

# Anomalous Diffusion

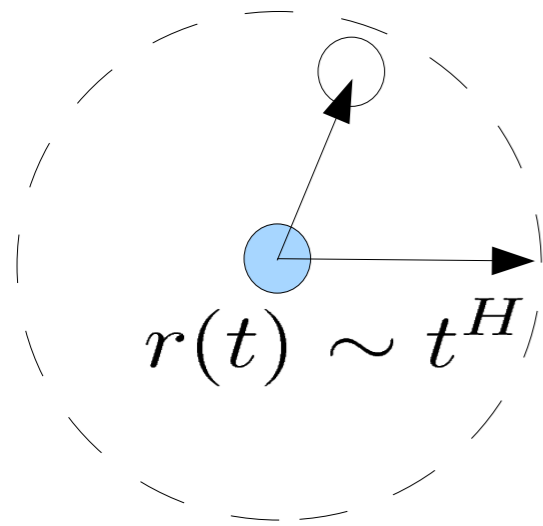
Alberto Rosso (LPTMS-Orsay)

Polymer translocation: A. Zoia (Saclay) + S. Majumdar

Perturbation theory: K. J. Wiese (ENS) + S. Majumdar

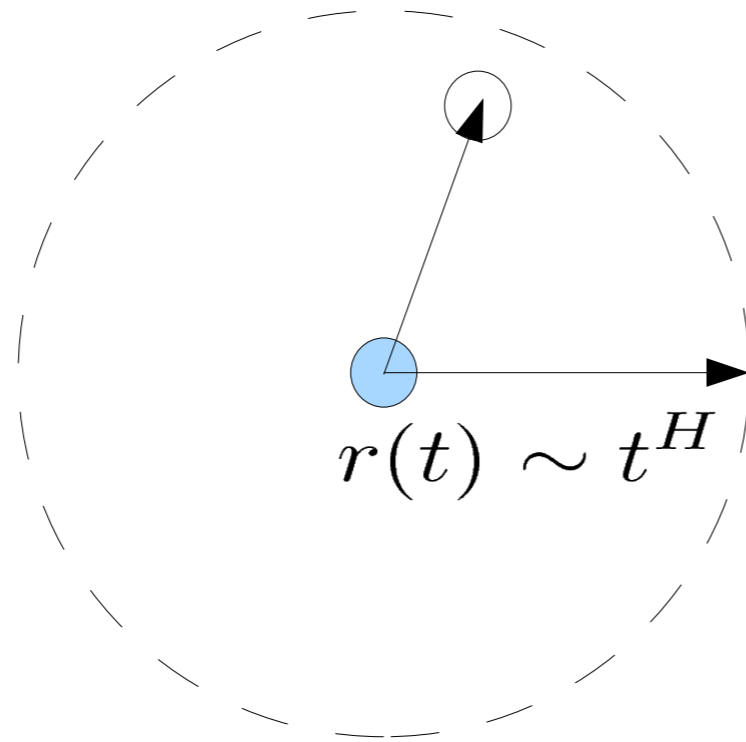
Longest excursion: R. Garcia (Bariloche) + G. Schehr

# Anomalous Diffusion



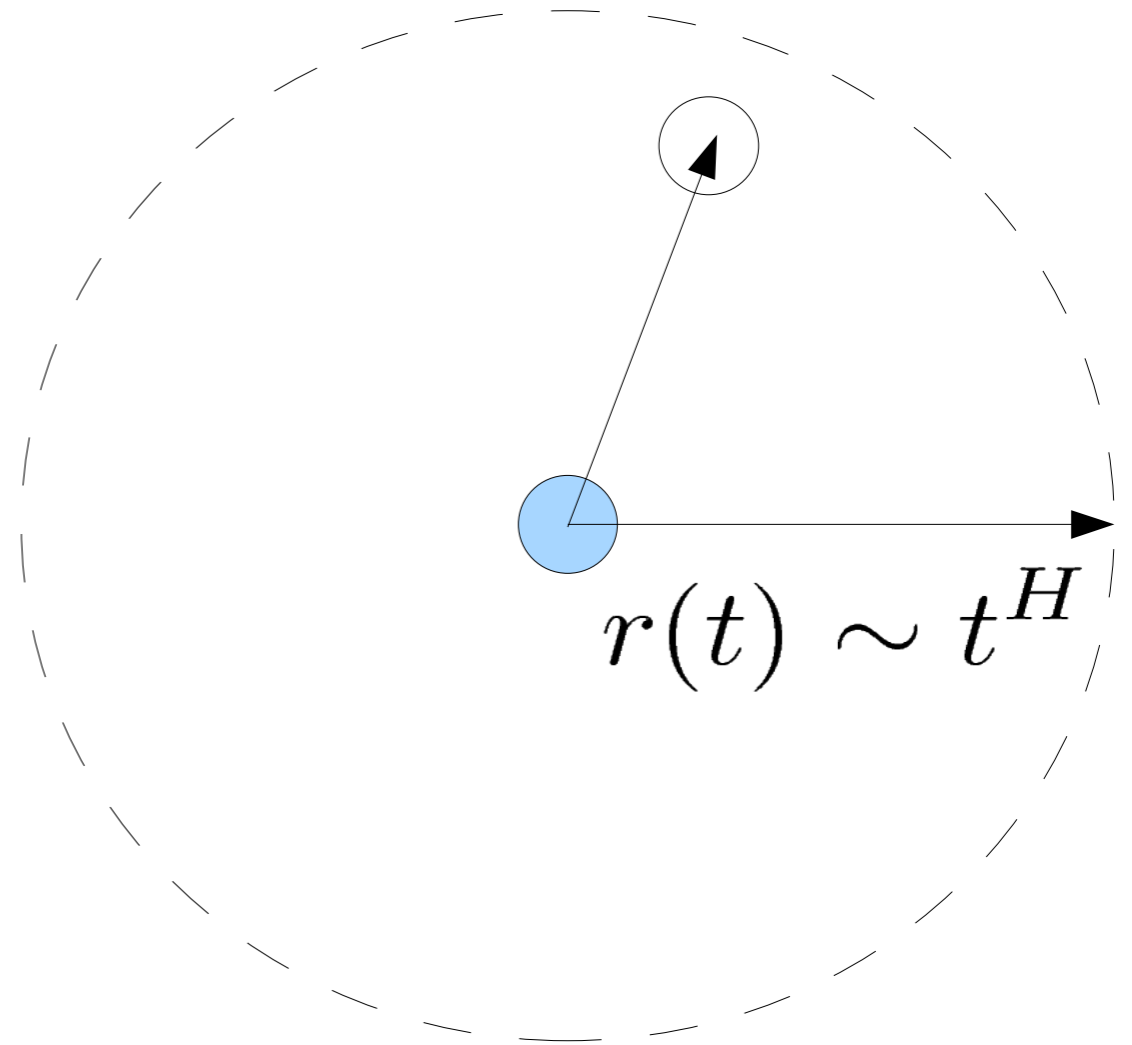
$$H < 1/2$$

Subdiffusion



$$H = 1/2$$

Diffusion

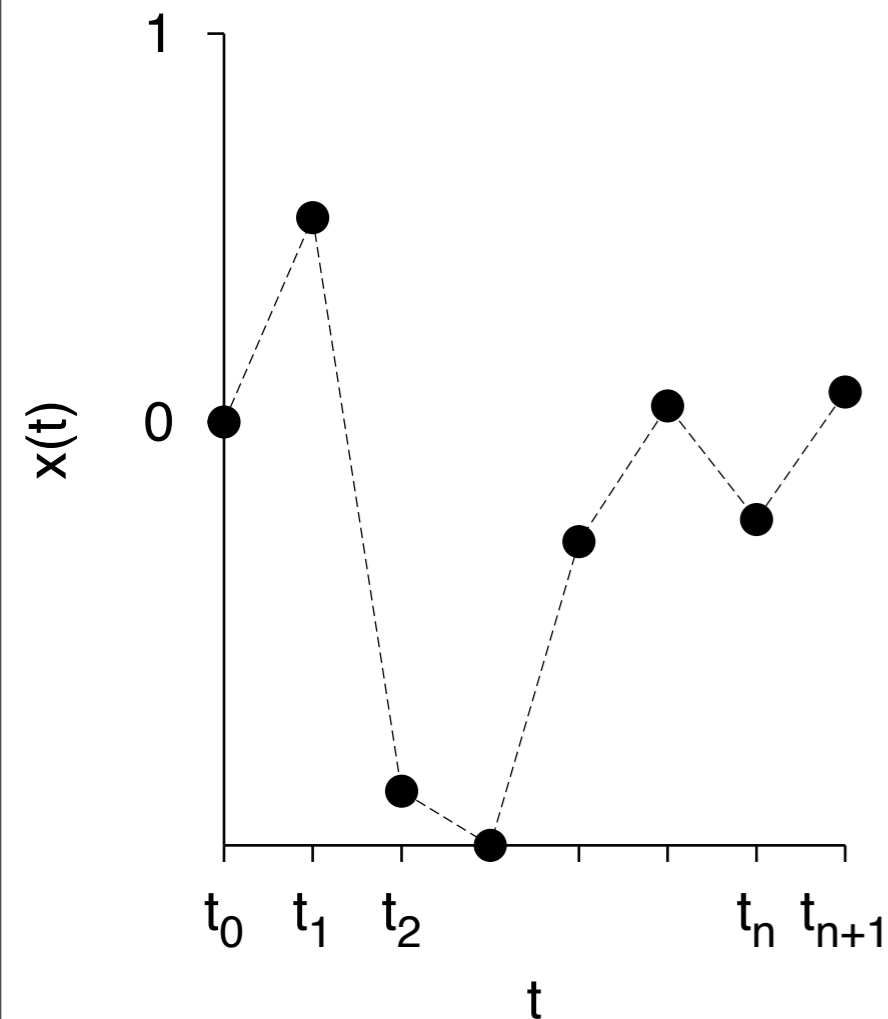


$$H > 1/2$$

Super-diffusion

# Central Limit Theorem

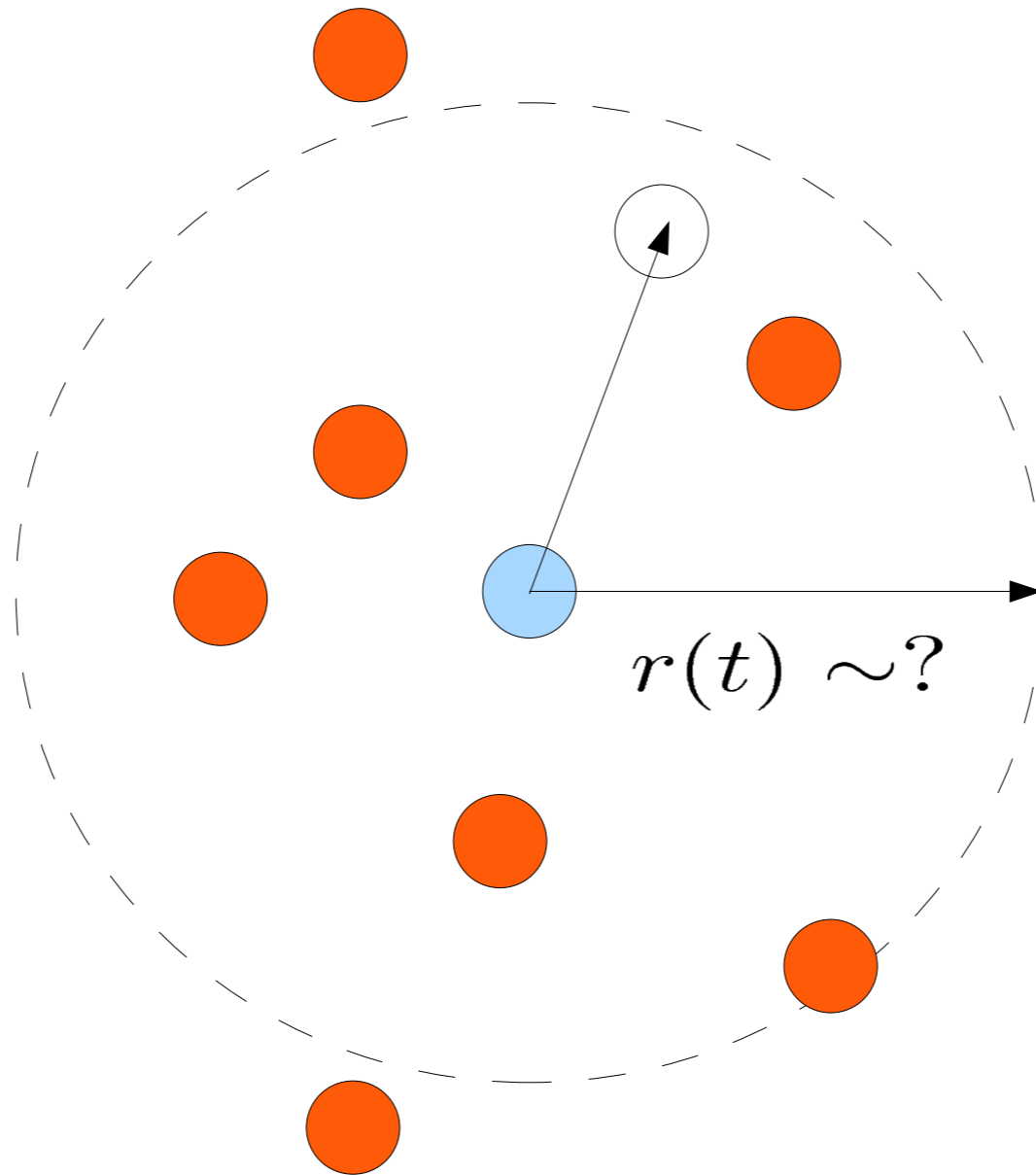
$$x(t) = \sum_{t'=1}^t \xi_{t'}$$



- *Identical*:  $\pi_{t'}(\xi) = \pi(\xi)$ . (Homogeneous)
- *Independent*:  $\langle \xi_{t_0} \xi_{t_0+t} \rangle = 0$ . (Markov)
- $\langle \xi^2 \rangle < \infty$ : Continuous process.

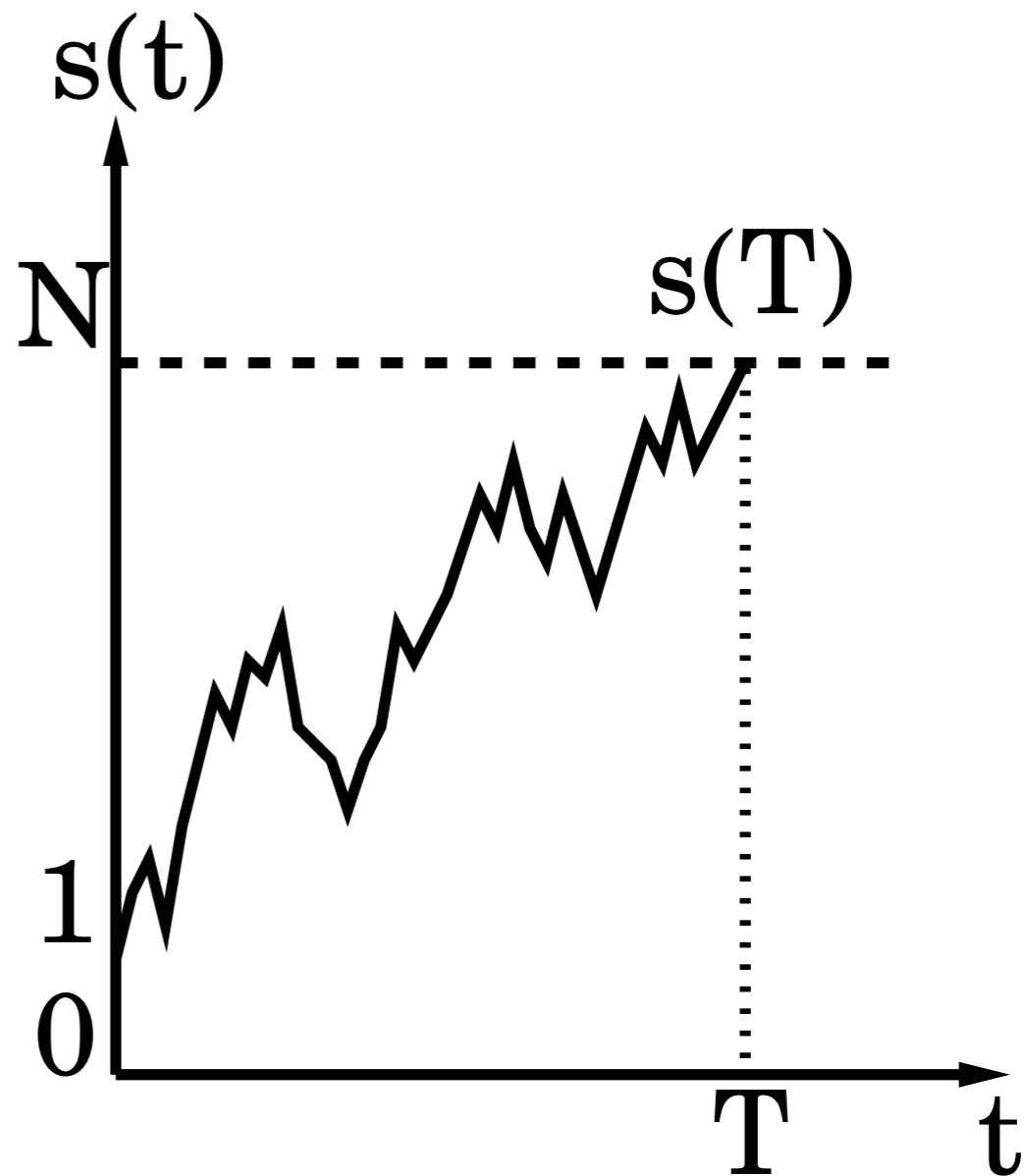
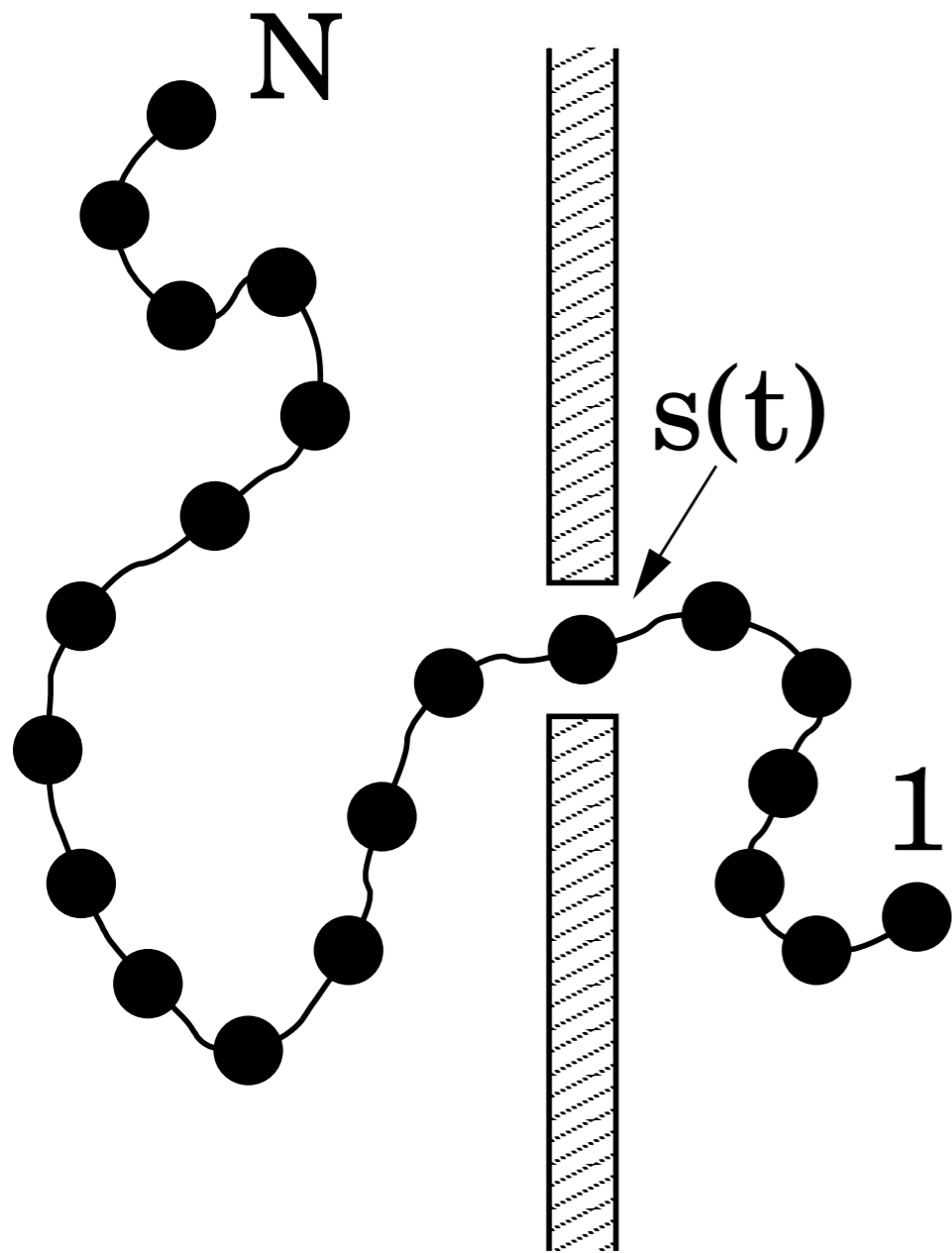
**Conclusion:**  $x(t)$  is Gaussian and  $x(t) \sim \sqrt{t}$

# Correlations

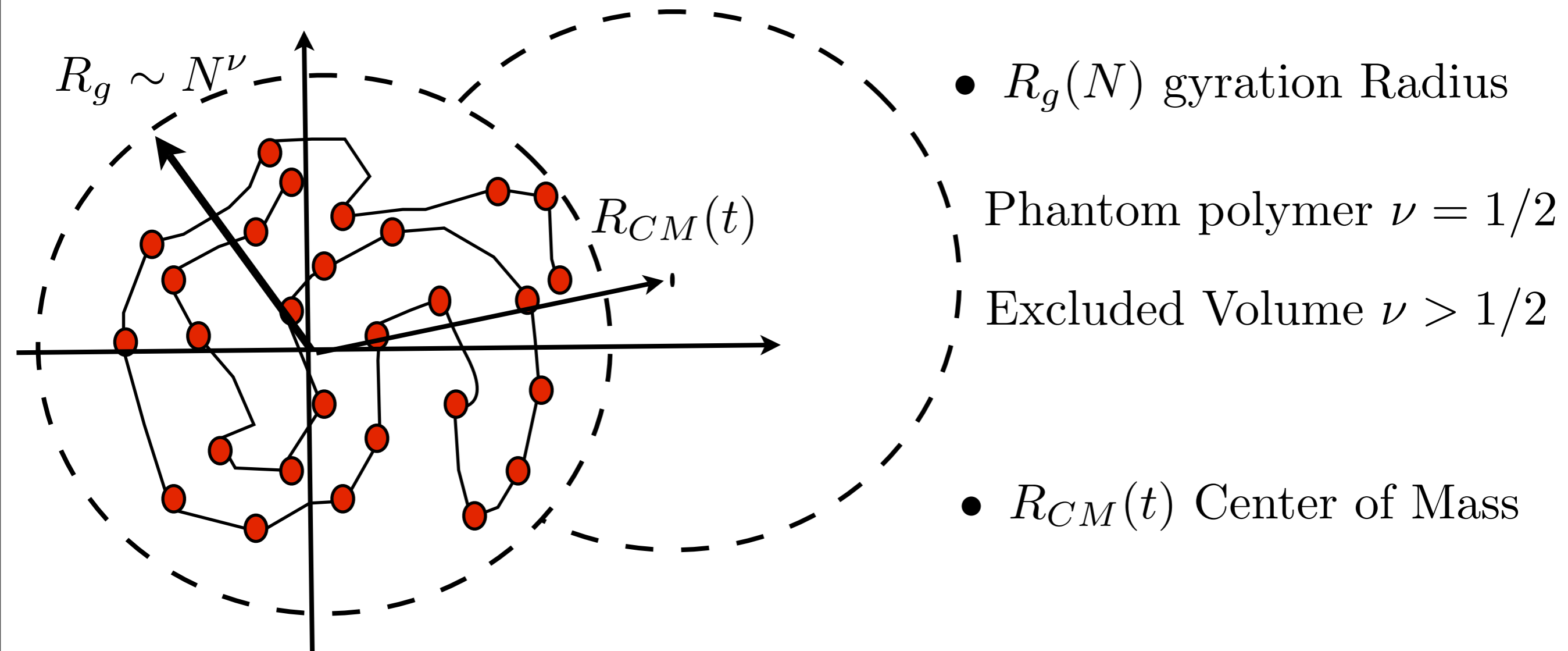


- jumps and waiting times are local
- colloids interact (strongly non-Markovian)

# Polymer Translocation



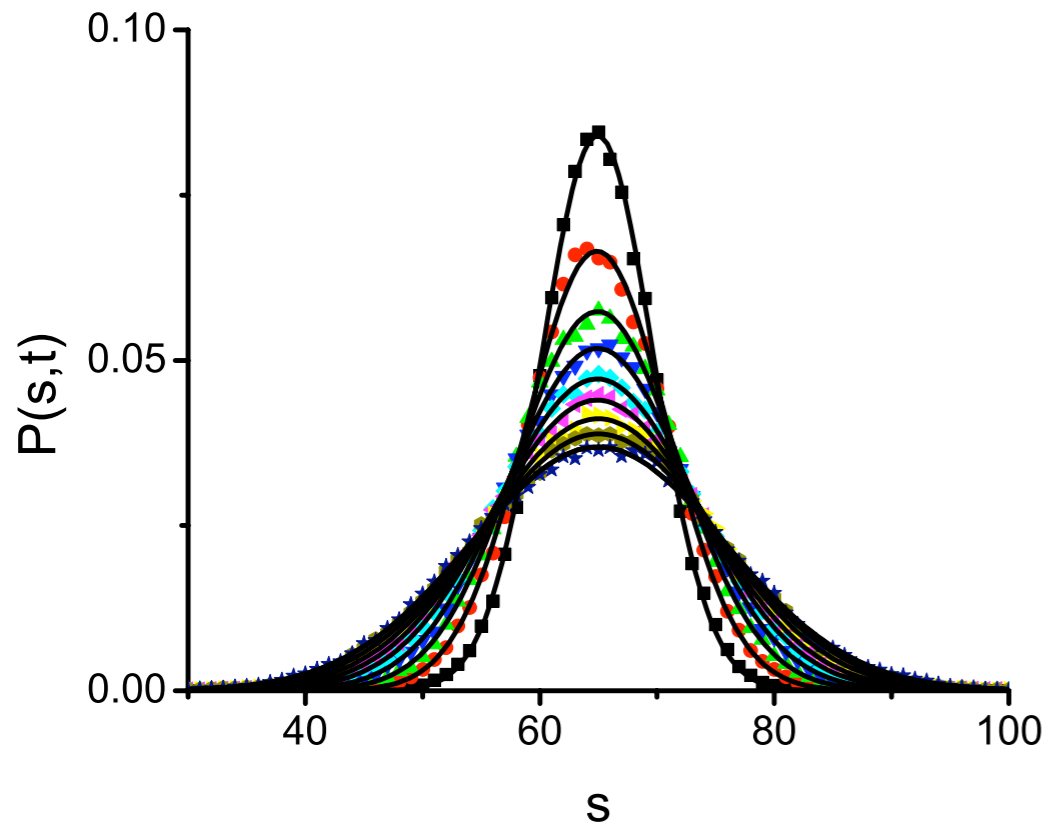
$$s(T) = N, \text{ if } s(t) \sim t^H \text{ then } T \sim N^{1/H}$$



$$\langle R_{CM}^2(t_{\text{eq.}}) \rangle = R_g^2 \quad \Rightarrow \quad \frac{D_1 t_{\text{eq.}}}{N} \sim N^{2\nu} \quad \Rightarrow \quad t_{\text{eq.}} \sim N^{2\nu+1}$$

$$T \gg t_{\text{eq.}} \quad \Rightarrow \quad H \leq \frac{1}{1+2\nu} \quad \text{numerically} \quad H \approx \frac{1}{1+2\nu}$$

# Fractional Brownian motion



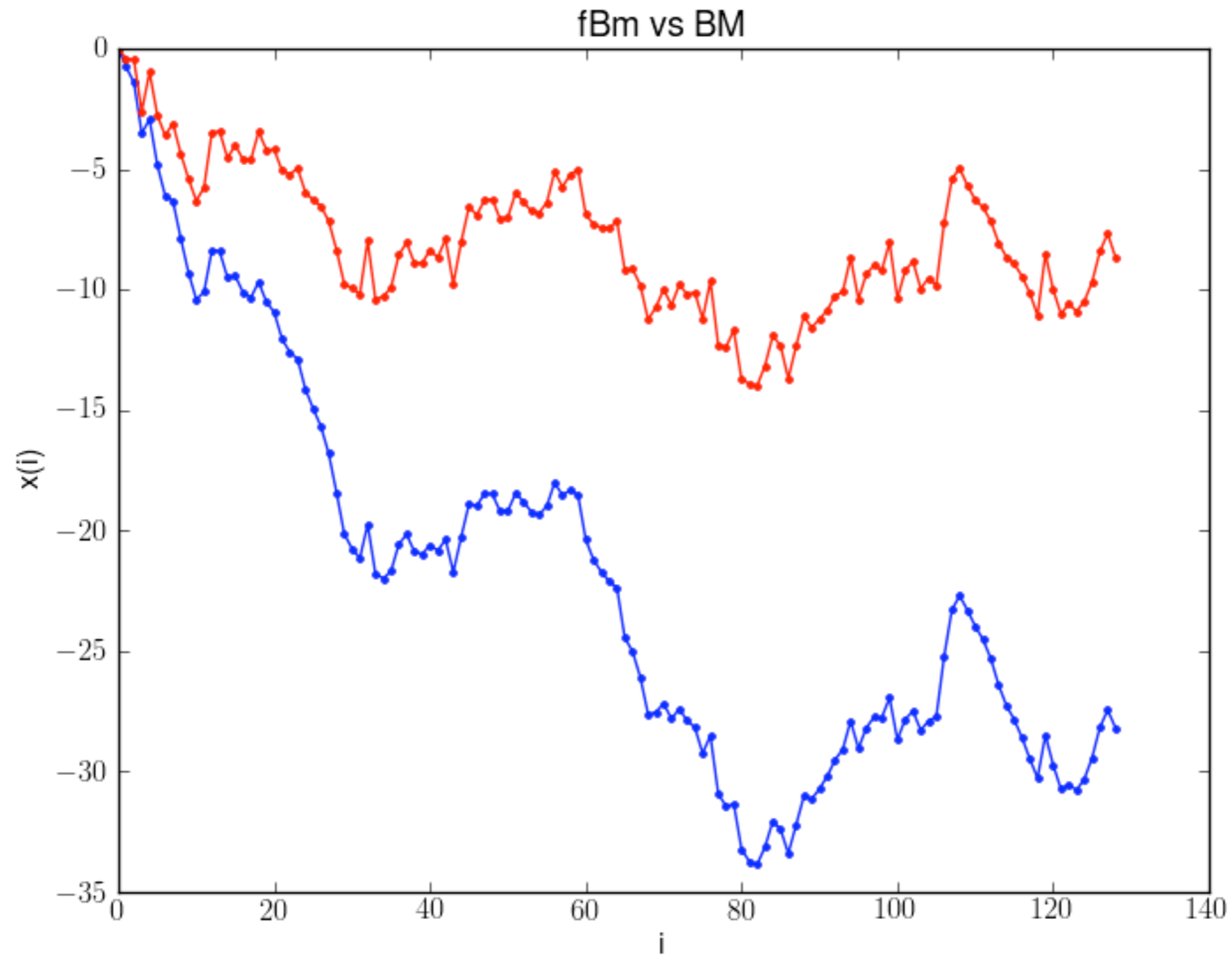
Monte Carlo simulation of polymer translocation in  $d=2$ ,  
Chatelain, Kantor, Kardar, PRE 78, 021129 (2008)

- Equilibrium with the solvent  
 $\Rightarrow$  stationary increments
- is a Gaussian process  
 $\Rightarrow$  local jumps
- $\Rightarrow$  non-Markov process

Gaussianity + self-affinity  $H = \frac{1}{2\nu+1}$  + stationary increments  
 $\Rightarrow$  fractional Brownian motion:

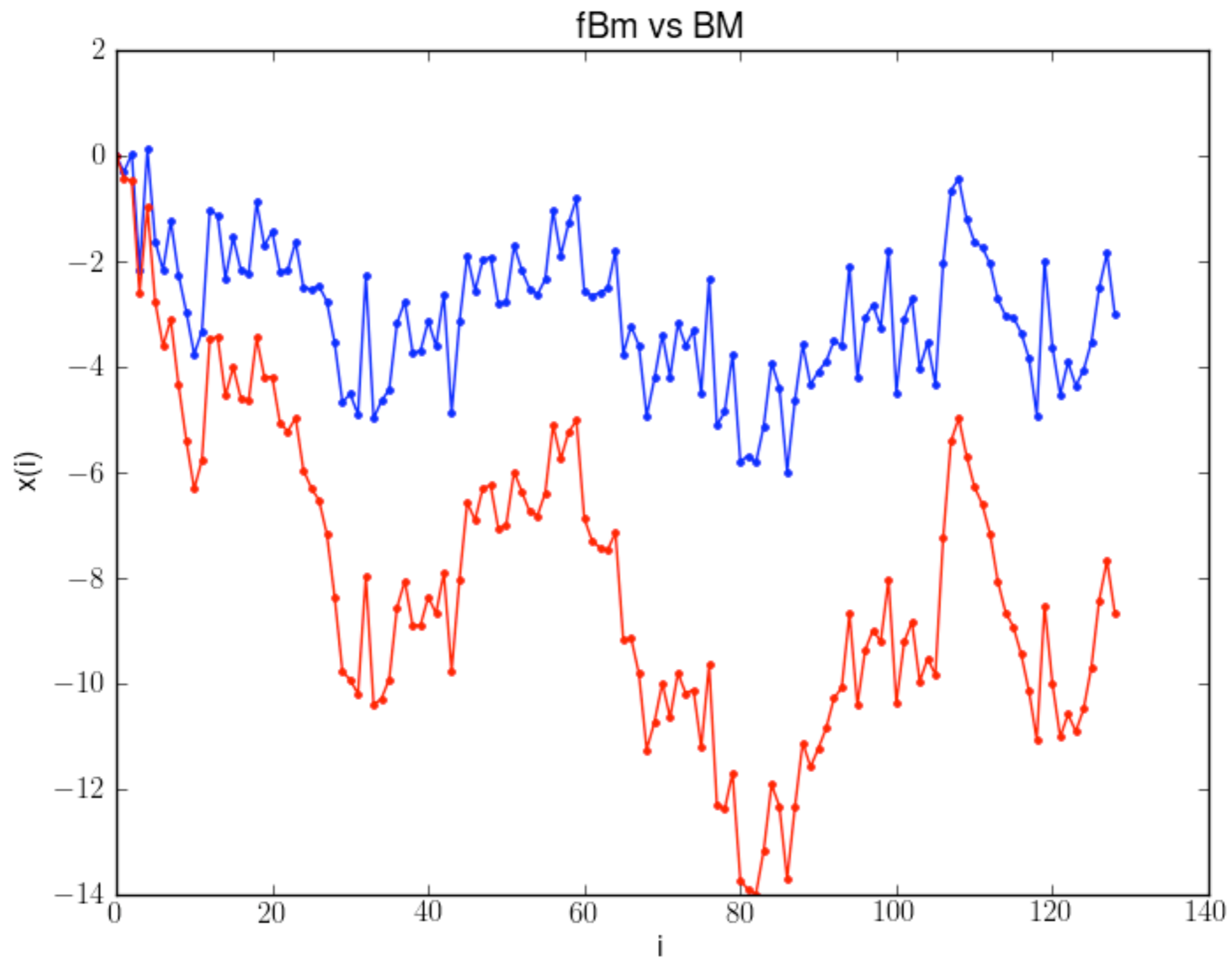
$$\langle s(t_1)s(t_2) \rangle \propto (t_1^{2H} + t_2^{2H} - |t_1 - t_2|^{2H}),$$

# H=3/4 Superdiffusion





# H=1/4 Subdiffusion

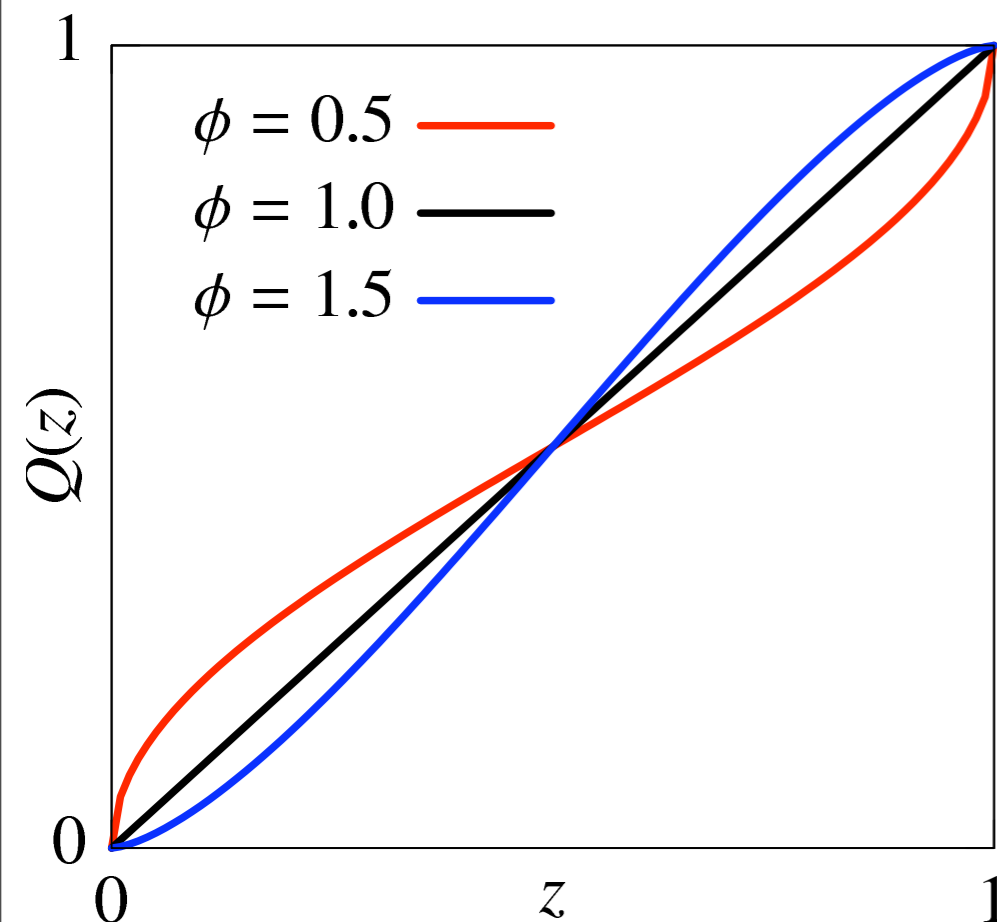
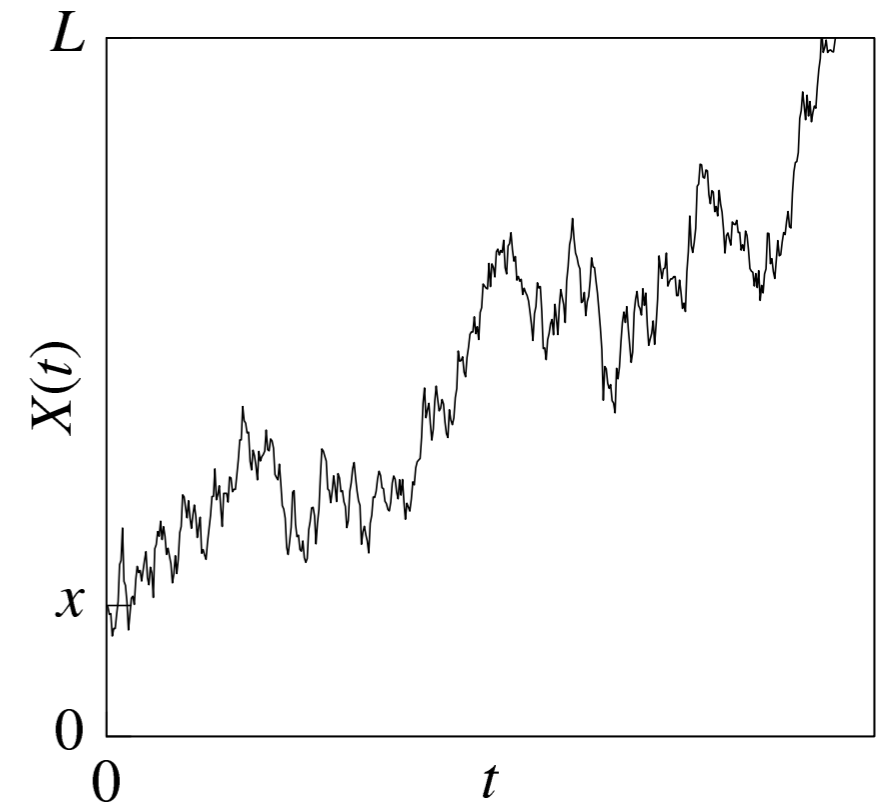


# Question I: A polymer chain will ultimately succeed in translocating through a pore ?

Hitting probability  $Q(x, L)$ :  
probability of exiting through  $L$

For self affine processes:

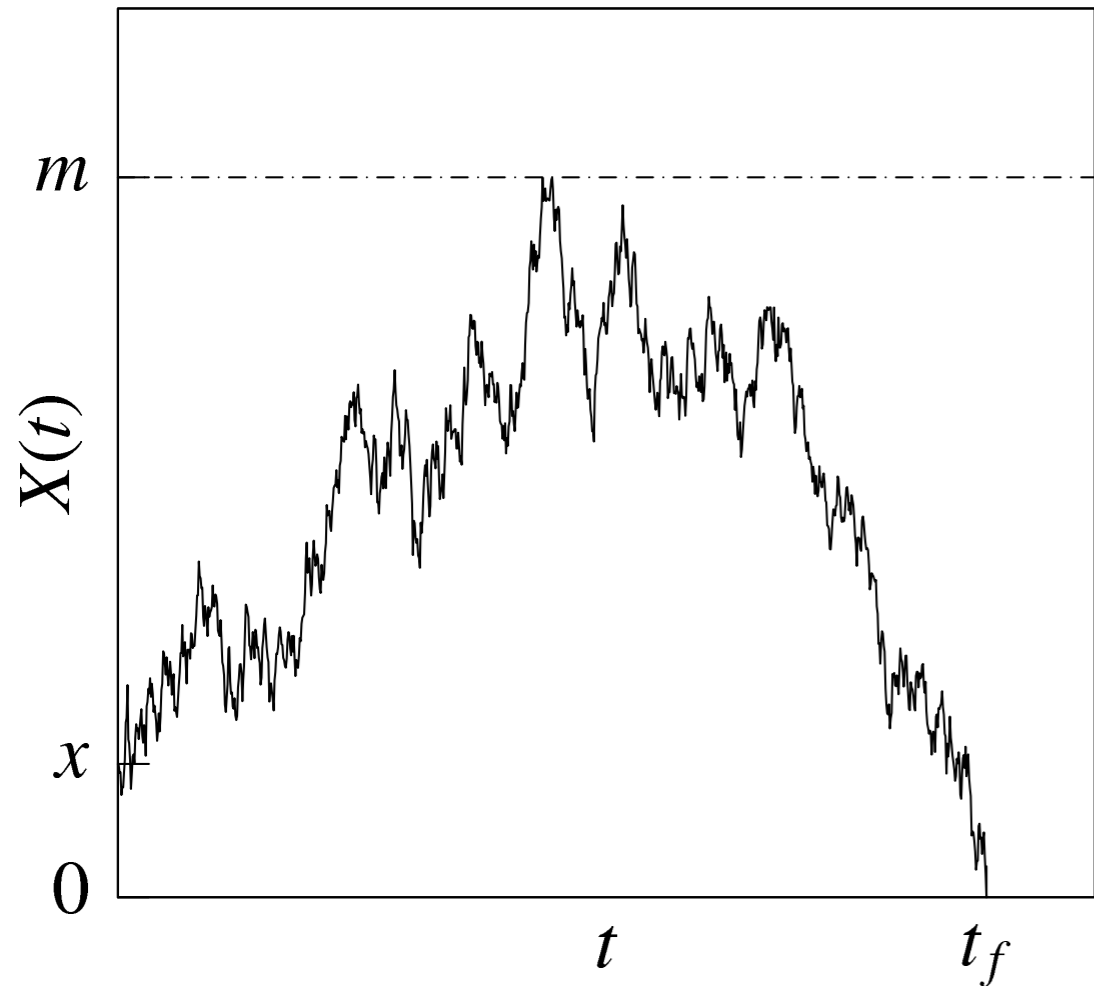
$$Q(x, L) = Q\left(z = \frac{x}{L}\right)$$



- Basic:  $Q(z) = 1 - Q(1 - z)$ ;  $Q(1/2) = 1/2$ ;  $Q(z) = 0$
- $H = 1/2$ :  $Q(z) = z$
- Expansion:  $Q(z) \sim c_1 z^\phi + \dots$

Translocation is enhanced or suppressed  
by excluded volume effects?

# A scaling argument: $\phi = \frac{\theta}{H}$



- $Q(x, L) = \text{Prob}[m > L]$

- Self-affinity I:  $m \sim t_f^H$

$$\implies Q(x, L) \simeq \text{Prob}[t_f^H > L]$$

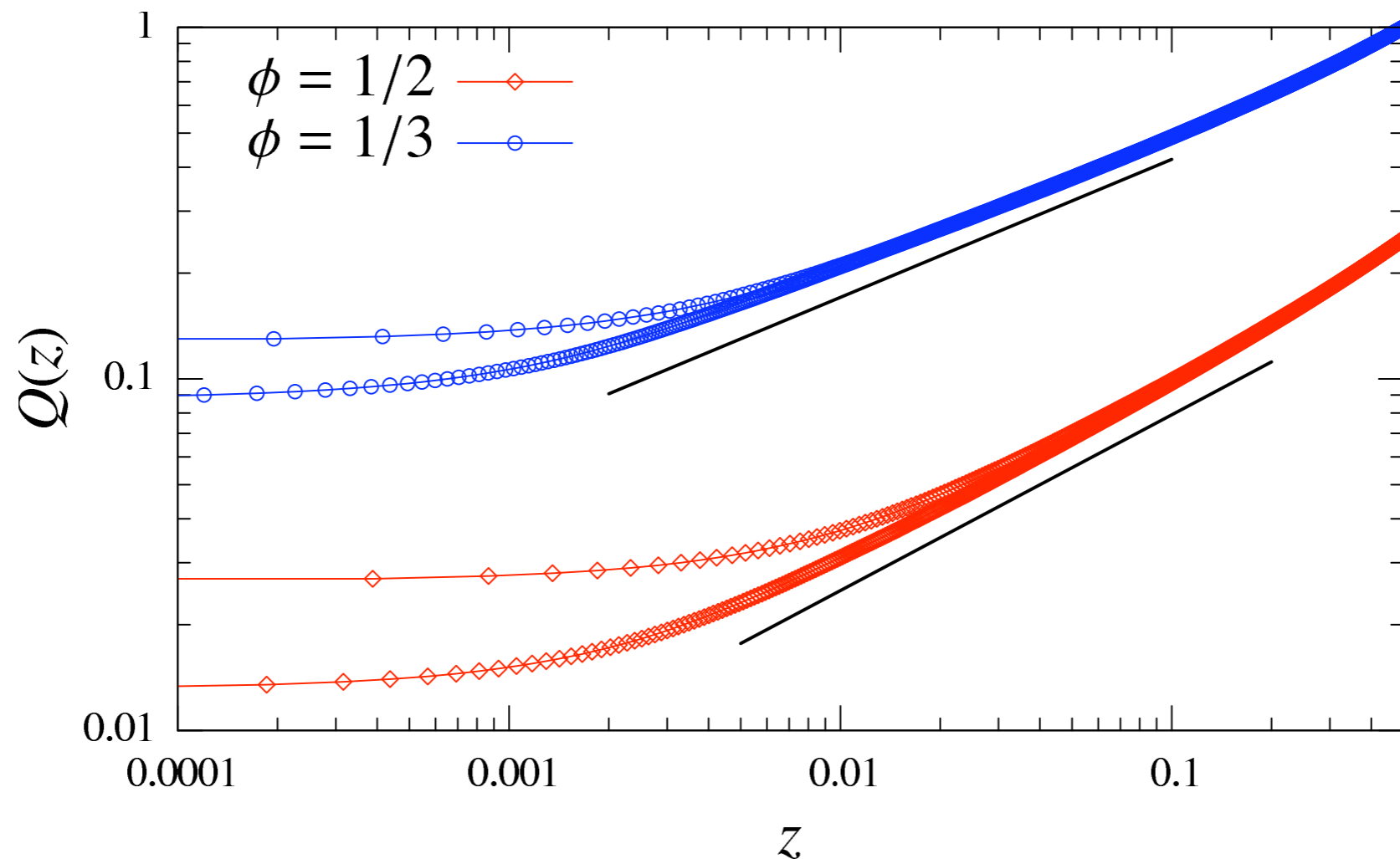
$$Q(x, L) \sim \text{Prob}[t_f > L^{\frac{1}{H}}]$$

- Survival probability:  $\text{Prob}[t_f > T] = S(x, T) \sim \frac{f(x)}{T^\theta}$   
 $\theta$  persistence exponent

- Self-affinity II:  $S(x, T) \sim \left(\frac{x}{T^H}\right)^{\frac{\theta}{H}}$

$$\implies Q(x, L) \sim \text{Prob}[t_f > L^{\frac{1}{H}}] = S(x, L^{\frac{1}{H}}) \sim \left(\frac{x}{L}\right)^{\frac{\theta}{H}}, \quad \phi = \frac{\theta}{H}$$

# Hitting probability: numerical test



Persistence of fBm in known  $\theta = 1 - H$  (see Krug et al.)

Prediction:  $\phi = \frac{\theta}{H} = \frac{1-H}{H}$

- Blue:  $H = 3/4 \longrightarrow \phi = 1/3$
- Red:  $H = 2/3 \longrightarrow \phi = 1/2$

**Conclusion: volume effects “suppress” Translocation**

# Previous results... recast using $\phi$

(i) Random acceleration process:  $\ddot{x} = \eta(t)$ ,  $x(0) = x$ ,  $\dot{x}(0) = 0$

Burkhardt:  $H = \frac{3}{2}$  and  $\theta = \frac{1}{4} \implies \phi = \frac{1}{6}$

$$Q(z = \frac{x}{L}) = I_z[\frac{1}{6}, \frac{1}{6}]. \quad \text{Where } I_z(\phi, \phi) = \frac{\Gamma(2\phi)}{\Gamma^2(\phi)} \int_0^z \frac{du}{[u(1-u)]^{1-\phi}}$$

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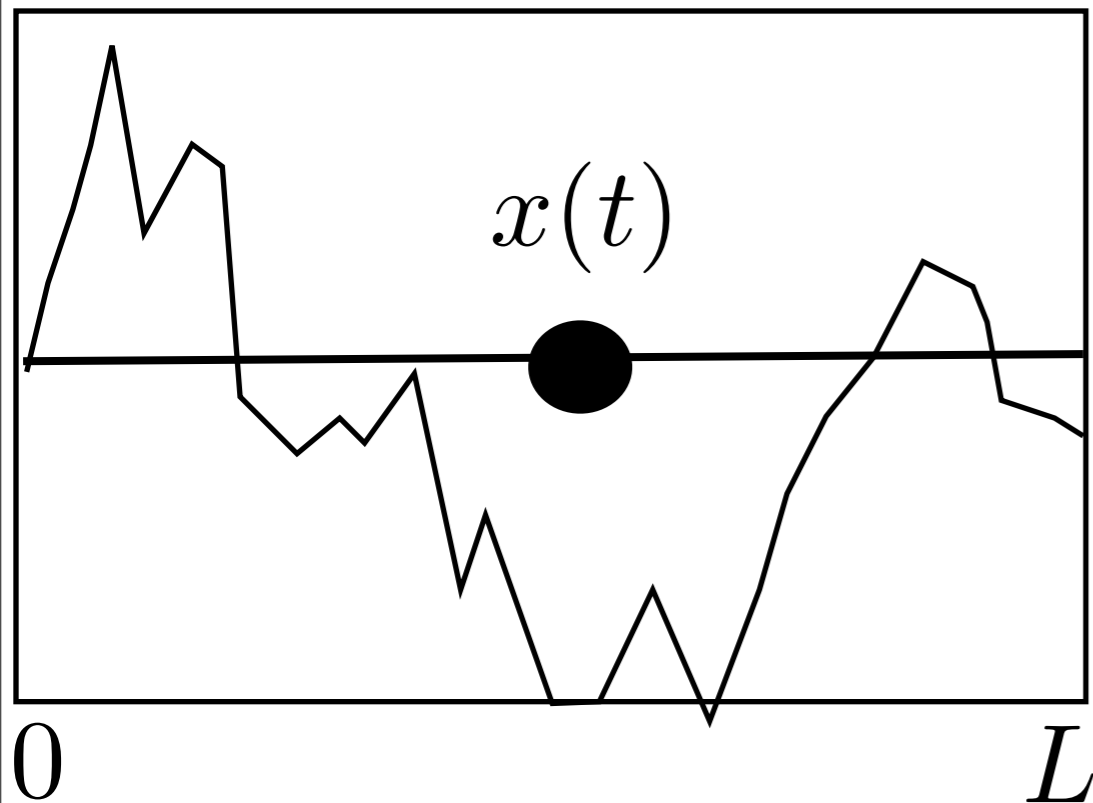
(ii) Lévy flights:  $x_{t+1} = x_t + \xi_t$  with  $\pi(\xi) \sim \xi^{-(\mu+1)}$

Sparre Andersen:  $H = \frac{1}{\mu}$  and  $\theta = \frac{1}{2} \implies \phi = \frac{\mu}{2}$

Widom ('61):  $Q(z = \frac{x}{L}) = I_z[\frac{\mu}{2}, \frac{\mu}{2}]$

# Sinai model

$V(x)$



$$t \sim e^{V(x)} \sim e^{\sqrt{x}} \Rightarrow x(t) \sim [\log(t)]^2$$

$$S(x_0, t) \sim [\log(t)]^{-1}$$

Using  $\tau = \log(t) \implies H = 2$  and  $\theta = 1$

From Backward Fokker Planck:  $Q(x, L) = \frac{\int_0^x e^{\beta V(x')} dx'}{\int_0^L e^{\beta V(x')} dx'}$

$$\overline{Q(z)} = \int_0^z dz' \overline{p_{eq}(z')}; \quad \text{From } \overline{p_{eq}(z)} = \frac{1}{\pi} \frac{1}{\sqrt{z(1-z)}}$$

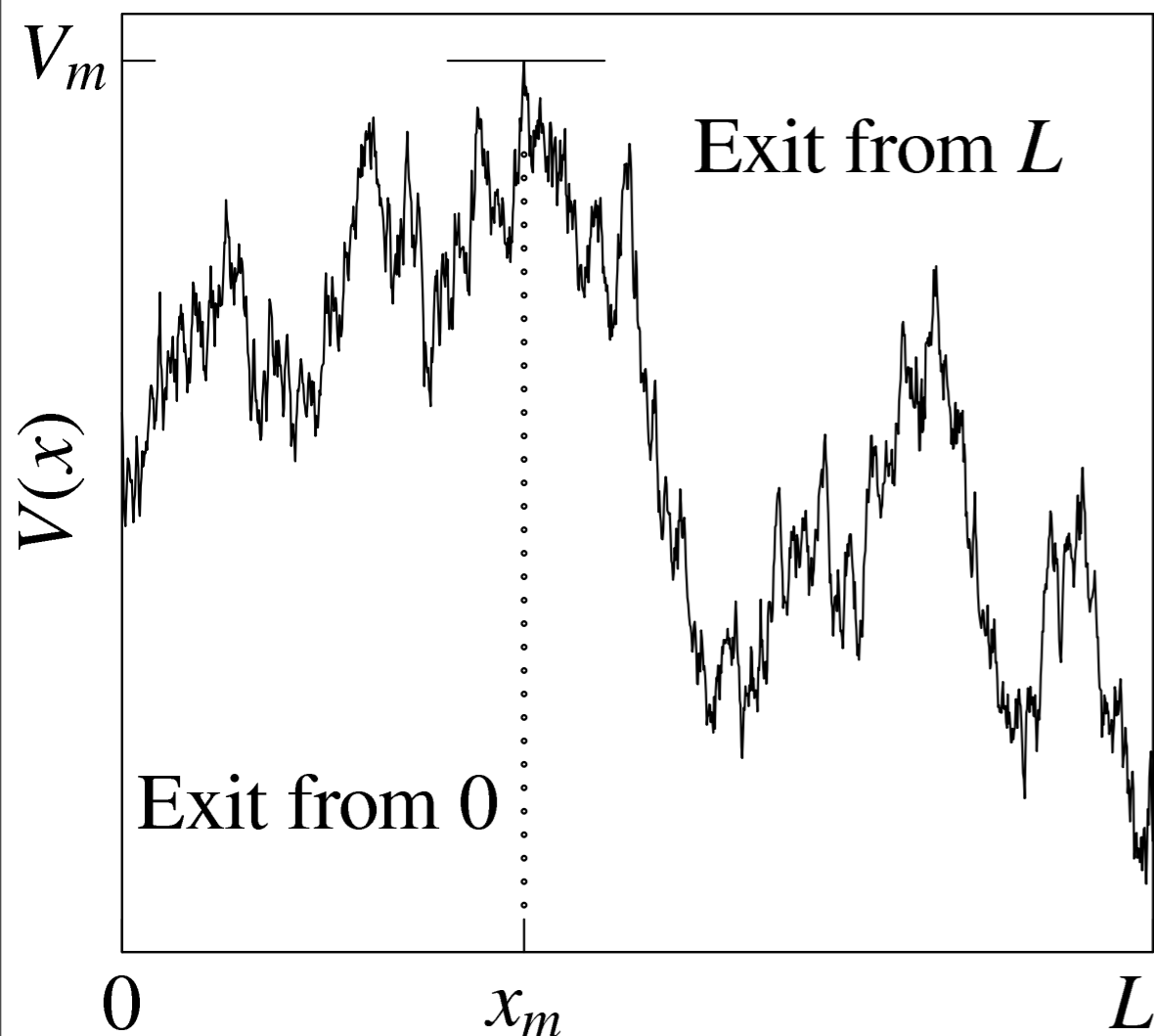
We deduce  $\overline{Q(z)} = I_z(1/2, 1/2) = \frac{2}{\pi} \arcsin(\sqrt{z}) \sim \sqrt{z}$

So that  $\phi = \theta/H = 1/2$

# Extreme statistics: maximum location in $V(x) \sim x^{H_V}$

$$p_{eq}(x, L) = \frac{e^{\beta V(x)}}{\int_0^L e^{\beta V(x')} dx'} \propto \left\{ \int_0^1 dz' \exp [\beta L^{H_V} \cdot (V(z') - V(z))] \right\}^{-1}$$

$$p_{eq}(x, L) \sim \delta(x - x_m) \implies \overline{p_{eq}(x, L)} \sim p_V(x_m, L) \quad [\text{Sinai: arcsine law}]$$



Fixed realization:  $Q(x, L) = \theta(x - x_m)$

Average  $\overline{Q(z = \frac{x}{L})} = \text{Prob}[z_m < z]$

If  $V(x) \sim x^{H_V}$ , using Arrhenius  $H = 1/H_V$

$$\text{Prob}[x_m < x \rightarrow 0] \sim \text{Prob}[V < 0 \text{ up to } L] \sim \frac{1}{L^{\theta_V}}$$

$$\text{So that } Q(x, L) \rightarrow \left(\frac{x}{L}\right)^{\theta_V}$$

**We Conclude  $\phi = \theta_V$  and  $\theta = H \cdot \phi = \theta_V / H_V$**

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$V(x)$  is a random acceleration process:  $H_V = 3/2$ ,  $\theta_V = 1/4$

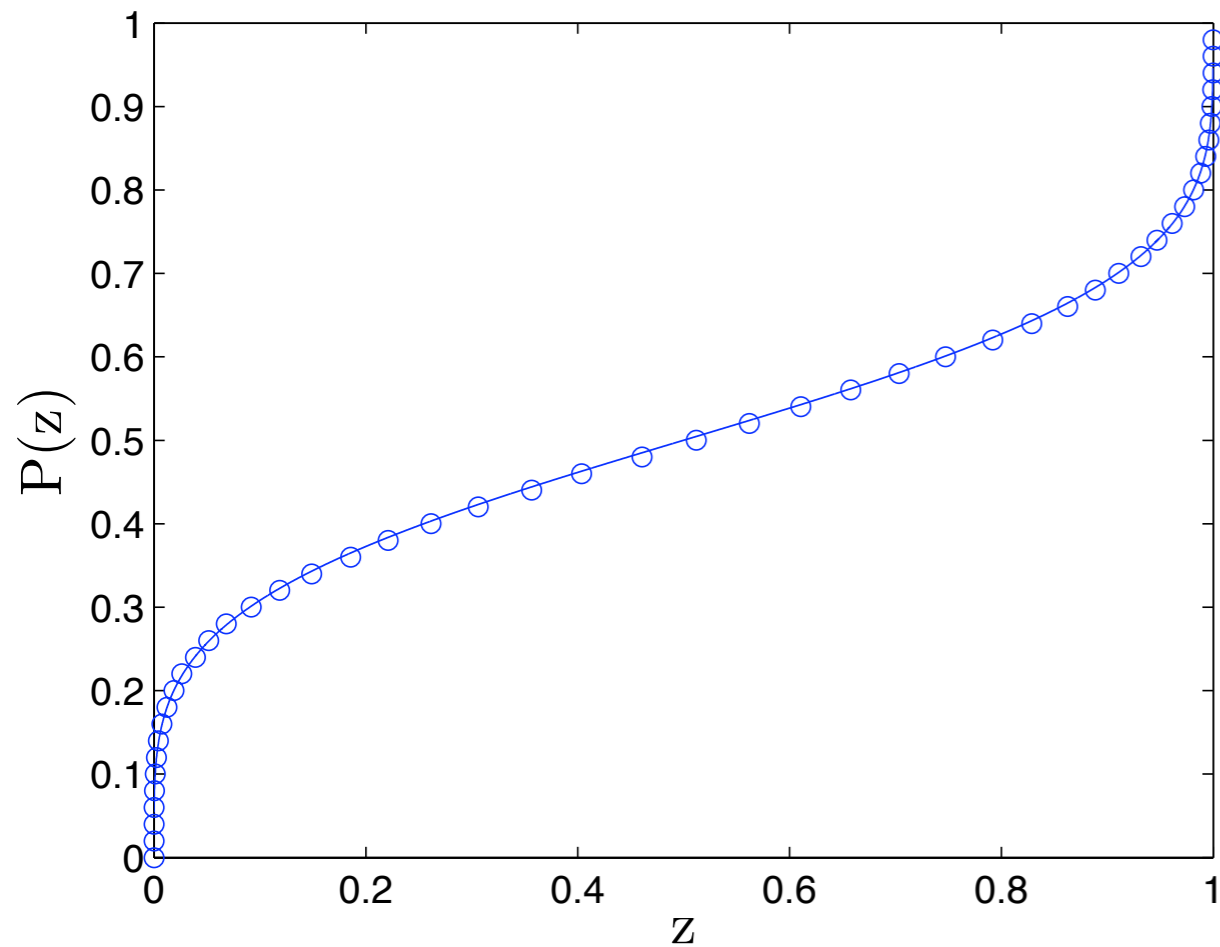
$$\text{Bridge case } (V'(L) = 0): \quad p(z_m) = \frac{\Gamma(1/2)}{\Gamma^2(1/4)} \frac{1}{(z_m(1-z_m))^{3/4}}$$

$$\text{Free case: } p(z_m) = \left(1 - \sqrt{\frac{3}{8}}\right) \delta(z_m - 1) + \frac{\sqrt{3}}{4\pi z_m^{3/4} (1-z_m)^{1/4}}$$

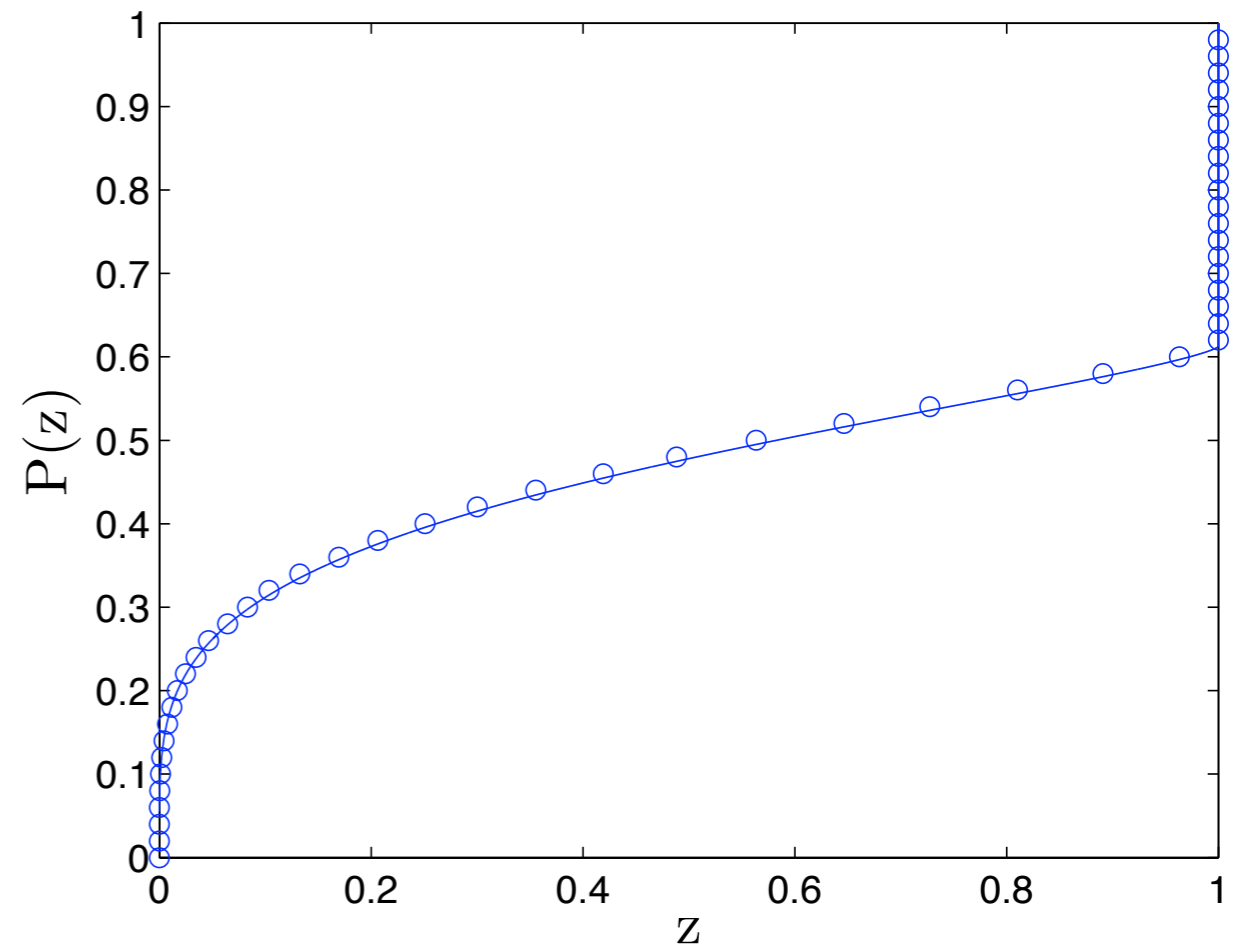
(See also Maximum location, for CTRW : Le Doussal and Schehr )



# Hitting Bridge

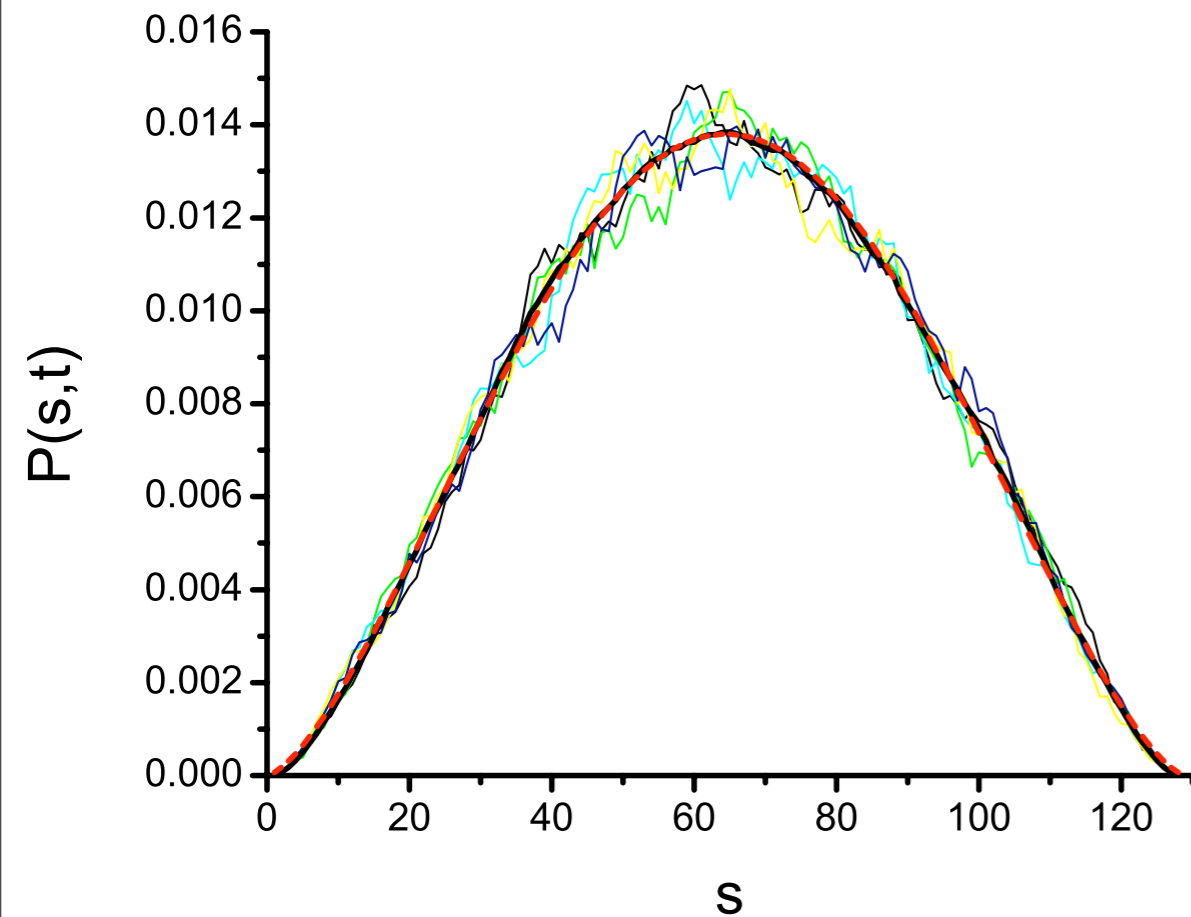


# Hitting Free

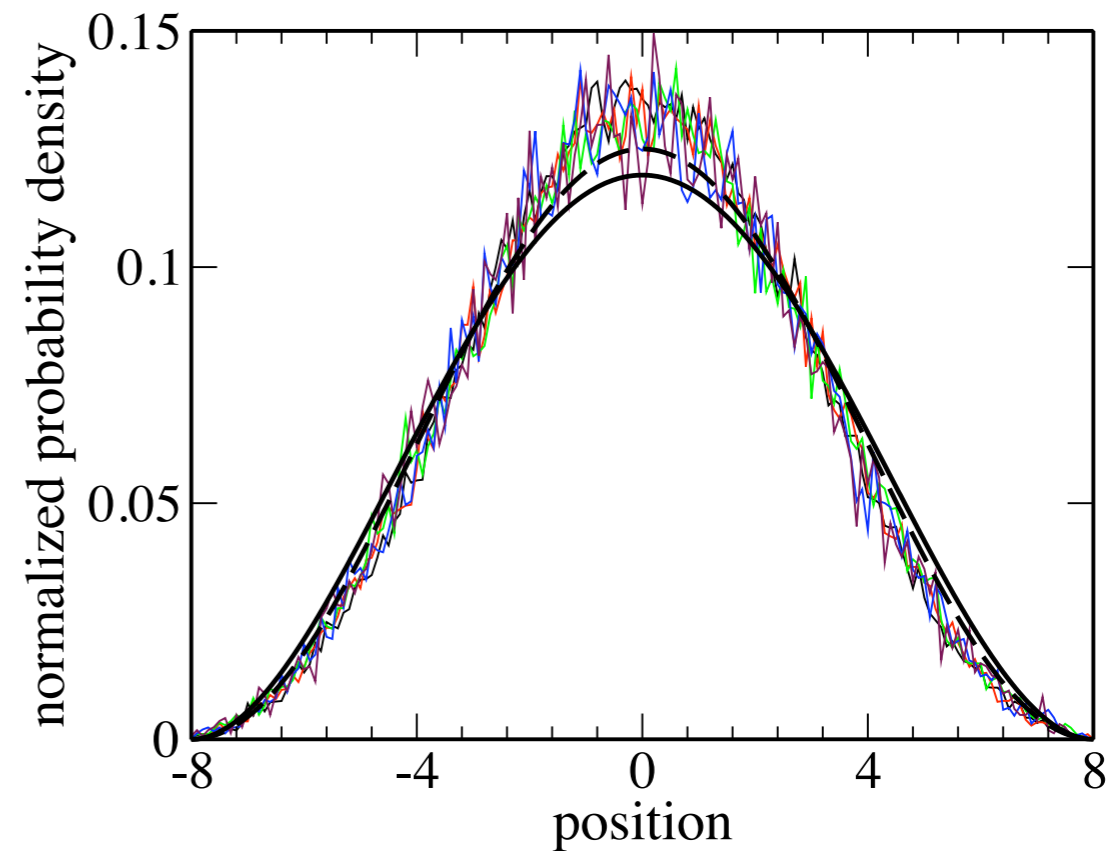


...and the persistence  $S(x_0, t) \sim \frac{1}{\log(t)^{\frac{1}{6}}}$

# “Anomalies” of anomalous diffusion



Monte Carlo simulation of polymer translocation in  $d=2$ ,  
Chatelain, Kantor, Kardar, PRE 78, 021129 (2008)



Monte Carlo simulation tagged monomer in a box ( $d=1$ )  
Kantor, Kardar, PRE 76, 061121 (2007)

$$d = 2, \quad \nu = \frac{3}{4}, \quad H = \frac{1}{2\nu + 1} = \frac{2}{5}$$

$$d = 1, \quad H = \frac{1}{4}$$

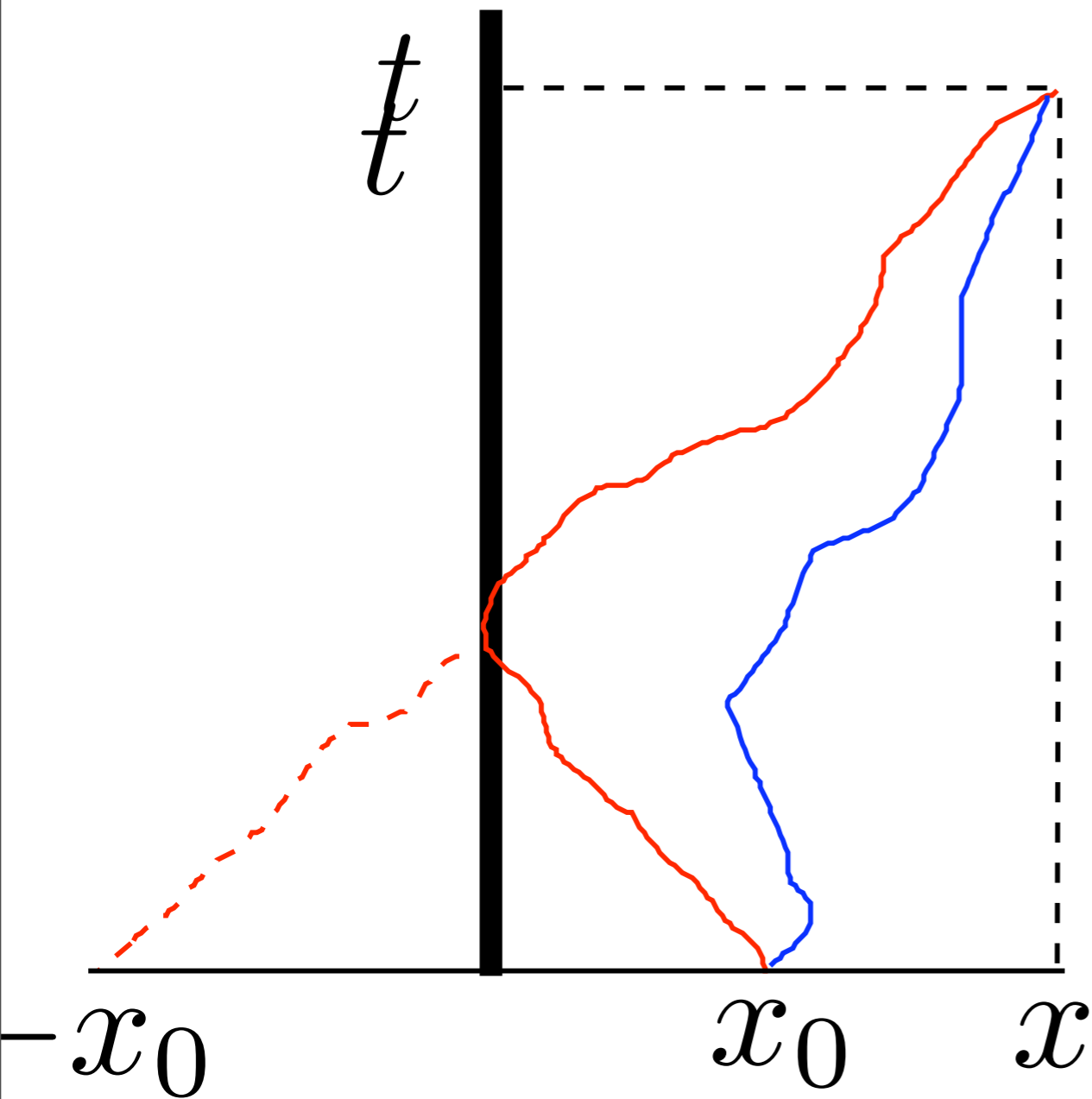
At large  $t$ ,  $P(s, t) \sim s^{1.44}$

At large  $t$ ,  $P(s, t) \sim s^\alpha$ , with  $\alpha > 2$

We predict  $\phi = \frac{1-H}{H} = 1.5$

We predict  $\phi = \frac{1-H}{H} = 3$

# Single Boundary: Images method



$$Z_+(x, x_0, t) = Z(x, x_0, t) - Z(x, -x_0, t)$$

After normalization

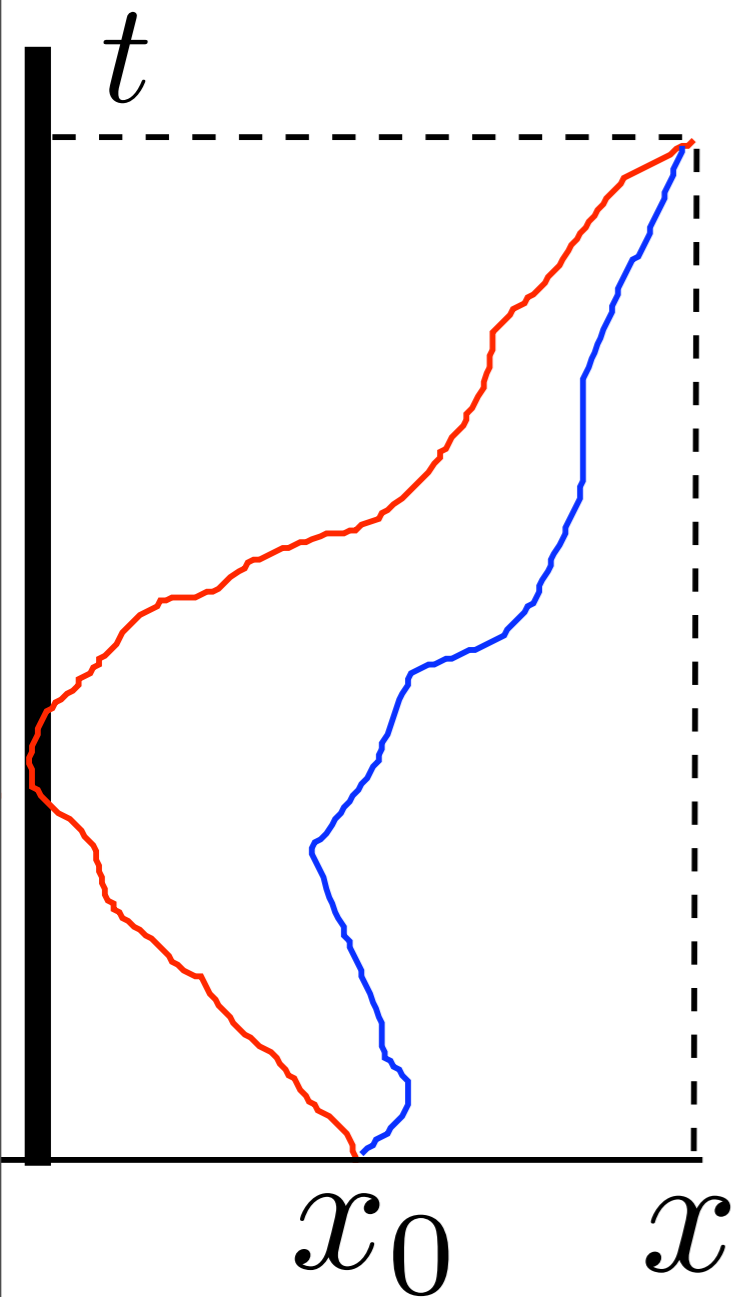
$$P_+(x, x_0, t) \xrightarrow{t \rightarrow \infty \text{ or } x_0 \rightarrow 0} P_+(x, t)$$

Self-Affinity III:  $y = \frac{x}{\sqrt{2Dt}}$

$$P_+(x, t) dx = R_+(y) dy = y e^{-\frac{y^2}{2}} dy$$

**Conclusion :**  $R_+(y) = y e^{-\frac{y^2}{2}}$

# Perturbation Theory



$$Z_+(x_0, x, t) = \int_{x(0)=x_0}^{x(t)=x} \mathcal{D}[x] e^{-\mathcal{S}[x]} \Theta[x]$$

$$P_+(x, t) = \lim_{x_0 \rightarrow 0} \frac{Z_+(x_0, x, t)}{\int_0^\infty dx Z_+(x_0, x, t)}$$

$$\mathcal{S}[x] = \int_0^t dt_1 \int_0^t dt_2 \frac{1}{2} x(t_1) G(t_1, t_2) x(t_2)$$

$$G^{-1}(t_1, t_2) = \langle x(t_1) x(t_2) \rangle$$

## Brownian motion

$$H = \frac{1}{2} \quad \Rightarrow \quad \langle x(t_1)x(t_2) \rangle = 2 \min(t_1, t_2) \quad \Rightarrow \quad \mathcal{S}^{(0)}[x] = \frac{1}{4} \int_0^t dt' (\partial_{t'} x)^2$$

## Fractional Brownian motion

$$H - \text{fBm} \quad \Rightarrow \quad \langle x(t_1)x(t_2) \rangle = t_1^{2H} + t_2^{2H} - |t_1 - t_2|^{2H} \quad \Rightarrow \quad \mathcal{S}[x] ??$$

## Perturbation

$$H = \frac{1}{2} + \epsilon \quad \Rightarrow \quad \langle x(t_1)x(t_2) \rangle = 2 \min(t_1, t_2) + \epsilon K(t_1, t_2) + O(\epsilon^2)$$

$$K(t_1, t_2) = 2 [t_1 \ln t_1 + t_2 \ln t_2 - |t_1 - t_2| \ln |t_1 - t_2|]$$

$$\mathcal{S}[x] = \int_0^t dt_1 \int_0^t dt_2 \frac{1}{2} x(t_1) G(t_1, t_2) x(t_2)$$

$$G = G^{(0)} - \epsilon G^{(0)} K G^{(0)}$$

$$\mathcal{S}[x] = \mathcal{S}^{(0)}[x] + \epsilon \mathcal{S}^{(1)}[x]$$

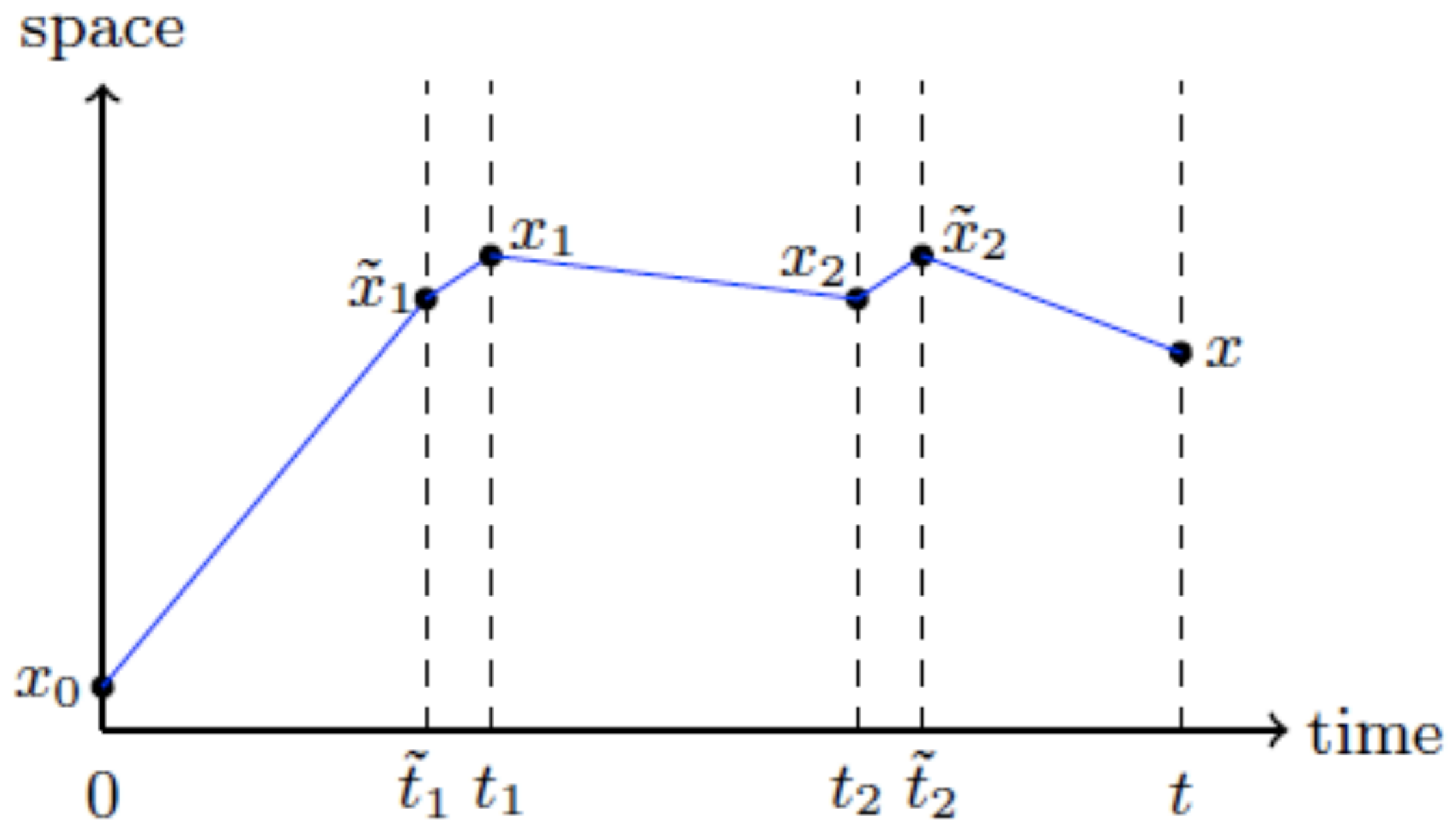
$$\mathcal{S}^{(1)}[x] \propto -\frac{1}{2} \int_0^t dt_1 \int_{t_1}^t dt_2 \frac{\partial_{t_1} x(t_1) \partial_{t_2} x(t_2)}{|t_1 - t_2|}$$

$$e^{-\mathcal{S}[x]} \sim e^{-\mathcal{S}^{(0)}[x]} \left( 1 + \epsilon \mathcal{S}^{(1)}[x] \right)$$

$$Z_+(x_0, x, t) = \int_{x(0)=x_0}^{x(t)=x} \mathcal{D}[x] e^{-\mathcal{S}[x]} \Theta[x]$$

$$Z_+(x_0, x, t) \sim Z_+^{(0)}(x_0, x, t) + \epsilon Z_+^{(1)}(x_0, x, t)$$

$$Z_+^{(1)}(x_0, x, t) = \int_{x(0)=x_0}^{x(t)=x} \mathcal{D}[x] \mathcal{S}^{(1)}[x] e^{-\mathcal{S}^{(0)}[x]} \Theta[x]$$



Brownian 2-points  
correlation function



# Final Result I

$$R_+(y) = R_+^{(0)}(y) [1 + \epsilon W(y) + O(\epsilon^2)]$$

$$W(y) = \frac{1}{6} y^4 {}_2F_2 \left( 1, 1; \frac{5}{2}, 3; \frac{y^2}{2} \right)$$

$$+ \pi(1 - y^2) \operatorname{erfi} \left( \frac{y}{\sqrt{2}} \right) + \sqrt{2\pi} e^{\frac{y^2}{2}} y$$

$$+ (y^2 - 2) [\log(2y^2) + \gamma_E] - 3y^2$$

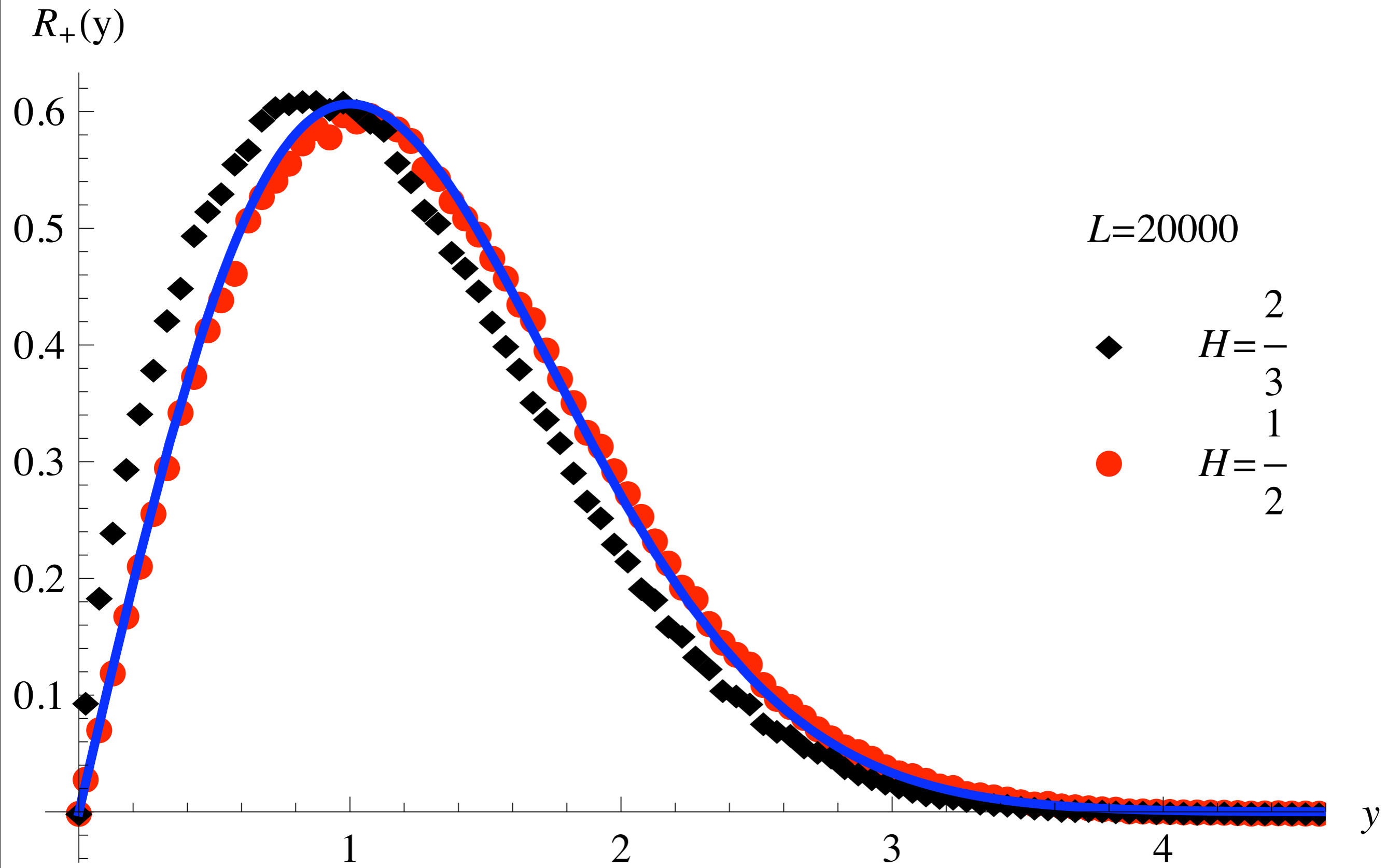
# Final Result II

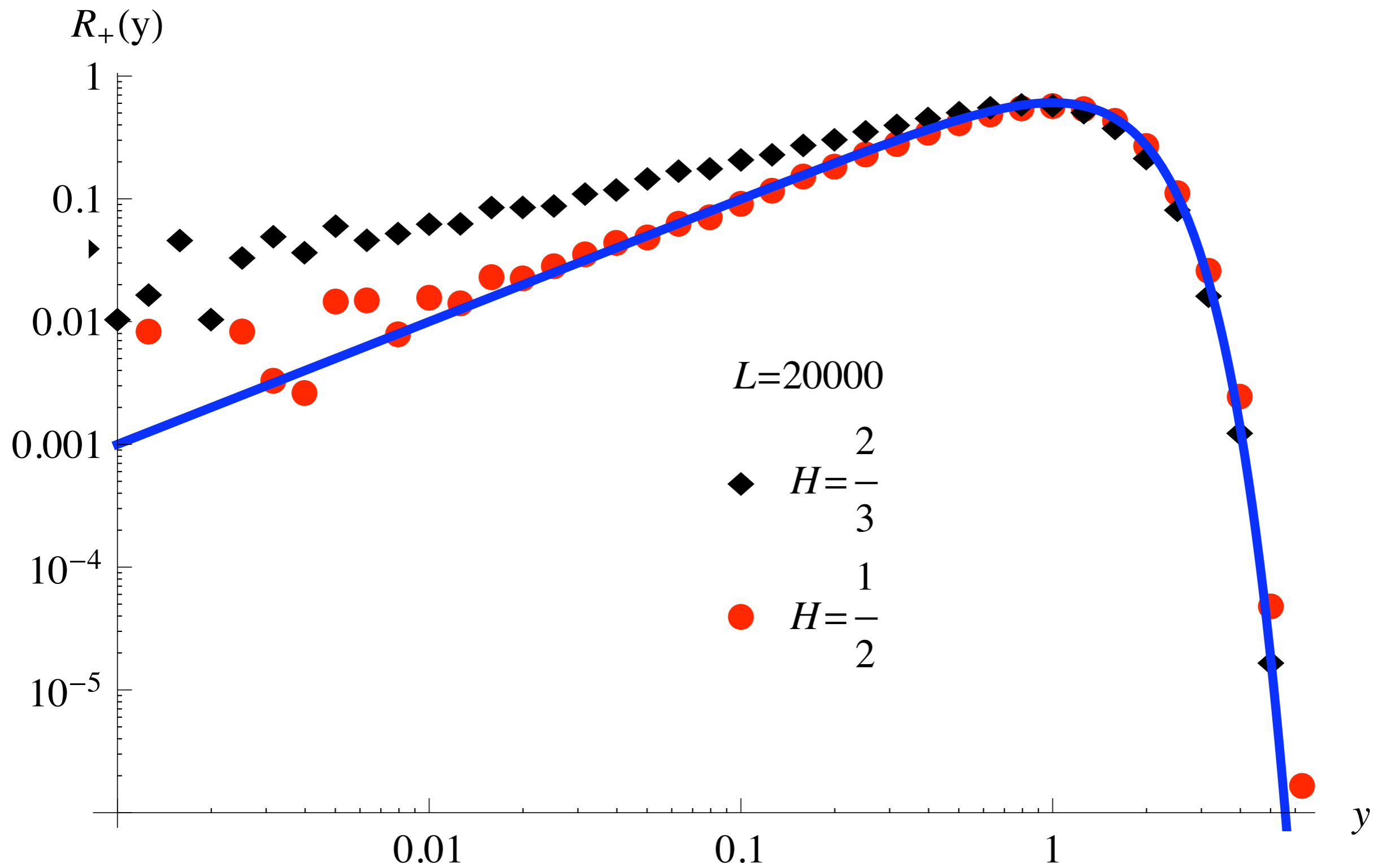
$$R_+(y) \xrightarrow{y \rightarrow 0} y^\phi$$

$$R_+(y) \xrightarrow{y \rightarrow \infty} y^\gamma e^{-\frac{y^2}{2}}$$

$$\phi = 1 - 4\epsilon + O(\epsilon^2) , \quad \gamma = 1 - 2\epsilon + O(\epsilon^2) .$$

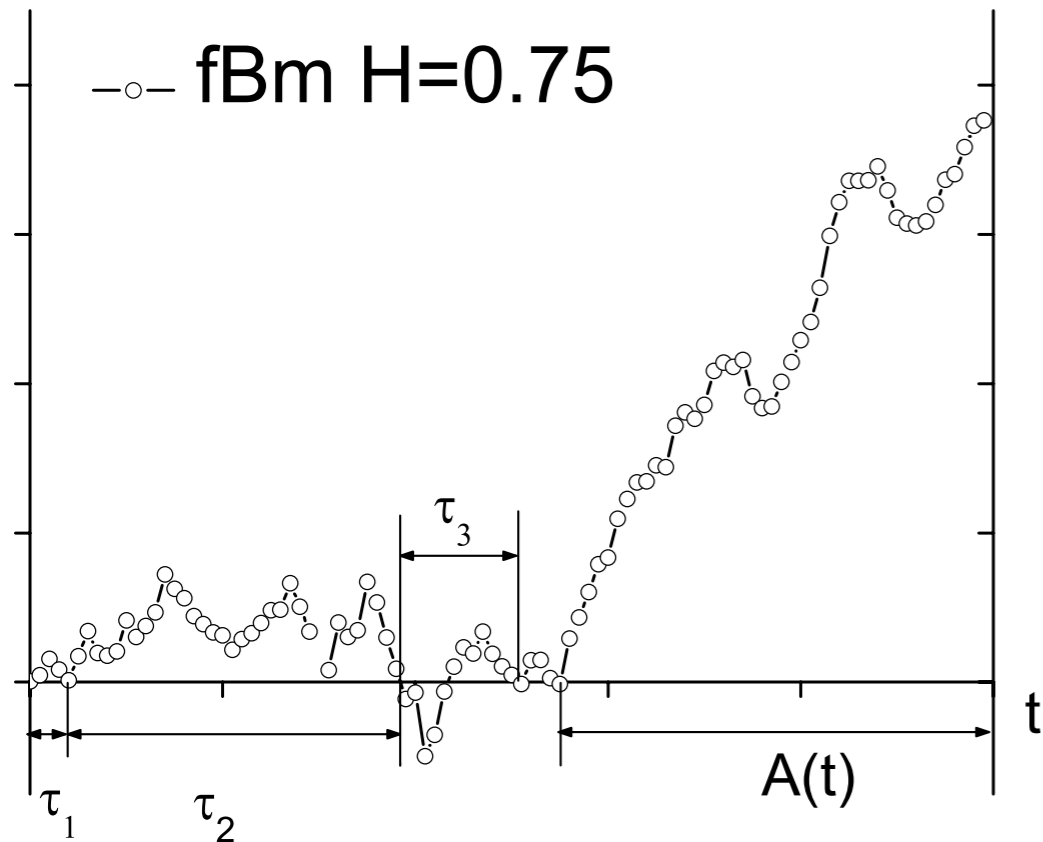
- $\epsilon$  expansion in agreement with the conjecture  $\phi = \frac{1-H}{H}$
- At large  $y$ , Free Gaussian Propagator
- + a New Exponent  $\gamma \neq \phi$





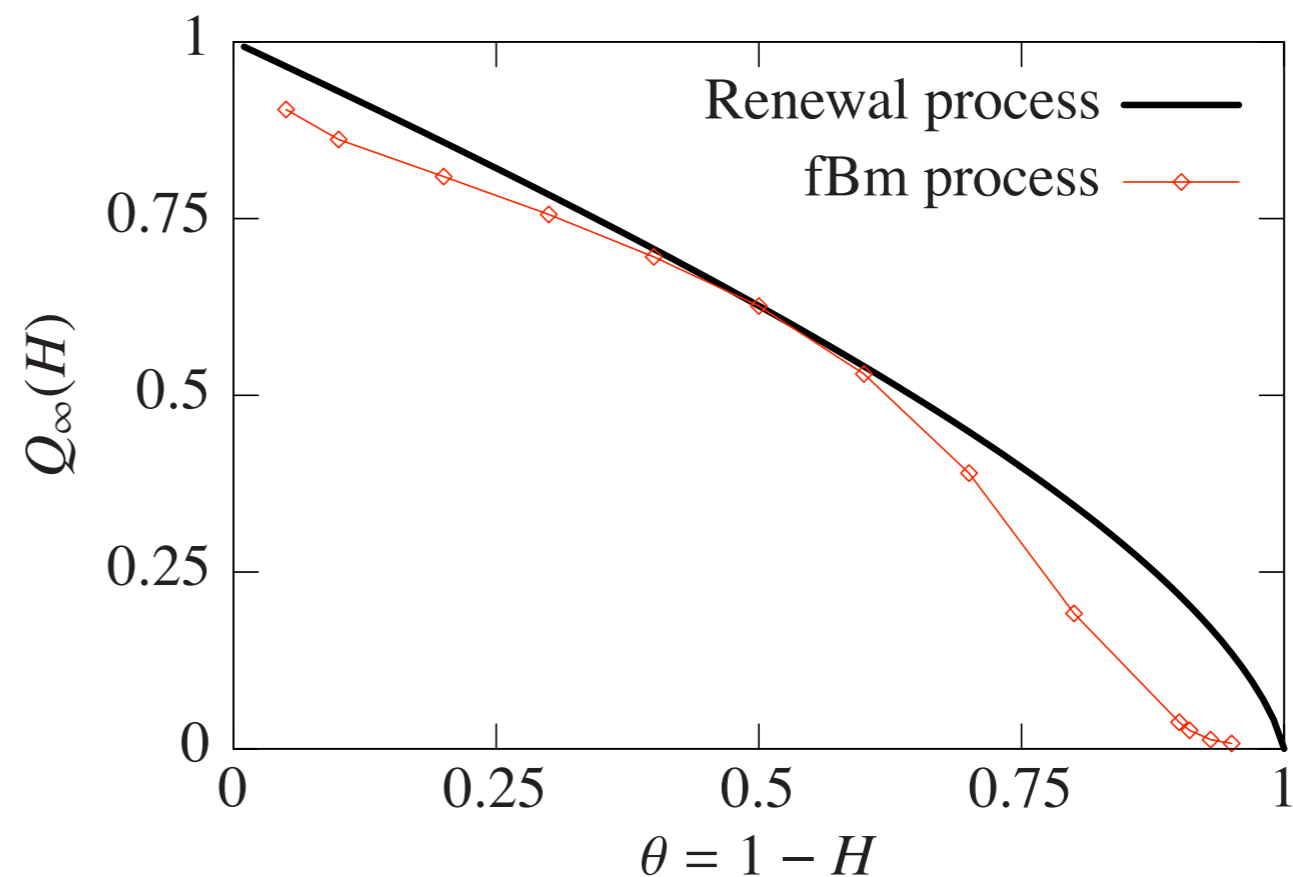
# Non Markovianity and extremes...

The probability  $Q(t)$  that  $A(t)$  is the *longest excursion*?

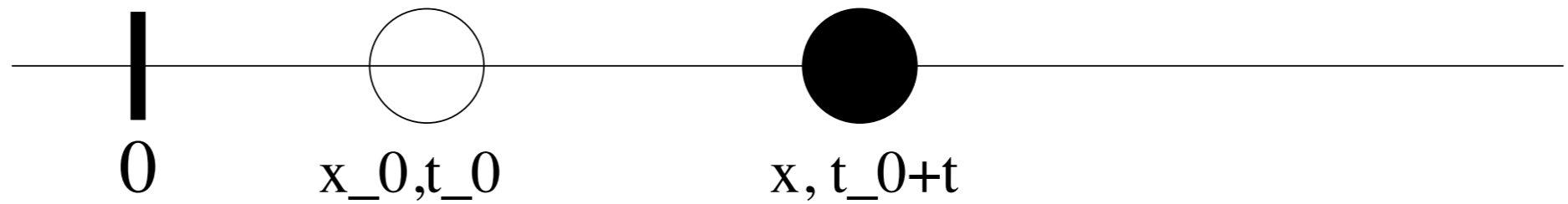


with Schehr and Garcia

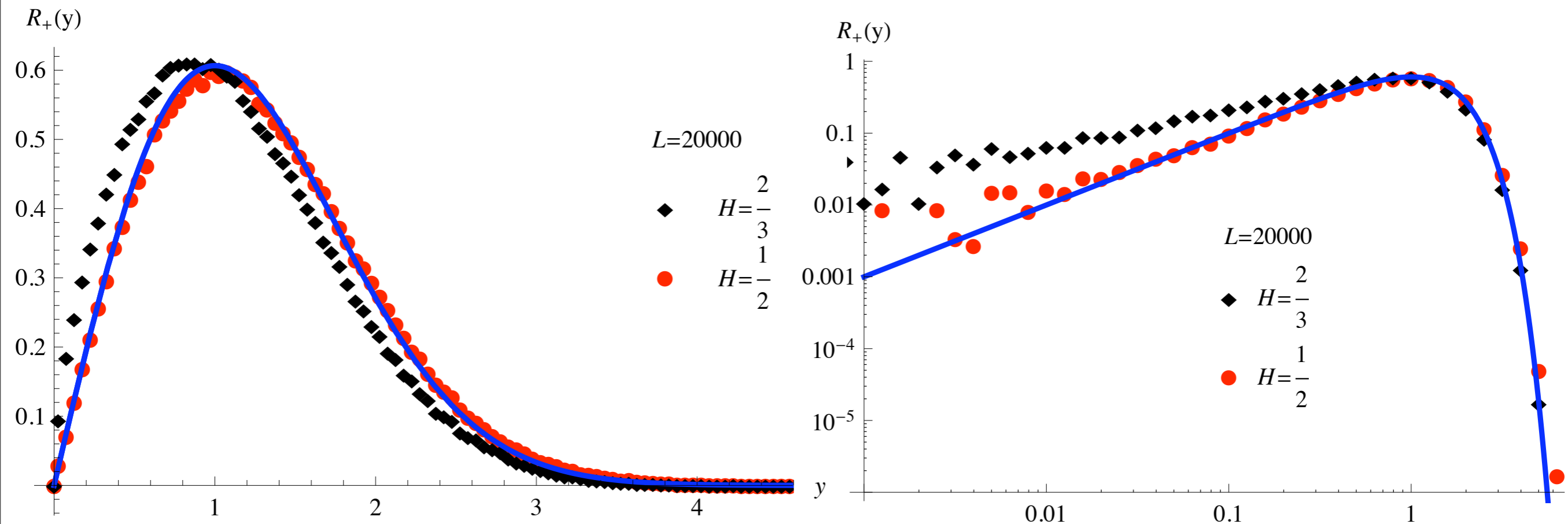
Exact solution for RENEWAL process  
[Godrèche, Majumdar, Schehr]



# Single Boundary



- Large time:  $P_+(x, t; x_0, t_0) \xrightarrow{t \rightarrow \infty} P_+(x, t)$
- Self Affinity III:  $P_+(x, t) = R_+(y)$  with  $y = \frac{x}{\sqrt{\langle x^2(t) \rangle}} \sim \frac{x}{t^H}$
- Brownian  $H = 1/2$  :  $R_+(y) = ye^{-\frac{y^2}{2}} \sim y + \dots$



# How to generate a correlated path:

- time discretization:  $x(t) \longrightarrow x_t = x_0 + \sum_{t'=1}^t \xi_{t'}$
- $\vec{\xi} = \{\xi_1, \xi_2, \dots, \xi_T\}$  is a vector of Gaussian numbers with a given covariance matrix  $\langle \xi_{t_1} \xi_{t_2} \rangle = C(|t_1 - t_2|)$
- we compute  $A = \sqrt{C}$  [ $T^3$  operations] and generate uncorrelated Gaussian numbers  $\vec{\epsilon} = \{\epsilon_1, \dots, \epsilon_T\}$
- Then the vector  $\vec{\xi}$  can be generated [ $T^2$  operations]:

$$\vec{\xi} = A\vec{\epsilon}$$

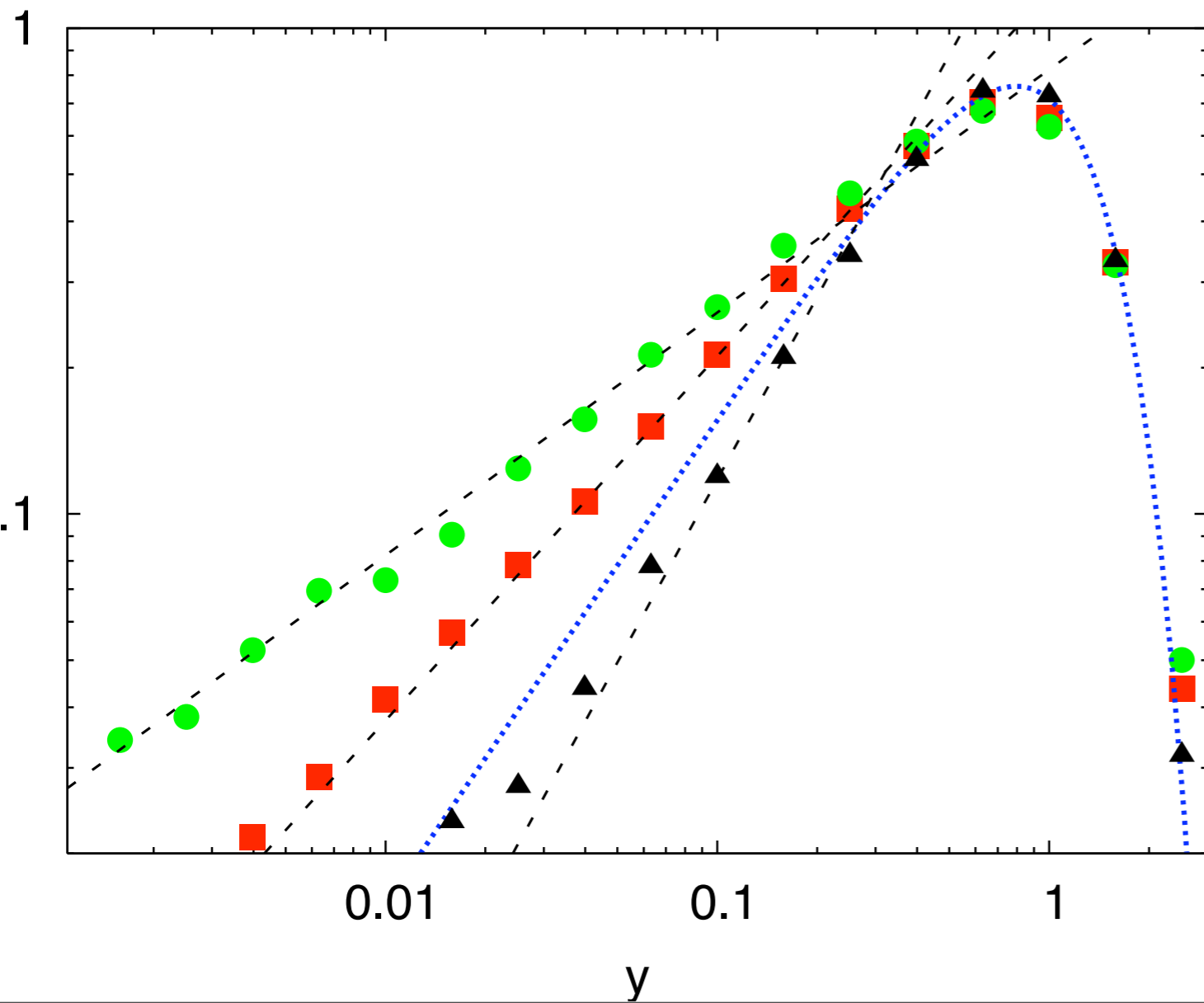
Proof:

$$\langle \xi_i \xi_j \rangle = \left\langle \sum_{i'} A_{i,i'} \epsilon_{i'} \sum_{j'} A_{j,j'} \epsilon_{j'} \right\rangle = \sum_{i',j'} A_{i,i'} A_{j,j'} \delta_{i',j'} = (A^2)_{i,j} = C_{i,j}$$

- Special algorithms if Covariance is Toeplitz matrix:  $C_{i,j} = C(|i - j|)$ .

# Single Boundary

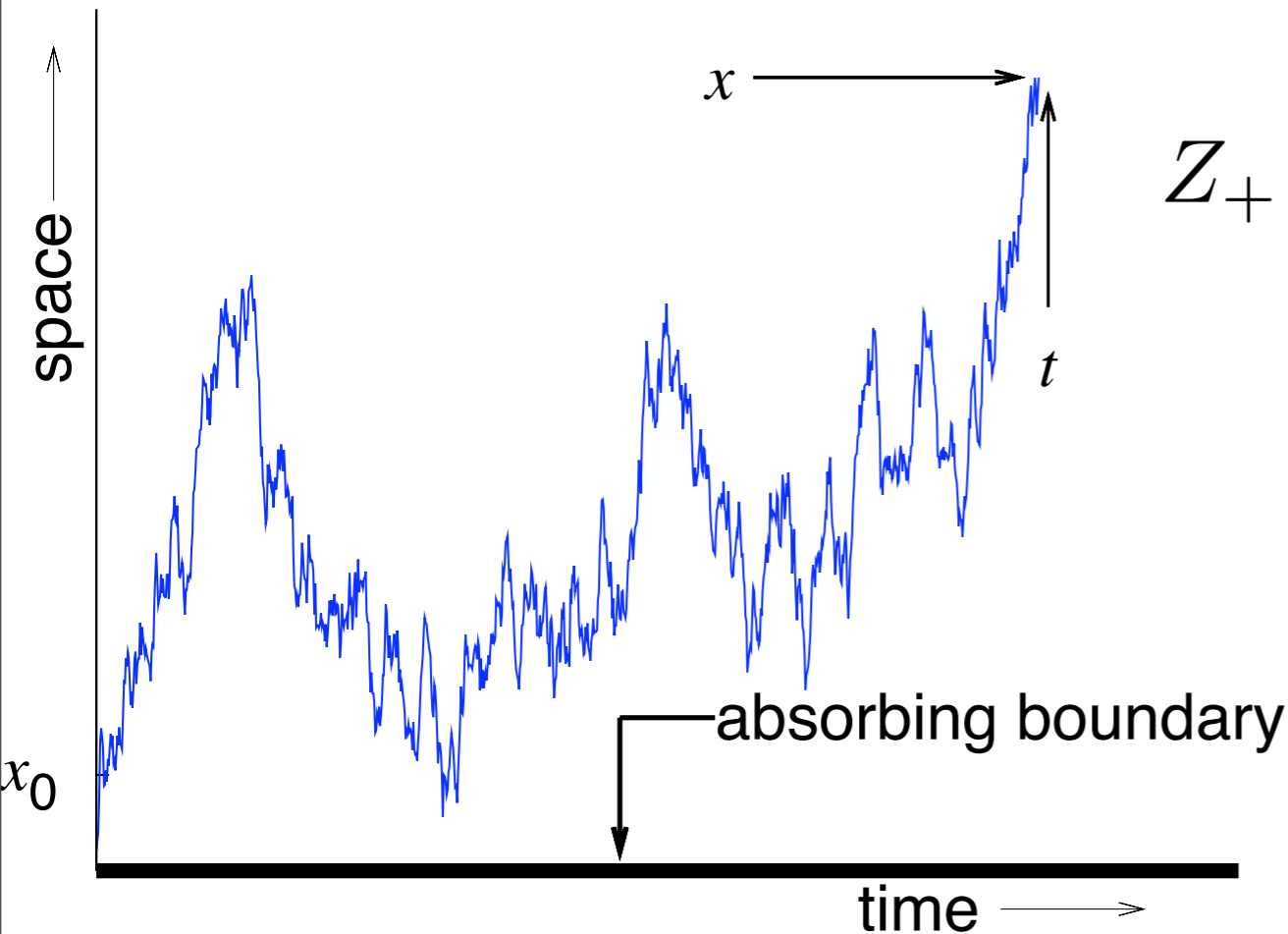
- In general we expect  $R_+(y) \sim y^\phi$
- For stationary increments, a scaling argument gives  $\phi = \frac{\theta}{H}$
- numerical simulations agrees



- Circle  $H = \frac{2}{3}$ ,  $\phi = \frac{1}{2}$
- Square  $H = \frac{4}{7}$ ,  $\phi = \frac{3}{4}$
- Line  $R_+(y) = ye^{-y^2/2}$
- Triangle  $H = \frac{4}{9}$ ,  $\phi = \frac{5}{4}$



# Perturbation Theory



$$Z_+(x_0, x, t) = \int_{x(0)=x_0}^{x(t)=x} \mathcal{D}[x] e^{-\mathcal{S}[x]} \Theta[x]$$

$$P_+(x, t) = \lim_{x_0 \rightarrow 0} \frac{Z_+(x_0, x, t)}{\int_0^\infty dx Z_+(x_0, x, t)}$$

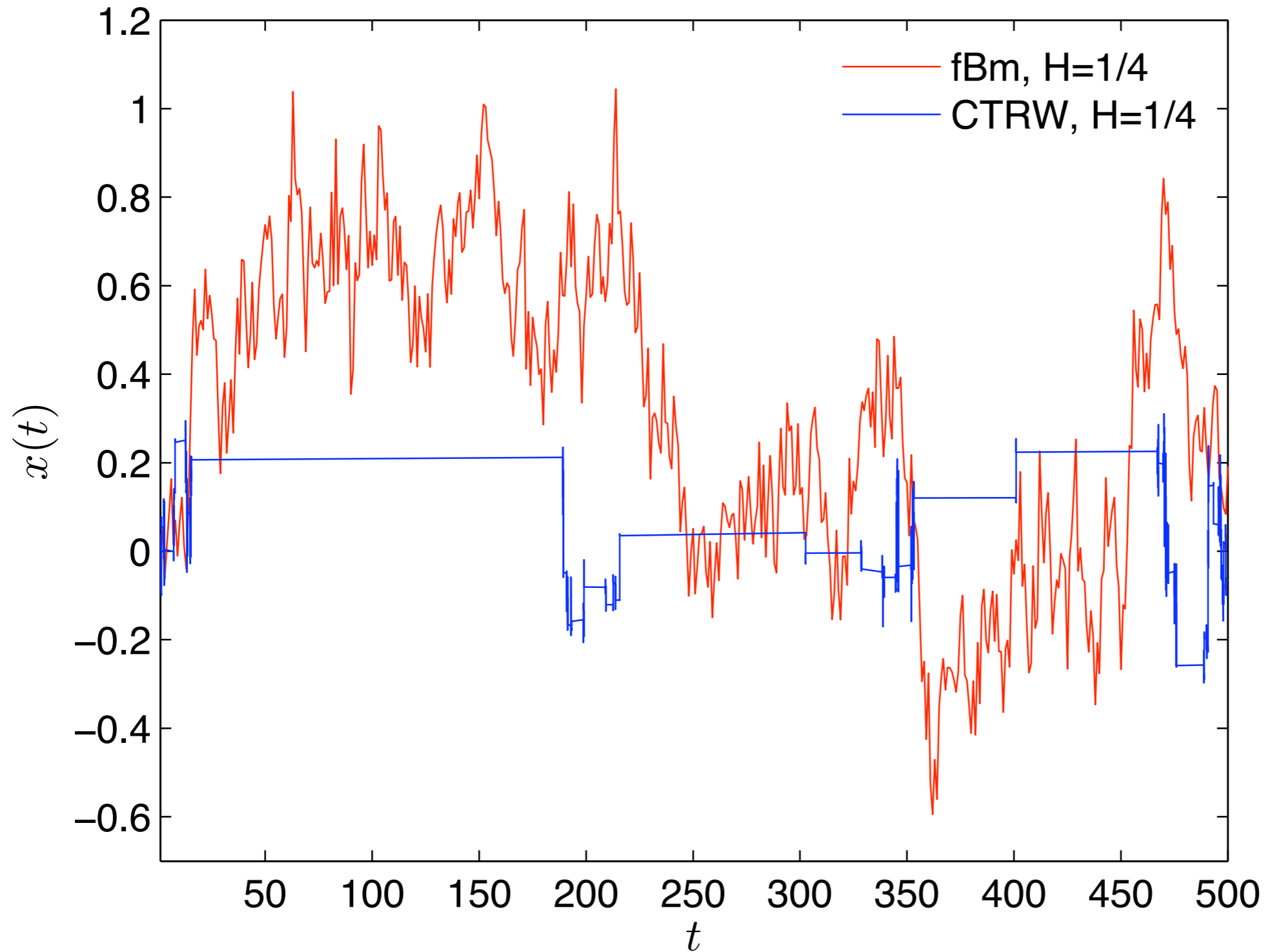
$$\mathcal{S}[x] = \int_0^t dt_1 \int_0^t dt_2 \frac{1}{2} x(t_1) G(t_1, t_2) x(t_2)$$

where  $G^{-1}(t_1, t_2) = \langle x(t_1) x(t_2) \rangle$  it is known

# Diffusion and Central Limit Theorem

- $\langle x^2(t) \rangle = 2Dt \propto t$  and  $D$  is diffusion constant
- $x(t)$  (for large  $t$ ) is a **Gaussian** process
- Propagator from  $x_0$  to  $x$ : 
$$P(x_0, x, t) = \frac{e^{-\frac{(x(t) - x_0)^2}{4Dt}}}{\sqrt{4\pi Dt}}$$

# And the Fractional Fokker Planck Equation (CTRW)?



# Anomalous Diffusion

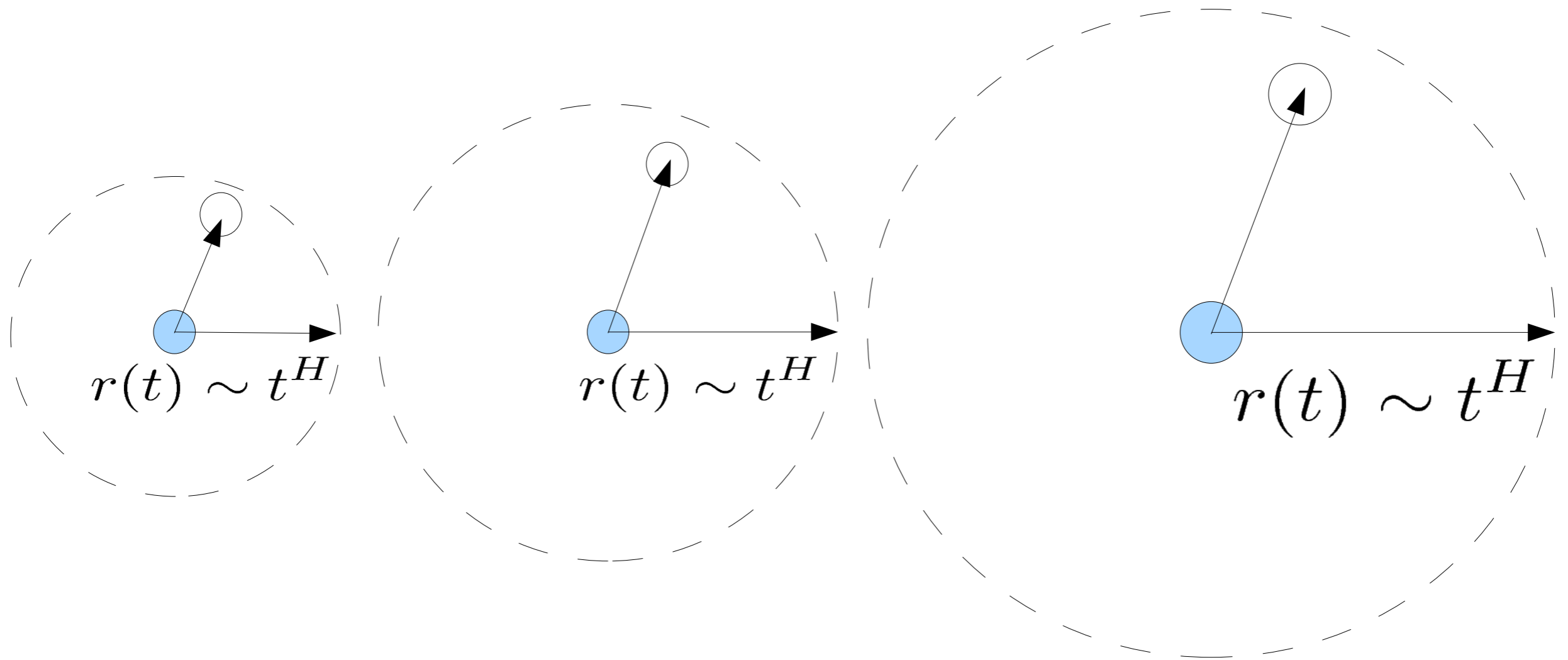
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Longest excursion: R. Garcia (Bariloche) + G. Schehr

# Anomalous Diffusion



$$H < 1/2$$

Subdiffusion

$$H = 1/2$$

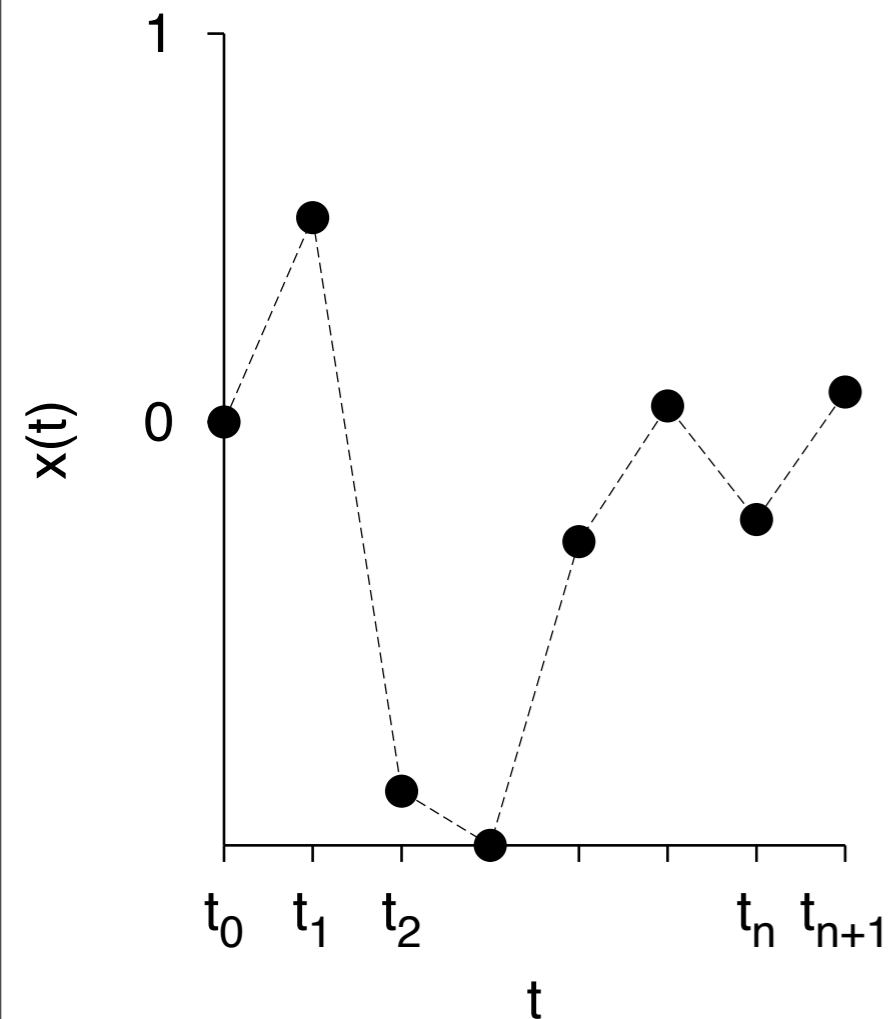
Diffusion

$$H > 1/2$$

Super-diffusion

# Central Limit Theorem

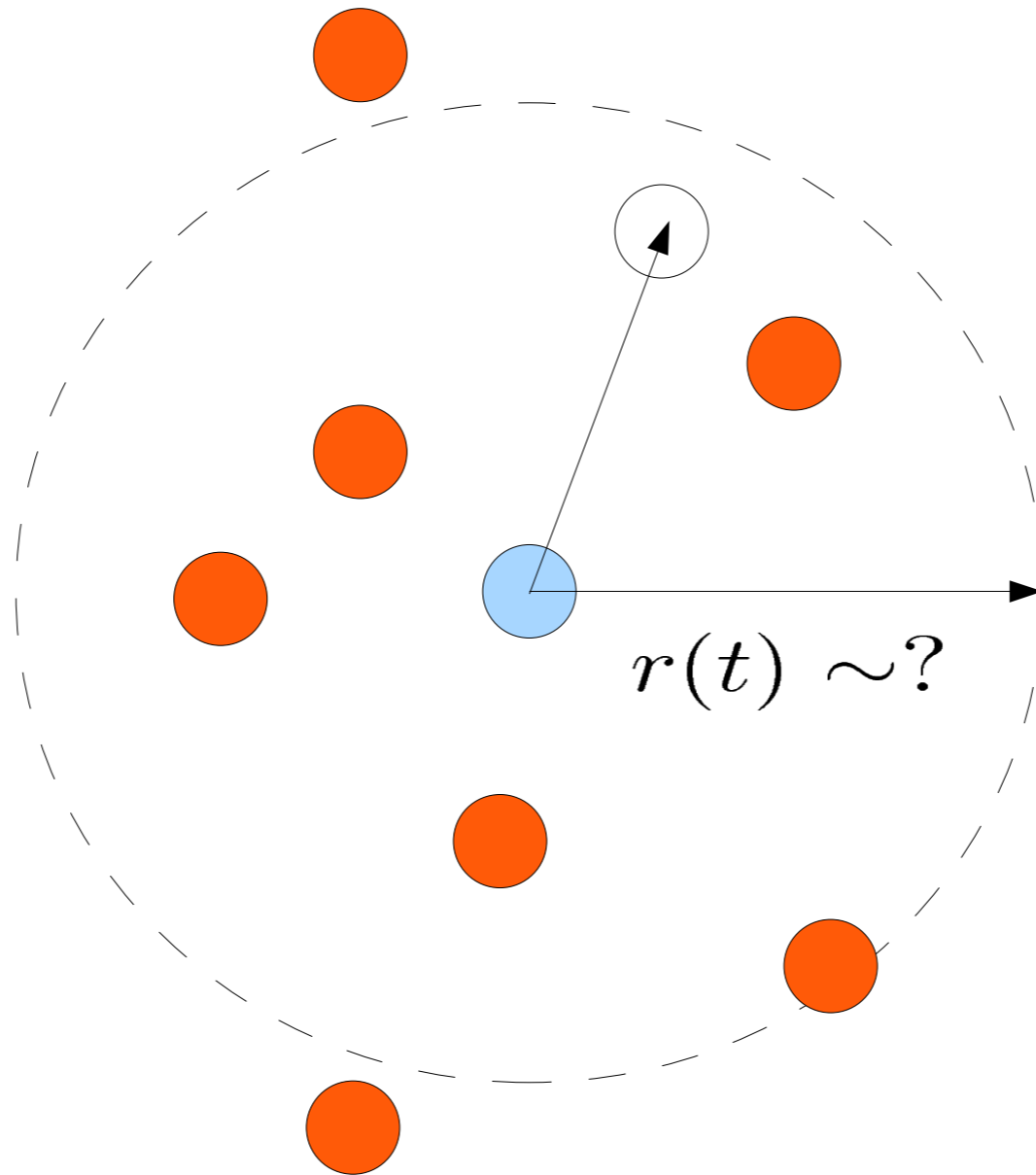
$$x(t) = \sum_{t'=1}^t \xi_{t'}$$



- *Identical*:  $\pi_{t'}(\xi) = \pi(\xi)$ . (Homogeneous)
- *Independent*:  $\langle \xi_{t_0} \xi_{t_0+t} \rangle = 0$ . (Markov)
- $\langle \xi^2 \rangle < \infty$ : Continuous process.

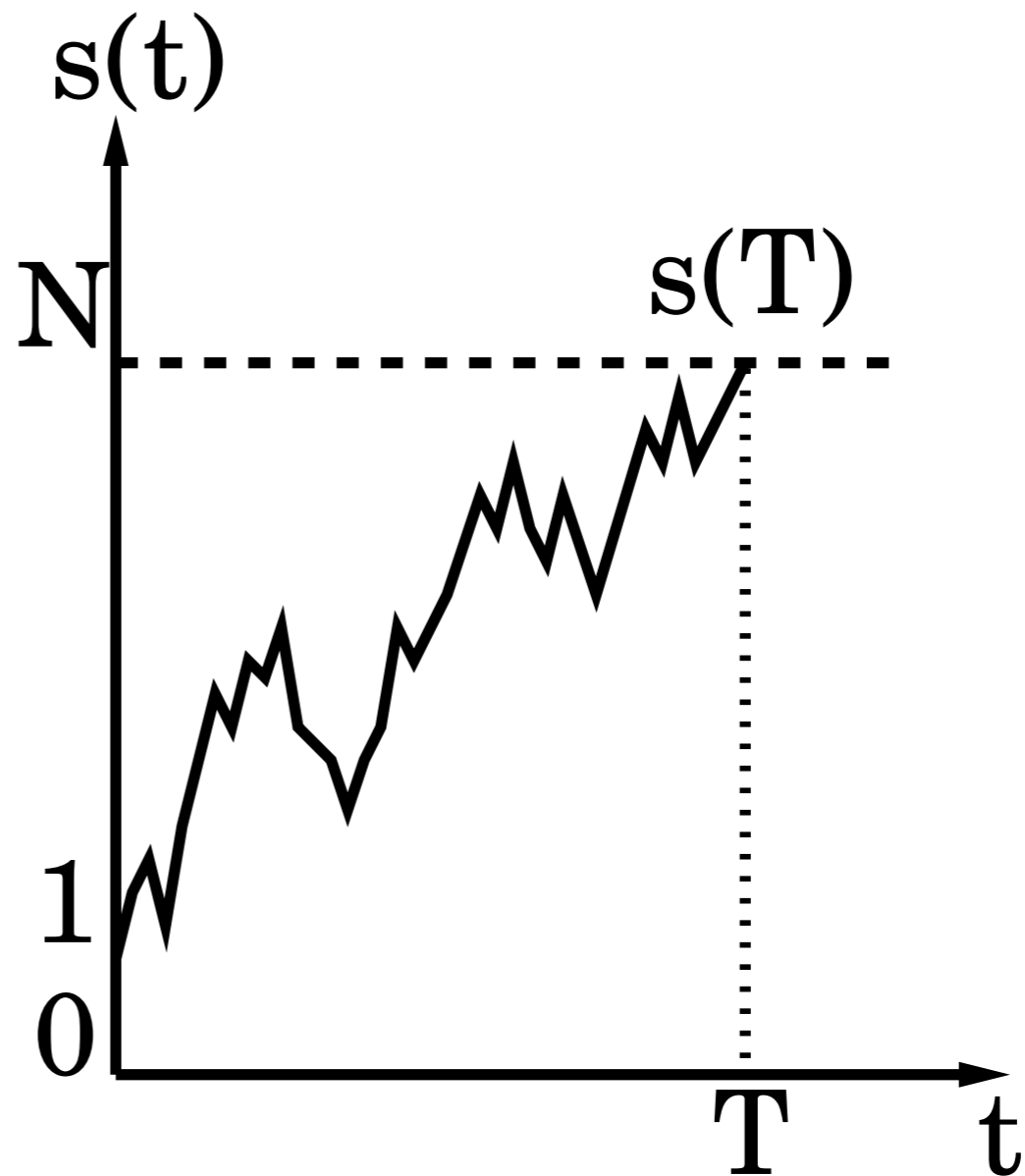
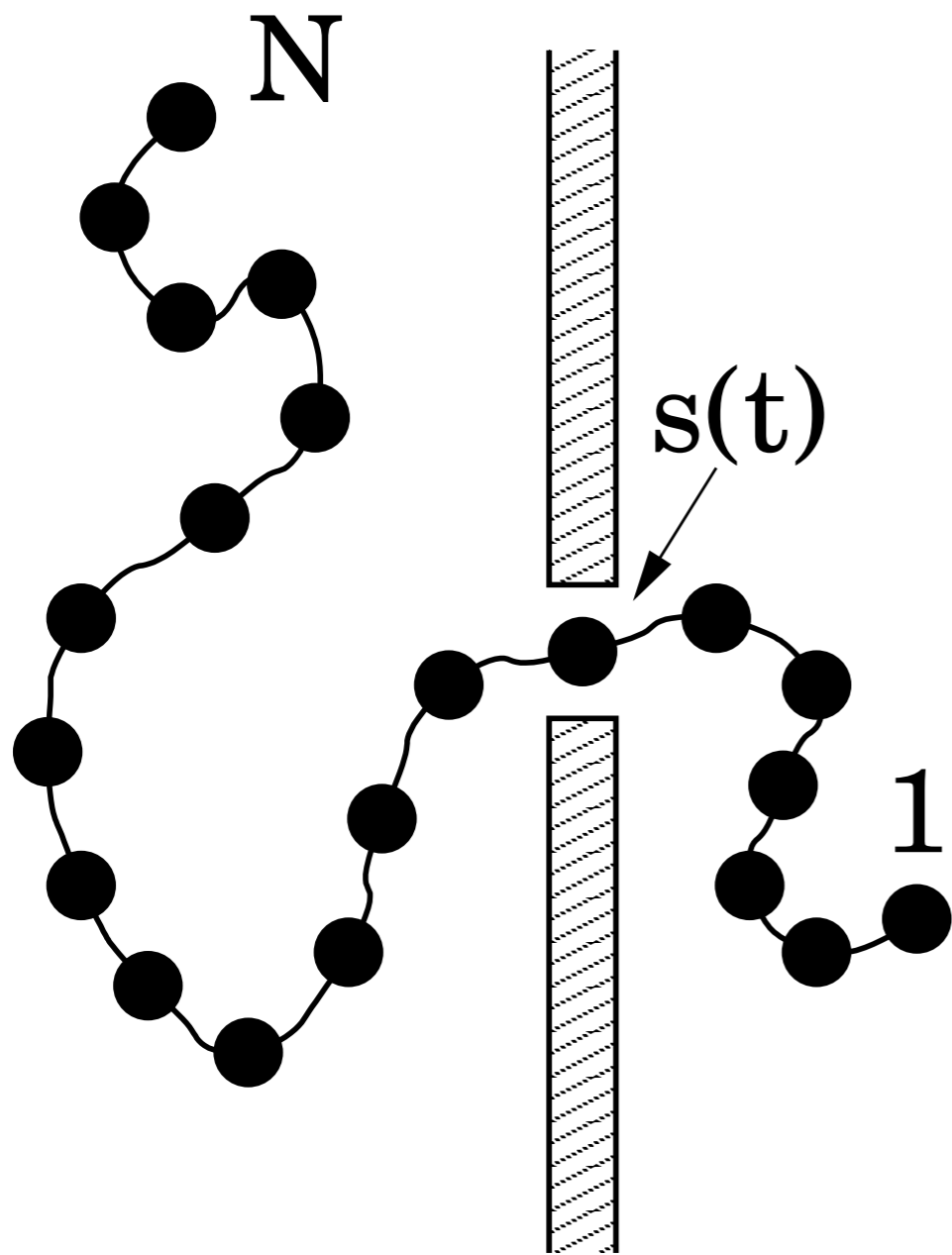
**Conclusion:**  $x(t)$  is Gaussian and  $x(t) \sim \sqrt{t}$

# Correlations



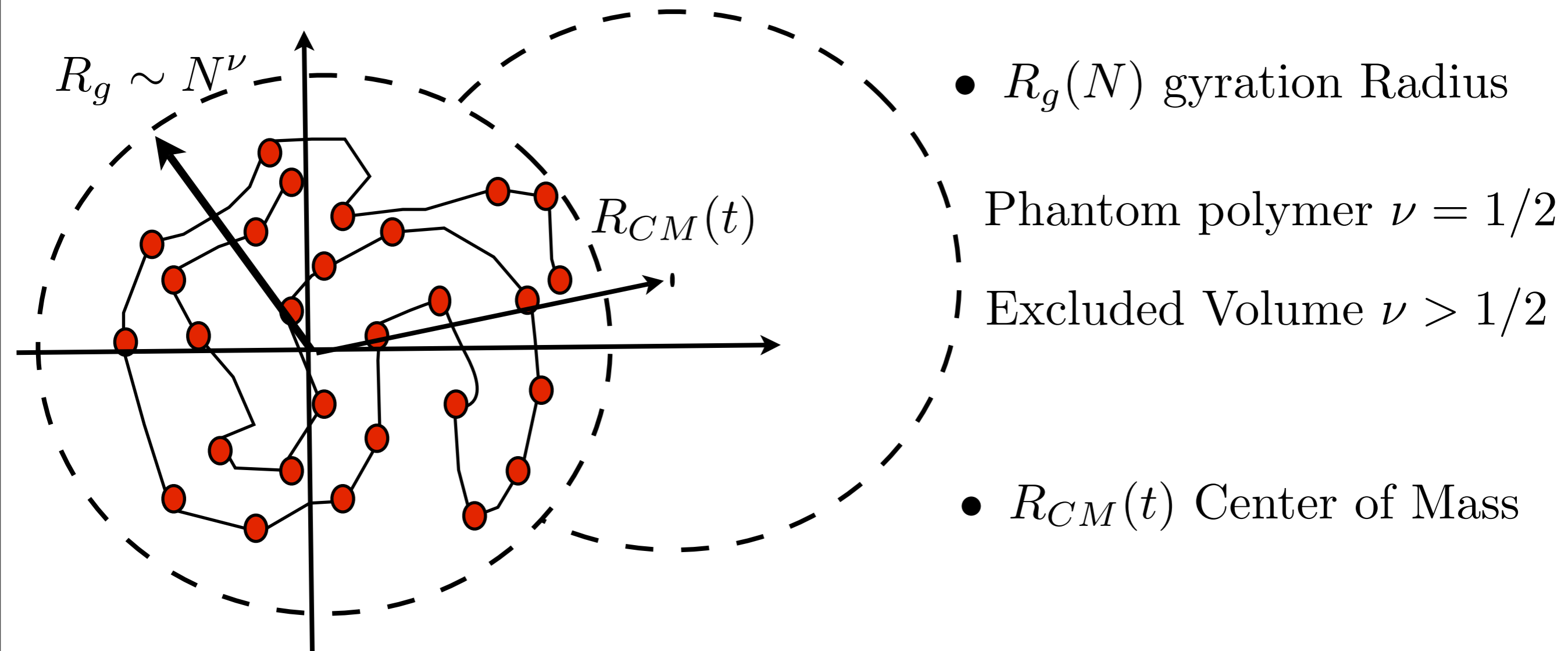
- jumps and waiting times are local
- colloids interact (strongly non-Markovian)

# Polymer Translocation



$$s(T) = N, \text{ if } s(t) \sim t^H \text{ then } T \sim N^{1/H}$$

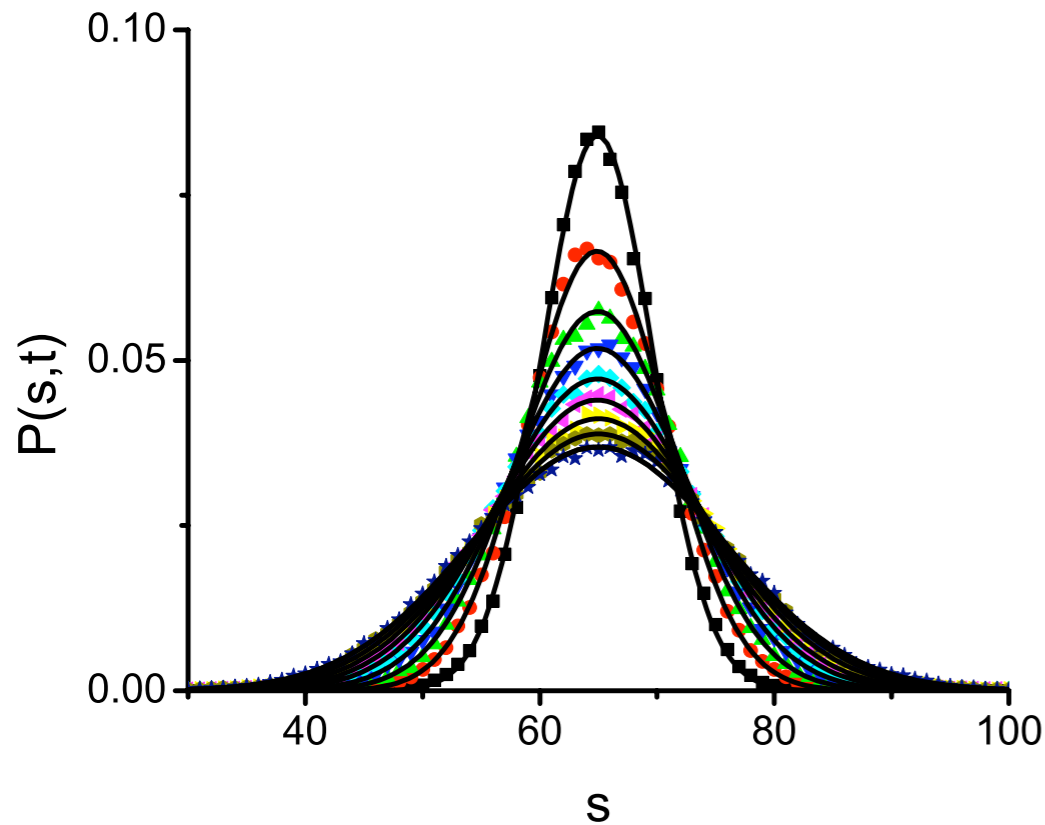




$$\langle R_{CM}^2(t_{eq.}) \rangle = R_g^2 \quad \Rightarrow \quad \frac{D_1 t_{eq.}}{N} \sim N^{2\nu} \quad \Rightarrow \quad t_{eq.} \sim N^{2\nu+1}$$

$$T \gg t_{eq.} \quad \Rightarrow \quad H \leq \frac{1}{1+2\nu} \quad \text{numerically} \quad H \approx \frac{1}{1+2\nu}$$

# Fractional Brownian motion



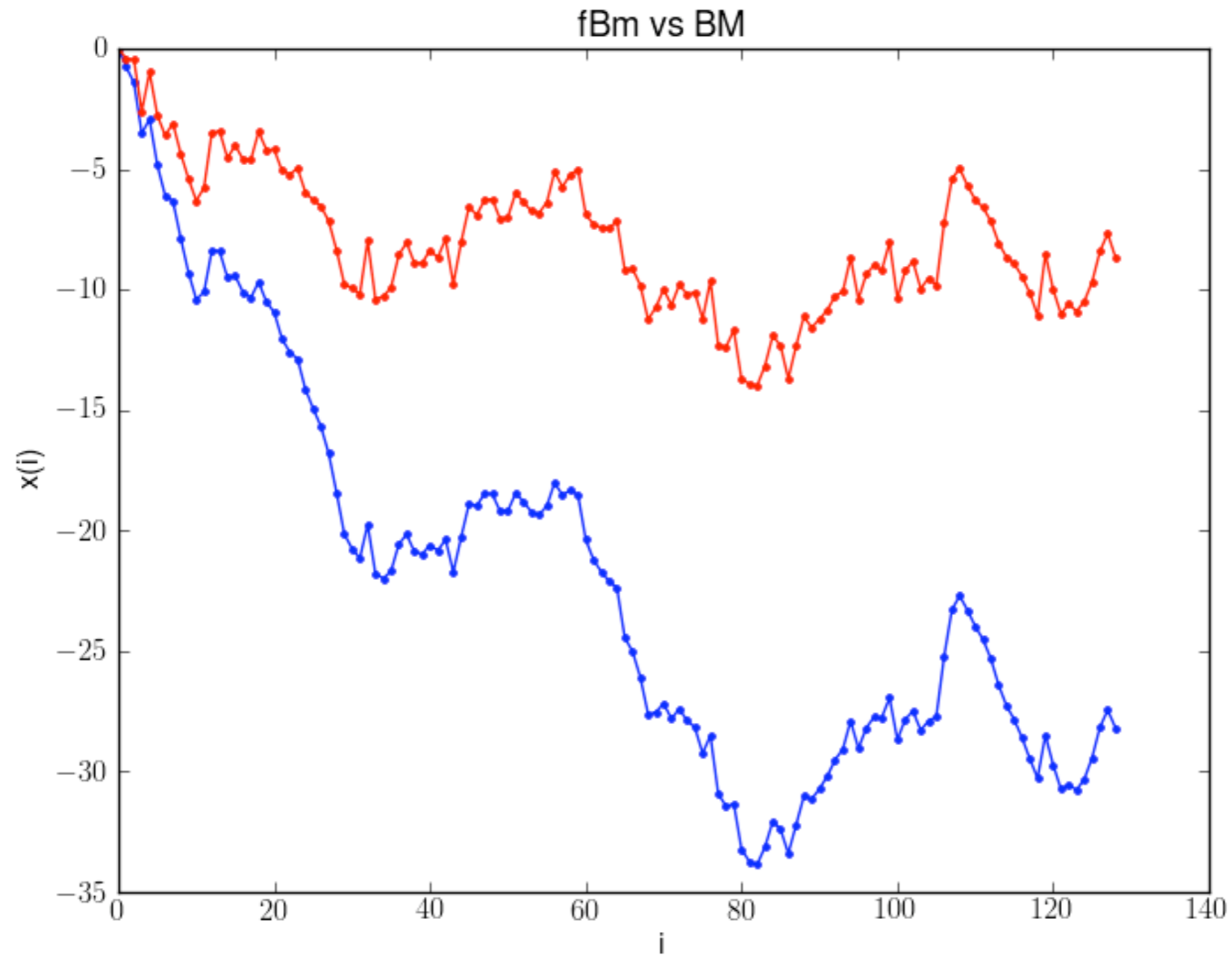
Monte Carlo simulation of polymer translocation in  $d=2$ ,  
Chatelain, Kantor, Kardar, PRE 78, 021129 (2008)

- Equilibrium with the solvent  
 $\Rightarrow$  stationary increments
- is a Gaussian process  
 $\Rightarrow$  local jumps
- $\Rightarrow$  non-Markov process

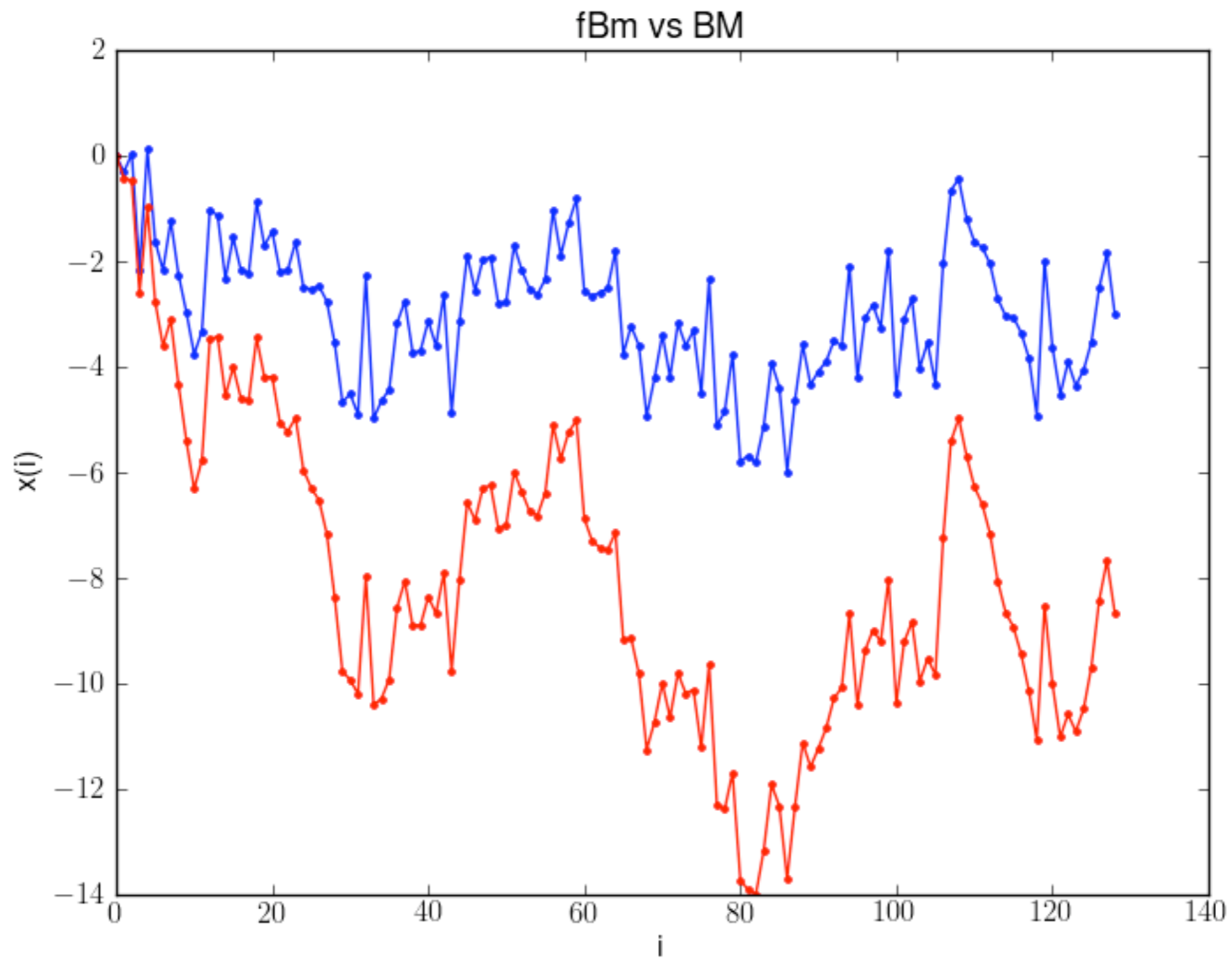
Gaussianity + self-affinity  $H = \frac{1}{2\nu+1}$  + stationary increments  
 $\Rightarrow$  fractional Brownian motion:

$$\langle s(t_1)s(t_2) \rangle \propto (t_1^{2H} + t_2^{2H} - |t_1 - t_2|^{2H}),$$

# H=3/4 Superdiffusion



# H=1/4 Subdiffusion

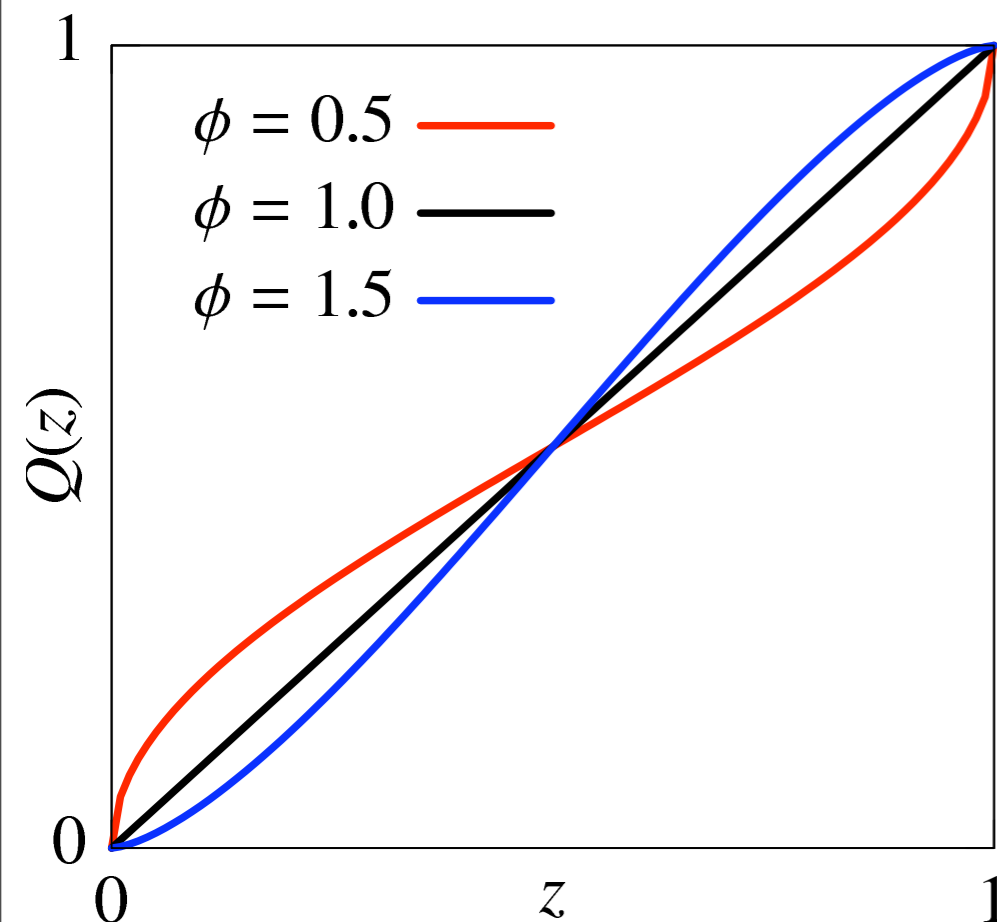
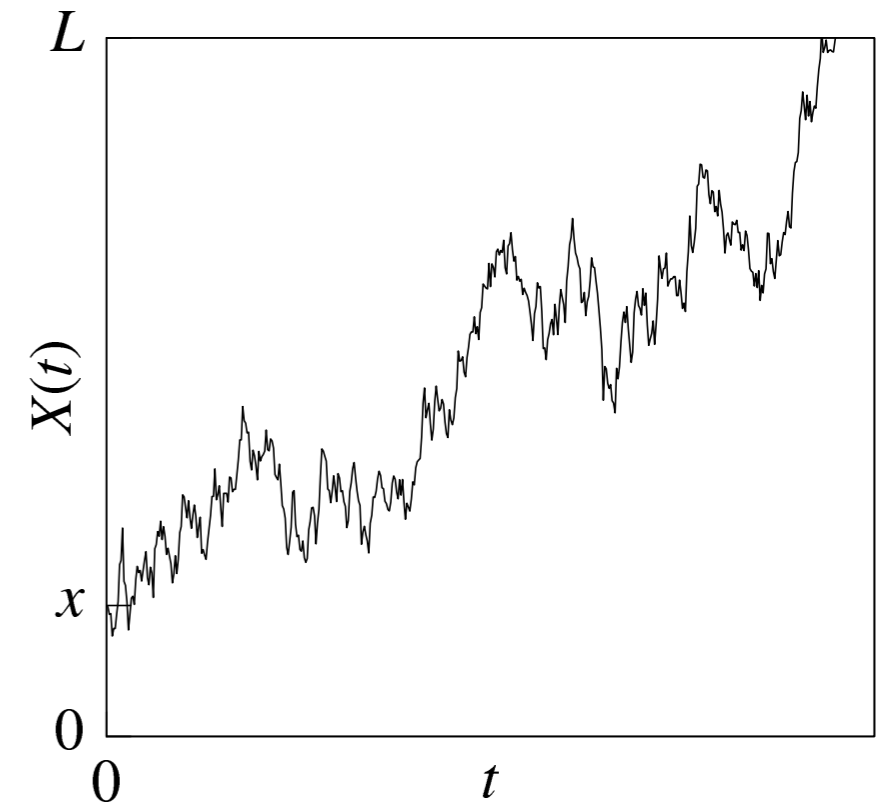


# Question I: A polymer chain will ultimately succeed in translocating through a pore ?

Hitting probability  $Q(x, L)$ :  
probability of exiting through  $L$

For self affine processes:

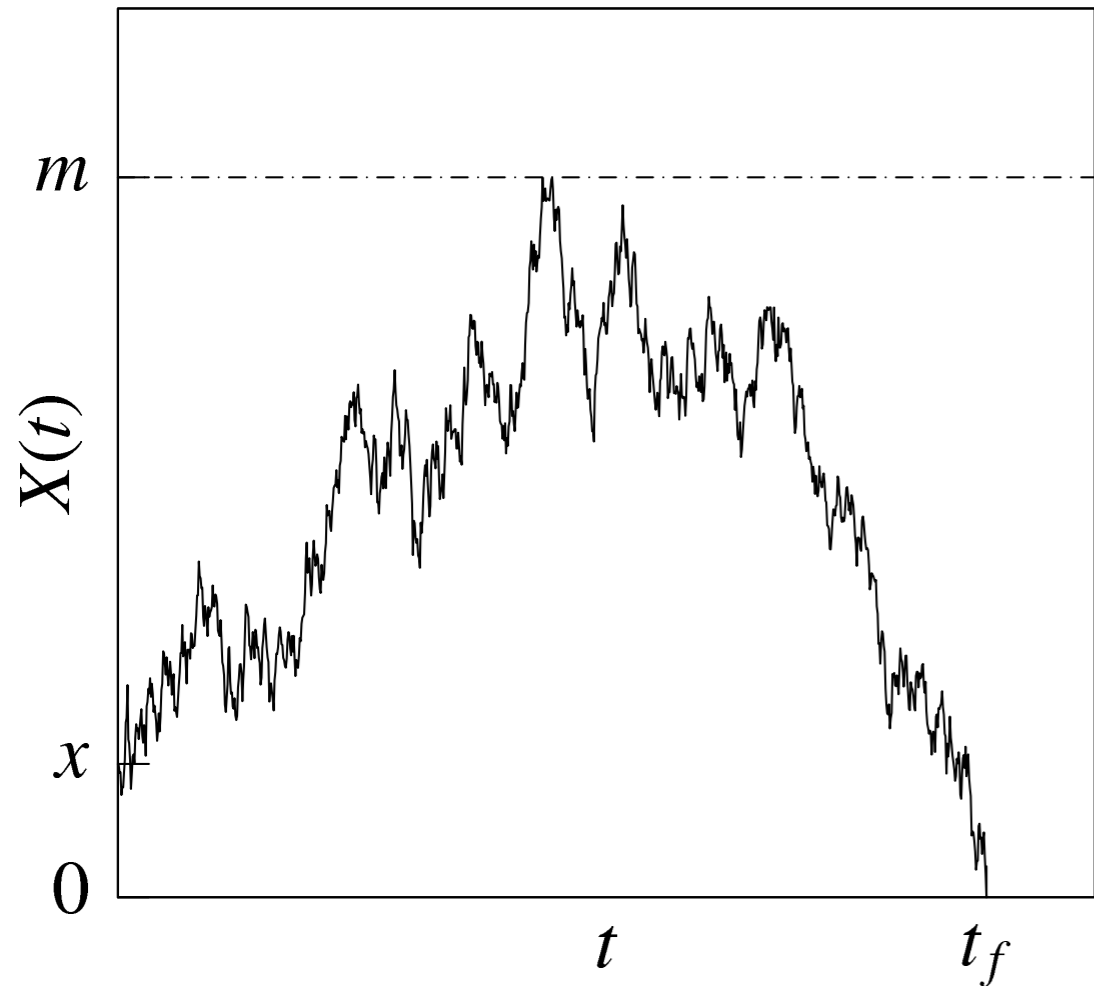
$$Q(x, L) = Q\left(z = \frac{x}{L}\right)$$



- Basic:  $Q(z) = 1 - Q(1 - z)$ ;  $Q(1/2) = 1/2$ ;  $Q(z) = 0$
- $H = 1/2$ :  $Q(z) = z$
- Expansion:  $Q(z) \sim c_1 z^\phi + \dots$

Translocation is enhanced or suppressed  
by excluded volume effects?

# A scaling argument: $\phi = \frac{\theta}{H}$



- $Q(x, L) = \text{Prob}[m > L]$

- Self-affinity I:  $m \sim t_f^H$

$$\implies Q(x, L) \simeq \text{Prob}[t_f^H > L]$$

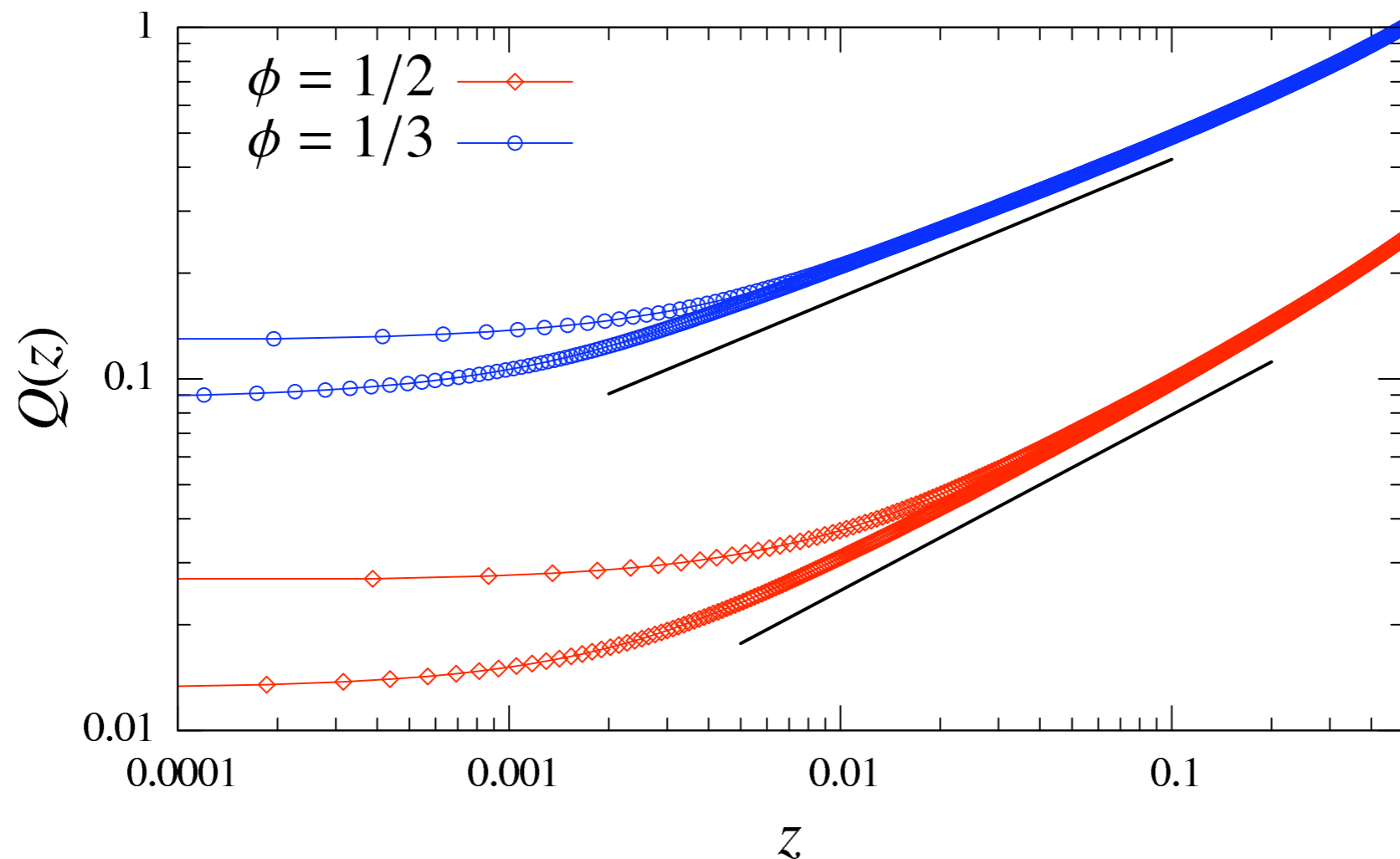
$$Q(x, L) \sim \text{Prob}[t_f > L^{\frac{1}{H}}]$$

- Survival probability:  $\text{Prob}[t_f > T] = S(x, T) \sim \frac{f(x)}{T^\theta}$   
 $\theta$  persistence exponent

- Self-affinity II:  $S(x, T) \sim \left(\frac{x}{T^H}\right)^{\frac{\theta}{H}}$

$$\implies Q(x, L) \sim \text{Prob}[t_f > L^{\frac{1}{H}}] = S(x, L^{\frac{1}{H}}) \sim \left(\frac{x}{L}\right)^{\frac{\theta}{H}}, \quad \phi = \frac{\theta}{H}$$

# Hitting probability: numerical test



Persistence of fBm in known  $\theta = 1 - H$  (see Krug et al.)

Prediction:  $\phi = \frac{\theta}{H} = \frac{1-H}{H}$

• Blue:  $H = 3/4 \longrightarrow \phi = 1/3$

• Red:  $H = 2/3 \longrightarrow \phi = 1/2$

**Conclusion: volume effects “suppress” Translocation**

# Previous results... recast using $\phi$

(i) Random acceleration process:  $\ddot{x} = \eta(t)$ ,  $x(0) = x$ ,  $\dot{x}(0) = 0$

Burkhardt:  $H = \frac{3}{2}$  and  $\theta = \frac{1}{4} \implies \phi = \frac{1}{6}$

$$Q(z = \frac{x}{L}) = I_z[\frac{1}{6}, \frac{1}{6}]. \quad \text{Where } I_z(\phi, \phi) = \frac{\Gamma(2\phi)}{\Gamma^2(\phi)} \int_0^z \frac{du}{[u(1-u)]^{1-\phi}}$$

---

(ii) Lévy flights:  $x_{t+1} = x_t + \xi_t$  with  $\pi(\xi) \sim \xi^{-(\mu+1)}$

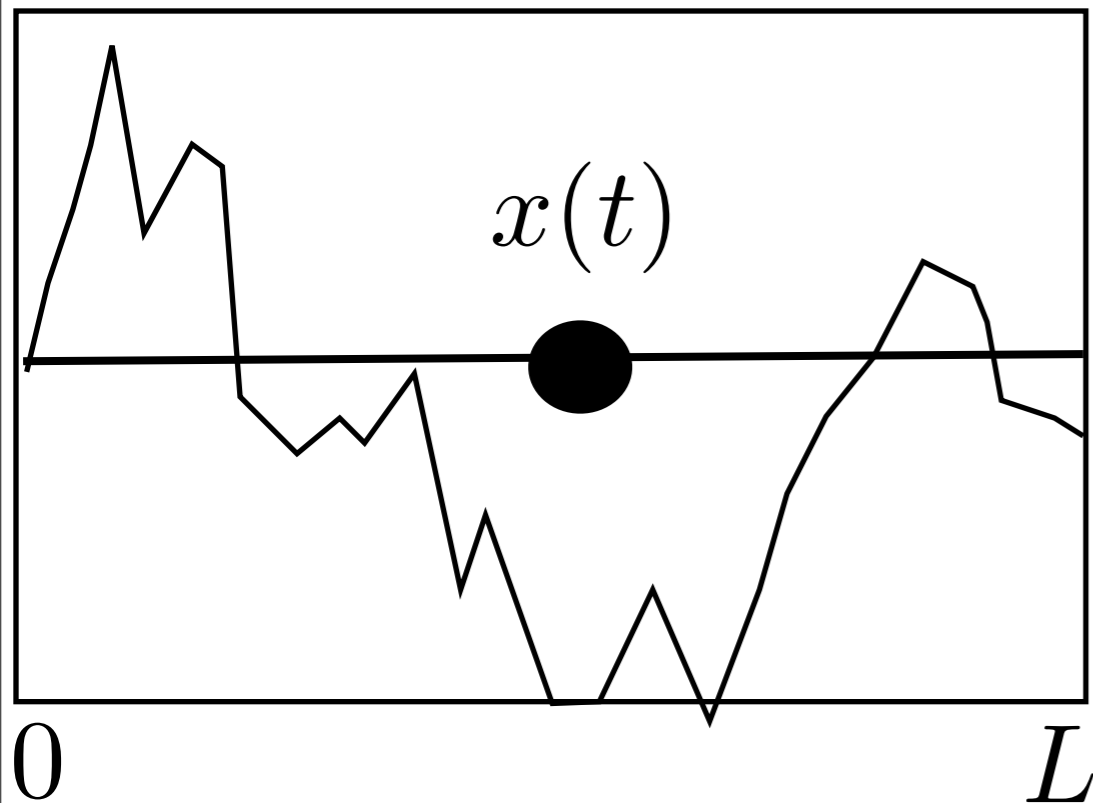
Sparre Andersen:  $H = \frac{1}{\mu}$  and  $\theta = \frac{1}{2} \implies \phi = \frac{\mu}{2}$

Widom ('61):  $Q(z = \frac{x}{L}) = I_z[\frac{\mu}{2}, \frac{\mu}{2}]$



# Sinai model

$V(x)$



$$t \sim e^{V(x)} \sim e^{\sqrt{x}} \Rightarrow x(t) \sim [\log(t)]^2$$

$$S(x_0, t) \sim [\log(t)]^{-1}$$

Using  $\tau = \log(t) \implies H = 2$  and  $\theta = 1$

From Backward Fokker Planck:  $Q(x, L) = \frac{\int_0^x e^{\beta V(x')} dx'}{\int_0^L e^{\beta V(x')} dx'}$

$$\overline{Q(z)} = \int_0^z dz' \overline{p_{eq}(z')}; \quad \text{From } \overline{p_{eq}(z)} = \frac{1}{\pi} \frac{1}{\sqrt{z(1-z)}}$$

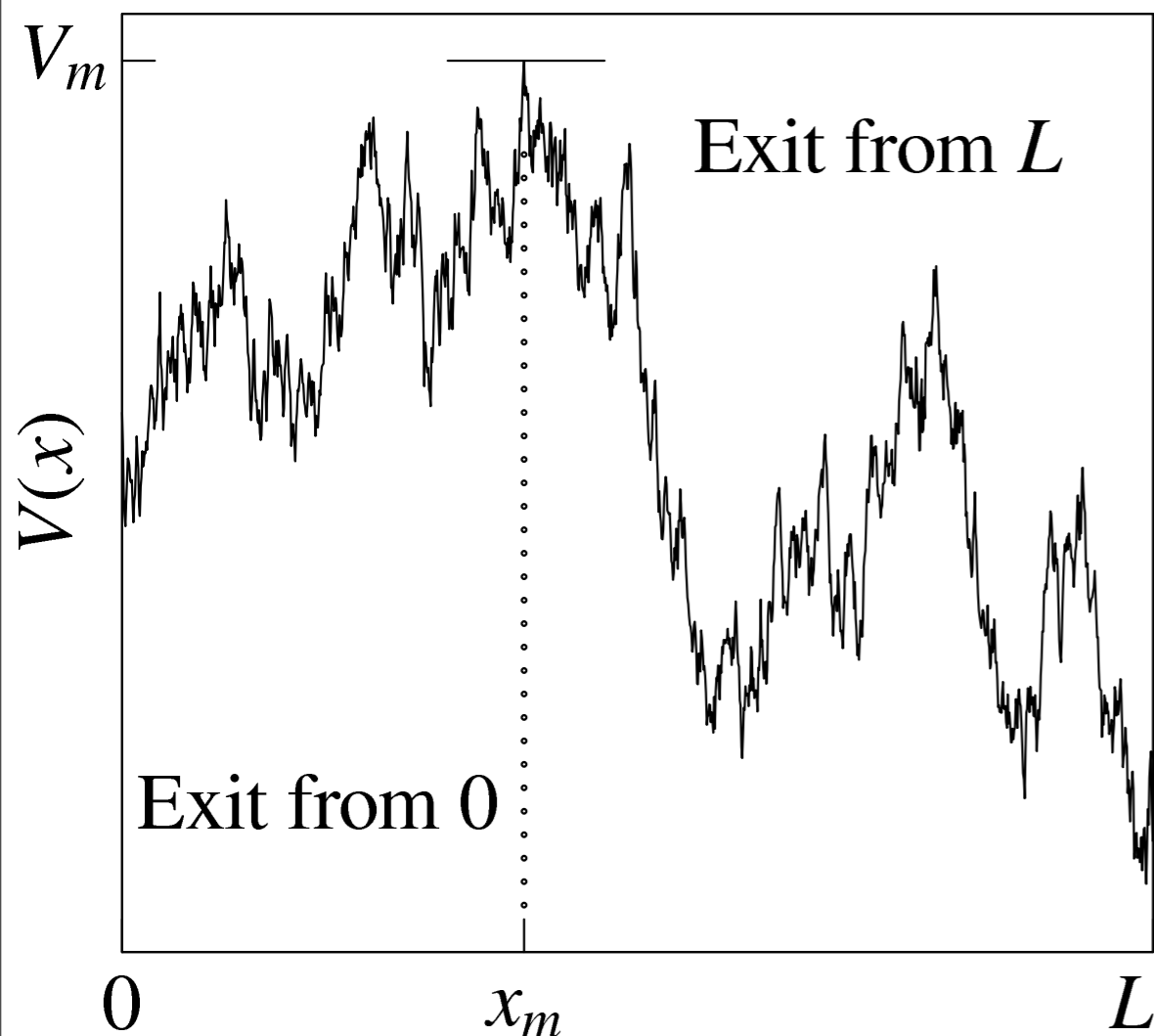
We deduce  $\overline{Q(z)} = I_z(1/2, 1/2) = \frac{2}{\pi} \arcsin(\sqrt{z}) \sim \sqrt{z}$

So that  $\phi = \theta/H = 1/2$

# Extreme statistics: maximum location in $V(x) \sim x^{H_V}$

$$p_{eq}(x, L) = \frac{e^{\beta V(x)}}{\int_0^L e^{\beta V(x')} dx'} \propto \left\{ \int_0^1 dz' \exp [\beta L^{H_V} \cdot (V(z') - V(z))] \right\}^{-1}$$

$$p_{eq}(x, L) \sim \delta(x - x_m) \implies \overline{p_{eq}(x, L)} \sim p_V(x_m, L) \quad [\text{Sinai: arcsine law}]$$



Fixed realization:  $Q(x, L) = \theta(x - x_m)$

Average  $\overline{Q(z = \frac{x}{L})} = \text{Prob}[z_m < z]$

If  $V(x) \sim x^{H_V}$ , using Arrhenius  $H = 1/H_V$

$$\text{Prob}[x_m < x \rightarrow 0] \sim \text{Prob}[V < 0 \text{ up to } L] \sim \frac{1}{L^{\theta_V}}$$

$$\text{So that } Q(x, L) \rightarrow \left(\frac{x}{L}\right)^{\theta_V}$$

**We Conclude  $\phi = \theta_V$  and  $\theta = H \cdot \phi = \theta_V / H_V$**

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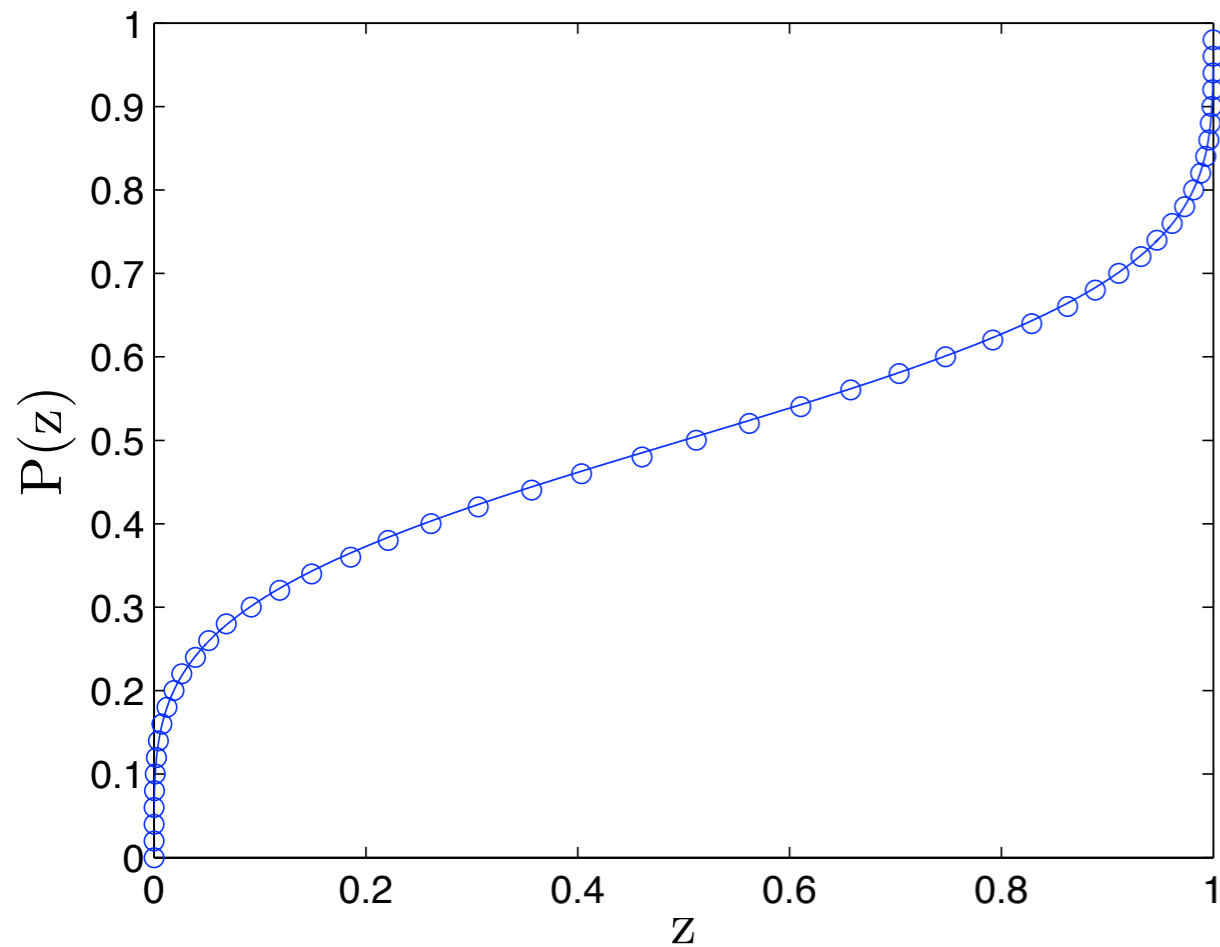
$V(x)$  is a random acceleration process:  $H_V = 3/2$ ,  $\theta_V = 1/4$

$$\text{Bridge case } (V'(L) = 0): \quad p(z_m) = \frac{\Gamma(1/2)}{\Gamma^2(1/4)} \frac{1}{(z_m(1-z_m))^{3/4}}$$

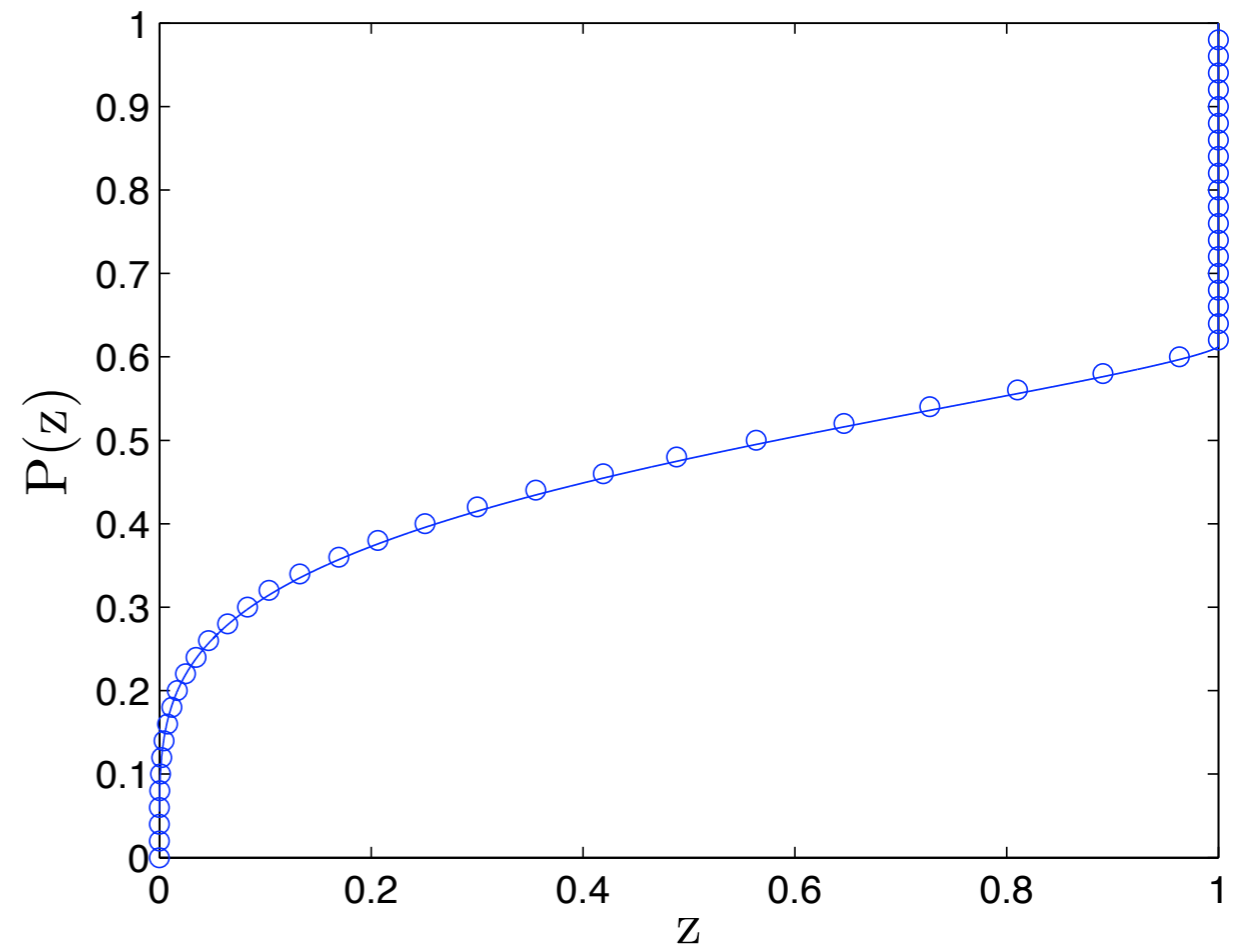
$$\text{Free case: } p(z_m) = \left(1 - \sqrt{\frac{3}{8}}\right) \delta(z_m - 1) + \frac{\sqrt{3}}{4\pi z_m^{3/4} (1-z_m)^{1/4}}$$

(See also Maximum location, for CTRW : Le Doussal and Schehr )

# Hitting Bridge

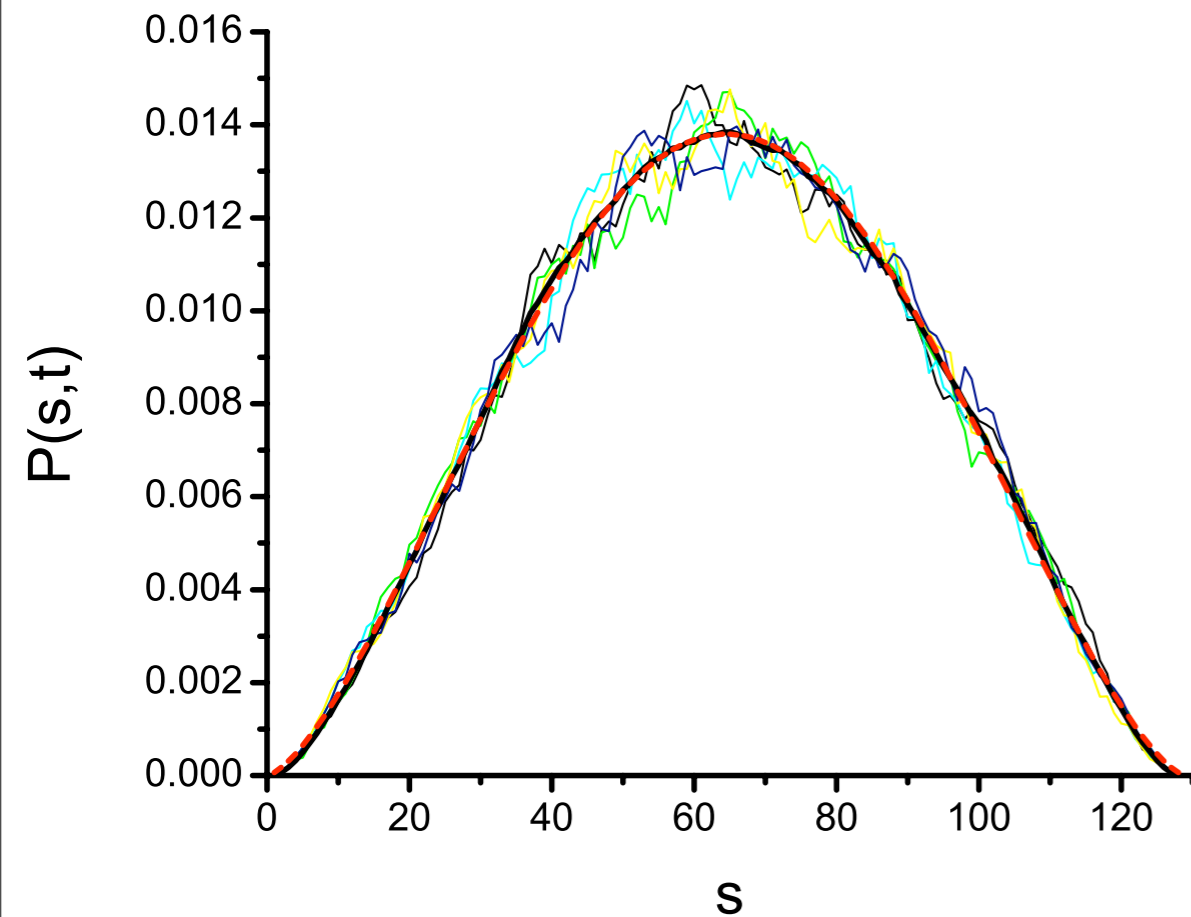


# Hitting Free

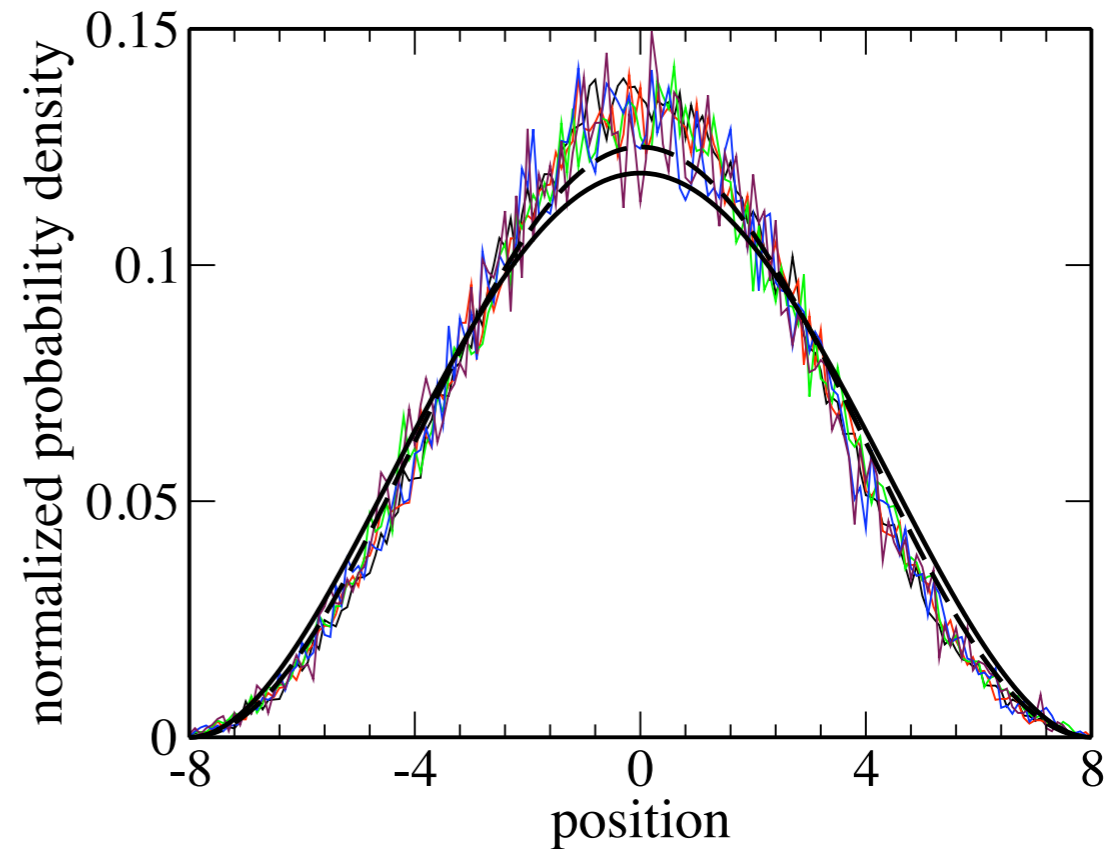


...and the persistence  $S(x_0, t) \sim \frac{1}{\log(t)^{\frac{1}{6}}}$

# “Anomalies” of anomalous diffusion



Monte Carlo simulation of polymer translocation in  $d=2$ ,  
Chatelain, Kantor, Kardar, PRE 78, 021129 (2008)



Monte Carlo simulation tagged monomer in a box ( $d=1$ )  
Kantor, Kardar, PRE 76, 061121 (2007)

$$d = 2, \quad \nu = \frac{3}{4}, \quad H = \frac{1}{2\nu + 1} = \frac{2}{5}$$

$$d = 1, \quad H = \frac{1}{4}$$

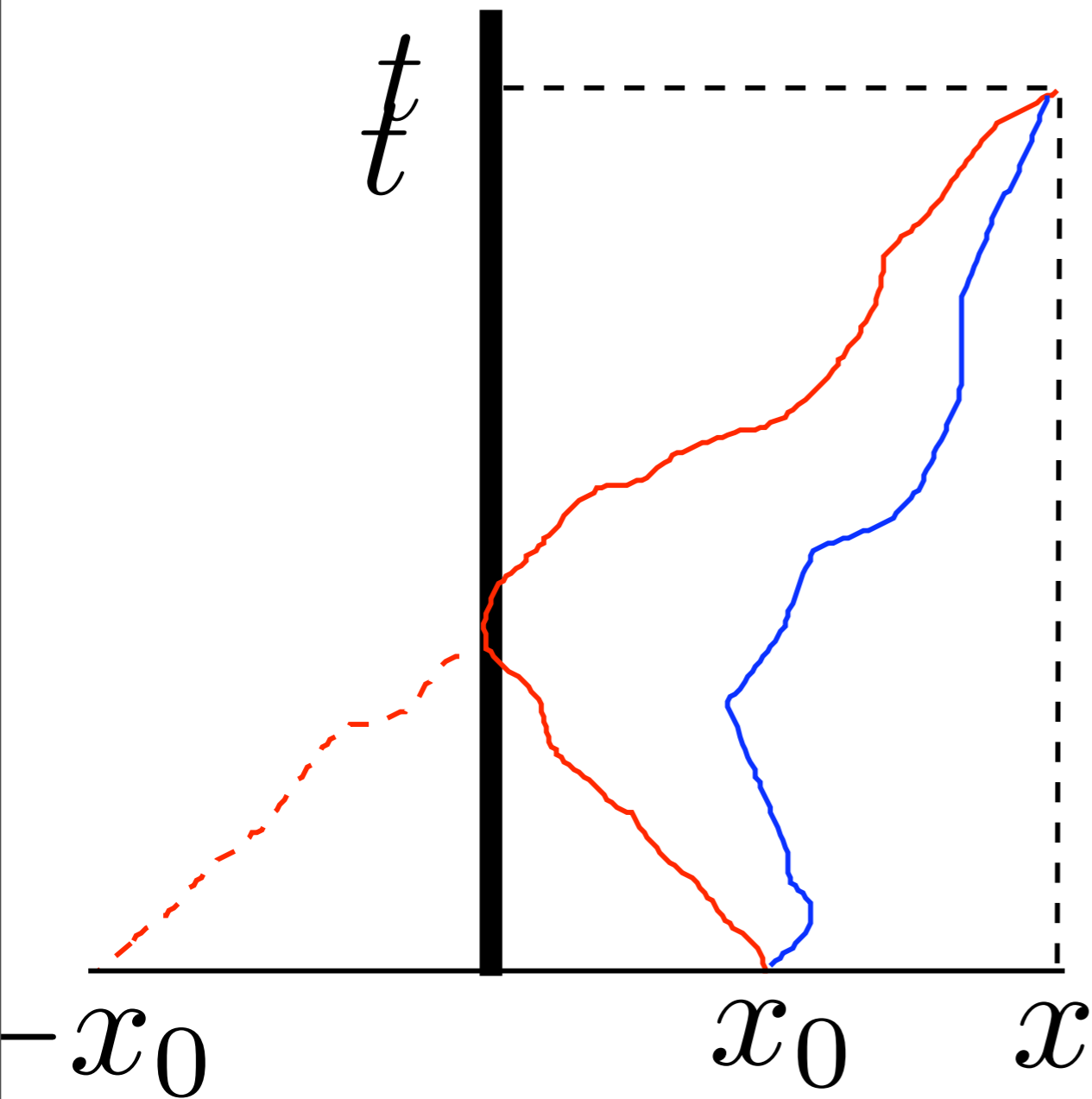
At large  $t$ ,  $P(s, t) \sim s^{1.44}$

At large  $t$ ,  $P(s, t) \sim s^\alpha$ , with  $\alpha > 2$

We predict  $\phi = \frac{1-H}{H} = 1.5$

We predict  $\phi = \frac{1-H}{H} = 3$

# Single Boundary: Images method



$$Z_+(x, x_0, t) = Z(x, x_0, t) - Z(x, -x_0, t)$$

After normalization

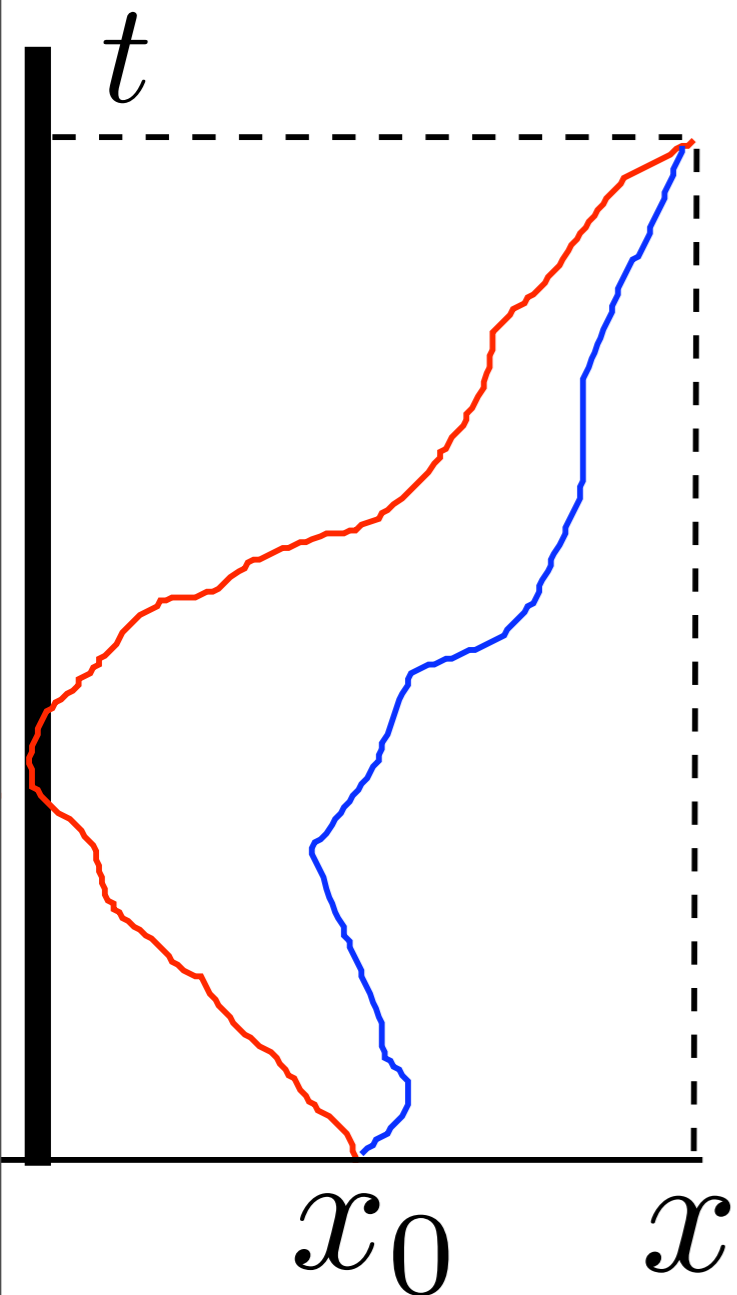
$$P_+(x, x_0, t) \xrightarrow{t \rightarrow \infty \text{ or } x_0 \rightarrow 0} P_+(x, t)$$

Self-Affinity III:  $y = \frac{x}{\sqrt{2Dt}}$

$$P_+(x, t) dx = R_+(y) dy = y e^{-\frac{y^2}{2}} dy$$

**Conclusion :**  $R_+(y) = y e^{-\frac{y^2}{2}}$

# Perturbation Theory



$$Z_+(x_0, x, t) = \int_{x(0)=x_0}^{x(t)=x} \mathcal{D}[x] e^{-\mathcal{S}[x]} \Theta[x]$$

$$P_+(x, t) = \lim_{x_0 \rightarrow 0} \frac{Z_+(x_0, x, t)}{\int_0^\infty dx Z_+(x_0, x, t)}$$

$$\mathcal{S}[x] = \int_0^t dt_1 \int_0^t dt_2 \frac{1}{2} x(t_1) G(t_1, t_2) x(t_2)$$

$$G^{-1}(t_1, t_2) = \langle x(t_1) x(t_2) \rangle$$

## Brownian motion

$$H = \frac{1}{2} \quad \Rightarrow \quad \langle x(t_1)x(t_2) \rangle = 2 \min(t_1, t_2) \quad \Rightarrow \quad \mathcal{S}^{(0)}[x] = \frac{1}{4} \int_0^t dt' (\partial_{t'} x)^2$$

## Fractional Brownian motion

$$H - \text{fBm} \quad \Rightarrow \quad \langle x(t_1)x(t_2) \rangle = t_1^{2H} + t_2^{2H} - |t_1 - t_2|^{2H} \quad \Rightarrow \quad \mathcal{S}[x] ??$$

## Perturbation

$$H = \frac{1}{2} + \epsilon \quad \Rightarrow \quad \langle x(t_1)x(t_2) \rangle = 2 \min(t_1, t_2) + \epsilon K(t_1, t_2) + O(\epsilon^2)$$

$$K(t_1, t_2) = 2 [t_1 \ln t_1 + t_2 \ln t_2 - |t_1 - t_2| \ln |t_1 - t_2|]$$



$$\mathcal{S}[x] = \int_0^t dt_1 \int_0^t dt_2 \frac{1}{2} x(t_1) G(t_1, t_2) x(t_2)$$

$$G = G^{(0)} - \epsilon G^{(0)} K G^{(0)}$$

$$\mathcal{S}[x] = \mathcal{S}^{(0)}[x] + \epsilon \mathcal{S}^{(1)}[x]$$

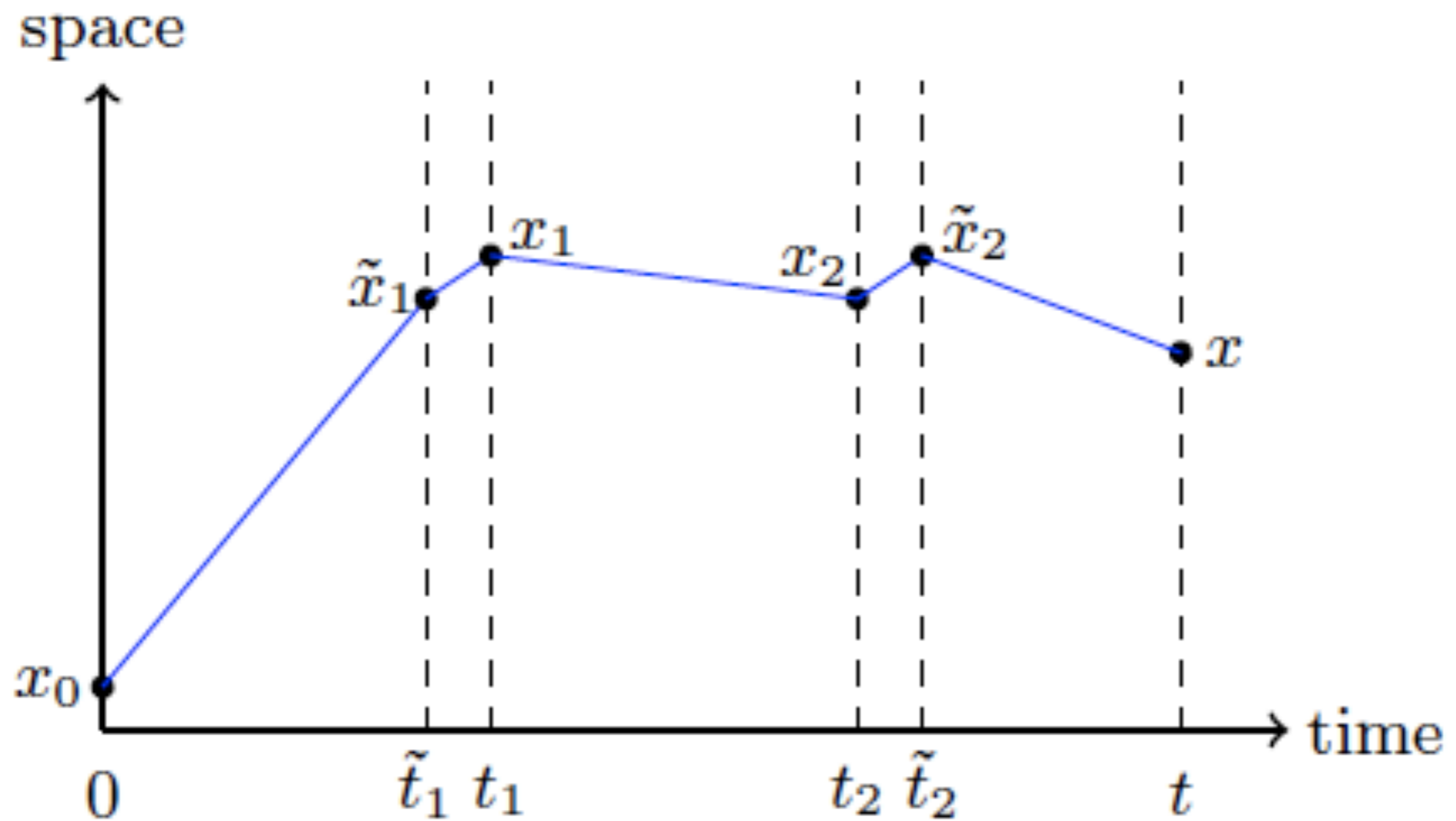
$$\mathcal{S}^{(1)}[x] \propto -\frac{1}{2} \int_0^t dt_1 \int_{t_1}^t dt_2 \frac{\partial_{t_1} x(t_1) \partial_{t_2} x(t_2)}{|t_1 - t_2|}$$

$$e^{-\mathcal{S}[x]} \sim e^{-\mathcal{S}^{(0)}[x]} \left( 1 + \epsilon \mathcal{S}^{(1)}[x] \right)$$

$$Z_+(x_0, x, t) = \int_{x(0)=x_0}^{x(t)=x} \mathcal{D}[x] e^{-\mathcal{S}[x]} \Theta[x]$$

$$Z_+(x_0, x, t) \sim Z_+^{(0)}(x_0, x, t) + \epsilon Z_+^{(1)}(x_0, x, t)$$

$$Z_+^{(1)}(x_0, x, t) = \int_{x(0)=x_0}^{x(t)=x} \mathcal{D}[x] \mathcal{S}^{(1)}[x] e^{-\mathcal{S}^{(0)}[x]} \Theta[x]$$



Brownian 2-points  
correlation function

# Final Result I

$$R_+(y) = R_+^{(0)}(y) [1 + \epsilon W(y) + O(\epsilon^2)]$$

$$W(y) = \frac{1}{6} y^4 {}_2F_2 \left( 1, 1; \frac{5}{2}, 3; \frac{y^2}{2} \right)$$

$$+ \pi(1 - y^2) \operatorname{erfi} \left( \frac{y}{\sqrt{2}} \right) + \sqrt{2\pi} e^{\frac{y^2}{2}} y$$

$$+ (y^2 - 2) [\log(2y^2) + \gamma_E] - 3y^2$$

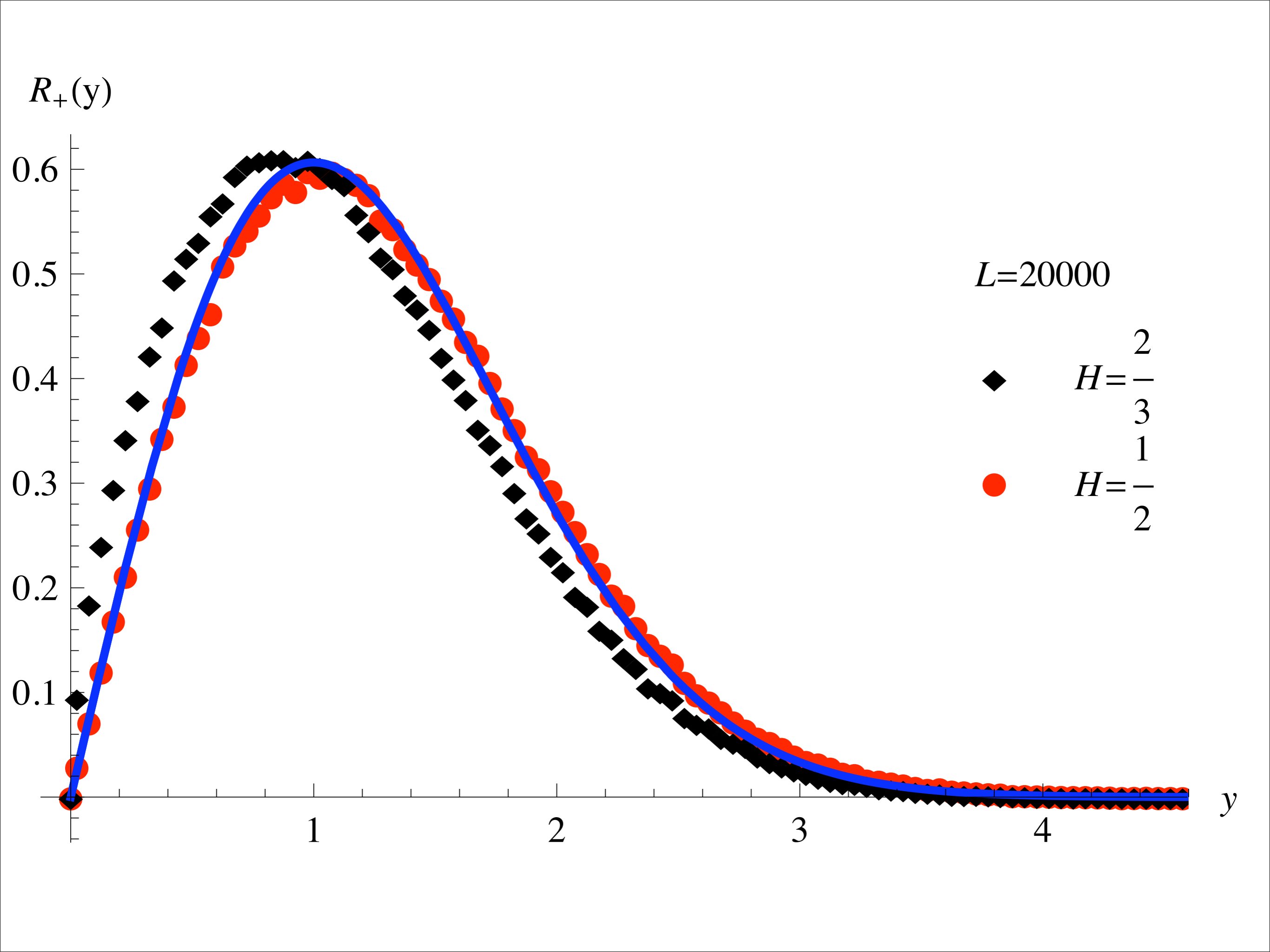
# Final Result II

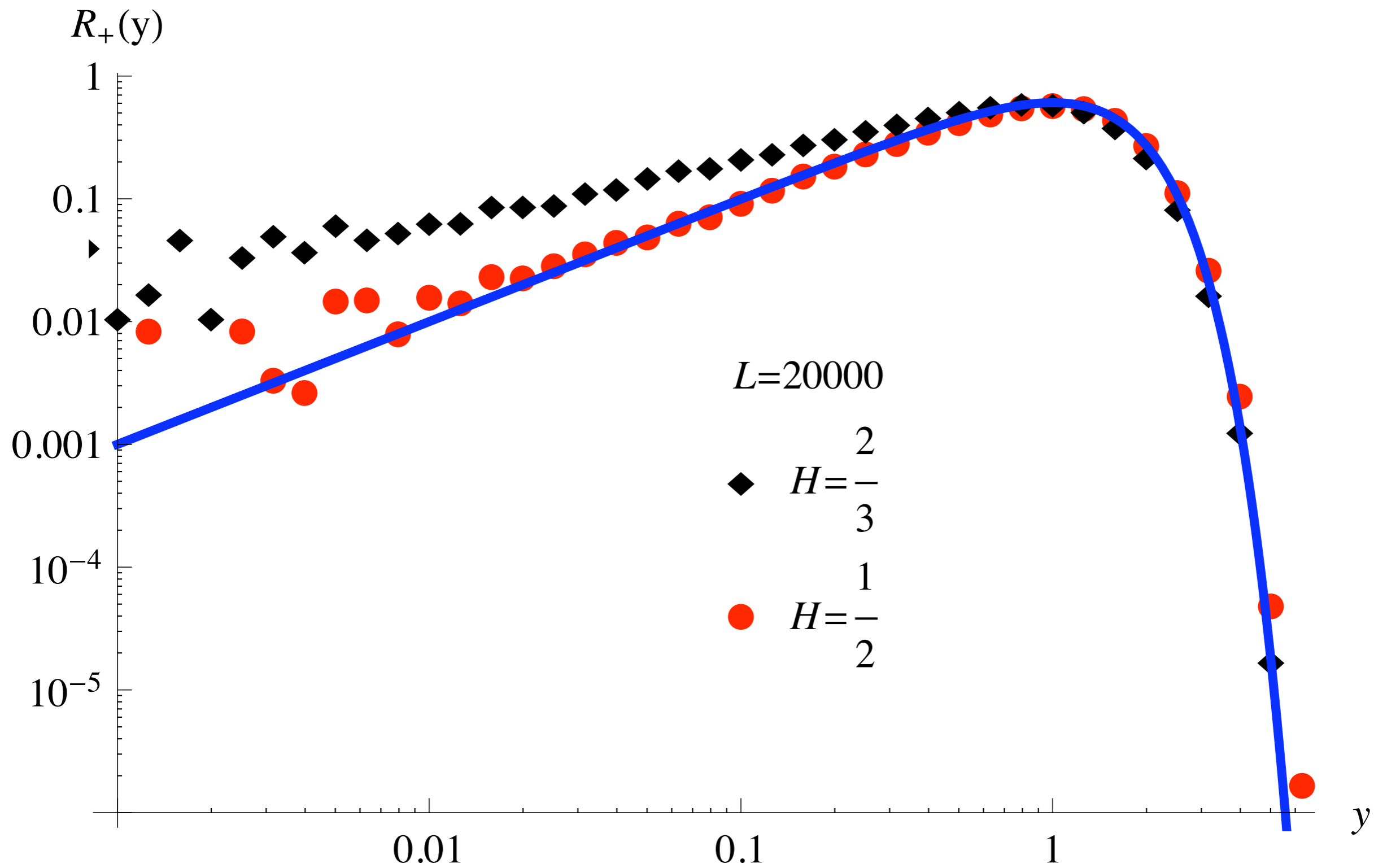
$$R_+(y) \xrightarrow{y \rightarrow 0} y^\phi$$

$$R_+(y) \xrightarrow{y \rightarrow \infty} y^\gamma e^{-\frac{y^2}{2}}$$

$$\phi = 1 - 4\epsilon + O(\epsilon^2) , \quad \gamma = 1 - 2\epsilon + O(\epsilon^2) .$$

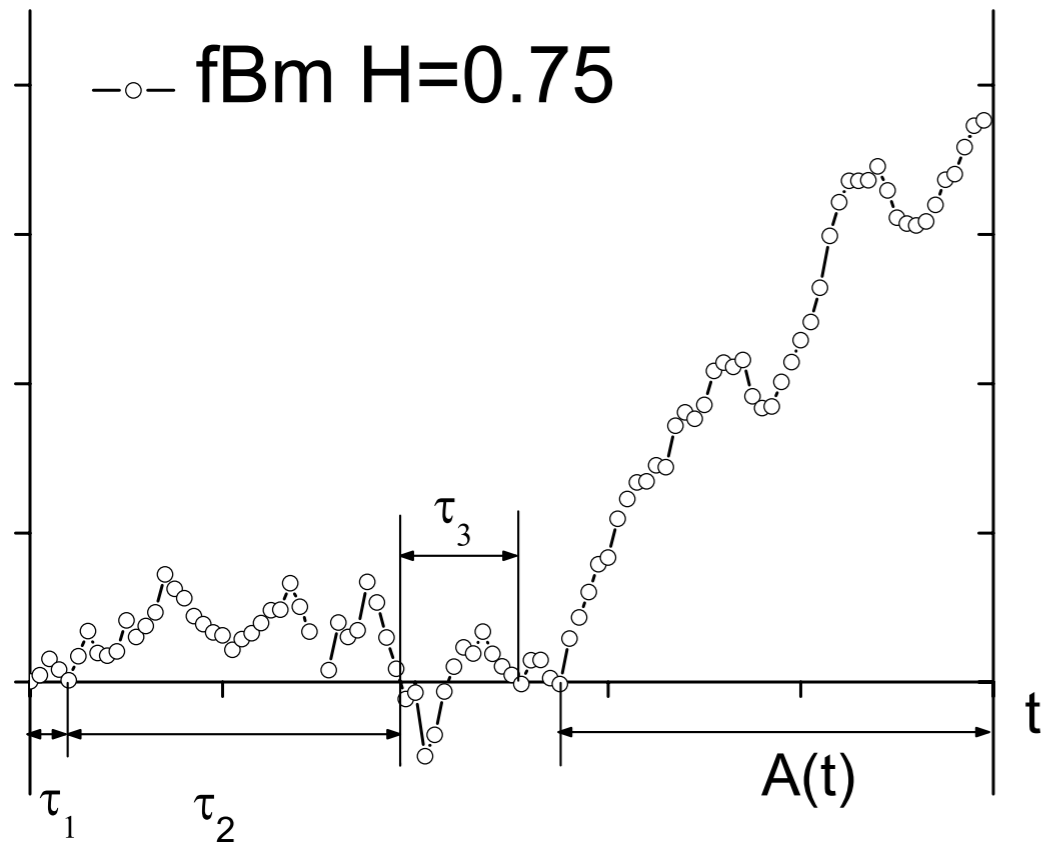
- $\epsilon$  expansion in agreement with the conjecture  $\phi = \frac{1-H}{H}$
- At large  $y$ , Free Gaussian Propagator
- + a New Exponent  $\gamma \neq \phi$





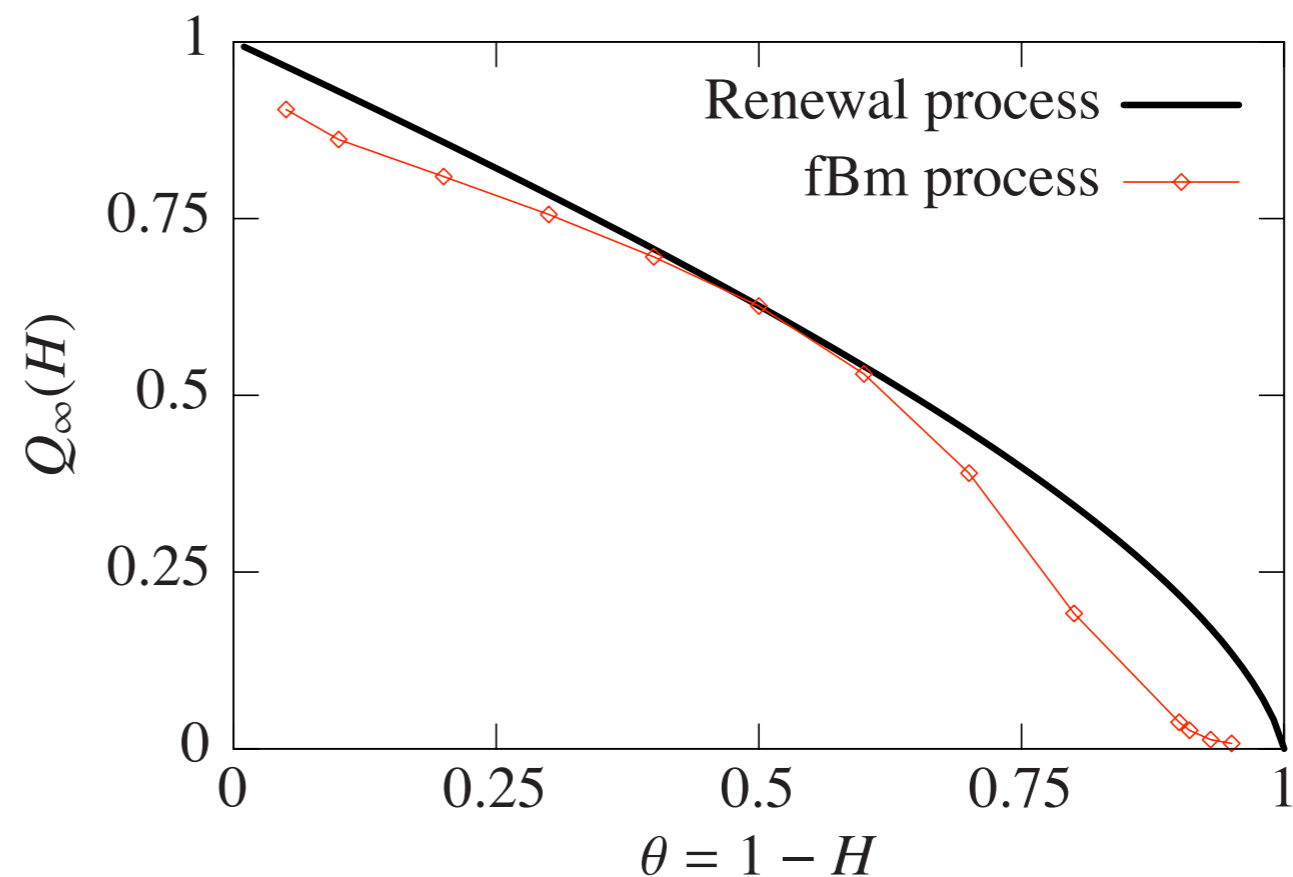
# Non Markovianity and extremes...

The probability  $Q(t)$  that  $A(t)$  is the *longest excursion*?



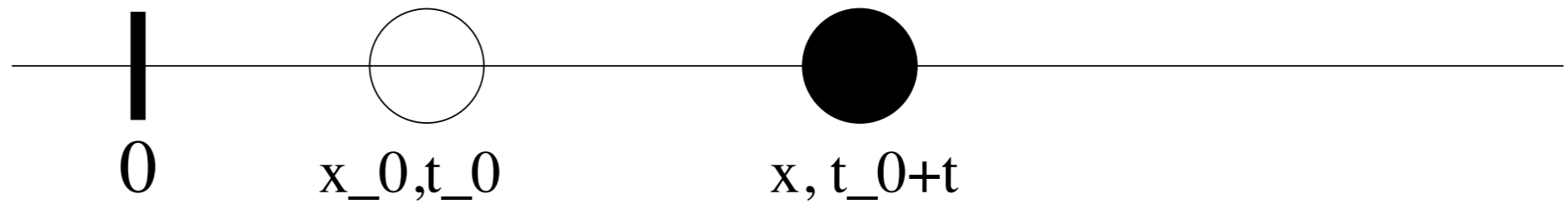
with Schehr and Garcia

Exact solution for RENEWAL process  
[Godrèche, Majumdar, Schehr]

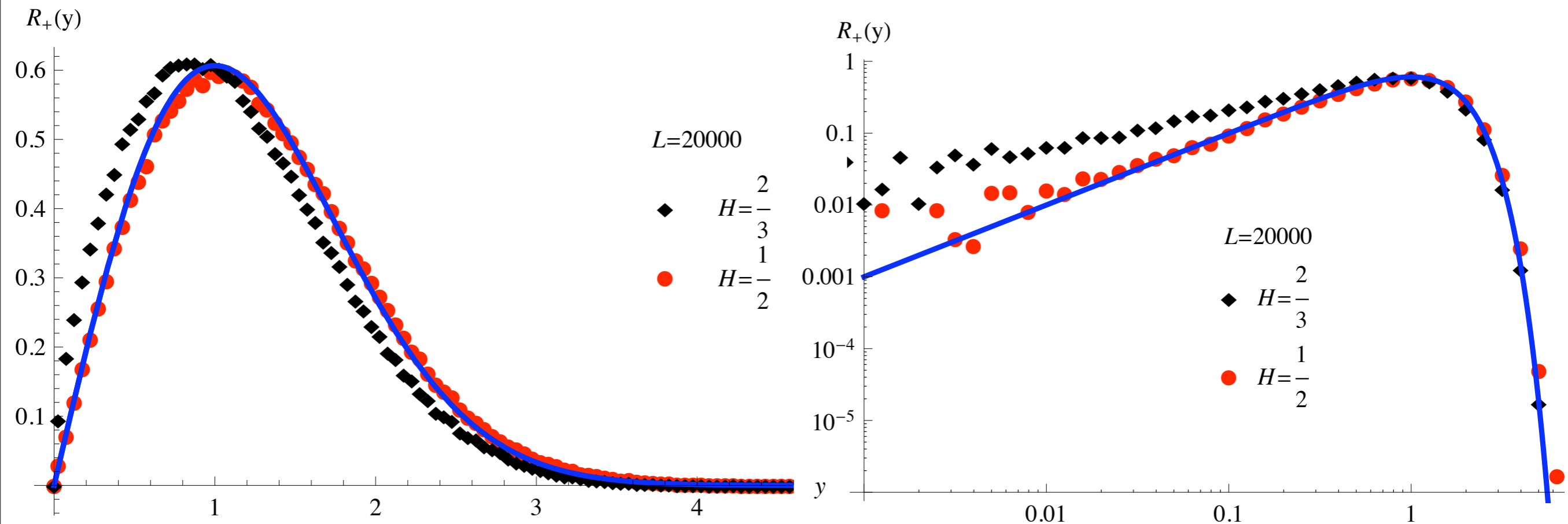




# Single Boundary



- Large time:  $P_+(x, t; x_0, t_0) \xrightarrow{t \rightarrow \infty} P_+(x, t)$
- Self Affinity III:  $P_+(x, t) = R_+(y)$  with  $y = \frac{x}{\sqrt{\langle x^2(t) \rangle}} \sim \frac{x}{t^H}$
- Brownian  $H = 1/2$  :  $R_+(y) = ye^{-\frac{y^2}{2}} \sim y + \dots$



# How to generate a correlated path:

- time discretization:  $x(t) \longrightarrow x_t = x_0 + \sum_{t'=1}^t \xi_{t'}$
- $\vec{\xi} = \{\xi_1, \xi_2, \dots, \xi_T\}$  is a vector of Gaussian numbers with a given covariance matrix  $\langle \xi_{t_1} \xi_{t_2} \rangle = C(|t_1 - t_2|)$
- we compute  $A = \sqrt{C}$  [ $T^3$  operations] and generate uncorrelated Gaussian numbers  $\vec{\epsilon} = \{\epsilon_1, \dots, \epsilon_T\}$
- Then the vector  $\vec{\xi}$  can be generated [ $T^2$  operations]:

$$\vec{\xi} = A\vec{\epsilon}$$

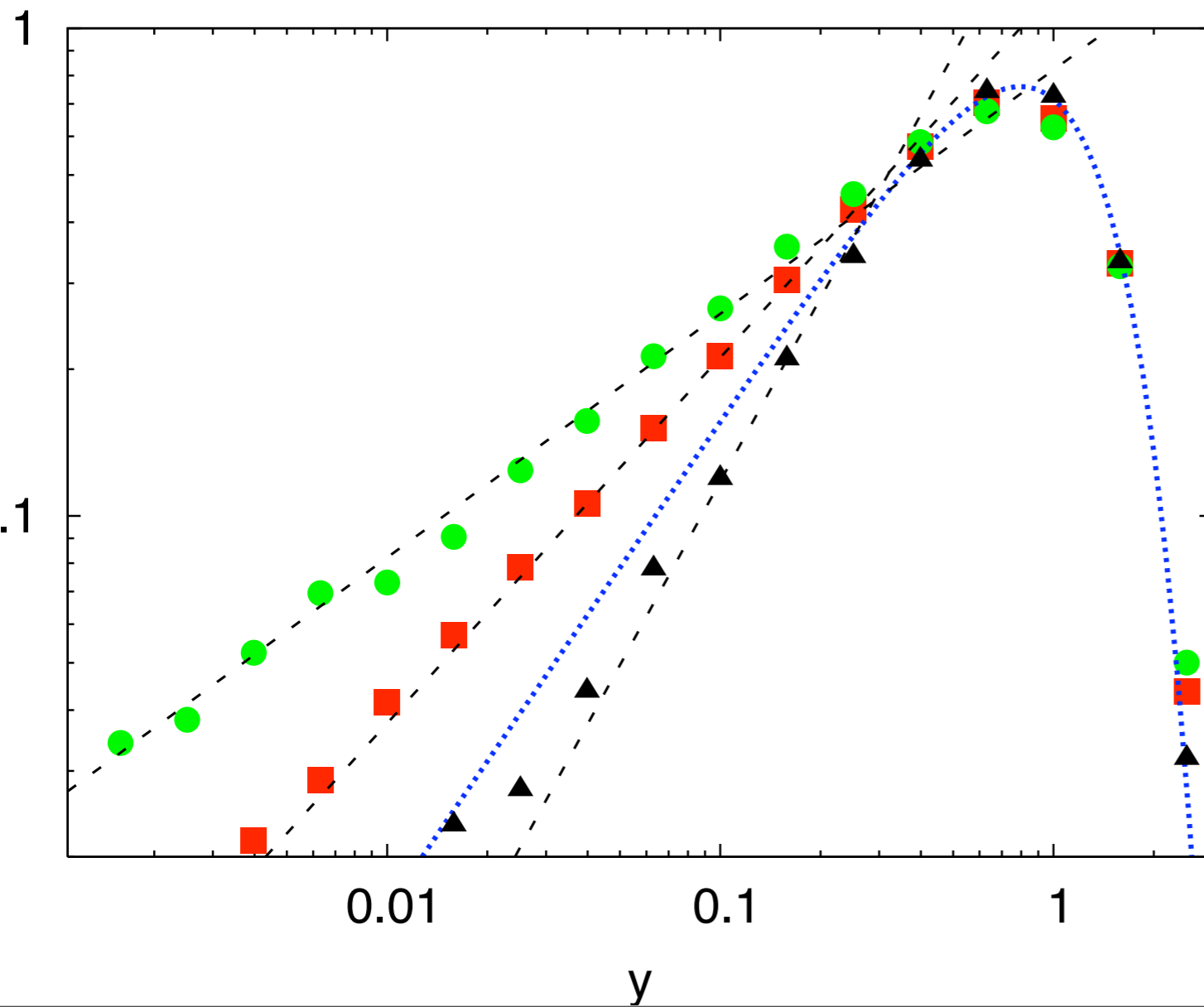
Proof:

$$\langle \xi_i \xi_j \rangle = \left\langle \sum_{i'} A_{i,i'} \epsilon_{i'} \sum_{j'} A_{j,j'} \epsilon_{j'} \right\rangle = \sum_{i',j'} A_{i,i'} A_{j,j'} \delta_{i',j'} = (A^2)_{i,j} = C_{i,j}$$

- Special algorithms if Covariance is Toeplitz matrix:  $C_{i,j} = C(|i - j|)$ .

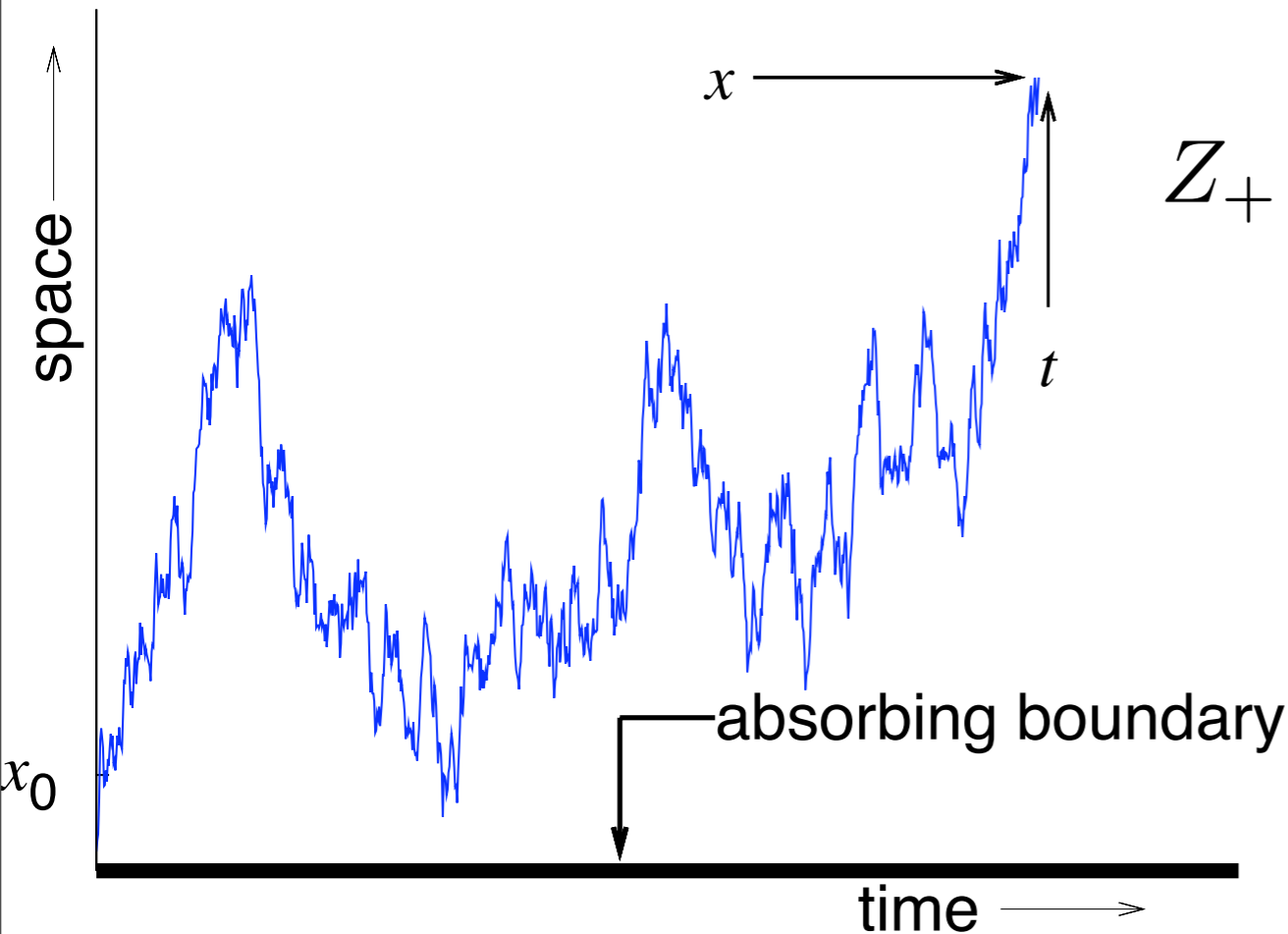
# Single Boundary

- In general we expect  $R_+(y) \sim y^\phi$
- For stationary increments, a scaling argument gives  $\phi = \frac{\theta}{H}$
- numerical simulations agrees



- Circle  $H = \frac{2}{3}$ ,  $\phi = \frac{1}{2}$
- Square  $H = \frac{4}{7}$ ,  $\phi = \frac{3}{4}$
- Line  $R_+(y) = ye^{-y^2/2}$
- Triangle  $H = \frac{4}{9}$ ,  $\phi = \frac{5}{4}$

# Perturbation Theory



$$Z_+(x_0, x, t) = \int_{x(0)=x_0}^{x(t)=x} \mathcal{D}[x] e^{-\mathcal{S}[x]} \Theta[x]$$

$$P_+(x, t) = \lim_{x_0 \rightarrow 0} \frac{Z_+(x_0, x, t)}{\int_0^\infty dx Z_+(x_0, x, t)}$$

$$\mathcal{S}[x] = \int_0^t dt_1 \int_0^t dt_2 \frac{1}{2} x(t_1) G(t_1, t_2) x(t_2)$$

where  $G^{-1}(t_1, t_2) = \langle x(t_1) x(t_2) \rangle$  it is known

# Diffusion and Central Limit Theorem

- $\langle x^2(t) \rangle = 2Dt \propto t$  and  $D$  is diffusion constant
- $x(t)$  (for large  $t$ ) is a **Gaussian** process
- Propagator from  $x_0$  to  $x$ : 
$$P(x_0, x, t) = \frac{e^{-\frac{(x(t) - x_0)^2}{4Dt}}}{\sqrt{4\pi Dt}}$$

# And the Fractional Fokker Planck Equation (CTRW)?

