{[Diagram: \bullet---circle---\bullet] + \text{[Diagram: \bullet---circle---overlapping circles---\bullet] + \dots}$$

- Generic excursion set bias for non-Gaussian walks:

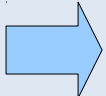
$$\langle \delta_h \delta_0 \rangle = \sum_{i=1}^N \left\langle \frac{\partial \delta_h}{\partial \delta_i} \right\rangle \langle \delta_i \delta_0 \rangle + \frac{1}{2} \sum_{i,j=1}^N \left\langle \frac{\partial^2 \delta_h}{\partial \delta_i \partial \delta_j} \right\rangle \langle \delta_i \delta_j \delta_0 \rangle + \dots$$

Adding non-Gaussianity: Bias

- With the two-step approximation:

$$\begin{aligned} \langle \delta_h \delta_0 \rangle &\simeq c_{10} \langle \delta \delta_0 \rangle + c_{11} \langle \delta' \delta_0 \rangle \\ &+ \frac{1}{2} \left[c_{20} \langle \delta^2 \delta_0 \rangle + 2 c_{21} \langle \delta' \delta \delta_0 \rangle + c_{22} \langle \delta'^2 \delta_0 \rangle \right] + \dots \end{aligned}$$

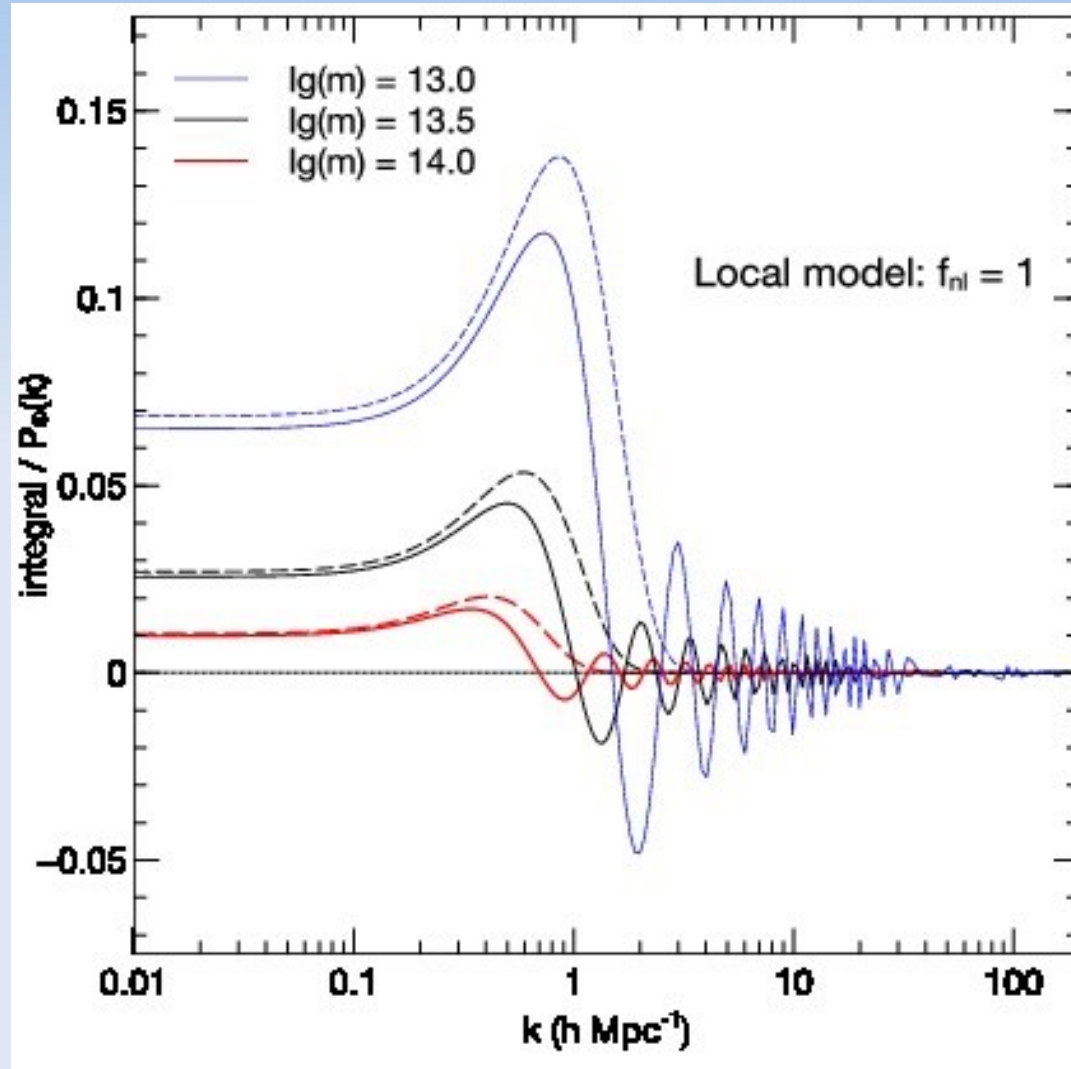
- Same definition of b_{nj} as before, but now with non-Gaussian $f(s)$


$$\Delta c_1(k) = \frac{2f_{\text{NL}}^{\text{local}}}{k^2 T(k)} \left[s c_{20} + c_{21} + \langle (\delta')^2 \rangle c_{22} + \mathcal{O}(k^2) \right]$$

- All coefficients matter at small k . Possibly, also some effects at $k \sim R$ (where the leading term starts decaying). Equilateral NG?

MM, Paranjape & Sheth (to appear)

Adding non-Gaussianity: Bias



MM, Paranjape & Sheth (to appear)

Conclusions

- Accurate solution of first passage of correlated random walks
- Full understanding of the excursion set approach to structure formation
- Simple rescaling of the spherical collapse barrier reproduces correctly the Gaussian mass function
- Straightforward non-perturbative inclusion of NG (can do Eulerian field!)
- Self-consistent predictions of bias functions and coefficients, new strategies to measure them in simulations
- To do: check against N-body simulations, generalization to excursion set theory of peaks
- Interesting possibilities (?) for primordial non-Gaussianity

Thanks!!