First-passage properties in diffusion with stochastic resetting

Satya N. Majumdar

Laboratoire de Physique Théorique et Modèles Statistiques, CNRS, Université Paris-Sud, France

• Motivation: Visual Search problem

- Motivation: Visual Search problem
- Diffusion with Resetting: A simple model
 - \Rightarrow new Nonequilibrium Steady State

- Motivation: Visual Search problem
- Diffusion with Resetting: A simple model
 - ⇒ new Nonequilibrium Steady State
- Target search by a single random walker with resetting
 - ⇒ optimal resetting rate that minimizes the mean first-passage time to the target

- Motivation: Visual Search problem
- Diffusion with Resetting: A simple model
 - ⇒ new Nonequilibrium Steady State
- Target search by a single random walker with resetting
 - ⇒ optimal resetting rate that minimizes the mean first-passage time to the target
- Various generalizations: space-dependent resetting rate random target position

- Motivation: Visual Search problem
- Diffusion with Resetting: A simple model
 - ⇒ new Nonequilibrium Steady State
- Target search by a single random walker with resetting
 - ⇒ optimal resetting rate that minimizes the mean first-passage time to the target
- Various generalizations: space-dependent resetting rate random target position
- Target search by multiple searchers with resetting
 - ⇒ power-law decay of the target persistence/survival prob. with nontrivial persistence exponent

- Motivation: Visual Search problem
- Diffusion with Resetting: A simple model
 - ⇒ new Nonequilibrium Steady State
- Target search by a single random walker with resetting
 - ⇒ optimal resetting rate that minimizes the mean first-passage time to the target
- Various generalizations: space-dependent resetting rate random target position
- Target search by multiple searchers with resetting
 - ⇒ power-law decay of the target persistence/survival prob. with nontrivial persistence exponent
- Summary and Conclusion

Search Problems

Search problems are ubiquitous in nature

- search for Holy Grail
- search for Higgs boson

```
...
```

- data search (Google)
- · animals searching for food
- protein searching for a binding site on a DNA
- Visual search: locating a face in the crowd

Search Problems

Search problems are ubiquitous in nature

- search for Holy Grail
- search for Higgs boson

```
...
```

- data search (Google)
- · animals searching for food
- protein searching for a binding site on a DNA
- Visual search: locating a face in the crowd

A robust class of models: Intermittent target search strategies combine

- (i) phases of slow motion (target detection)
- (ii) phases of fast motion (searcher relocates but not reactive)

```
[O. Bénichou et. al. Rev. Mod. Phys. 83, 81 (2011)]
```

Visual search: a face in a crowd



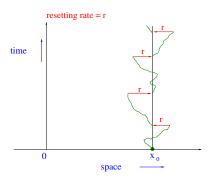
Visual search in psychology

```
PPP
        \mathbf{P}
          PPP
    ₽
            PP
PPPP
        \mathbf{P}
          PPP
    \mathbf{P}
            PP
 P
\mathbf{B}
  P
    P
      P
        P
          P
            PP
PPPP
        PPPPP
PPPP
        P
          \mathbf{P}
            PPPP
 PPP
        PР
            PP
PPPPPPPP
```

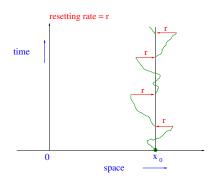
Search via diffusion and resetting

The Model and its Steady State

Diffusion with stochastic resetting: The model



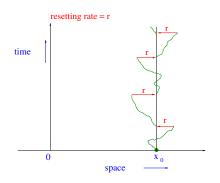
Diffusion with stochastic resetting: The model



Dynamics: In a small time interval Δt

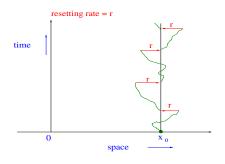
$$x(t + \Delta t) = x_0$$
 with prob. $r\Delta t$ (resetting)
= $x(t) + \eta(t) \Delta t$ with prob. $1 - r\Delta t$ (diffusion)

Diffusion with stochastic resetting: The model

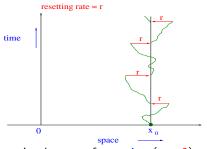


Dynamics: In a small time interval Δt

$$x(t+\Delta t)=x_0$$
 with prob. $r\Delta t$ (resetting)
$$=x(t)+\eta(t)\,\Delta t \quad \text{with prob. } 1-r\Delta t \quad \text{(diffusion)}$$
 $\eta(t) \to \text{Gaussian white noise: } \langle \eta(t) \rangle = 0 \text{ and } \langle \eta(t)\eta(t') \rangle = 2\,D\,\delta(t-t')$ [M.R. Evans & S.M., PRL, 106, 160601 (2011)]



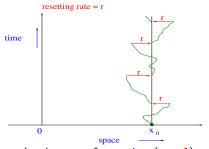
$$p(x, t) \rightarrow \text{prob.}$$
 density at time t , given $p(x, 0) = \delta(x - x_0)$



$$p(x,t) \rightarrow \text{prob.}$$
 density at time t , given $p(x,0) = \delta(x-x_0)$

• In absence of resetting (r = 0):

$$p(x,t) = \frac{1}{\sqrt{4\pi Dt}} \exp[-(x-x_0)^2/4Dt]$$



$$p(x,t) \rightarrow \text{prob.}$$
 density at time t , given $p(x,0) = \delta(x-x_0)$

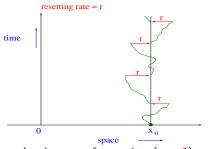
• In absence of resetting (r = 0):

$$p(x,t) = \frac{1}{\sqrt{4\pi D t}} \exp[-(x-x_0)^2/4Dt]$$

• In presence of resetting (r > 0):

Fokker-Planck equation:

$$p(x, t + \Delta t) = [1 - r \Delta t] \langle p(x - \eta(t) \Delta t, t) \rangle + [r \Delta t] \delta(x - x_0)$$



$$p(x,t) \to \text{prob.}$$
 density at time t ,
given $p(x,0) = \delta(x-x_0)$

• In absence of resetting (r = 0):

$$p(x,t) = \frac{1}{\sqrt{4\pi D t}} \exp[-(x-x_0)^2/4Dt]$$

• In presence of resetting (r > 0):

Fokker-Planck equation:

$$p(x, t + \Delta t) = [1 - r \Delta t] \langle p(x - \eta(t) \Delta t, t) \rangle + [r \Delta t] \delta(x - x_0)$$

$$\Rightarrow \left[\partial_t p = D \partial_x^2 p - r p(x, t) + r \delta(x - x_0) \right]$$

• Fokker-Planck Eq: $\partial_t p = D \partial_x^2 p - r p(x,t) + r \delta(x-x_0)$

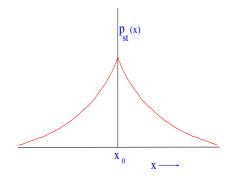
- Fokker-Planck Eq: $\partial_t p = D \partial_x^2 p r p(x,t) + r \delta(x-x_0)$
- stationary $(t \to \infty)$ solution: $D \frac{d^2 p_{\rm st}(x)}{dx^2} r p_{\rm st}(x) + r \delta(x x_0) = 0$

- Fokker-Planck Eq: $\partial_t p = D \partial_x^2 p r p(x,t) + r \delta(x-x_0)$
- stationary $(t \to \infty)$ solution: $D \frac{d^2 p_{\rm st}(x)}{dx^2} r p_{\rm st}(x) + r \delta(x x_0) = 0$

Exact solution
$$\rightarrow \left[p_{\rm st}(x) = \frac{\alpha_0}{2} \, \exp[-\alpha_0 \, |x-x_0|] \right]$$
 with $\alpha_0 = \sqrt{r/D}$

- Fokker-Planck Eq: $\partial_t p = D \partial_x^2 p r p(x, t) + r \delta(x x_0)$
- stationary $(t \to \infty)$ solution: $D \frac{d^2 p_{\rm st}(x)}{dx^2} r p_{\rm st}(x) + r \delta(x x_0) = 0$

Exact solution
$$\rightarrow \left[p_{\rm st}(x) = \frac{\alpha_0}{2} \, \exp[-\alpha_0 \, |x-x_0|] \right]$$
 with $\alpha_0 = \sqrt{r/D}$

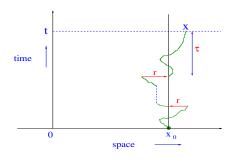


- → nonequilibrium steady state
- \Rightarrow current carrying with detailed balance \rightarrow violated

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

effective potential:

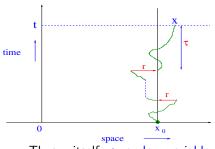
$$V_{\rm eff}(x) = \alpha_0 |x - x_0|$$



 $au
ightarrow ext{time}$ since the last resetting during which free diffusion \Rightarrow

If
$$0 < \tau < t$$

 $p(x,t) = \frac{1}{\sqrt{4\pi D \tau}} \exp[-(x - x_0)^2/4D\tau]$
If $\tau > t$ (no resetting in $[0,t]$)
 $p(x,t) = \frac{1}{\sqrt{4\pi D t}} \exp[-(x - x_0)^2/4Dt]$



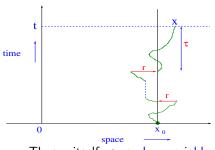
 $au
ightarrow ext{time}$ since the last resetting during which free diffusion \Rightarrow

If
$$0 < \tau < t$$

 $p(x,t) = \frac{1}{\sqrt{4\pi D \tau}} \exp[-(x - x_0)^2/4D\tau]$
If $\tau > t$ (no resetting in $[0,t]$)
 $p(x,t) = \frac{1}{\sqrt{4\pi D t}} \exp[-(x - x_0)^2/4Dt]$

• Thus τ itself \rightarrow random variable

$$Prob.[\tau < t] = r e^{-r\tau} \quad \text{for } 0 \le \tau < t$$
$$= e^{-rt} \quad \text{for } \tau > t \text{ (no resetting in [0, t])}$$



au o time since the last resetting during which free diffusion \Rightarrow

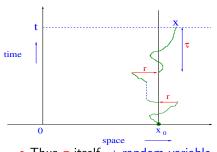
If
$$0 < \tau < t$$

 $p(x,t) = \frac{1}{\sqrt{4\pi D \tau}} \exp[-(x - x_0)^2/4D\tau]$
If $\tau > t$ (no resetting in $[0,t]$)
 $p(x,t) = \frac{1}{\sqrt{4-D t}} \exp[-(x - x_0)^2/4Dt]$

• Thus τ itself \rightarrow random variable

Prob.
$$[\tau < t] = r e^{-r\tau}$$
 for $0 \le \tau < t$
= e^{-rt} for $\tau > t$ (no resetting in $[0, t]$)

 Averaging over τ gives $p(x,t) = \int_0^t d\tau (r e^{-r\tau}) G(x,\tau) + e^{-rt} G(x,t)$ where $G(x,\tau) = \text{diffusion propagator} = \frac{1}{\sqrt{4\pi D \tau}} \exp[-(x-x_0)^2/4D\tau]$



au o time since the last resetting during which free diffusion ⇒

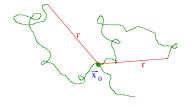
If
$$0 < \tau < t$$

 $p(x,t) = \frac{1}{\sqrt{4\pi D \tau}} \exp[-(x - x_0)^2/4D\tau]$
If $\tau > t$ (no resetting in $[0,t]$)
 $p(x,t) = \frac{1}{\sqrt{4\pi D t}} \exp[-(x - x_0)^2/4Dt]$

• Thus τ itself \rightarrow random variable

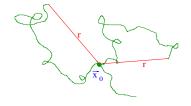
Prob.
$$[\tau < t] = r e^{-r\tau}$$
 for $0 \le \tau < t$
= e^{-rt} for $\tau > t$ (no resetting in $[0, t]$)

- Averaging over τ gives $p(x,t) = \int_0^t d\tau (r e^{-r\tau}) G(x,\tau) + e^{-rt} G(x,t)$ where $G(x,\tau) = \text{diffusion propagator} = \frac{1}{\sqrt{4\pi D \tau}} \exp[-(x-x_0)^2/4D\tau]$
- as $t \to \infty$, $p_{\rm st}(x) = \frac{\alpha_0}{2} \exp[-\alpha_0 |x x_0|]$ with $\alpha_0 = \sqrt{r/D}$



particle starting at \vec{x}_0 diffuses in d dim. and resets to \vec{x}_0 with rate r $p(\vec{x},t) \rightarrow \text{prob.}$ density. at time t

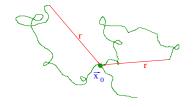
• Fokker-Planck Eq. :
$$\partial_t p = D \nabla^2 p(\vec{x},t) - r p(\vec{x},t) + r \delta(\vec{x} - \vec{x}_0)$$



particle starting at \vec{x}_0 diffuses in d dim. and resets to \vec{x}_0 with rate r $p(\vec{x}, t) \rightarrow \text{prob. density. at time } t$

- Fokker-Planck Eq. : $\partial_t p = D \nabla^2 p(\vec{x},t) r p(\vec{x},t) + r \delta(\vec{x} \vec{x}_0)$
- stationary solution:

$$p_{\rm st}(\vec{x}) = \frac{(\alpha_0)^d}{(2\pi)^d} \left[\alpha_0 |\vec{x} - \vec{x}_0| \right]^{\nu} K_{\nu} \left(\alpha_0 |\vec{x} - \vec{x}_0| \right)$$

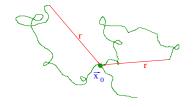


particle starting at \vec{x}_0 diffuses in d dim. and resets to \vec{x}_0 with rate r $p(\vec{x}, t) \rightarrow \text{prob. density. at time } t$

- Fokker-Planck Eq. : $\partial_t p = D \nabla^2 p(\vec{x},t) r p(\vec{x},t) + r \delta(\vec{x} \vec{x}_0)$
- stationary solution:

$$p_{\rm st}(\vec{x}) = \frac{(\alpha_0)^d}{(2\pi)^d} \left[\alpha_0 |\vec{x} - \vec{x}_0| \right]^{\nu} K_{\nu} \left(\alpha_0 |\vec{x} - \vec{x}_0| \right)$$

where $\alpha_0 = \sqrt{r/D}$, $\nu = 1 - d/2$ and $K_{\nu}(z) \rightarrow$ modified Bessel function



particle starting at \vec{x}_0 diffuses in d dim. and resets to \vec{x}_0 with rate r $p(\vec{x},t) \rightarrow \text{prob. density. at time } t$

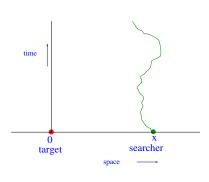
- Fokker-Planck Eq. : $\partial_t p = D \nabla^2 p(\vec{x},t) r p(\vec{x},t) + r \delta(\vec{x} \vec{x}_0)$
- stationary solution:

$$p_{\rm st}(\vec{x}) = \frac{(\alpha_0)^d}{(2\pi)^d} \left[\alpha_0 |\vec{x} - \vec{x}_0| \right]^{\nu} K_{\nu} \left(\alpha_0 |\vec{x} - \vec{x}_0| \right)$$

where $\alpha_0 = \sqrt{r/D}$, $\nu = 1 - d/2$ and $K_{\nu}(z) \rightarrow$ modified Bessel function

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

II: Persistence and First-passage properties



 $Q(x,t) \rightarrow$ persistence/survival prob. of the target

backward Fokker-Planck equation

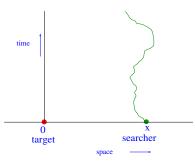
$$\partial_t Q(x,t) = D\partial_x^2 Q(x,t)$$
 for $x \ge 0$

boundary cond. :

$$Q(x = 0, t) = 0$$
 and $Q(x \to \infty, t) = 1$

intial cond. :

$$Q(x, t = 0) = 1 \text{ for } x > 0$$



Q(x,t) o persistence/survival prob. of the target

backward Fokker-Planck equation

$$\partial_t Q(x,t) = D\partial_x^2 Q(x,t)$$
 for $x \ge 0$

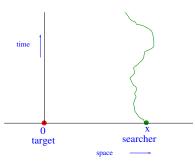
boundary cond. :

$$Q(x = 0, t) = 0$$
 and $Q(x \to \infty, t) = 1$

intial cond. :

$$Q(x, t = 0) = 1 \text{ for } x > 0$$

• exact solution for survival prob. : $Q(x,t) = erf(|x|/\sqrt{4Dt})$



 $Q(x,t) \rightarrow \text{persistence/survival prob.}$ of the target

backward Fokker-Planck equation

$$\partial_t Q(x,t) = D\partial_x^2 Q(x,t)$$
 for $x \ge 0$

boundary cond. :

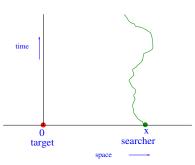
$$Q(x = 0, t) = 0$$
 and $Q(x \rightarrow \infty, t) = 1$

intial cond. :

$$Q(x, t = 0) = 1 \text{ for } x > 0$$

- exact solution for survival prob. : $Q(x,t) = erf(|x|/\sqrt{4Dt})$
- first-passage prob.:

$$F(x,t) = -\partial_t Q(x,t) = \frac{x}{\sqrt{4\pi Dt^3}} \exp[-x^2/4Dt] \xrightarrow{t \to \infty} t^{-3/2}$$



 $Q(x, t) \rightarrow \text{persistence/survival prob. of}$ the target

backward Fokker-Planck equation

$$\partial_t Q(x,t) = D\partial_x^2 Q(x,t)$$
 for $x \ge 0$

boundary cond. :

$$Q(x = 0, t) = 0$$
 and $Q(x \to \infty, t) = 1$

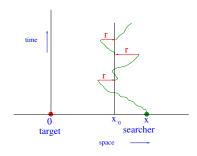
intial cond. :

$$Q(x, t = 0) = 1 \text{ for } x > 0$$

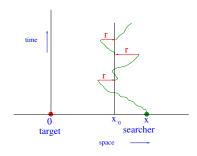
- exact solution for survival prob. : $Q(x,t) = erf(|x|/\sqrt{4Dt})$
- first-passage prob.:

$$F(x,t) = -\partial_t Q(x,t) = \frac{x}{\sqrt{4\pi Dt^3}} \exp[-x^2/4Dt] \xrightarrow{t \to \infty} t^{-3/2}$$

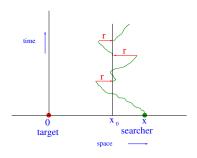
• mean capture time $\to \bar{T} = \int_0^\infty t F(x, t) dt = \infty$



• starting position $x \to$ 'variable', resetting to x_0 with rate r

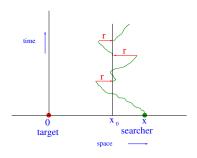


- starting position $x \rightarrow$ 'variable', resetting to x_0 with rate r
- $Q(x,t) \rightarrow \text{persistence/survival prob.}$ of the target



- starting position $x \to$ 'variable', resetting to x_0 with rate r
- $Q(x,t) \rightarrow \text{persistence/survival prob.}$ of the target
- backward Fokker-Planck Eq.:

$$\partial_t Q(x,t) = D\partial_x^2 Q(x,t) - r Q(x,t) + r Q(x_0,t)$$
 for $x \ge 0$



- starting position $x \rightarrow$ 'variable', resetting to x_0 with rate r
- $Q(x, t) \rightarrow \text{persistence/survival prob.}$ of the target
- backward Fokker-Planck Eq.:

$$\partial_t Q(x,t) = D\partial_x^2 Q(x,t) - r Q(x,t) + r Q(x_0,t)$$
 for $x \ge 0$ boundary cond. : $Q(x=0,t) = 0$ and $Q(x \to \infty,t) \to finite$ intial cond. : $Q(x,t=0) = 1$ for $x > 0$

• Laplace transform: $\tilde{Q}(x,s) = \int_0^\infty Q(x,t) e^{-st} dt$

- Laplace transform: $\tilde{Q}(x,s) = \int_0^\infty Q(x,t) e^{-st} dt$
- Exact solution:

$$\tilde{Q}(x,s) = \frac{1+r\,\tilde{Q}(x_0,s)}{r+s}\,\left[1-\exp\left(-\sqrt{(r+s)/D}\,x\right)\right]$$

- Laplace transform: $\tilde{Q}(x,s) = \int_0^\infty Q(x,t) e^{-st} dt$
- Exact solution:

$$\tilde{Q}(x,s) = \frac{1+r\,\tilde{Q}(x_0,s)}{r+s}\,\left[1-\exp\left(-\sqrt{(r+s)/D}\,x\right)\right]$$

- Laplace transform: $\tilde{Q}(x,s) = \int_0^\infty Q(x,t) e^{-st} dt$
- Exact solution:

$$\tilde{Q}(x,s) = \frac{1+r\,\tilde{Q}(x_0,s)}{r+s}\,\left[1-\exp\left(-\sqrt{(r+s)/D}\,x\right)\right]$$

$$\tilde{Q}(x_0, s) = \frac{1 - \exp\left(-\sqrt{(r+s)/D} x_0\right)}{s + r \exp\left(-\sqrt{(r+s)/D} x_0\right)}$$

- Laplace transform: $\tilde{Q}(x,s) = \int_0^\infty Q(x,t) e^{-st} dt$
- Exact solution:

$$\tilde{Q}(x,s) = \frac{1+r\,\tilde{Q}(x_0,s)}{r+s}\,\left[1-\exp\left(-\sqrt{(r+s)/D}\,x\right)\right]$$

$$\tilde{Q}(x_0, s) = \frac{1 - \exp\left(-\sqrt{(r+s)/D} x_0\right)}{s + r \exp\left(-\sqrt{(r+s)/D} x_0\right)}$$

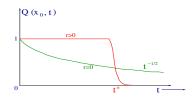
$$\Rightarrow$$
 As $t \to \infty$, $Q(x_0, t) \approx \exp\left[-r t e^{-\sqrt{r/D} x_0}\right]$

- Laplace transform: $\tilde{Q}(x,s) = \int_0^\infty Q(x,t) \, e^{-s \, t} \, dt$
- Exact solution:

$$ilde{Q}(x,s) = rac{1+r\, ilde{Q}(x_0,s)}{r+s}\,\left[1-\exp\left(-\sqrt{(r+s)/D}\,x
ight)
ight]$$

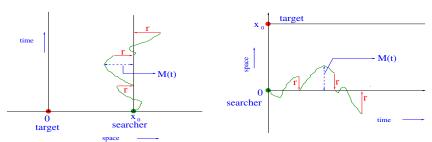
$$\tilde{Q}(x_0, s) = \frac{1 - \exp\left(-\sqrt{(r+s)/D} x_0\right)}{s + r \exp\left(-\sqrt{(r+s)/D} x_0\right)}$$

$$\Rightarrow$$
 As $t \to \infty$, $Q(x_0, t) \approx \exp\left[-r t e^{-\sqrt{r/D} x_0}\right]$

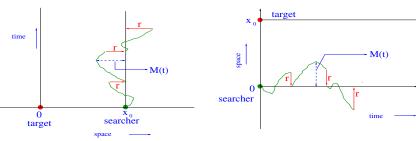


For
$$r>0$$
, $Q(x_0,t) pprox \exp[-t/t^*]$ where $t^*pprox (1/r)\,e^{\sqrt{r/D}\,x_0}$

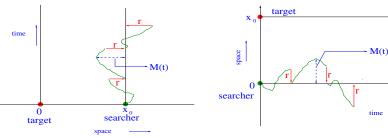
ROTATE & SHIFT



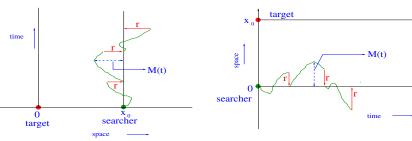
• $M(t) \rightarrow \text{maximum of the process up to time } t$



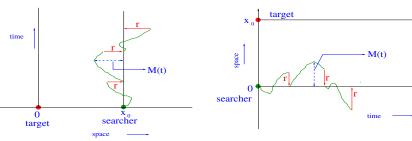
- $M(t) \rightarrow \text{maximum of the process up to time } t$
- Survival prob. $Q(x_0, t) \equiv \text{Prob.}[M(t) \le x_0]$



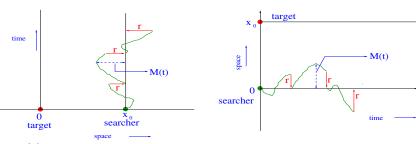
- $M(t) \rightarrow \text{maximum of the process up to time } t$
- Survival prob. $Q(x_0, t) \equiv \text{Prob.}[M(t) \le x_0]$
- Correlation time $\tau = 1/r$



- $M(t) \rightarrow \text{maximum of the process up to time } t$
- Survival prob. $Q(x_0, t) \equiv \text{Prob.}[M(t) \le x_0]$
- Correlation time $\tau=1/r\to N_{\rm eff}=t/\tau=rt$ effectively independent blocks in the time interval [0,t]

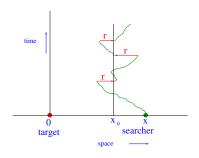


- $M(t) \rightarrow \text{maximum of the process up to time } t$
- Survival prob. $Q(x_0, t) \equiv \text{Prob.}[M(t) \le x_0]$
- Correlation time $\tau = 1/r \rightarrow N_{\rm eff} = t/\tau = rt$ effectively independent blocks in the time interval [0,t]
- $Q(x_0, t) \approx \exp\left[-rt e^{-\sqrt{r/D}x_0}\right]$



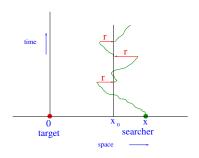
- $M(t) o ext{maximum of the process up to time } t$
- Survival prob. $Q(x_0, t) \equiv \text{Prob.}[M(t) \le x_0]$
- Correlation time $\tau = 1/r \rightarrow N_{\rm eff} = t/\tau = rt$ effectively independent blocks in the time interval [0, t]
- $Q(x_0, t) \approx \exp\left[-rt \, e^{-\sqrt{r/D}\,x_0}
 ight] \approx \exp\left[-N_{\mathrm{eff}} \, e^{-\sqrt{r/D}\,x_0}
 ight]$
- ⇒ classical Gumbel distribution for the maximum of a set of N_{eff} exponentially distributed independent random variables

Mean capture/search time



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

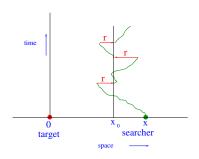
Mean capture/search time



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

$$ar{T}(r, x_0) = rac{1}{r} \left[\exp\left(\sqrt{r/D} x_0\right) - 1
ight]$$

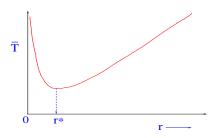
Mean capture/search time



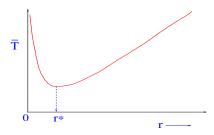
mean capture time:
$$\bar{T} = \int_0^\infty t \left[-\partial_t Q(x_0, t) \right] dt = \tilde{Q}(x_0, s = 0)$$

$$\overline{T}(r, x_0) = \frac{1}{r} \left[\exp\left(\sqrt{r/D} x_0\right) - 1 \right]$$

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

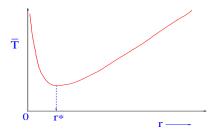


$$\bar{T}(r, x_0) = \frac{1}{r} \left[\exp\left(\sqrt{r/D} x_0\right) - 1 \right]$$



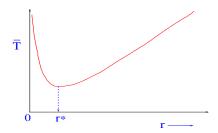
$$ar{\mathcal{T}}(r, x_0) = rac{1}{r} \left[\exp\left(\sqrt{r/D} \, x_0\right) - 1
ight]$$

• For fixed x_0 and D, the mean capture time $\overline{T}(r, x_0)$ diverges as $r \to 0$ and also as $r \to \infty$



$$ar{\mathcal{T}}(r, x_0) = rac{1}{r} \left[\exp\left(\sqrt{r/D} \, x_0\right) - 1
ight]$$

- For fixed x_0 and D, the mean capture time $\overline{T}(r, x_0)$ diverges as $r \to 0$ and also as $r \to \infty$
- As a function of r, $\bar{T}(r, x_0)$ has a minimum at $r = r^*$



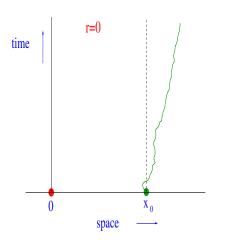
$$ar{\mathcal{T}}(r, x_0) = rac{1}{r} \, \left[\exp \left(\sqrt{r/D} \, x_0
ight) - 1
ight]$$

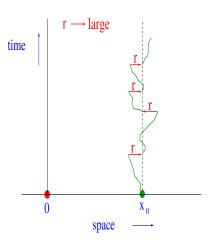
- For fixed x_0 and D, the mean capture time $\overline{T}(r, x_0)$ diverges as $r \to 0$ and also as $r \to \infty$
- As a function of r, $\bar{T}(r, x_0)$ has a minimum at $r = r^*$ optimal resetting rate r^* is given by:

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

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

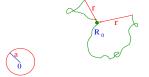
Typical trajectories for $r \to 0$ and $r \to \infty$







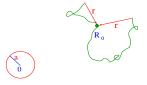
stationary target of radius a at 0 in d > 2



stationary target of radius a at 0 in d > 2

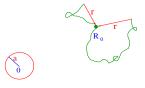
searcher starts at $R_0 > a$, diffuses, and resets with rate r

• $Q(R_0,t) \rightarrow$ survival prob. of the target starting at a radial distance R_0



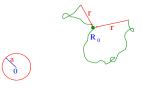
stationary target of radius a at 0 in d > 2

- $Q(R_0,t) \to \text{survival prob.}$ of the target starting at a radial distance R_0
- Laplace trasform $\tilde{Q}(R_0, s) = \int_0^\infty Q(R_0, t) e^{-st} dt$ is obtained by solving the *d*-dim. backward Fokker-Planck Eq.



stationary target of radius a at 0 in d > 2

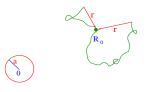
- $Q(R_0,t) o$ survival prob. of the target starting at a radial distance R_0
- Laplace trasform $\tilde{Q}(R_0,s)=\int_0^\infty Q(R_0,t)\,e^{-s\,t}\,dt$ is obtained by solving the d-dim. backward Fokker-Planck Eq.
- mean capture time: $\bar{T} = \tilde{Q}(R_0, s = 0)$



stationary target of radius a at 0 in d > 2

- $Q(R_0,t) o$ survival prob. of the target starting at a radial distance R_0
- Laplace trasform $\tilde{Q}(R_0,s)=\int_0^\infty Q(R_0,t)\,e^{-s\,t}\,dt$ is obtained by solving the d-dim. backward Fokker-Planck Eq.
- mean capture time: $\bar{T} = \tilde{Q}(R_0, s = 0)$

$$\boxed{\bar{T}(r,R_0) = \frac{1}{r} \left[\left(\frac{a}{R_0} \right)^{\nu} \frac{K_{\nu}(a\sqrt{r/D})}{K_{\nu}(R_0\sqrt{r/D})} - 1 \right]} \text{ where } \nu = 1 - d/2$$



stationary target of radius a at 0 in d > 2

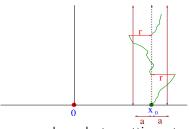
searcher starts at $R_0 > a$, diffuses, and resets with rate r

- ullet $Q(R_0,t) o$ survival prob. of the target starting at a radial distance R_0
- Laplace trasform $\tilde{Q}(R_0,s)=\int_0^\infty Q(R_0,t)\,e^{-s\,t}\,dt$ is obtained by solving the d-dim. backward Fokker-Planck Eq.
- mean capture time: $\bar{T} = \tilde{Q}(R_0, s = 0)$

$$\overline{T}(r,R_0) = \frac{1}{r} \left[\left(\frac{a}{R_0} \right)^{\nu} \frac{K_{\nu}(a\sqrt{r/D})}{K_{\nu}(R_0\sqrt{r/D})} - 1 \right] \text{ where } \nu = 1 - d/2$$

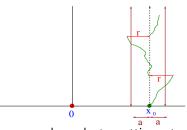
• Once again, there is an optimal r^* that minimizes $\bar{T}(r,R_0)$ in all d

[M.R. Evans and S.M., J. Phys. A: Math. Theo. 47, 285001 (2014)]



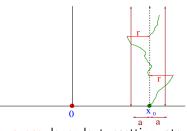
stationary target at 0, searcher diffuses and resets to x_0 with rate r only if it goes outside the box $[x_0 - a, x_0 + a]$ \longrightarrow otherwise no resetting

• space-dependent resetting rate r(x): What is the optimization strategy?



stationary target at 0, searcher diffuses and resets to x_0 with rate r only if it goes outside the box $[x_0 - a, x_0 + a]$ \longrightarrow otherwise no resetting

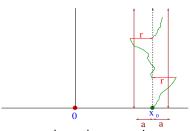
- space-dependent resetting rate r(x): What is the optimization strategy?
- When the target position is drawn randomly from $P_{\text{target}}(x)$, where should the particle reset?



stationary target at 0, searcher diffuses and resets to x_0 with rate r only if it goes outside the box $[x_0 - a, x_0 + a]$ \longrightarrow otherwise no resetting

- space-dependent resetting rate r(x): What is the optimization strategy?
- When the target position is drawn randomly from $P_{\text{target}}(x)$, where should the particle reset?
- If the reset position is chosen randomly from a distribution $P_{\text{reset}}(x)$, what is the optimal $P_{\text{reset}}(x)$ for a given target distribution $P_{\text{target}}(x)$?

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



stationary target at 0, searcher diffuses and resets to x_0 with rate r only if it goes outside the box $[x_0 - a, x_0 + a]$ \longrightarrow otherwise no resetting

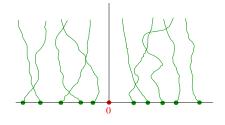
- space-dependent resetting rate r(x): What is the optimization strategy?
- When the target position is drawn randomly from $P_{\text{target}}(x)$, where should the particle reset?
- If the reset position is chosen randomly from a distribution $P_{\text{reset}}(x)$, what is the optimal $P_{\text{reset}}(x)$ for a given target distribution $P_{\text{target}}(x)$?

 [M.R. Evans & S.M., J. Phys. A: Math. Theo. 44, 435001 (2011)]
- Nonequilibrium reset dynamics vs. equilibrium dynamics
 "Nonequilibrium beats equilibrium"

[M.R. Evans, S.M. & K. Mallick, J. Phys. A: Math. Theor. 46, 185001 (2013)]

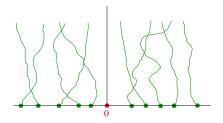
III: Multiparticle Problem

Target search by multiple searchers d=1



stationary target at 0 surrounded by a sea of independent searchers (traps), initially distributed with uniform density ρ

Target search by multiple searchers d=1

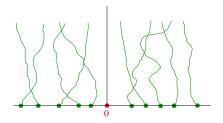


stationary target at 0 surrounded by a sea of independent searchers (traps), initially distributed with uniform density ρ

• target survival probability: $P_s(t) = \prod_{i=1}^N Q(x_i, t)$

 $Q(x_i, t) \rightarrow$ prob. that the *i*-th searcher starting initially at x_i does not hit the origin up to time t

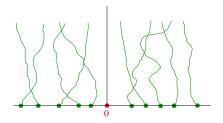
Target search by multiple searchers d = 1



stationary target at 0 surrounded by a sea of independent searchers (traps), initially distributed with uniform density ρ

- target survival probability: P_s(t) = ∏^N_{i=1} Q(x_i, t)
 Q(x_i, t) → prob. that the *i*-th searcher starting initially at x_i does not hit the origin up to time t
- ullet Average surv. prob. of the target: $\langle P_s(t)
 angle = \left\langle \prod_{i=1}^N Q(x_i,t)
 ight
 angle$
 - $\langle \rangle \rightarrow$ average over x_i 's each drawn independently and uniformly from a box [-L/2, L/2]

Target search by multiple searchers d=1



stationary target at 0 surrounded by a sea of independent searchers (traps), initially distributed with uniform density ρ

- target survival probability: P_s(t) = ∏^N_{i=1} Q(x_i, t)
 Q(x_i, t) → prob. that the *i*-th searcher starting initially at x_i does not hit the origin up to time t
- Average surv. prob. of the target: $\langle P_s(t) \rangle = \left\langle \prod_{i=1}^N Q(x_i, t) \right\rangle$ $\langle \rangle \rightarrow$ average over x_i 's each drawn independently and uniformly from a box [-L/2, L/2]
- Eventually $N \to \infty$ and $L \to \infty$ with their ratio $N/L = \rho$ fixed

•
$$\langle P_s(t) \rangle = \left\langle \prod_{i=1}^N Q(x_i, t) \right\rangle$$

•
$$\langle P_s(t) \rangle = \left\langle \prod_{i=1}^N Q(x_i, t) \right\rangle = \prod_{i=1}^N \left[1 - \left\langle (1 - Q(x_i, t)) \right\rangle \right]$$

$$\bullet \langle P_s(t) \rangle = \left\langle \prod_{i=1}^N Q(x_i, t) \right\rangle = \prod_{i=1}^N \left[1 - \left\langle (1 - Q(x_i, t)) \right\rangle \right]$$

$$= \left[1 - \frac{1}{L} \int_{-L/2}^{L/2} (1 - Q(x, t)) dx \right]^N$$

$$\bullet \langle P_s(t) \rangle = \left\langle \prod_{i=1}^N Q(x_i, t) \right\rangle = \prod_{i=1}^N \left[1 - \left\langle (1 - Q(x_i, t)) \right\rangle \right]$$

$$= \left[1 - \frac{1}{L} \int_{-L/2}^{L/2} (1 - Q(x, t)) dx \right]^N$$

$$\Rightarrow \left| \langle P_s(t) \rangle = \exp \left[-2 \rho \int_0^\infty (1 - Q(x, t)) \ dx \right] \right|$$

$$\bullet \langle P_s(t) \rangle = \left\langle \prod_{i=1}^N Q(x_i, t) \right\rangle = \prod_{i=1}^N \left[1 - \left\langle (1 - Q(x_i, t)) \right\rangle \right]$$

$$= \left[1 - \frac{1}{L} \int_{-L/2}^{L/2} (1 - Q(x, t)) dx \right]^N$$

$$\Rightarrow \left| \langle P_s(t) \rangle = \exp \left[-2 \rho \int_0^\infty (1 - Q(x, t)) \ dx \right] \right|$$

 \rightarrow a rather general result

$$\bullet \langle P_s(t) \rangle = \left\langle \prod_{i=1}^N Q(x_i, t) \right\rangle = \prod_{i=1}^N \left[1 - \left\langle (1 - Q(x_i, t)) \right\rangle \right]$$

$$= \left[1 - \frac{1}{L} \int_{-L/2}^{L/2} (1 - Q(x, t)) dx \right]^N$$

$$\Rightarrow \left| \langle P_s(t) \rangle = \exp \left[-2 \rho \int_0^\infty (1 - Q(x, t)) \ dx \right] \right|$$

 \rightarrow a rather general result

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

$$|\langle P_s(t)
angle = \exp\left[-4\,
ho\,\sqrt{Dt/\pi}
ight]|
ightarrow ext{stretched exponential decay}$$

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

•
$$\langle P_s(t) \rangle = \exp\left[-2\rho \int_0^\infty (1 - Q(x, t)) dx\right] = \exp\left[-2\rho E[M(t)]\right]$$

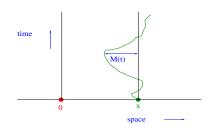
- $\langle P_s(t) \rangle = \exp\left[-2\rho \int_0^\infty (1 Q(x, t)) dx\right] = \exp\left[-2\rho E[M(t)]\right]$
- $E[M(t)] = \int_0^\infty (1 Q(x, t)) dx \rightarrow \text{expected maximum } M(t) \text{ of the trap process starting at the origin}$
- ⇒ general result valid for any trap process

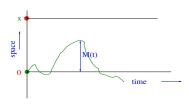
```
(J. Franke and S.M., JSTAT, P05024 (2012))
```

- $\langle P_s(t) \rangle = \exp\left[-2\rho \int_0^\infty (1 Q(x, t)) dx\right] = \exp\left[-2\rho E[M(t)]\right]$
- $E[M(t)] = \int_0^\infty (1 Q(x, t)) dx \rightarrow \text{expected maximum } M(t) \text{ of the trap process starting at the origin}$
- ⇒ general result valid for any trap process

(J. Franke and S.M., JSTAT, P05024 (2012))

Q(x,t) = surv. prob. of the trap starting at $x \equiv \text{Prob.}[M(t) \le x]$ ROTATE & SHIFT

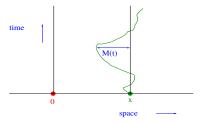


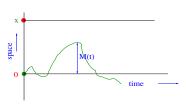


- $\langle P_s(t) \rangle = \exp\left[-2\rho \int_0^\infty (1 Q(x, t)) dx\right] = \exp\left[-2\rho E[M(t)]\right]$
- $E[M(t)] = \int_0^\infty (1 Q(x, t)) dx \rightarrow \text{expected maximum } M(t) \text{ of the trap process starting at the origin}$
- ⇒ general result valid for any trap process

(J. Franke and S.M., JSTAT, P05024 (2012))

Q(x,t) = surv. prob. of the trap starting at $x \equiv \text{Prob.}[M(t) \le x]$ ROTATE & SHIFT

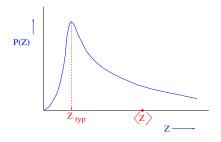




• Several exact results for E[M(t)] for subdiffusive and superdiffusive (Lévy flights) processes (J. Franke and S.M., 2012)

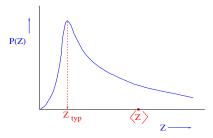
• average of a random variable may be different from typical

average of a random variable may be different from typical



 $P(Z)
ightarrow ext{highly peaked at } oldsymbol{Z_{ ext{typ}}}$ but has a long tail such that $\langle Z
angle >> oldsymbol{Z_{ ext{typ}}}$

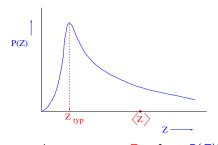
average of a random variable may be different from typical



• how to extract Z_{typ} from P(Z)?

 $P(Z)
ightarrow ext{highly peaked at $Z_{
m typ}$}$ but has a long tail such that $\langle Z
angle >> Z_{
m typ}$

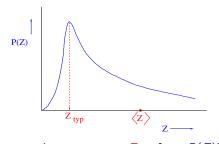
average of a random variable may be different from typical



 $P(Z) \rightarrow$ highly peaked at $Z_{\rm typ}$ but has a long tail such that $\langle Z \rangle >> Z_{\rm typ}$

- how to extract Z_{typ} from P(Z)?
- One simple prescription: compute $\langle \ln Z \rangle = \int \ln Z P(Z) dx$

• average of a random variable may be different from typical



 $P(Z) \rightarrow$ highly peaked at $Z_{\rm typ}$ but has a long tail such that $\langle Z \rangle >> Z_{\rm typ}$

- how to extract Z_{typ} from P(Z)?
- One simple prescription: compute $\langle \ln Z \rangle = \int \ln Z \, P(Z)) \, dx \approx \ln Z_{\rm typ}$

$$\Rightarrow \overline{Z_{\mathrm{typ}} pprox \mathrm{exp}\left[\langle \mathrm{ln}\, Z
angle
ight]}$$

•
$$P_s(t) = \prod_{i=1}^N Q(x_i, t)$$

- $P_s(t) = \prod_{i=1}^N Q(x_i, t)$
- $P_s^{\mathrm{typ}}(t) = \exp\left[\langle \ln\left(\prod_{i=1}^N Q(x_i, t)\right) \rangle\right]$

$$\bullet P_s(t) = \prod_{i=1}^N Q(x_i, t)$$

•
$$P_s^{\mathrm{typ}}(t) = \exp\left[\left\langle \ln\left(\prod_{i=1}^N Q(x_i,t)\right)\right\rangle\right] = \exp\left[\sum_{i=1}^N \left\langle \ln Q(x_i,t)\right\rangle\right]$$

•
$$P_s(t) = \prod_{i=1}^N Q(x_i, t)$$

•
$$P_s^{\mathrm{typ}}(t) = \exp\left[\langle \ln\left(\prod_{i=1}^N Q(x_i, t)\right) \rangle\right] = \exp\left[\sum_{i=1}^N \langle \ln Q(x_i, t) \rangle\right]$$

= $\exp\left[\frac{N}{L} \int_{-L/2}^{L/2} \ln Q(x, t) \, dx\right]$

•
$$P_s(t) = \prod_{i=1}^N Q(x_i, t)$$

•
$$P_s^{\text{typ}}(t) = \exp\left[\left\langle \ln\left(\prod_{i=1}^N Q(x_i, t)\right)\right\rangle\right] = \exp\left[\sum_{i=1}^N \left\langle \ln Q(x_i, t)\right\rangle\right]$$

= $\exp\left[\frac{N}{L} \int_{-L/2}^{L/2} \ln Q(x, t) dx\right]$

$$\Rightarrow P_s^{\text{typ}}(t) = \exp\left[2\rho \int_0^\infty \ln Q(x,t) \, dx\right]$$

•
$$P_s(t) = \prod_{i=1}^N Q(x_i, t)$$

•
$$P_s^{\text{typ}}(t) = \exp\left[\left\langle \ln\left(\prod_{i=1}^N Q(x_i, t)\right)\right\rangle\right] = \exp\left[\sum_{i=1}^N \left\langle \ln Q(x_i, t)\right\rangle\right]$$

= $\exp\left[\frac{N}{L} \int_{-L/2}^{L/2} \ln Q(x, t) dx\right]$

$$\Rightarrow P_s^{\text{typ}}(t) = \exp\left[2\rho \int_0^\infty \ln Q(x,t) \, dx\right]$$

ullet to be compared to $igl\langle P_s(t)
angle = \exp\left[-2\,
ho\,\int_0^\infty \left(1-Q({\sf x},t)
ight)
ight]$

- $P_s(t) = \prod_{i=1}^N Q(x_i, t)$
- $P_s^{\text{typ}}(t) = \exp\left[\left\langle \ln\left(\prod_{i=1}^N Q(x_i, t)\right)\right\rangle\right] = \exp\left[\sum_{i=1}^N \left\langle \ln Q(x_i, t)\right\rangle\right]$ = $\exp\left[\frac{N}{L} \int_{-L/2}^{L/2} \ln Q(x, t) dx\right]$

$$\Rightarrow P_s^{\text{typ}}(t) = \exp\left[2\rho \int_0^\infty \ln Q(x,t) \, dx\right]$$

- ullet to be compared to $igg \langle P_s(t)
 angle = \exp\left[-2\,
 ho\,\int_0^\infty \left(1-Q({\sf x},t)
 ight)
 ight]$
- For diffusive searchers: $Q(x, t) = erf(|x|/\sqrt{4Dt})$

- $P_s(t) = \prod_{i=1}^N Q(x_i, t)$
- $P_s^{\text{typ}}(t) = \exp\left[\left\langle \ln\left(\prod_{i=1}^N Q(x_i, t)\right)\right\rangle\right] = \exp\left[\sum_{i=1}^N \left\langle \ln Q(x_i, t)\right\rangle\right]$ = $\exp\left[\frac{N}{L} \int_{-L/2}^{L/2} \ln Q(x, t) dx\right]$

$$\Rightarrow P_s^{\text{typ}}(t) = \exp\left[2\rho \int_0^\infty \ln Q(x,t) \, dx\right]$$

- ullet to be compared to $igg \langle P_s(t)
 angle = \exp \left[-2 \,
 ho \, \int_0^\infty \left(1 Q(x,t)
 ight)
 ight]$
- For diffusive searchers: $Q(x, t) = \text{erf}(|x|/\sqrt{4Dt})$

$$\Rightarrow \qquad \langle P_s(t) \rangle = \exp\left[-4 \, \rho \, \sqrt{Dt/\pi} \right]$$

$$P_s^{\mathrm{typ}}(t) = \exp\left[-4 \, \rho \, b \, \sqrt{Dt} \right]$$
where $b = -\int_0^\infty \ln \operatorname{erf}(z) \, dz = 1.03442 \dots$

$$\bullet P_s(t) = \prod_{i=1}^N Q(x_i, t)$$

•
$$P_s^{\mathrm{typ}}(t) = \exp\left[\left\langle \ln\left(\prod_{i=1}^N Q(x_i, t)\right)\right\rangle\right] = \exp\left[\sum_{i=1}^N \left\langle \ln Q(x_i, t)\right\rangle\right]$$

= $\exp\left[\frac{N}{L} \int_{-L/2}^{L/2} \ln Q(x, t) dx\right]$

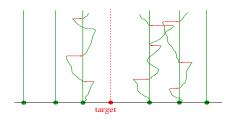
$$\Rightarrow \left| P_s^{\mathrm{typ}}(t) = \exp \left[2 \rho \int_0^\infty \ln Q(x, t) \, dx \right] \right|$$

- ullet to be compared to $igg \langle P_s(t)
 angle = \exp \left[-2 \,
 ho \, \int_0^\infty \left(1 Q(x,t)
 ight)
 ight]$
- For diffusive searchers: $Q(x, t) = \text{erf}(|x|/\sqrt{4Dt})$

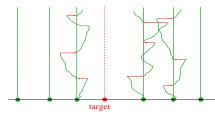
$$\Rightarrow \qquad \langle P_s(t) \rangle = \exp\left[-4 \, \rho \, \sqrt{Dt/\pi} \right]$$

$$P_s^{\mathrm{typ}}(t) = \exp\left[-4 \, \rho \, b \, \sqrt{Dt} \right]$$
where $b = -\int_0^\infty \ln \operatorname{erf}(\mathbf{z}) \, dz = 1.03442 \dots$

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



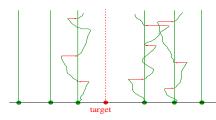
stationary target at 0 surrounded by a sea of independent diffusive searchers or traps (each with reset rate r), initially distributed with uniform density ρ



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

• Avg. survival prob. of the target:

$$\langle P_s(t) \rangle = \exp\left[-2\rho E[M(t)]\right]$$
 where $E[M(t)] = \int_0^\infty (1 - Q(x, t)) dx$



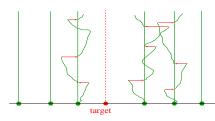
stationary target at 0 surrounded by a sea of independent diffusive searchers or traps (each with reset rate r), initially distributed with uniform density ρ

• Avg. survival prob. of the target:

$$\langle P_s(t) \rangle = \exp\left[-2 \, \rho \, E[M(t)]\right] \text{ where } E[M(t)] = \int_0^\infty \left(1 - Q(x,t)\right) \, dx$$

• Using
$$\tilde{Q}(x_0, s) = \int_0^\infty Q(x, t) e^{-s t} dt = \frac{1 - \exp(-\sqrt{(r+s)/D} x_0)}{s + r \exp(-\sqrt{(r+s)/D} x_0)}$$

$$\Rightarrow E[M(t)] \sim \sqrt{D/r} \ln t$$
 for large t



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

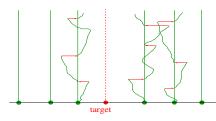
Avg. survival prob. of the target:

$$\langle P_s(t) \rangle = \exp\left[-2 \, \rho \, E[M(t)]\right] \text{ where } E[M(t)] = \int_0^\infty \left(1 - Q(x,t)\right) \, dx$$

• Using
$$\tilde{Q}(x_0, s) = \int_0^\infty Q(x, t) e^{-s t} dt = \frac{1 - \exp(-\sqrt{(r+s)/D} x_0)}{s + r \exp(-\sqrt{(r+s)/D} x_0)}$$

$$\Rightarrow E[M(t)] \sim \sqrt{D/r} \ln t$$
 for large t

•
$$\Rightarrow$$
 power-law decay for avg. survival prob. $|\langle P_s(t) \rangle \sim t^{-2\,\rho\,\sqrt{D/r}}|$ as $t \to \infty$ (Evans and S.M., 2011)



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

• Avg. survival prob. of the target:

$$\langle P_s(t) \rangle = \exp\left[-2\,\rho\,E[M(t)]\right]$$
 where $E[M(t)] = \int_0^\infty \left(1 - Q(x,t)\right)\,dx$

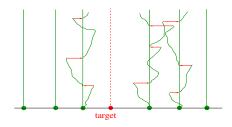
• Using
$$\tilde{Q}(x_0, s) = \int_0^\infty Q(x, t) e^{-s t} dt = \frac{1 - \exp\left(-\sqrt{(r+s)/D} x_0\right)}{s + r \exp\left(-\sqrt{(r+s)/D} x_0\right)}$$

$$\Rightarrow E[M(t)] \sim \sqrt{D/r} \ln t$$
 for large t

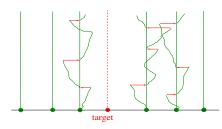
⇒ power-law decay for avg. survival prob.

$$raket{\langle P_s(t)
angle \sim t^{-2\,
ho\,\sqrt{D/r}}}$$
 as $t o\infty$ (Evans and S.M., 2011)

• As $r \to 0$, one gets back: $\langle P_s(t) \rangle \sim \exp \left[-4 \, \rho \, \sqrt{D t / \pi} \right]$

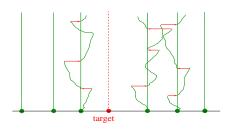


stationary target at 0 surrounded by a sea of independent diffusive searchers or traps (each with reset rate r), initially distributed with uniform density ρ



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

$$ullet$$
 Avg. surv. prob. decays as: $\overline{\langle P_s(t)
angle \sim t^{-2\,
ho\,\sqrt{D/r}}}$ as t

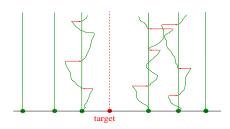


stationary target at 0 surrounded by a sea of independent diffusive searchers or traps (each with reset rate r), initially distributed with uniform density ρ

• Avg. surv. prob. decays as:
$$|\langle P_s(t) \rangle \sim t^{-2\rho\sqrt{D/r}}|$$
 as $t \to \infty$

• In contrast, Typical surv. prob.: $P_s^{\text{typ}}(t) = \exp \left[2 \rho \int_0^\infty \ln Q(x,t) \, dx \right]$

decays as:
$$\left[P_s^{\mathrm{typ}}(t) \sim \exp\left[-8 \left(1 - \ln 2 \right) \sqrt{r \, D} \,
ho \, t \right] \right]$$
 as $t o \infty$



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

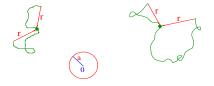
- ullet Avg. surv. prob. decays as: $\left|\langle P_s(t)
 angle \sim t^{-2\,
 ho\,\sqrt{D/r}}
 ight|$ as $t o\infty$
- In contrast, Typical surv. prob.: $P_s^{\mathrm{typ}}(t) = \exp\left[2\,\rho\,\int_0^\infty \ln\,Q(x,t)\,dx\right]$ decays as: $P_s^{\mathrm{typ}}(t) \sim \exp\left[-8\,(1-\ln2)\,\sqrt{r\,D}\,\rho\,t\right]$ as $t\to\infty$
- In presence of resetting (r > 0): $P_s^{\text{typ}}(t) << \langle P_s(t) \rangle$

Rare trajectories dominate the average



stationary target of radius a at 0 in d > 2

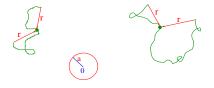
searchers diffuse and reset with rate r independently



stationary target of radius a at 0 in d > 2

searchers diffuse and reset with rate r independently

• Average survival prob. of the target for large *t*:

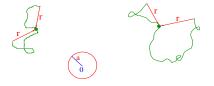


stationary target of radius a at 0 in d > 2

searchers diffuse and reset with rate r independently

Average survival prob. of the target for large t:

$$oxed{\langle P_{s}(t)
angle \sim A \exp \left[-c \,
ho \, (\ln t)^d
ight]} ext{ where } c = (\pi D/r)^{d/2}/\Gamma(1+d/2)$$



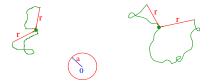
stationary target of radius a at 0 in d > 2

searchers diffuse and reset with rate r independently

Average survival prob. of the target for large t:

$$oxed{\langle P_{\mathfrak{s}}(t)
angle \sim A\, ext{exp}\left[-c\,
ho\,(ext{ln}\,t)^d
ight]}\,\, ext{where}\,\, c=(\pi D/r)^{d/2}/\Gamma(1+d/2)$$

• Typical survival prob. of the target for large t:



stationary target of radius a at 0 in d > 2

searchers diffuse and reset with rate *r* independently

• Average survival prob. of the target for large t:

$$oxed{\langle P_{\mathfrak{s}}(t)
angle \sim A\, ext{exp}\left[-c\,
ho\,(ext{ln}\,t)^d
ight]}\,\, ext{where}\,\, c=(\pi D/r)^{d/2}/\Gamma(1+d/2)$$

• Typical survival prob. of the target for large t:

$$\overline{ig|_s^{ ext{typ}}(t)\sim ext{exp}\left[-\lambda_d\,
ho\,t
ight]}$$
 where $\lambda_d o$ non-universal constant



stationary target of radius a at 0 in d > 2

searchers diffuse and reset with rate *r* independently

Average survival prob. of the target for large t:

$$oxed{\langle P_s(t)
angle \sim A\, ext{exp}\left[-c\,
ho\,(ext{ln}\,t)^d
ight]} ext{ where } c=(\pi D/r)^{d/2}/\Gamma(1+d/2)$$

• Typical survival prob. of the target for large t:

$$oxed{P_s^{
m typ}(t)\sim \exp\left[-\lambda_d\,
ho\,t
ight]}$$
 where $\lambda_d o$ non-universal constant

• As in one dimension: $P_s^{\mathrm{typ}}(t) << \langle P_s(t) \rangle$

Rare trajectories dominate the average

[M.R. Evans and S.M., J. Phys. A: Math. Theo. 47, 285001 (2014)]

ullet Diffusion with stochastic resetting \to new nonequilibrium steady state in all dimensions

 Diffusion with stochastic resetting → new nonequilibrium steady state in all dimensions

• Search of a stationary target via diffusion+resetting \rightarrow efficient mean search time $\bar{T}(r)$ has a minimum at an optimal resetting rate r^*

 Diffusion with stochastic resetting → new nonequilibrium steady state in all dimensions

Search of a stationary target via diffusion+resetting → efficient
mean search time T

(r) has a minimum at an optimal resetting rate r*

• In presence of multiple searchers each resetting to their initial positions $\langle P_s(t) \rangle >> P_s^{\rm typ}(t) \Rightarrow$ rare trajectories dominate the average

 Diffusion with stochastic resetting → new nonequilibrium steady state in all dimensions

• Search of a stationary target via diffusion+resetting \rightarrow efficient mean search time $\bar{T}(r)$ has a minimum at an optimal resetting rate r^*

• In presence of multiple searchers each resetting to their initial positions $\langle P_s(t) \rangle >> P_s^{\rm typ}(t) \Rightarrow$ rare trajectories dominate the average

For example, in d = 1, for large time t,

$$\langle P_s(t) \rangle \sim t^{-2\rho\sqrt{D/r}}$$

$$P_s^{
m typ}(t) \sim \exp\left[-8\left(1-\ln2
ight)\sqrt{r\,D}\,
ho\,t
ight]$$

Various generalisations:

Lévy flights with resetting

[L. Kusmierz, S.M., S. Sabhapandit, G. Schehr (2014)]

Various generalisations:

Lévy flights with resetting

```
[L. Kusmierz, S.M., S. Sabhapandit, G. Schehr (2014)]
```

- Resetting dynamics of spatially extended system
 - → fluctuating interfaces

```
[S. Gupta, S.M. & G. Schehr, PRL, 112, 220601 (2014)]
```

Various generalisations:

Lévy flights with resetting

```
[L. Kusmierz, S.M., S. Sabhapandit, G. Schehr (2014)]
```

- Resetting dynamics of spatially extended system
 - → fluctuating interfaces

```
[S. Gupta, S.M. & G. Schehr, PRL, 112, 220601 (2014)]
```

• space-dependent resetting rate, random target and reset positions, Eq. vs. Non-eq. dynamics ...

Various generalisations:

Lévy flights with resetting

```
[L. Kusmierz, S.M., S. Sabhapandit, G. Schehr (2014)]
```

- Resetting dynamics of spatially extended system
 - → fluctuating interfaces

```
[S. Gupta, S.M. & G. Schehr, PRL, 112, 220601 (2014)]
```

• space-dependent resetting rate, random target and reset positions, Eq. vs. Non-eq. dynamics ...

Resetting \rightarrow rich and interesting static and dynamic phenomena

Collaborators and References

Collaborators:

- M. R. Evans (Edinburgh University, UK)
- S. Gupta (LPTMS, Orsay, France)
- L. Kusmierz (Inst. of Phys., Krakow, Poland)
- K. Mallick (Saclay, France)
- S. Sabhapandit (RRI, Bangalore, India)
- G. Schehr (LPTMS, Orsay, France)
- J. Whitehouse (Edinburgh University, UK)

References:

- M.R. Evans and S.N. Majumdar, Phys. Rev. Lett. 106, 160601 (2011).
- M.R. Evans and S.N. Majumdar, J. Phys. A: Math. Theor. 44, 435001 (2011).
- M. R. Evans, S.N. Majumdar, K. Mallick, J. Phys. A: Math. Theor. 46, 185001 (2013).
- J. Whitehouse, M. R. Evans, and S. N. Majumdar, Phys. Rev. E 87, 022118 (2013).
- S. Gupta, S.N. Majumdar, G. Schehr, Phys. Rev. Lett. 112, 220601 (2014).
- M.R. Evans and S. N. Majumdar J. Phys. A: Math. Theor. 47, 285001 (2014).

References

For a recent extensive survey on persistence from physics perspectives, see:

"Persistence and First-Passage Properties in Non-equilibrium Systems"

A.J. Bray, S.N. Majumdar and G. Schehr,

Adv. in Phys. 62, 225-361 (2013)

also available at arXiv: 1304.1195