

Random Walks in the Sky (remembering your steps!)

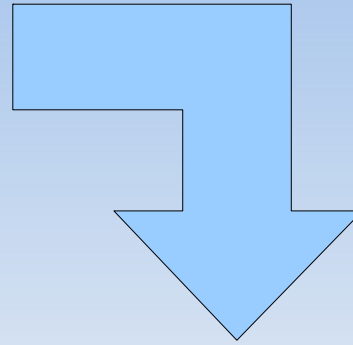
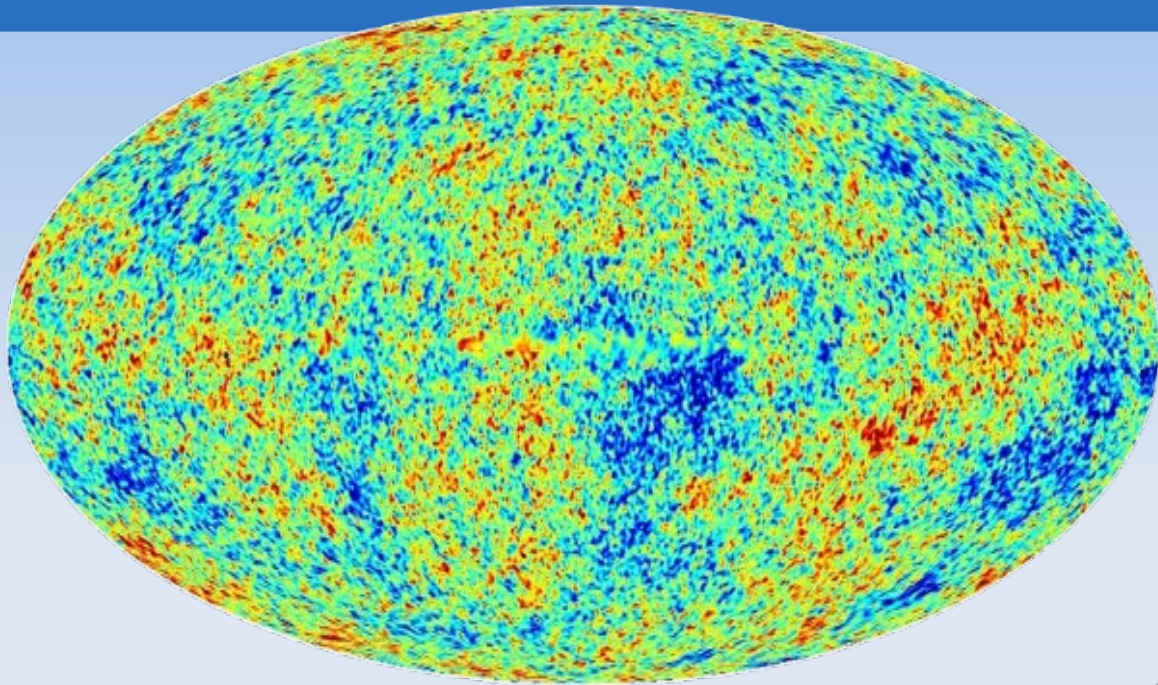
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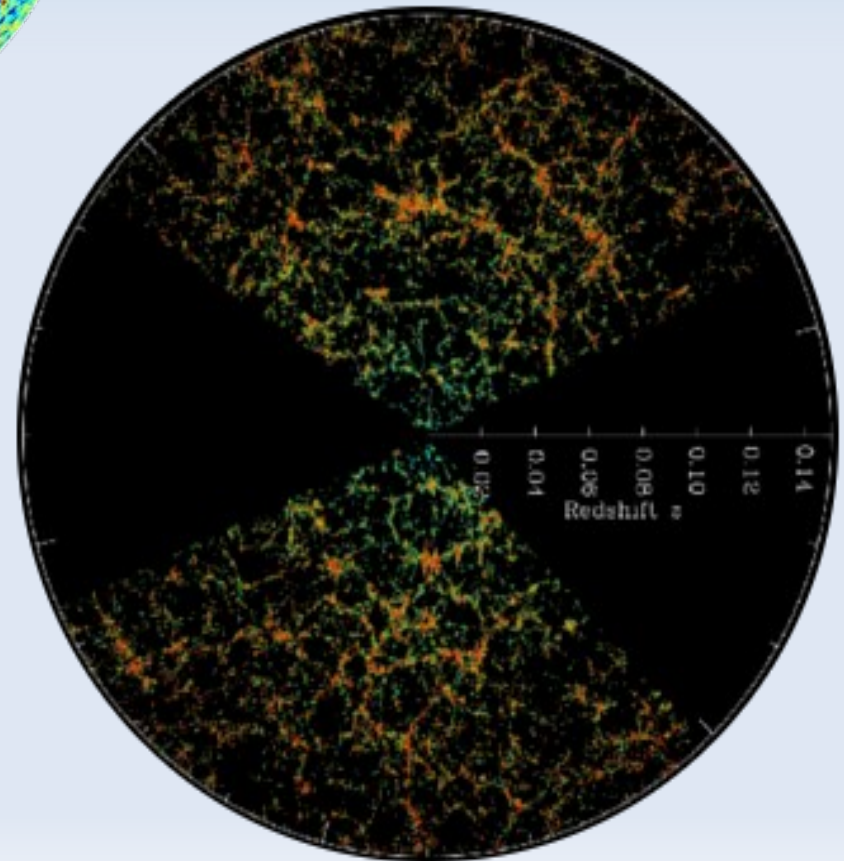
In collaboration with A. Paranjape and R. Sheth
(arXiv: 1201.3876, 1205.3401, 1305.0724, 1306.0551)



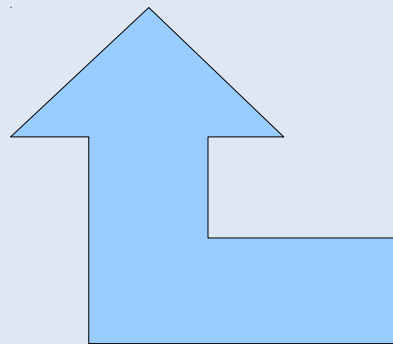
Formation of structures



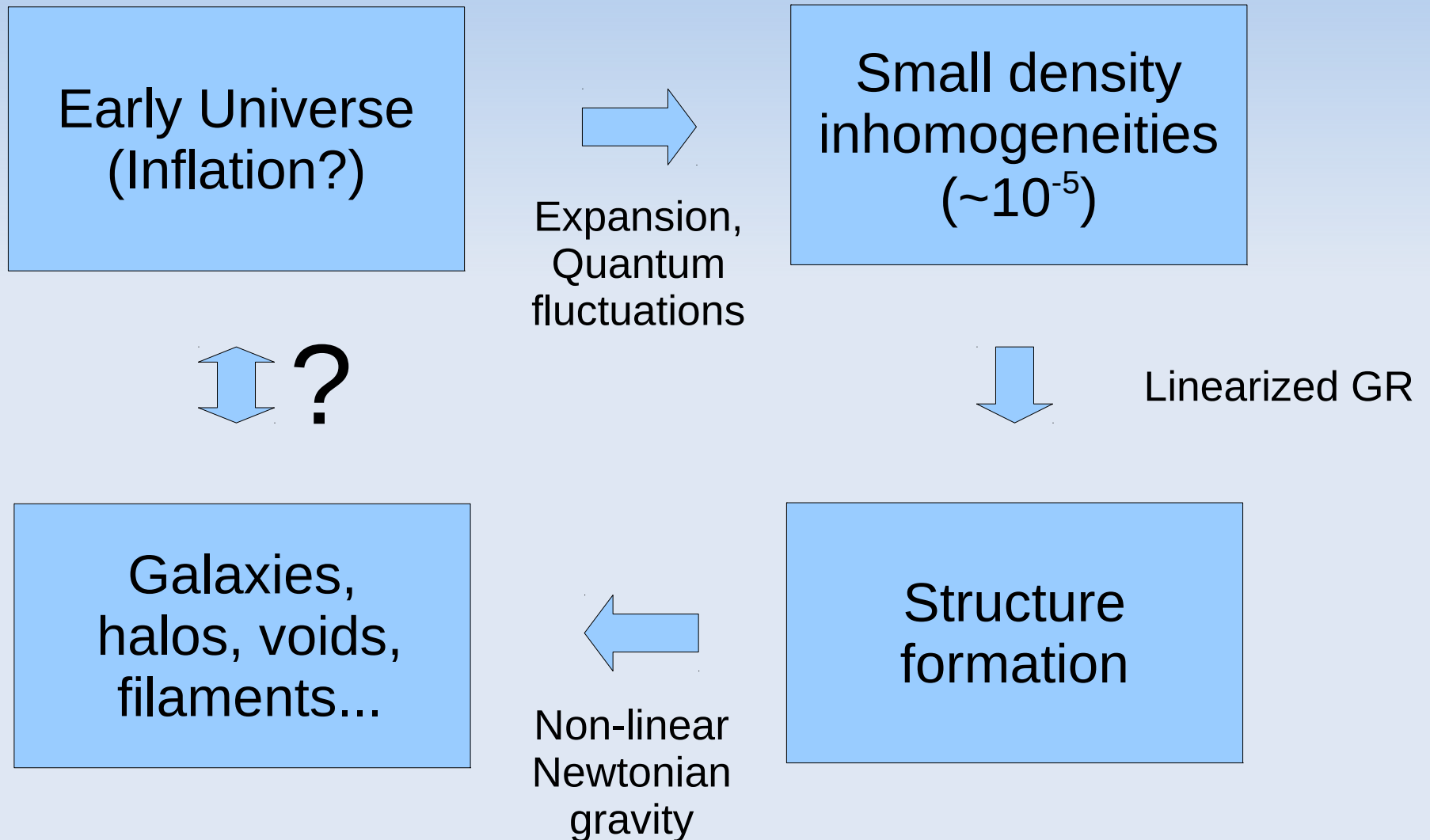
Halos evolve
from critical
overdensities



Can use halos
to reconstruct
the primordial
distribution



Formation of structures



Halo Mass Function and Bias

- How many halos of given mass (mass function)
- Correlation of halo counts with underlying matter field (halo bias)

GOALS:

- Analytical predictions for halo statistics reproducing N-body sims
- Information on initial conditions and matter/energy balance from data:

Properties of Early Universe
Energy and matter content
 Λ vs Quintessence vs ModGrav

Outline

Part ONE

- The excursion set approach
- Analytical progress for mass functions: the UPWARDS approximation

Part TWO

- Bias from excursion sets (a simple way!)

Excursion set theory

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

$$\delta_R(\mathbf{x}) \equiv \frac{1}{V_R} \int d^3y W_R(\mathbf{y} - \mathbf{x}) \delta(\mathbf{y}) \geq B$$

- Halo mass M proportional to the volume $V \sim R^3$ of the patch.

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

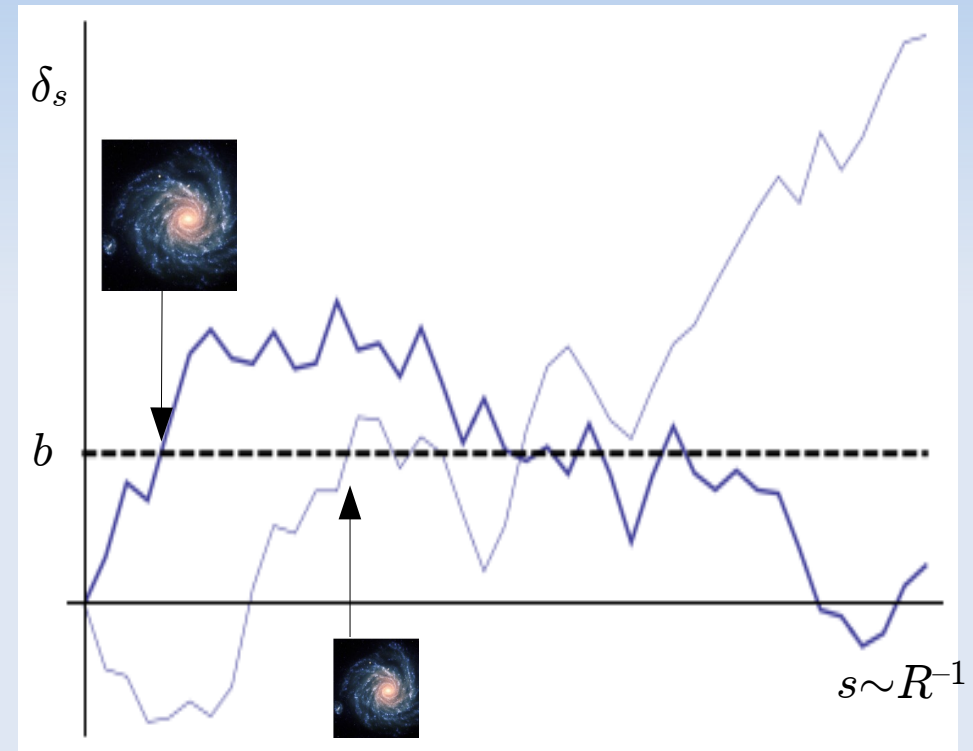
- If B from spherical collapse in Einstein-DeSitter, then $B = \delta_c = 1.686$. But in general B is sensitive to Ω_m , Ω_Λ , w , baryons, neutrinos, modifications of gravity... It depends on scale and redshift
- Correlations depend on matter power spectrum $P(k)$ and choice of W

Excursion set theory

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

FIRST PASSAGE PROBLEM!

CORRELATED STEPS!
No known solution
NEED BETTER MATHS



- Abundance $n(M) \longleftrightarrow$ first crossing probability $f(s)$ at scale $s(M)$
- Oversimplified picture, working surprisingly well

First crossing distribution

- Probability of ANY crossing at s :

$$f(s) = \frac{d}{ds} \langle \vartheta(\delta - b(s)) \rangle \quad \text{Press \& Schechter (1974)}$$

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

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

First crossing distribution

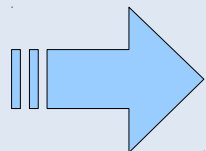
- However: strongly correlated walks are less affected (less zig-zags)

Paranjape, Lam & Sheth (2011)

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

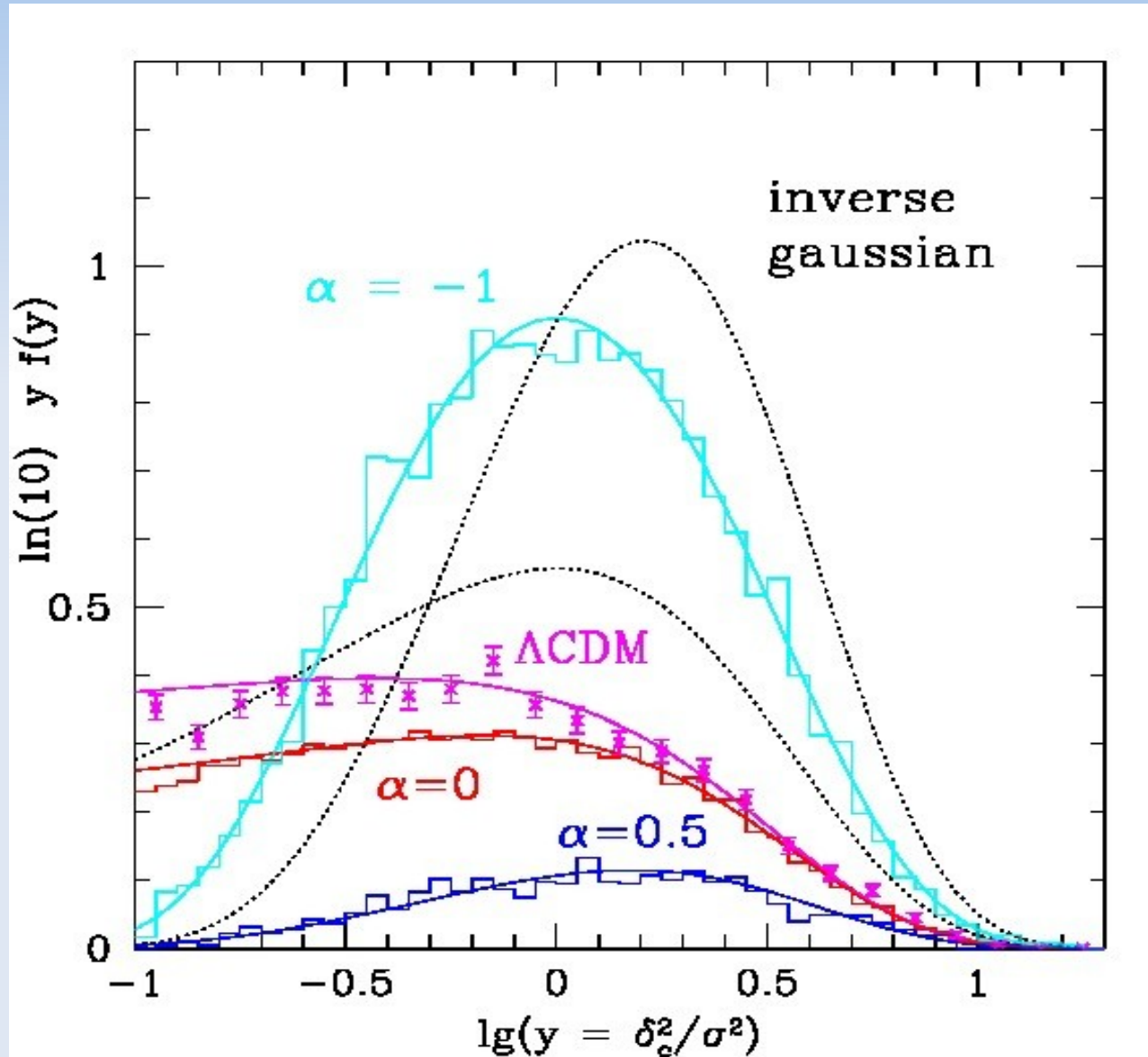
$$f(s) = \left\langle \left[\frac{d}{ds} \vartheta(\delta_s - B) \right] \vartheta(\delta'_s - B') \right\rangle = \int_{B'}^{\infty} dv (v - B') p(B, v; s)$$

MM & Sheth (2012)



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

First crossing distribution



MM & Sheth (2012)

Upward mobility, back-substitution

- To reach any $\delta > b$ at s , must cross at some $S < s$:

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

- Do the **UPWARDS** approximation $\delta'(S) \geq b'(S)$ and solve for $f(S)$

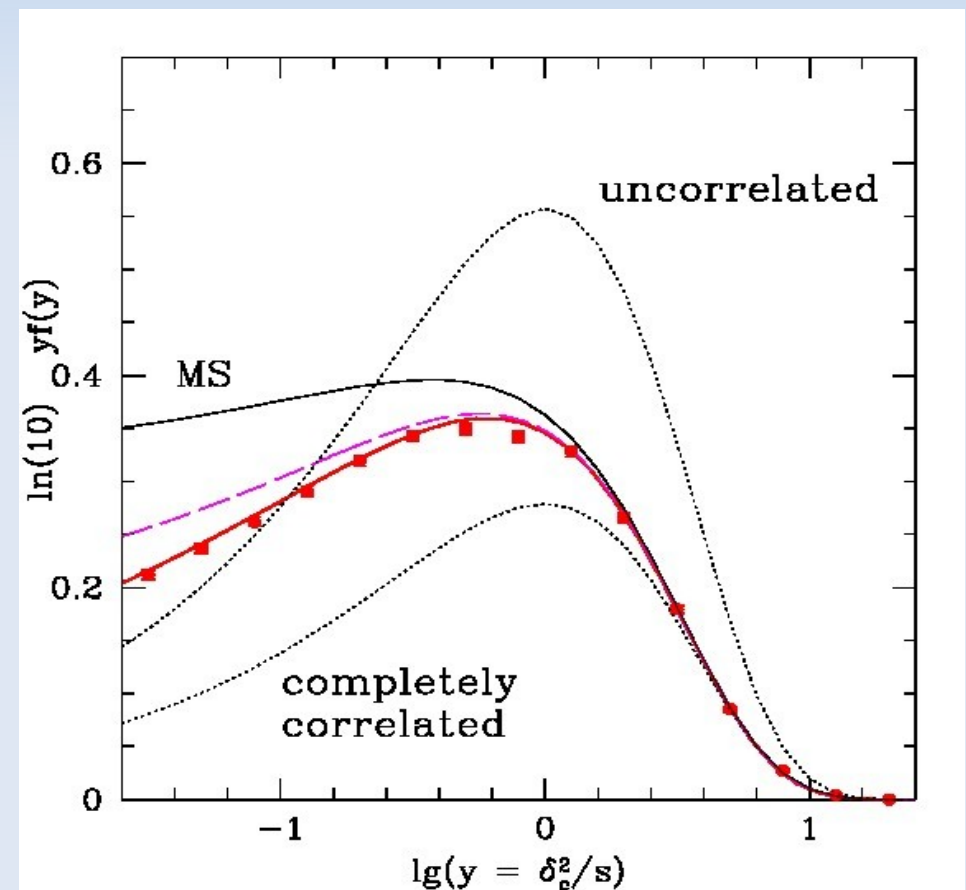
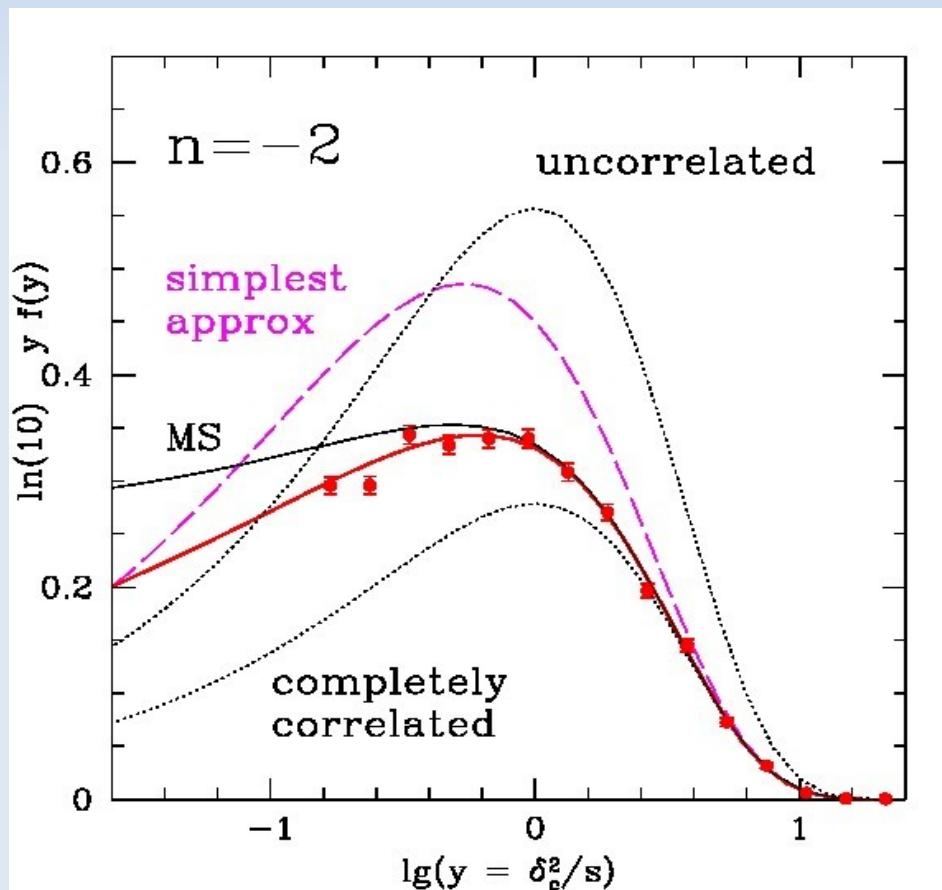
$$p(\delta \geq b, s | \text{first}, S) \quad \Rightarrow \quad p(\delta \geq b, s | \text{up}, S)$$

- Now solve for $f(S)$

MM & Sheth (2013)

Upward mobility, back-substitution

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



MM & Sheth (2013)

Upward mobility, back-substitution

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

- 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

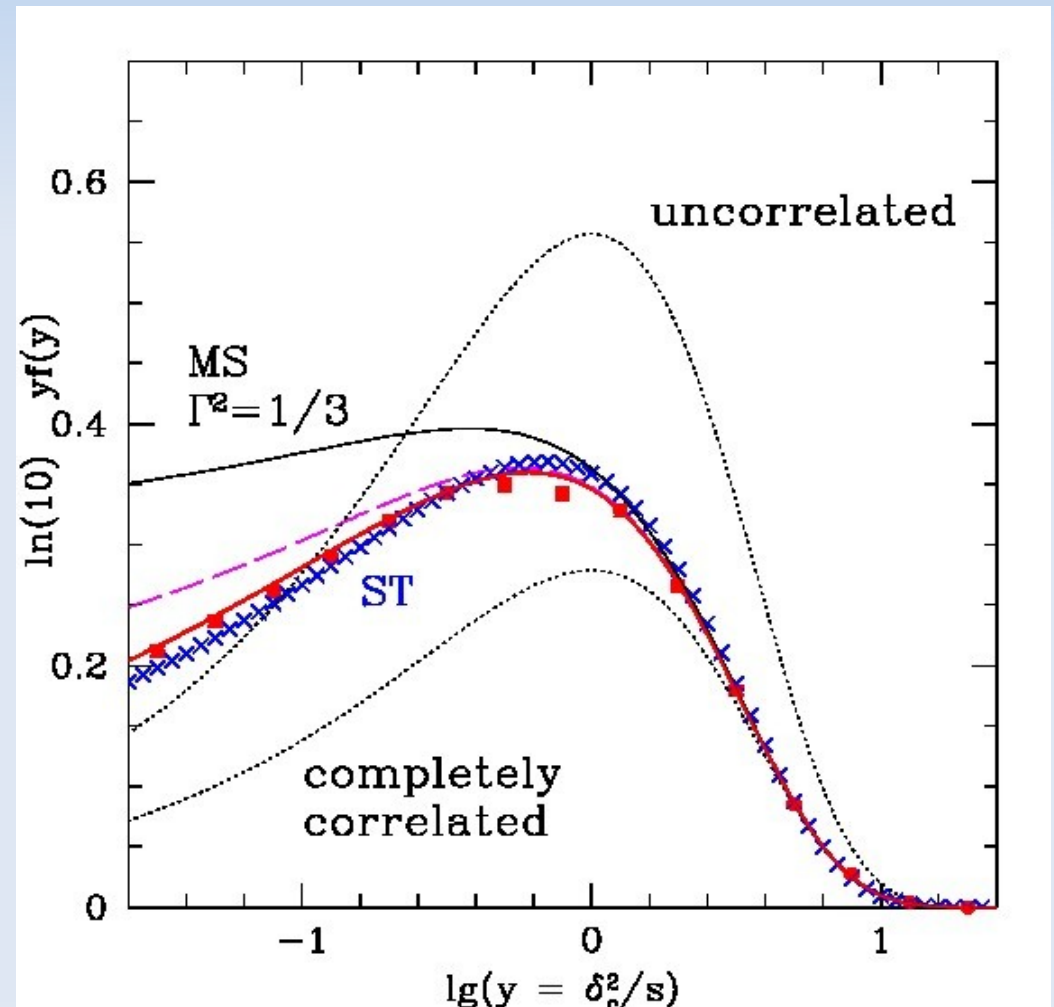
$$p(\delta \geq b, s) = \int_0^s dS f(S) p(\delta \geq b, s | \text{up}, S)$$

HOWEVER:

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

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

IT WORKS!!



MM & Sheth (2013)

Adding non-Gaussianity: MF

Same formalism for non-Gaussian initial conditions:

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

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

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

- Residual NG corrections are small: OK as perturbations

MM & Sheth (2013)

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:

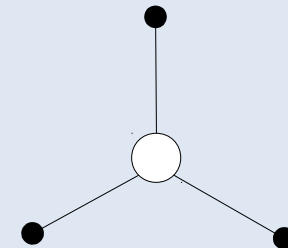
$$\delta_h(\mathbf{x}) = \sum_{k=0}^{\infty} \frac{1}{k!} \int d^3 y_1 \dots d^3 y_k b_k(\mathbf{x} - \mathbf{y}_1, \dots, \mathbf{x} - \mathbf{y}_k) \delta(\mathbf{y}_1) \dots \delta(\mathbf{y}_k)$$

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:

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

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

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

- 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:

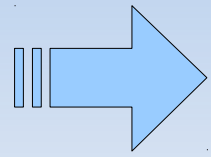
$$b_1(\mathbf{x} - \mathbf{y}) = -\frac{1}{f(s)} \left[W_R(\mathbf{x} - \mathbf{y}) \frac{\partial}{\partial B} + W'_R(\mathbf{x} - \mathbf{y}) \frac{\partial}{\partial B'} \right] f(s)$$

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

MM, Paranjape & Sheth (to appear)

Halo Bias from Excursion Sets

- What should one measure?



$$\langle \delta_h \delta_0 \rangle = b_{10}^{(f)} \langle \delta \delta_0 \rangle + b_{11}^{(f)} \langle \delta' \delta_0 \rangle$$

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

MM, Paranjape & Sheth (2012)

- The coefficients are straightforward:

$$b_{nk}^{(f)} = \frac{(-1)^n}{f(s)} \frac{\partial^{n-k}}{\partial B^{n-k}} \frac{\partial^k}{\partial B'^k} f(s)$$

with $f(s) = \int_{B'}^{\infty} dv (v - B') p(B, v)$

MM, Paranjape & Sheth (to appear)

Halo Bias from Excursion Sets

- Linear bias in Fourier space:

$$b_1(k) = b_{10}^{(f)} \underbrace{W(kR)}_{\sim 1} + b_{11}^{(f)} \underbrace{2sW'(kR)}_{\sim k^2 R^2}$$

- Quadratic bias:

$$b_2(k_1, k_2) \simeq b_{20}^{(f)} + b_{21}^{(f)} (k_1^2 + k_2^2) R^2 + b_{22}^{(f)} k_1^2 k_2^2 R^4$$

- 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:

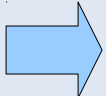
$$\langle \delta_h \delta_0 \rangle = \sum_{i=1}^N \left\langle \frac{\partial \delta_h}{\partial \delta_i} \right\rangle \langle \delta_i \delta_0 \rangle + \frac{1}{2} \sum_{i,j=1}^N \left\langle \frac{\partial^2 \delta_h}{\partial \delta_i \partial \delta_j} \right\rangle \langle \delta_i \delta_j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

Adding non-Gaussianity: Bias

- With the two-step approximation:

$$\langle \delta_h \delta_0 \rangle \simeq b_{10}^{(f)} \langle \delta \delta_0 \rangle + b_{11}^{(f)} \langle \delta' \delta_0 \rangle \\ + \frac{1}{2} \left[b_{20}^{(f)} \langle \delta^2 \delta_0 \rangle + 2 b_{21}^{(f)} \langle \delta' \delta \delta_0 \rangle + b_{22}^{(f)} \langle \delta'^2 \delta_0 \rangle \right] + \dots$$

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


$$\Delta b_1(k) = \frac{2f_{\text{NL}}^{\text{local}}}{k^2 T(k)} \left[s b_{20}^{(f)} + b_{21}^{(f)} + \langle (\delta')^2 \rangle b_{22}^{(f)} + \mathcal{O}(k^2) \right]$$

- 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)

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!!