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Memory matters: random walks to galaxies

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Memory matters: random walks to galaxies

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In collaboration with A. Paranjape and R. Sheth
(arXiv: 1201.3876, 1205.3401 and work in progress)

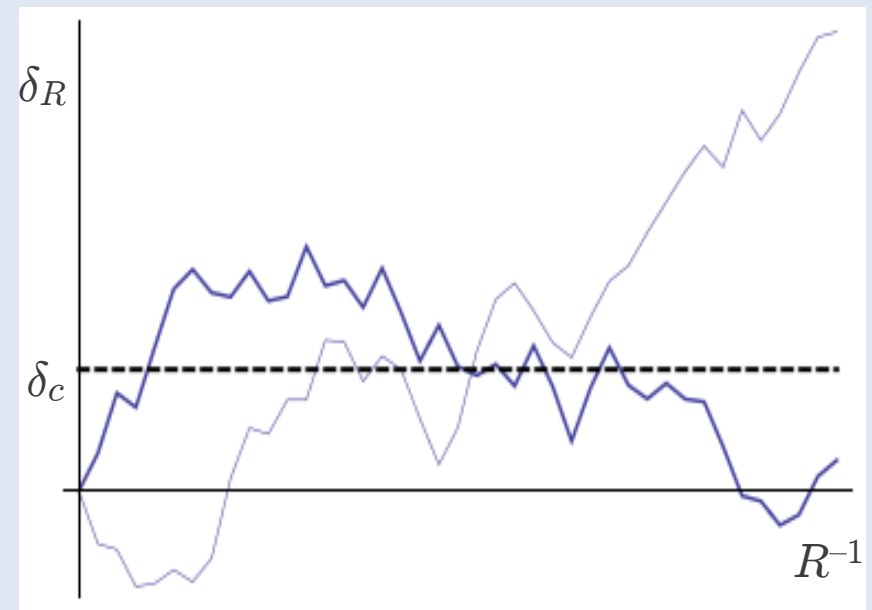
ICTP, July 2012

Excursion set theory

- Halos evolve from “dense enough” patches in the initial matter density distribution
- Mass of each halo is proportional to the volume of its “dense enough” initial patch

$$\delta_R(x) \equiv \frac{1}{V} \int d^3y W\left(\frac{y-x}{R}\right) \delta_{lin}(y)$$

$$s(R) = \langle \delta_R(x) \rangle$$



First crossing distribution

- Abundance $n(M) \leftrightarrow$ absorption rate of walks at scale $s(R)$

$$f(s) = \frac{d}{ds} \int_{B(s)}^{+\infty} d\delta p(\delta; s) \quad \text{Press \& Schechter (1974)}$$

- Not any crossing, but FIRST crossing (cloud-in-cloud problem)

Bond et al. (1991)

- Path integral of trajectories with N discrete steps of width ΔS :

$$f(s) = \frac{1}{\Delta S} \int_{-\infty}^B d\delta_1 \cdots d\delta_{N-1} \int_B^{+\infty} d\delta_N p(\delta_1, \dots, \delta_N)$$

cf. Maggiore & Riotto (2010)

- Strongly correlated walks are less affected by cloud-in-cloud

Paranjape, Lam & Sheth (2011)

First crossing distribution

- Only the last two steps are really important:

$$f(s) \simeq \int_{B'(s)}^{+\infty} dv (v - B') p(v, B) \quad \left[v = \frac{d\delta}{ds} \right]$$

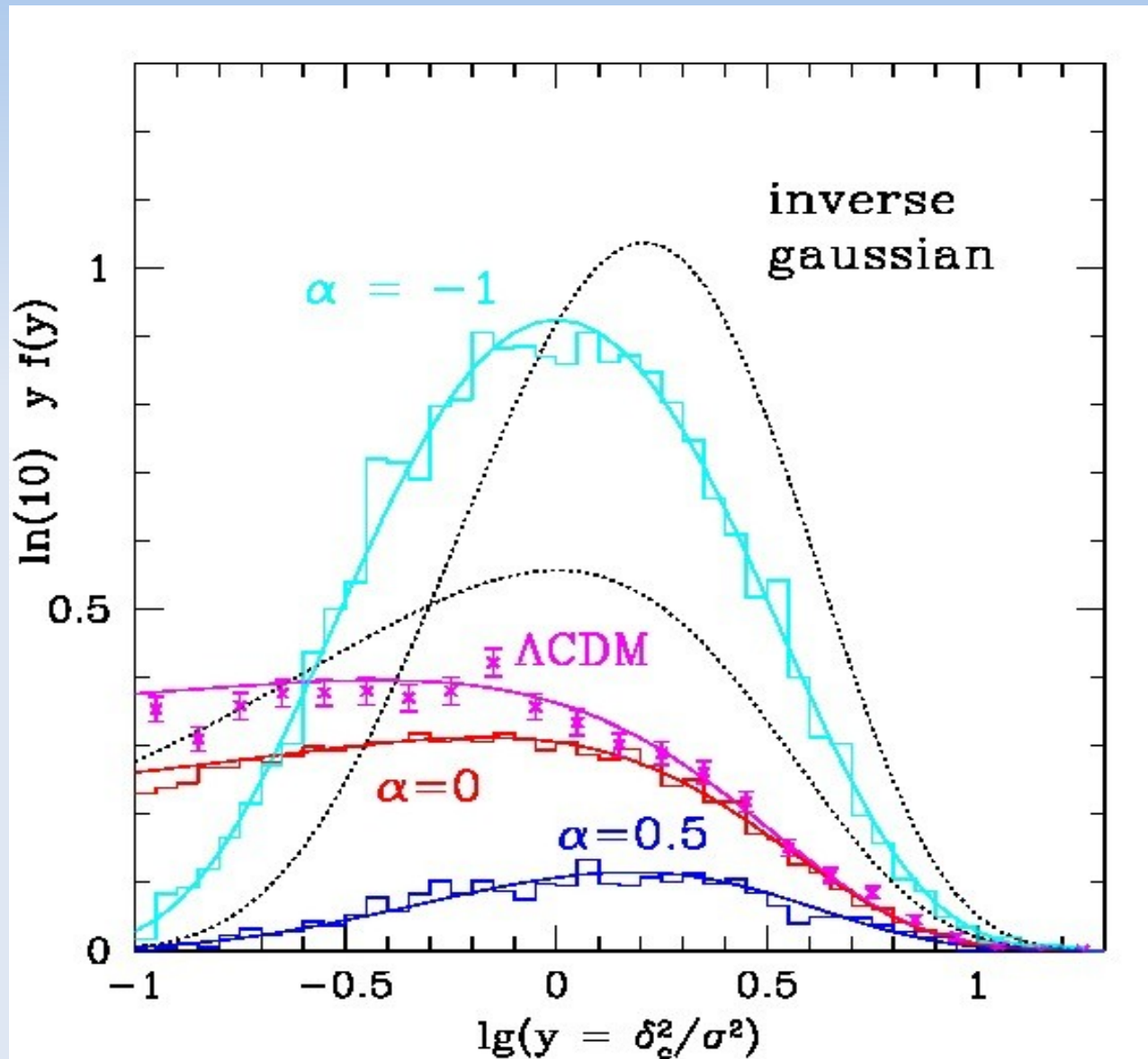
MM & Sheth (2012)

- Correlated Gaussian walks (Gaussian initial conditions and any smooth filter)

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

$$\beta = B(s)/\sqrt{s} \quad ; \quad \nu_*^2 = \beta'^2 / \langle (\delta/\sqrt{s})'^2 \rangle$$

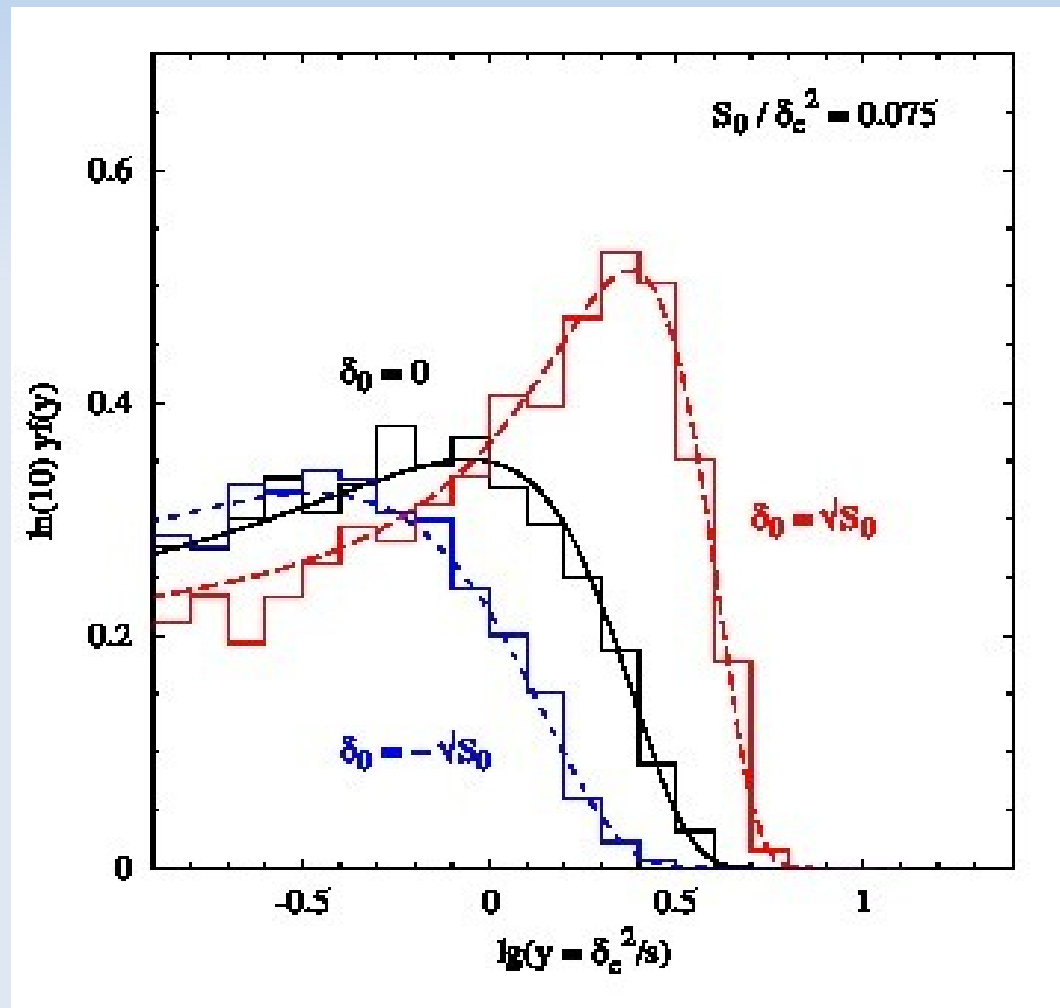
First crossing distribution



MM & Sheth (2012)

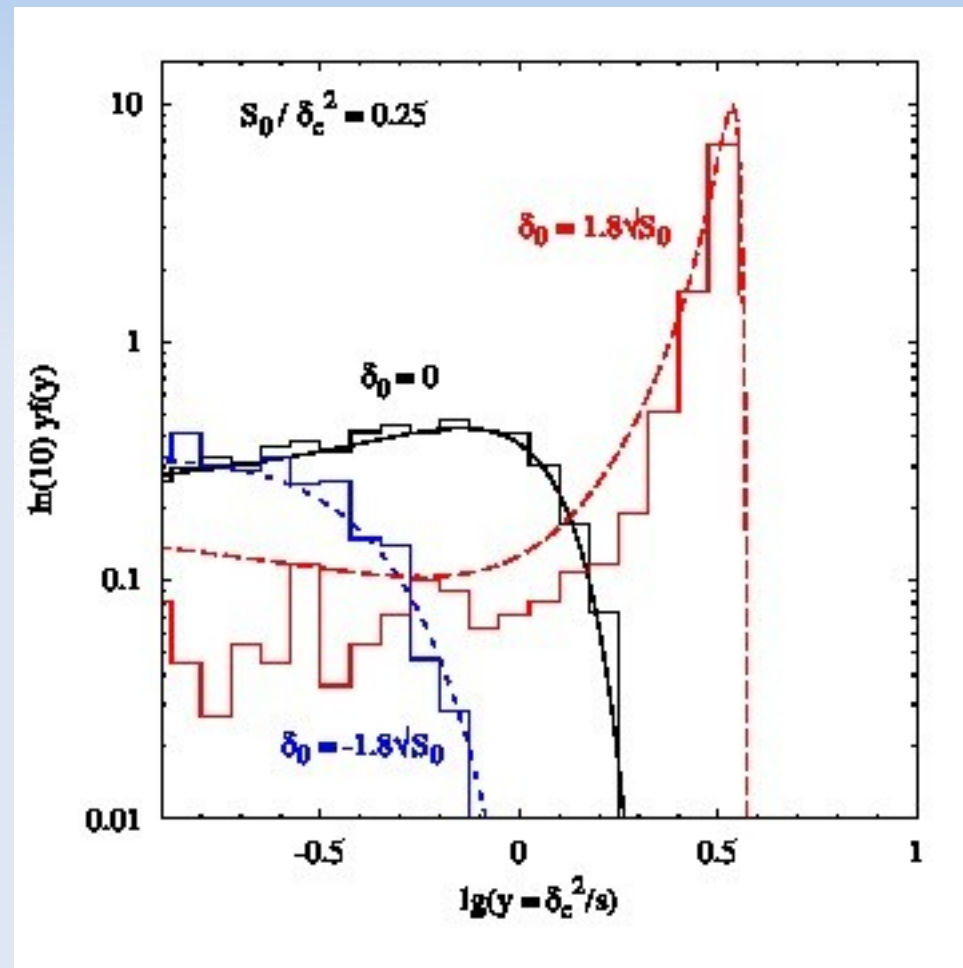
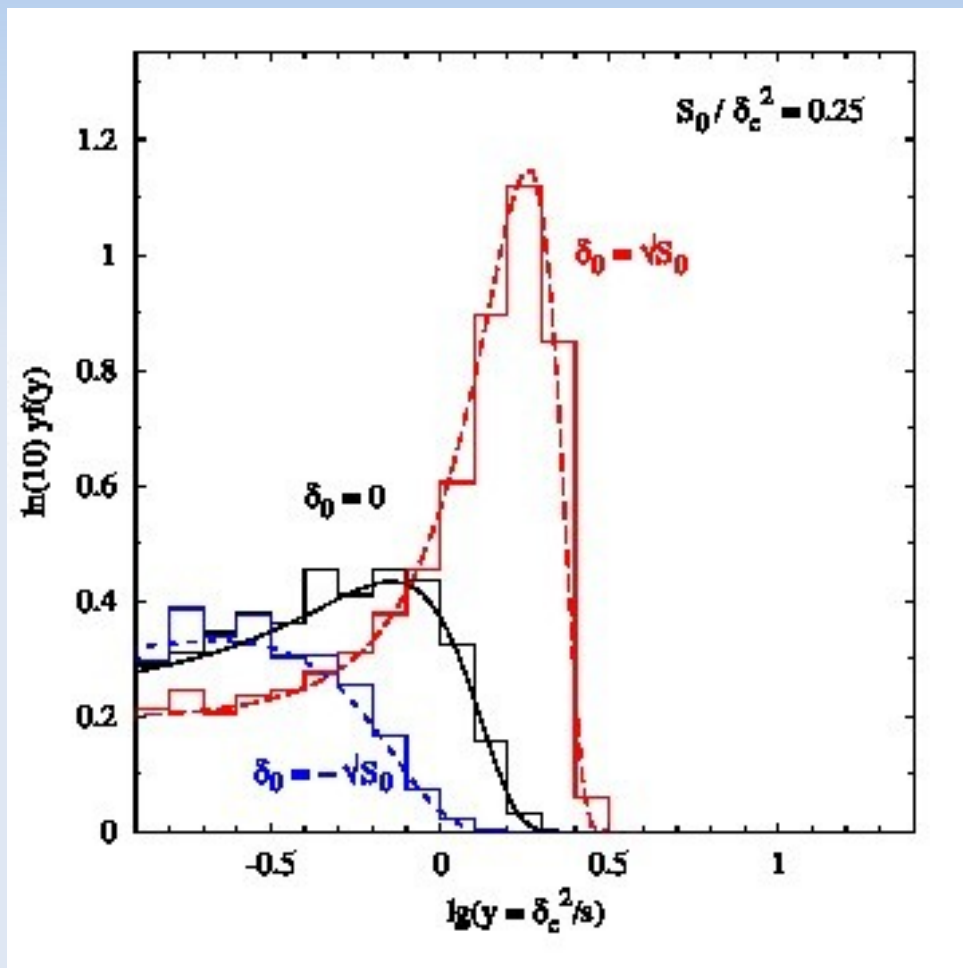
Conditional FC distribution

- Same formalism for walks that went through δ_0 at $s(R_0) = s_0$



MM, Paranjape & Sheth (2012)

Conditional FC distribution

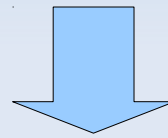


MM, Paranjape & Sheth (2012)

Halo Bias

- Suppose the most generic dependence on the matter field:

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



e.g. Matsubara (2011)

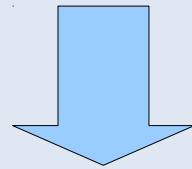
- Can compute connected correlation functions

$$\begin{aligned} & \langle \delta_h(\mathbf{x}) \delta_0(\mathbf{z}_1) \dots \delta_0(\mathbf{z}_n) \rangle_c \\ &= \int d^3\mathbf{x}_1 \dots d^3\mathbf{x}_n \underbrace{\left\langle \frac{\delta^n \delta_h(\mathbf{x})}{\delta\delta(\mathbf{x}_1) \dots \delta\delta(\mathbf{x}_n)} \right\rangle}_{\text{Bias functions}} \prod_{j=1}^n \langle \delta(\mathbf{x}_j) \delta_0(\mathbf{z}_j) \rangle \end{aligned}$$

Halo Bias from Excursion Sets

Define the halo random density field:

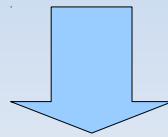
$$1 + \delta_h(x) = \frac{1}{f(s)\Delta s} \vartheta(B - \delta_1(x)) \cdots \vartheta(B - \delta_{N-1}(x)) \vartheta(\delta_N(x) - B)$$



$$\langle \delta_h \delta_0^n \rangle_c = \sum_{i_1, \dots, i_n=1}^N \left\langle \frac{\partial^n \delta_h}{\partial \delta_{i_1} \cdots \partial \delta_{i_n}} \right\rangle \langle \delta_{i_1} \delta_0 \rangle \cdots \langle \delta_{i_n} \delta_0 \rangle$$

Halo Bias from Excursion Sets

Two-step approx: $1 + \delta_h \simeq \frac{\vartheta(B - \delta_{N-1})\vartheta(\delta_N - B)}{f(s)\Delta s}$



$$\langle \delta_g \delta_0^n \rangle_c = \frac{(-1)^n}{f(s)} \int_0^\infty dv v \left(\langle \delta \delta_0 \rangle \frac{\partial}{\partial B} + \langle \delta' \delta_0 \rangle \frac{\partial}{\partial v} \right)^n p(B, v)$$

MM, Paranjape & Sheth (2012)

For instance:

$$\langle \delta_g \delta_0 \rangle = \langle \delta \delta_0 \rangle b_{10}^{(f)} + \langle \delta' \delta_0 \rangle 2s b_{11}^{(f)}$$

$$\langle \delta_g \delta_0^2 \rangle_c = \langle \delta \delta_0 \rangle^2 b_{20}^{(f)} + \langle \delta \delta_0 \rangle \langle \delta' \delta_0 \rangle 4s b_{21}^{(f)} + \langle \delta' \delta_0 \rangle^2 4s^2 b_{22}^{(f)}$$

Bias coefficients

- The highest coefficient looks the same:

$$b_{n0} = \frac{1}{f(s)} \left(-\frac{\partial}{\partial B} \right)^n f(s)$$

but now $f(s)$ is the result for correlated walks

- Moreover, there are non-trivial relations among the coefficients:

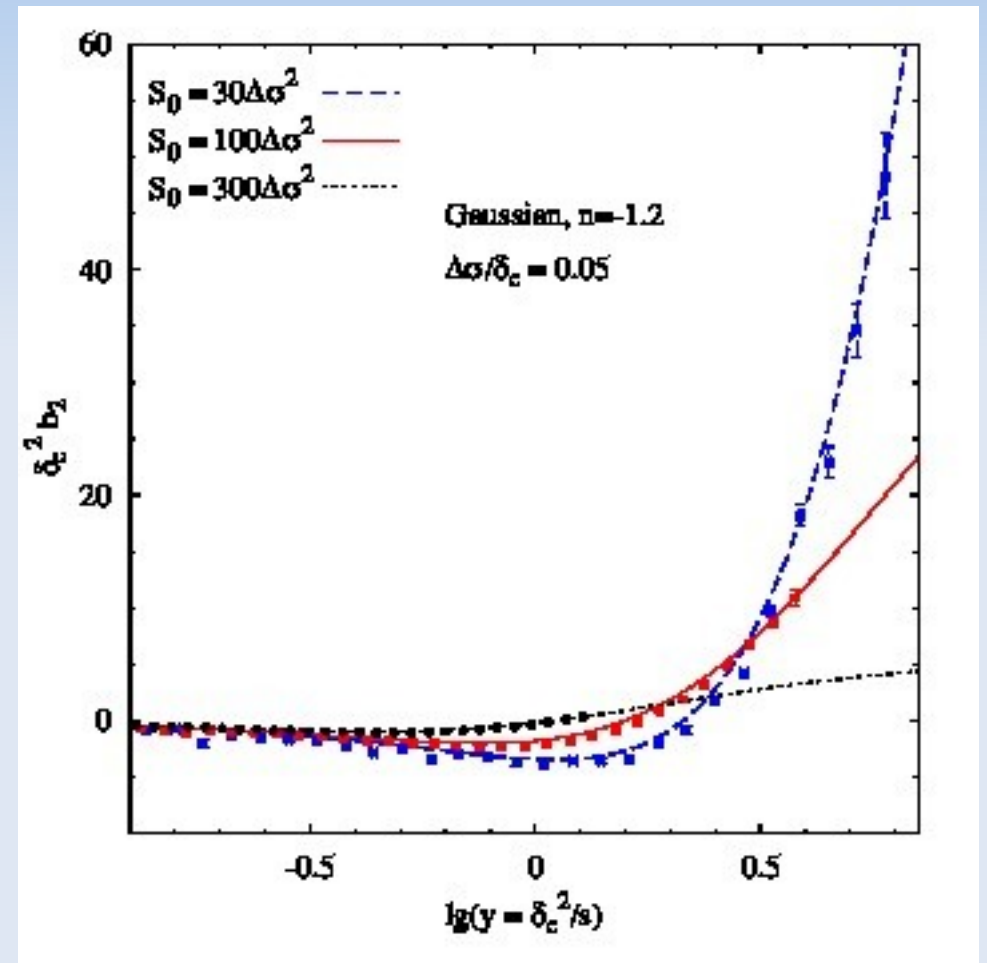
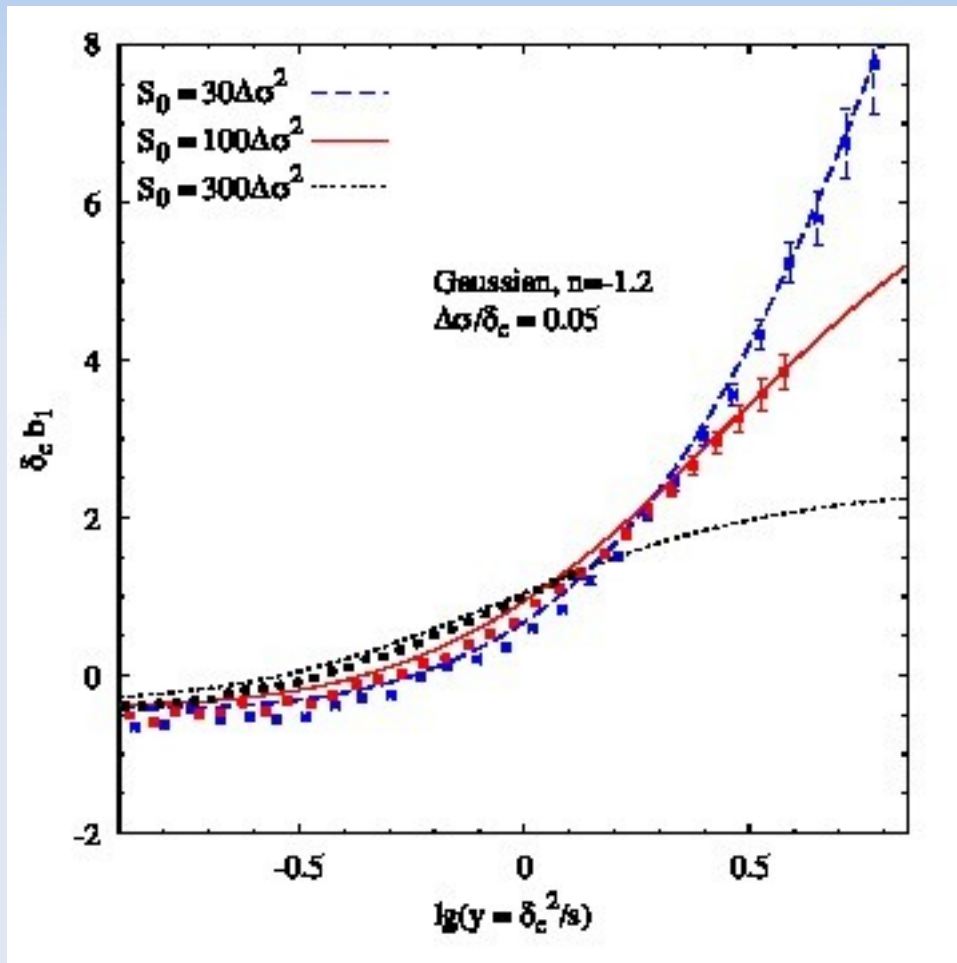
$$b_{11}B = \nu^2 - b_{10}B$$

$$b_{21}B^2 = \nu^2(b_{10}B - 1) - b_{20}B^2$$

$$b_{22}B^2 = b_{20}B^2 + \nu^2(\nu^2 - 2b_{10}B + 1)$$

same relations as from peak theory (see Desjacques et al 2010)

Bias coefficients



MM, Paranjape & Sheth (2012)

Fourier Space Bias

- Linear bias:

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

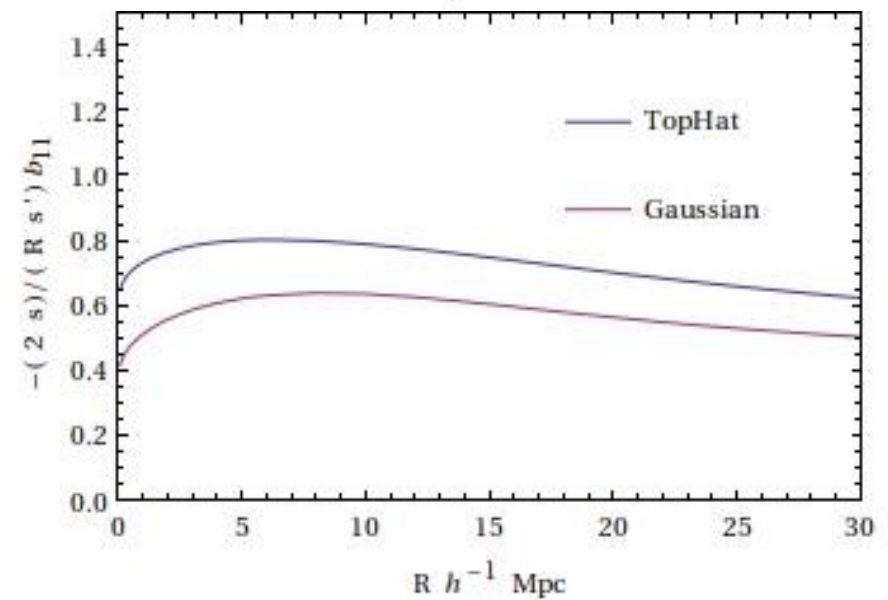
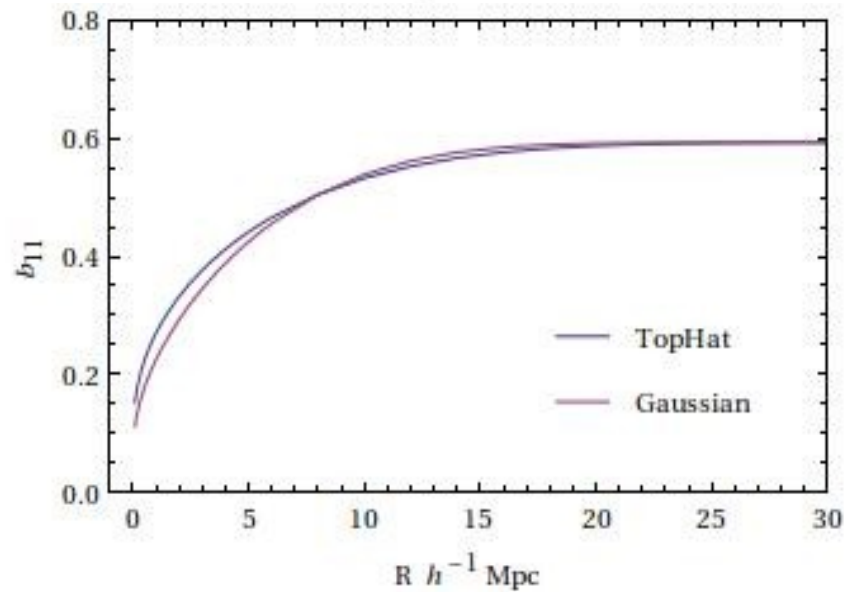
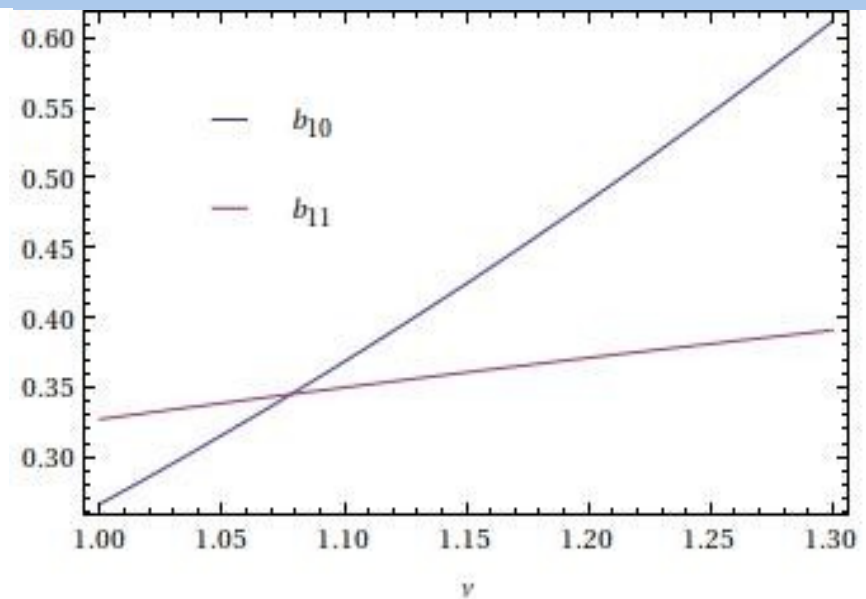
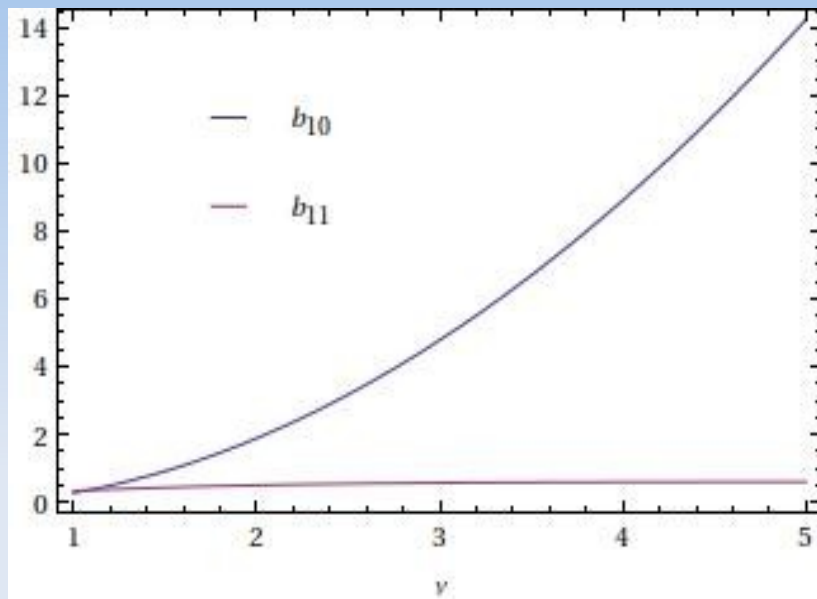
- Quadratic bias:

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

Gaussian bias is k -dependent too!

MM, Paranjape & Sheth (2012)

Fourier Space Bias



Adding non-Gaussianity: MF

Same formalism applies to non-Gaussian initial conditions

$$f(s) = \left[\frac{d}{ds} \int_{B(s)}^{\infty} d\delta p(\delta; s) \right] \frac{1 + \operatorname{erf}(\nu_*/\sqrt{2})}{2} + p(B; s) \left[\frac{e^{-\nu_*^2/2}}{2\Gamma\sqrt{2\pi s}} + \dots \right]$$

Full NG distribution same Gaussian factors

The diagram shows two arrows originating from the text below. One arrow points from 'Full NG distribution' to the term $\left[\frac{d}{ds} \int_{B(s)}^{\infty} d\delta p(\delta; s) \right]$ in the equation. The other arrow points from 'same Gaussian factors' to the term $p(B; s) \left[\frac{e^{-\nu_*^2/2}}{2\Gamma\sqrt{2\pi s}} + \dots \right]$ in the equation. The terms $p(\delta; s)$ and $p(B; s)$ in the equation are highlighted with blue circles.

Same dependence of f on $p(\delta; s)$ as in the Gaussian case

Adding non-Gaussianity: Bias

- Generic excursion set bias for non-Gaussian walks:

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

- With the two-step approximation:

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

Same dependence of b_{ij} on f and $p(\delta)$

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

Conclusions

- Accurate approximation of correlated random walks
- Systematic treatment of corrections
- (First?) satisfactory mathematical understanding of the excursion set approach to structure formation
- Straightforward non-perturbative inclusion of NG
- Predictions for bias coefficients, new strategies to measure them in simulations
- To do: check against N-body simulations
 - generalisation to excursion set theory of peaks
- Other applications: virus diffusion, stock market, ...

THANKS!