

# Diffusion with Stochastic Resetting

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## *Collaborators:*

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J. Franke (Univ. of Cologne, Germany)

Refs: [Phys. Rev. Lett. 106, 160601 \(2011\)](#)  
[J. Phys. A: Math. Theor. 44, 435001 \(2011\)](#)  
[J. Stat. Mech.: Theor. & Exp. P05024 \(2012\)](#)

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  - ⇒ **power-law** decay of the target **survival probability**  
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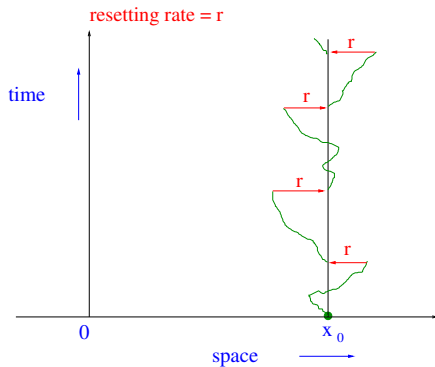
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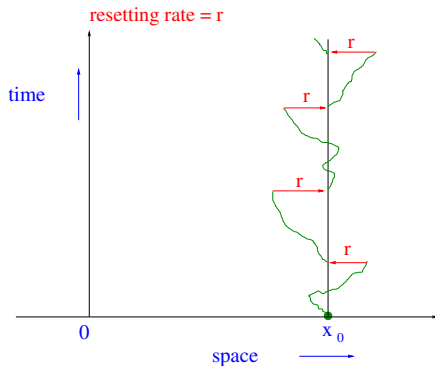
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- Summary and Conclusion

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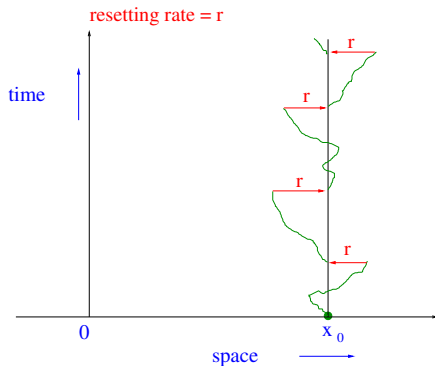
Dynamics: In a small time interval  $\Delta t$

$$x(t + \Delta t) = x_0 \quad \text{with prob. } r\Delta t \quad \text{(resetting)}$$

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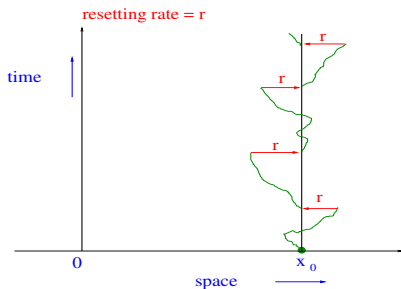
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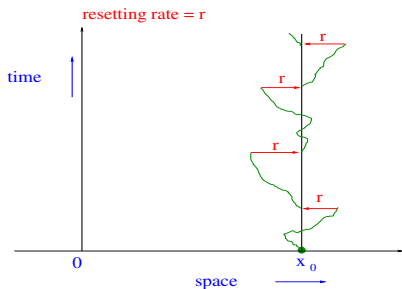
$\eta(t) \rightarrow$  Gaussian white noise:  $\langle \eta(t) \rangle = 0$  and  $\langle \eta(t)\eta(t') \rangle = 2D\delta(t - t')$

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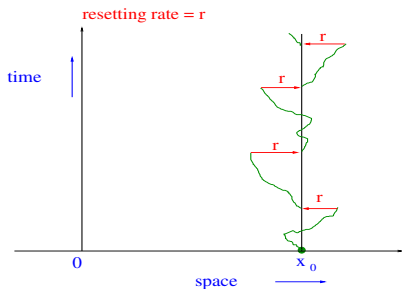


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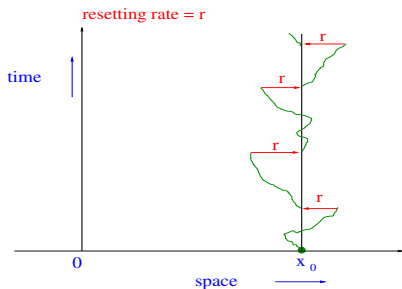
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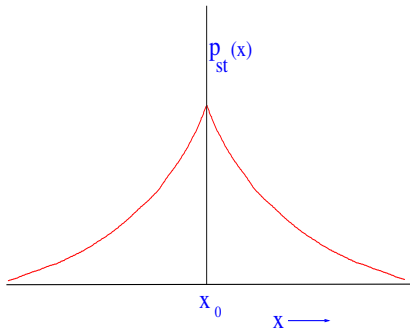


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$\rightarrow$  nonequilibrium steady state

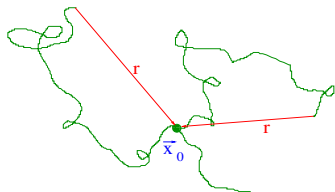
$\Rightarrow$  current carrying with  
detailed balance  $\rightarrow$  violated

$$p_{\text{st}}(x) = \alpha_0 \exp[-V_{\text{eff}}(x)]$$

effective potential:

$$V_{\text{eff}}(x) = \alpha_0 |x - x_0|$$

# Generalization to **higher** dimensions

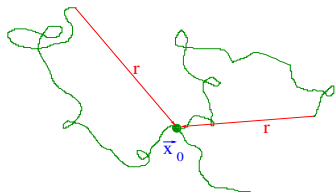


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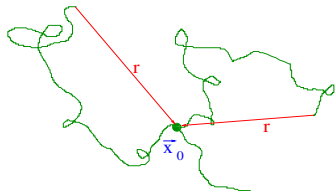
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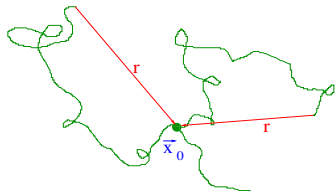
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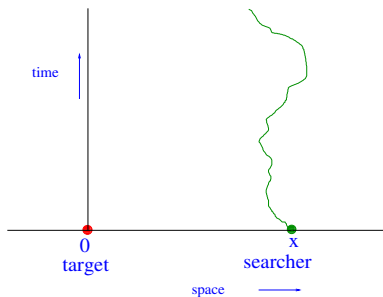
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- For  $d = 1$ , it reduces to  $\rightarrow p_{\text{st}}(x) = \frac{\alpha_0}{2} \exp[-\alpha_0 |x - x_0|]$

# Search of a fixed target by a purely **diffusive** searcher in $d = 1$ without resetting



$Q(x, t) \rightarrow$  survival prob. of the target  
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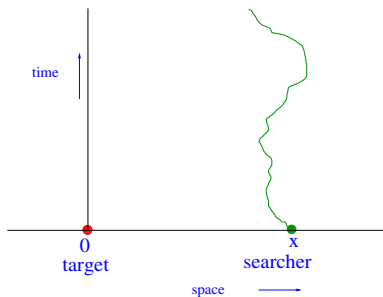
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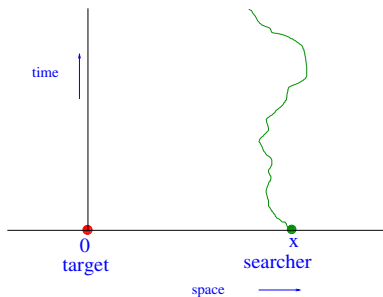
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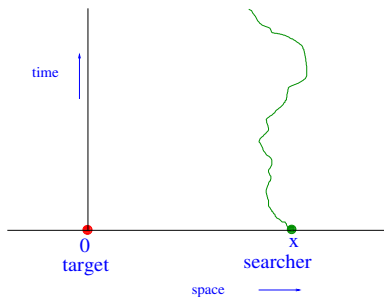
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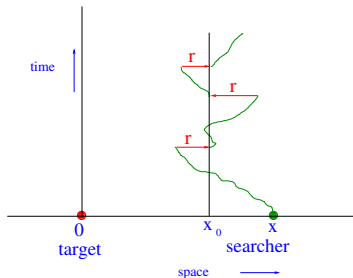
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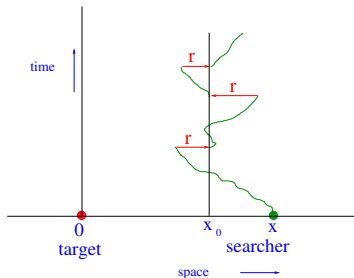
• mean capture time  $\rightarrow \bar{T} = \int_0^\infty t F(x, t) dt = \infty$

# Target search via diffusion with **resetting**



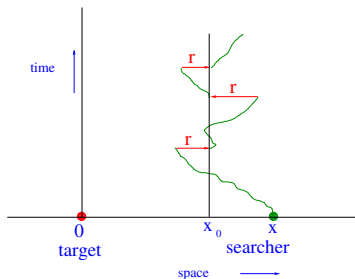
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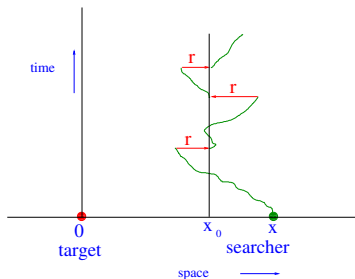
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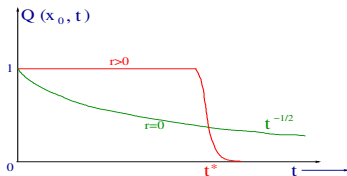
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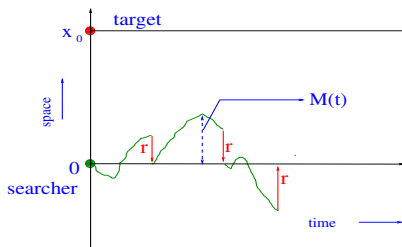
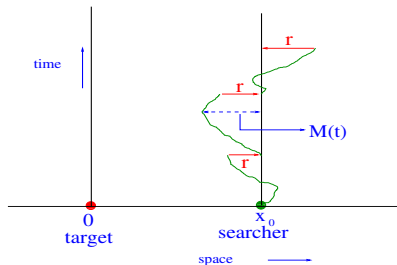


For  $r > 0$ ,  $Q(x_0, t) \approx \exp[-t/t^*]$

where  $t^* \approx (1/r) e^{\sqrt{r/D} x_0}$

# Survival Probability $\longleftrightarrow$ Extreme Value Statistics

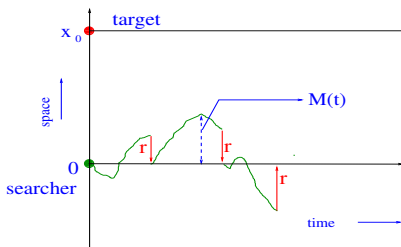
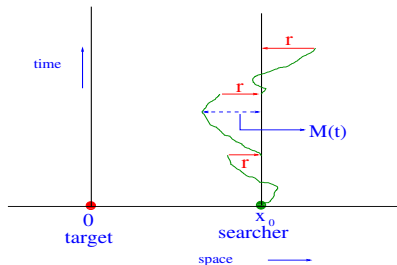
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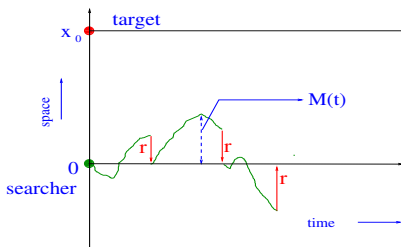
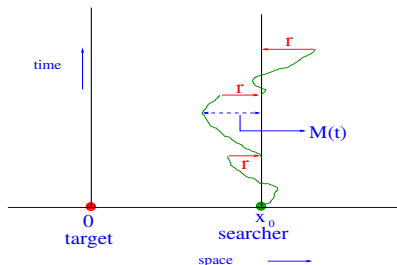
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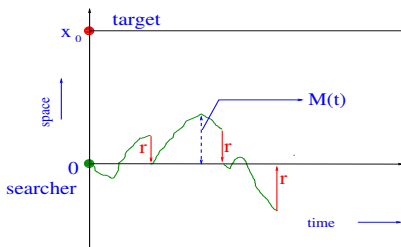
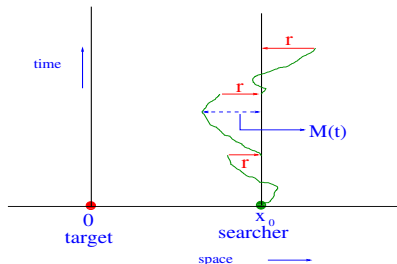


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## ROTATE & SHIFT

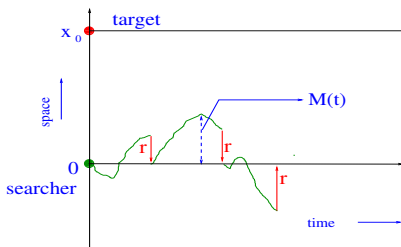
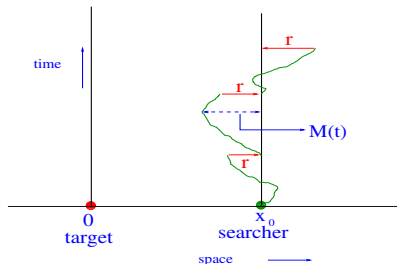


- $M(t) \rightarrow$  maximum of the process up to time  $t$
- Survival prob.  $Q(x_0, t) \equiv \text{Prob.}[M(t) \leq x_0]$
- Correlation time  $\tau = 1/r \rightarrow N_{\text{eff}} = t/\tau = rt$  effectively independent blocks in the time interval  $[0, t]$
- $Q(x_0, t) \approx \exp \left[ -rt e^{-\sqrt{r/D} x_0} \right]$



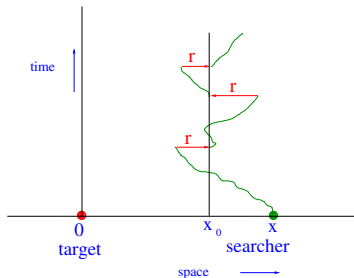
# Survival Probability $\longleftrightarrow$ Extreme Value Statistics

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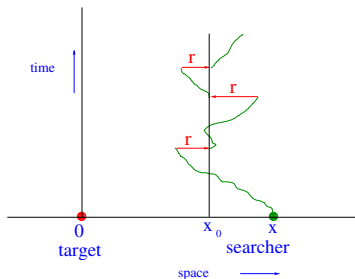
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- $\Rightarrow$  classical Gumbel distribution for the maximum of a set of  $N_{\text{eff}}$  exponentially distributed independent random variables

# Mean capture/search time



mean capture time:  $\bar{T} = \int_0^\infty t [-\partial_t Q(x_0, t)] dt = \tilde{Q}(x_0, s = 0)$

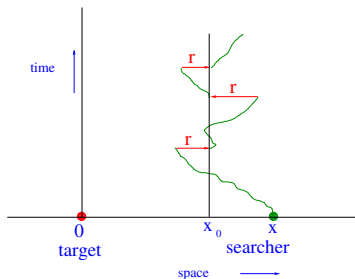
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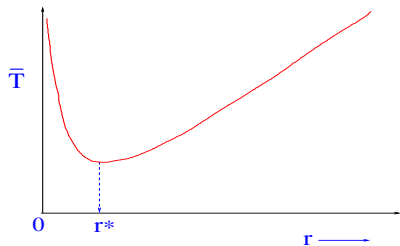


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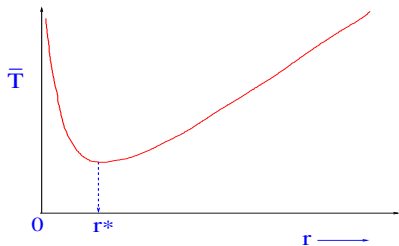
$\Rightarrow$  mean capture time is  $\infty$  for  $r = 0$ , but finite when  $r > 0$

# Optimal resetting rate



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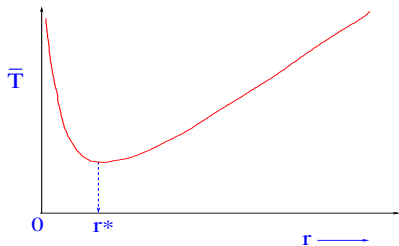
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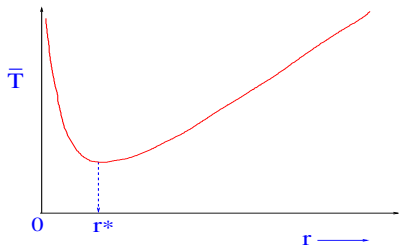
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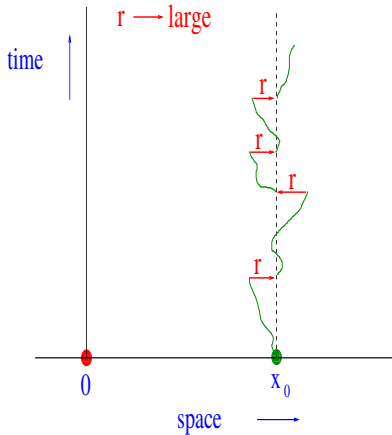
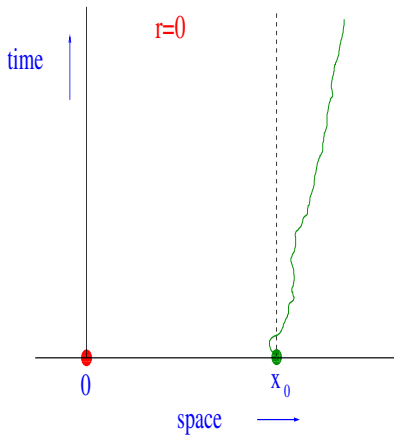
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optimal resetting rate  $r^*$  is given by:

$$r^* = \gamma^2 \frac{D}{x_0^2} \quad \text{where} \quad \gamma - 2(1 - e^{-\gamma}) = 0 \Rightarrow \gamma = 1.59362 \dots$$

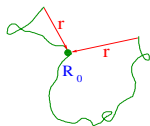
(M.R. Evans and S.M., Phys. Rev. Lett. 106, 160601 (2011))



# Typical trajectories for $r \rightarrow 0$ and $r \rightarrow \infty$



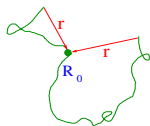
# Target search via diffusion with **resetting** in $d > 1$



stationary target of radius  $a$  at  $0$  in  
 $d > 2$

searcher starts at  $R_0 > a$ , diffuses, and  
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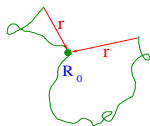


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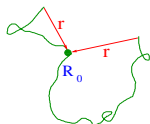


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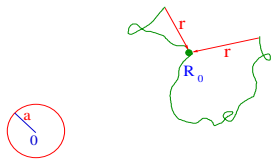


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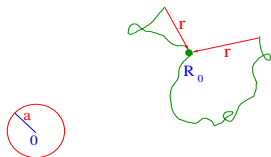
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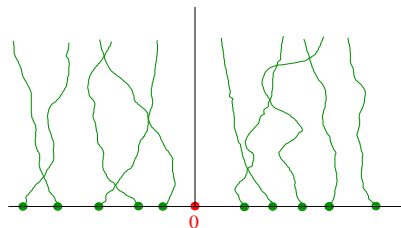
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- Once again, there is an optimal  $r^*$  that minimizes  $\bar{T}(r, R_0)$  in all  $d$

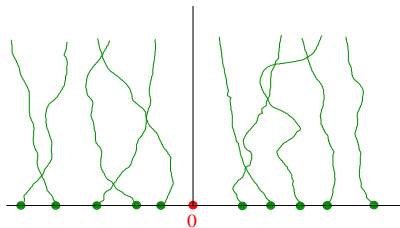
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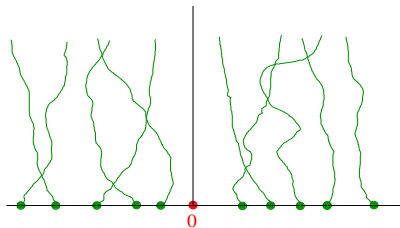
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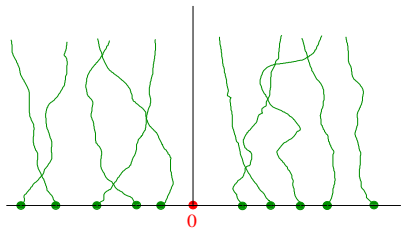
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- Eventually  $N \rightarrow \infty$  and  $L \rightarrow \infty$  with their ratio  $N/L = \rho$  fixed

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- For **diffusive** searchers without resetting:  $Q(x, t) = \text{erf}(|x|/\sqrt{4Dt})$

$$\langle P_s(t) \rangle = \exp \left[ -4 \rho \sqrt{Dt/\pi} \right] \rightarrow \text{stretched exponential decay}$$

(Zumofen, Klafter, Blumen '83, Tachiya '83, Burlatsky & Ovchinnikov '87 )

# Link to Extreme Value Statistics

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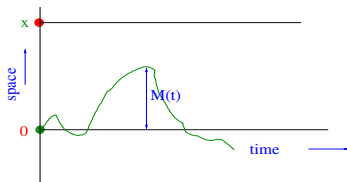
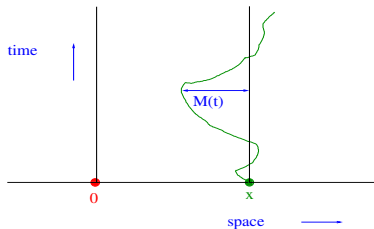
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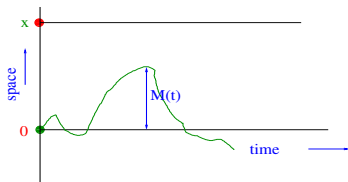
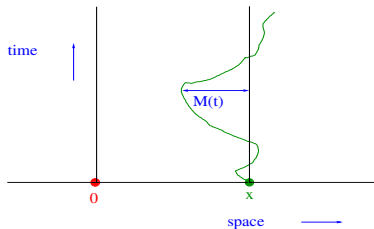


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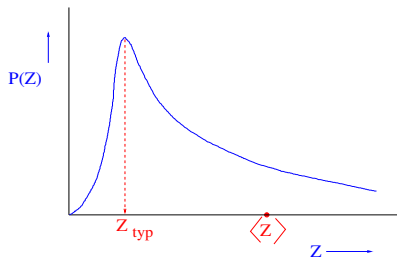
- Several exact results for  $E[M(t)]$  for **subdiffusive** and **superdiffusive** (Lévy flights) processes (J. Franke and S.M., 2012)

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- average of a random variable may be different from typical

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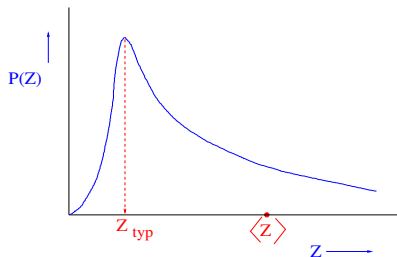
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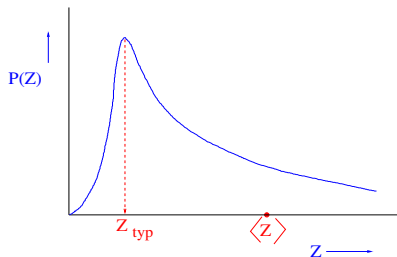
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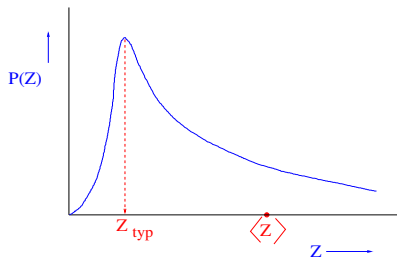
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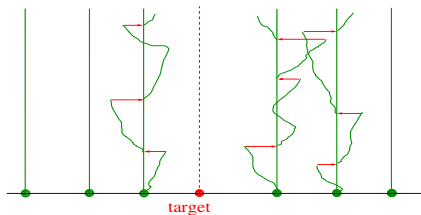
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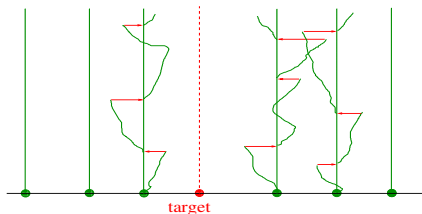
- $\Rightarrow \langle P_s(t) \rangle \sim P_s^{\text{typ}}(t)$  and both decay **stretched-exponentially**

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stationary target at 0 surrounded by a sea of independent diffusive searchers or traps (each with reset rate  $r$ ), initially distributed with uniform density  $\rho$

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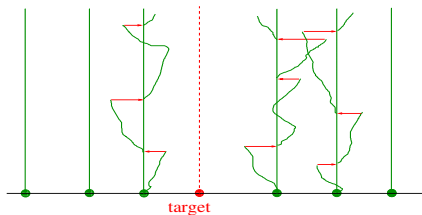


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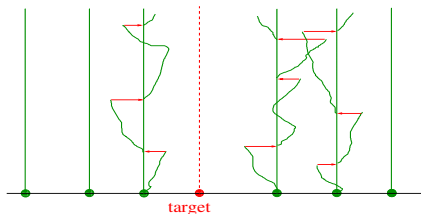
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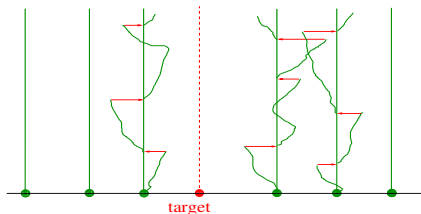
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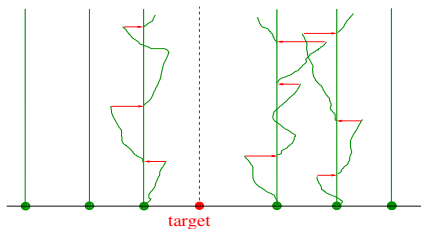
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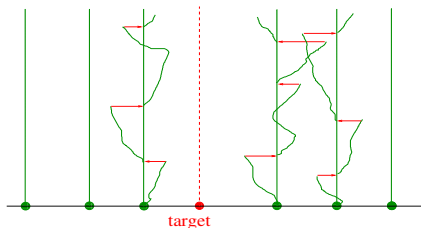


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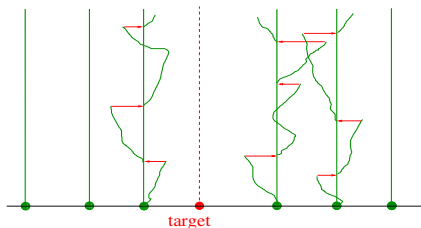
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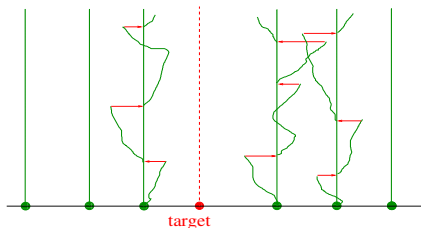
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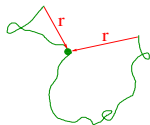


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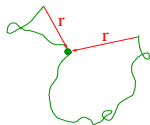
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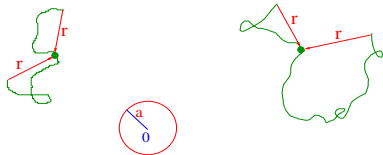


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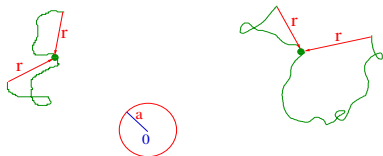
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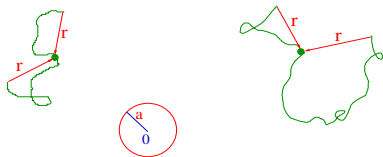
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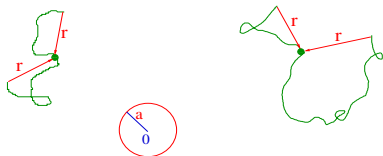
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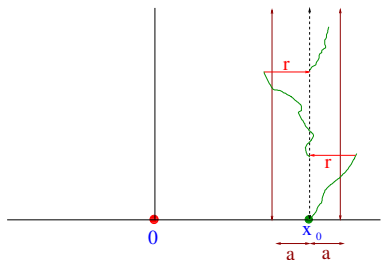
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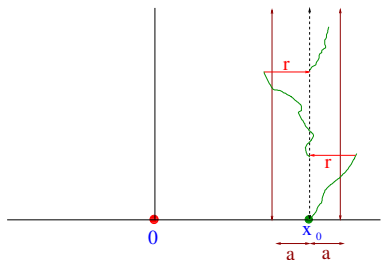
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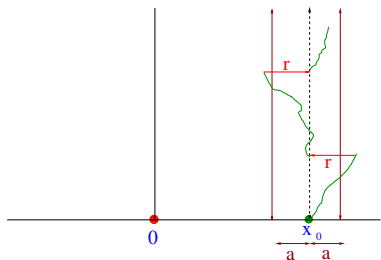
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(M.R. Evans & S.M., J. Phys. A: Math. Theo. 44, 435001 (2011))

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