

I Diffusion with Stochastic Resetting

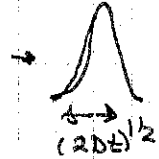
- IA Motivations (i) Searching for a target - local moves (diffusion) "intermittent search" + long range (resetting)  
 (ii) resetting a process  $\xrightarrow{t_0}$  initial condition  $\rightarrow$  NESS

IB Preliminaries - Diffusion

Diffusive Particle  $x_{t+\Delta t} = x_t + \xi$   $P(\xi) = \frac{1}{\sqrt{4\pi D \Delta t}} e^{-\frac{\xi^2}{4D \Delta t}}$



Diffusion Eq<sup>n</sup> for pdf  $\frac{\partial P(x,t|x_0)}{\partial t} = D \frac{\partial^2 P(x,t|x_0)}{\partial x^2}$  (1)  
 i.c.  $\delta(x-x_0)$   $\Rightarrow$  sol<sup>n</sup>  $P(x,t) = \frac{1}{\sqrt{4\pi Dt}} \exp\left(-\frac{(x-x_0)^2}{4Dt}\right)$  (2)



Sol<sup>n</sup> of (1) by L.T  $\tilde{P}(x,s|x_0) = \int_0^\infty dt e^{-st} P(x,t|x_0)$

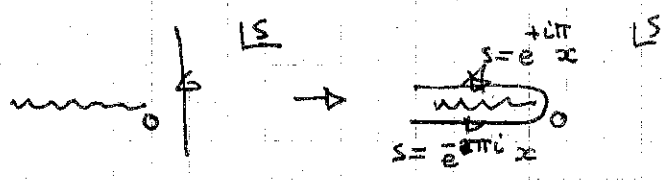
$D \frac{\partial^2 \tilde{P}}{\partial x^2} - s \tilde{P} = -\delta(x-x_0)$  (3)

$\rightarrow \tilde{P} = \frac{1}{2(sD)^{1/2}} e^{-(\frac{s}{D})^{1/2} |x-x_0|}$  (4)

Invert by 'magic integral'  $\int_0^\infty t^{\nu-1} e^{-\frac{\beta}{t} - st} dt = 2 \left(\frac{\beta}{s}\right)^{\nu/2} K_\nu(2\sqrt{\beta s})$  (5)

$K_{1/2}^{(\beta)} = K_{-1/2}^{(\beta)}(z) = \sqrt{\frac{\pi}{z}} e^{-z} z^{-1/2}$  then  $\nu=1/2$   $\beta = \frac{(x-x_0)^2}{4D} \rightarrow$  (2)

or invert by Bromwich contour  
 $P(x,t) = \frac{1}{2\pi i} \int_{\gamma-i\infty}^{\gamma+i\infty} e^{+st} \tilde{P}(x,s) ds$



IC Preliminaries - Survival probability



bc.  $P(x_T, t) = 0 \forall t$

absorbing target  $x_T$

take  $x_T = 0$

Approaches (i) solve (1) + bc.  $P(x,t|x_0) = \frac{1}{\sqrt{4\pi Dt}} \left[ \exp\left(-\frac{(x-x_0)^2}{4Dt}\right) - \exp\left(-\frac{(x+x_0)^2}{4Dt}\right) \right]$

Survival prob  $q(t|x_0) = \int_0^\infty dx P(x,t|x_0) = \text{erf}\left(\frac{x_0}{2\sqrt{Dt}}\right)$  (6)

(ii)

(i) use Backward Eq<sup>n</sup>  $\frac{\partial P}{\partial t} = D \frac{\partial^2 P}{\partial x_0^2} (x, t | x_0)$

b.c.  $P(x, 0 | x_0) = \delta(x - x_0)$   
 $P(x, t | 0) = 0$

survival prob  $\frac{\partial q}{\partial t} = D \frac{\partial^2 q}{\partial x_0^2}$

$q(0 | x_0) = 1$   
 $q(t | 0) = 0$

L.T  $D \frac{\partial^2 \tilde{q}}{\partial x_0^2} - s \tilde{q} = -1$   
 $-(s/D)^{1/2} x_0$

sol<sup>n</sup>  $\tilde{q}(s | x_0) = \frac{1 - e^{-\dots}}{s}$

Large t asymptotics ~ small s

$\tilde{q}(s | x_0) \approx \frac{1}{s^{1/2}} \frac{x_0}{D^{1/2}} + \text{const} + O(s^{1/2})$

$L^{-1} \left[ \frac{1}{s^\alpha} \right] = \frac{1}{\Gamma(\alpha)} t^{\alpha-1}$

$q(t | x_0) \approx \frac{x_0}{D^{1/2} \Gamma(1/2)} t^{-1/2} + O(t^{-3/2})$

Meantime to absorption  $T(x_0) = \int_0^\infty (-\frac{\partial q}{\partial t}) t dt = [q t]_0^\infty + \int_0^\infty q(t | x_0) dt = \tilde{q}(0 | x_0) \rightarrow \infty$

(iii) Renewal Eq<sup>n</sup>

$P(x, t | x_0) = \int_0^t dt' \underbrace{F(x, t' | x_0)}_{\text{PDF for 1st time to reach } x} P(x, t-t' | x)$

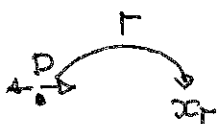
$F(x, t | x_0) = -\frac{\partial q(t | x_0)}{\partial t}$

L.T.  $\tilde{P}(x, s | x_0) = \tilde{F}(x, s | x_0) \tilde{P}(x, s | x)$

④  $\rightarrow \tilde{F}$  . ⑤  $\rightarrow F(x, t | x_0)$

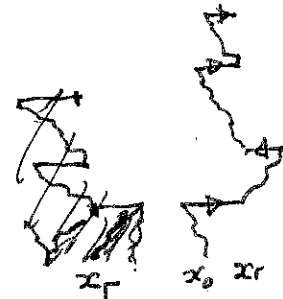
OMIT

ID Diffusion with Stochastic resetting



$x_{t+\Delta t} = x_t + \frac{\Delta x}{2}$  prob  $1 - r\Delta t$   
 $= x_r$  prob  $r\Delta t$

resetting is Poisson Process



$\frac{\partial P}{\partial t} (x, t | x_0, x_r) = D \frac{\partial^2 P}{\partial x^2} - rP + r \delta(x - x_r)$

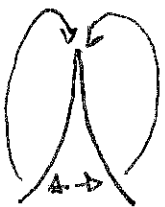
Stationary State  
 NESS Circulation of prob

$D \frac{\partial^2 P}{\partial x^2} - rP = -r \delta(x - x_r)$

$P = \frac{\alpha_0}{2} e^{-\alpha_0 |x - x_r|}$  ⑧  
 $\alpha_0 = \left(\frac{r}{D}\right)^{1/2}$

c.f. ③

c.f. ④



resetting current

diffusive current

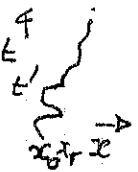
relaxation to NESS

renewal equation  
(suppress  $x_r$ )

$$P(x, t | x_0) = e^{-rt} G_D(x, t | x_0) + r \int_0^t dt' e^{-r(t-t')} G_D(x, t-t' | x_r) \quad (10)$$

no resets

$t' = \text{time of last reset}$



$G_D$  is  $\frac{1}{\sqrt{4\pi Dt}}$  exp  $-\frac{(x-x_0)^2}{4Dt}$

$$t \rightarrow \infty \quad P^*(x | x_r) = r \int_0^\infty dt' e^{-rt'} G_D(x, t' | x_r) \quad \text{recovers (9)}$$

large  $t$ : evaluate  $\int_0^t dt' e^{-r(t-t')} \frac{e^{-\frac{(x-x_r)^2}{4D(t-t')}}}{\sqrt{4\pi D(t-t')}} \quad \text{by saddle point}$

take  $x - x_r = O(t)$  let  $t - t' = y t$

$$\int_0^1 dy e^{-\left[ \frac{r}{D} y + \frac{(x-x_r)^2}{4D y} \right]}$$

if  $|x - x_r| > t (4Dr)^{1/2}$

$y = 1$  i.e.  $t' = 0$  dominates

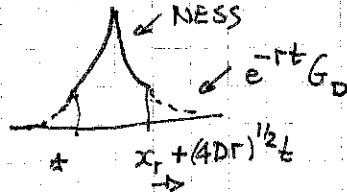
$$\approx e^{-rt} G_D(x, t | x_r)$$

if  $|x - x_r| < t (4Dr)^{1/2}$

$y = \frac{|x - x_r|}{t} (4Dr)^{1/2}$  dominates

$$\approx \frac{\sqrt{r/D}}{2} e^{-\frac{|x-x_r|}{D} (r/D)^{1/2} t}$$

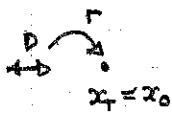
so ~~equation~~ front travels with  $v = (4Dr)^{1/2}$   
"stationarity"



IE Diffusion + resetting + absorbing target

[here take  $x_r = x_0$  to simplify]  
 $x_r = 0$

$x$   
 $x_r$   
absorbing



Use renewal eq<sup>n</sup>

$$P(x, t | x_0) = e^{-rt} G_D(x, t | x_0) + r \int_0^t dt' e^{-r(t-t')} G_D(x, t-t' | x_0) \int_0^\infty dx' P(x', t' | x_0) \quad (11)$$

survival prob  $q(t' | x_r)$

$G$  is now propagator with absorbing bdy

int wrt  $x$

$$q(x, t | x_r) = e^{-rt} q_D(t | x_0) + r \int_0^t dt' e^{-r(t-t')} q_D(t-t' | x_0) q(t' | x_0)$$

L.T

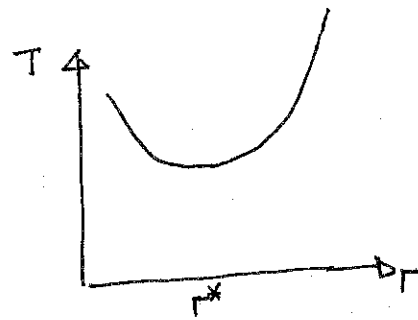
$$\tilde{q}(s | x_0) = \hat{q}_D(s+r | x_0) + r \hat{q}_D(s+r | x_0) \tilde{q}(s | x_0)$$

$$\tilde{q}(s | x_r) = \frac{\hat{q}_D(s+r | x_0)}{1 - r \hat{q}_D(s+r | x_0)} = \frac{1 - e^{-\alpha x_0}}{s + r e^{-\alpha x_0}} \quad \alpha = \left( \frac{r+s}{D} \right)^{1/2} \quad (12)$$

MTA  $\tilde{q}(s=0) = T = \frac{e^{-\alpha_0 x_0} - 1}{r} \quad (13)$

$\alpha_0 = \left( \frac{r}{D} \right)^{1/2}$

$$T = \frac{e^{\lambda_0 x_0} - 1}{\Gamma}$$



as  $r \rightarrow \Gamma$

$$T(0) \rightarrow \infty$$

$$T(\infty) \rightarrow \infty$$

$$\text{min at } \frac{dT}{dr} = 0$$

$$\frac{dT}{dr} = 0 \Rightarrow$$

$$\frac{x}{s} = 1 - e^{-y}$$

$$y = \lambda_0 x_0 = \left(\frac{\lambda_0}{\Gamma}\right) \frac{x_0}{\Gamma} = \frac{\text{distance to target}}{\text{typical length diffused between resets}}$$

unique sol<sup>n</sup>

$$y^* = 1.5936 \dots$$

optimal ratio of lengths

← Finish lecture I here →

IF / Diff + reset + target Survival prob

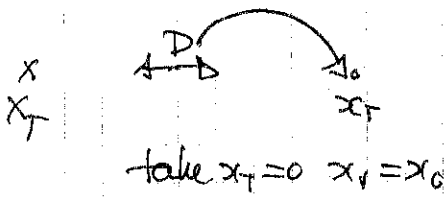
$$\tilde{q}(s|x_0) = \frac{1 - e^{-\lambda x_0}}{s + \Gamma e^{-\lambda x_0}}$$

(12)

Inversion

# Lecture II

Recap

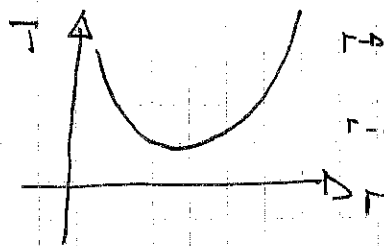


Survival prob LIT  
 $q(s|x_0) = \frac{1 - e^{-\lambda x_0}}{s + re^{-\lambda x_0}}$  (12)

$\lambda = \left(\frac{s+r}{D}\right)^{1/2}$

$T = \hat{q}(0|x_0) = \frac{e^{-\lambda_0 x_0}}{r}$  (13)

$\lambda_0 = \left(\frac{r}{D}\right)^{1/2}$



$r \rightarrow 0 \quad T \sim \frac{x_0}{\sqrt{r}} \rightarrow \infty$   
 $r \rightarrow \infty \quad T \sim \frac{e^{-\lambda_0 x_0}}{r} \rightarrow 0$

$\frac{dT}{dr} = 0 \Rightarrow \frac{y}{2} = 1 - e^{-y} \quad y = \lambda_0 x_0 = \frac{x_0}{(D/r)^{1/2}} = \frac{\text{distance to target}}{\text{typical length diffused between resets}}$

unique solution  $y > 0$

$y \approx 1.5936$  optimal ratio of lengths

IF Diff + reset + target survival prob

Inversion  $q(t|x_0) = \frac{1}{2\pi i} \int_{\gamma-i\infty}^{\gamma+i\infty} ds e^{st} \frac{1 - e^{-\lambda x_0}}{s + re^{-\lambda x_0}}$

poles at  $s_0 + re^{-\left(\frac{r+s_0}{D}\right)^{1/2} x_0} = 0$  (14) gives dominant contribution

b.p.t  $\frac{1}{2} - \gamma t$  [check]  
 subdominant

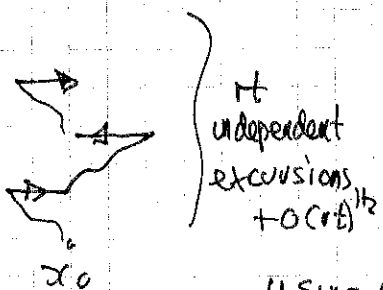
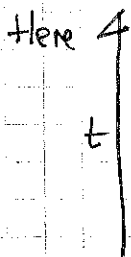
$\rightarrow q(t|x_0) \approx e^{s_0 t} [1 - \exp^{-\left(\frac{r+s_0}{D}\right)^{1/2} x_0}]$

if  $|s_0| \ll 1 \quad s_0 \approx -re^{-y} \quad q(t|x_0) \approx e^{-rt} e^{-y}$

Aside on Gumbel dist<sup>n</sup>

$N$  iidrv with pdf  $f(x) \quad P_N(M) = \text{prob max} < M = \left[ \int_{-\infty}^M dx f(x) \right]^N$   
 $= \exp N \ln [1 - \int_M^{\infty} dx f(x)] \approx \exp \left[ -N \int_M^{\infty} dx f(x) \right]$

if  $f(x) \sim A e^{-ax} \quad x \text{ large} \quad P_N(M) \approx \exp \left[ -\frac{NA}{a} e^{-aM} \right]$  Gumbel



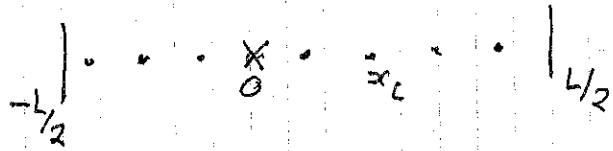
max excursion to left  $< x_0$

$q \approx \mathbb{P} \left[ \int_0^{x_0} dt e^{-rt} r q_D(t|x_0) \right]^N$   
 $= \left[ 1 - e^{-x_0 \left(\frac{r}{D}\right)^{1/2}} \right]^N$  (not average over duration)  
 $\approx \exp[-rt e^{-y}]$

using (7)

## II Many Searchers

IIA Many independent searchers beginning and resetting to  $x_i$



$i=1 \dots N$   
 $p(x_i) dx_i = \frac{dx_i}{L}$  uniform

$\rho = \frac{N}{L}$  density

Survival prob. of target at origin  
 need to average over  $\{x_i\}$

$Q(t | \{x_i\}) = \prod_{i=1}^N q(t | x_i)$

Aside "Quenched" and "annealed" averages

Example r.v.  $x = e^{aN}$  prob  $1-p$   $p = e^{-cN}$   
 $= e^{bN}$   $p$   $N$  large.

clearly 'typical'  $x = e^{aN}$

but  $E[x] = (1 - e^{-cN})e^{aN} + e^{(b-c)N} \approx e^{(b-c)N}$  if  $b < c > a$ .

but  $E[\ln x] = aN [1 - e^{-cN}] + bN e^{-cN} \approx aN$  "annealed average"

so  $e^{\langle \ln x \rangle}$  gives typical value "quenched average"

"annealed" ave  $\langle Q(t | \{x_i\}) \rangle = \prod_{i=1}^N \langle q(t | x_i) \rangle = \exp N \ln \langle q(t | x_i) \rangle_x$

$\langle q(t | x) \rangle_x = \int_{-\infty}^{\infty} dx \frac{1}{L} \int_{-L/2}^{L/2} dx' q(t | x') = \frac{2}{L} \int_0^{L/2} dx [1 - (1 - q(t | x))]$   
 $= 1 - \frac{2}{L} \int_0^{L/2} dx (1 - q(t | x))$

$\therefore Q^{av} \approx \exp N \ln [1 - \frac{2}{L} \int_0^{L/2} dx (1 - q(t | x))] \xrightarrow{N, L \rightarrow \infty} \exp -2\rho \int_0^{\infty} dx (1 - q(t | x))$   
 $\xrightarrow{\frac{N}{L} = \rho} \approx \exp -2\rho I_1(t)$

$Q^{typ} = \langle \exp \langle \ln Q \rangle \rangle = \exp N \langle \ln q \rangle_x = \exp -2\rho I_2$

$I_2(t) = - \int_0^{\infty} dx \ln q(t | x) \quad [ > 0 ]$

— half time break —

## II B Diffusive searchers (no resetting)

$$q(t|x_0) = \text{erf}\left(\frac{x_0}{2\sqrt{Dt}}\right) \quad (6) \quad I_1 = 2\sqrt{Dt} \int_0^\infty du \text{erf}(u) = 2\sqrt{\frac{Dt}{\pi}} \text{erfc}(x_0)$$

$$I_2 = -2\sqrt{Dt} \int_0^\infty du \ln \text{erf}(u) \quad \text{converges} \quad 1 - \text{erf}(u)$$

$$Q^{\text{av, typ}} = \exp\left[-\lambda^{\text{av, typ}} \sqrt{Dt}\right] \quad \text{stretched exponential decay}$$

III with resetting

$$I_1 = \int_0^\infty dx (1 - q(t|x_0)) \quad \text{L.T. } \hat{I}_1 = \int_0^\infty dx \left(\frac{1}{s} - \hat{q}(s|x_0)\right)$$

$$= \int_0^\infty dx \left[ \frac{(s+r)e^{-sx}}{s(s+r)e^{-sx}} \right] = \frac{r+s}{sr} \ln\left(\frac{s+r}{s}\right) \quad \text{sing at } s=0 \text{ furthest to right}$$

logarithm  $\sim$  small  $s$   $\sim -\frac{1}{s} \ln s + \mathcal{O}(1/s)$

$$\text{Now } \int_0^\infty dt e^{-st} \ln t = \frac{1}{s} \int_0^\infty du e^{-u} \ln \frac{u}{s} = -\frac{\gamma}{s} - \frac{\ln s}{s} \quad \gamma = \int_0^\infty du e^{-u} \ln u$$

$$L^{-1}\left[-\frac{\ln s}{s}\right] = \ln t + \gamma_E \quad I_1 \approx +\left(\frac{D}{r}\right)^{1/2} [\ln t + \gamma_E]$$

$$Q_{\text{av}} \sim \text{const} \exp\left(-2\rho\left(\frac{D}{r}\right)^{1/2} \ln t\right) = \text{const } t^{-2\rho\left(\frac{D}{r}\right)^{1/2}} \quad \text{Power law decay (slow)}$$

exponent  $\left(\frac{D}{r}\right)^{1/2} / \rho$   $\frac{\text{diffusive length}}{\text{spacing between searchers}}$

$$\frac{I_2}{2} = -\int_0^\infty dx \text{erf}\left(\frac{x}{2\sqrt{Dt}}\right) = \sqrt{Dt} \int_0^\infty du \ln \text{erf}(u)$$

$$I_2 = -\int_0^\infty dx_0 \ln [q(x_0, t)] \quad t \gg 1 \quad q(x_0, t) \sim e^{-S_0(x_0 t)} - \left(\frac{S_0}{r}\right)^{1/2} x_0$$

$$= -t \int_0^\infty dx_0 S_0(x_0) \quad u = -\frac{x_0}{t} \quad u = \exp\left[-y(1-u)^{1/2}\right] \quad y = x_0 x_0$$

$$\frac{I_2}{2} = (Dr)^{1/2} \int_0^\infty dy u(y) = (Dr)^{1/2} \left\{ \int_0^1 du \left| \frac{dy}{du} \right| u \right\} \quad y = -\frac{\ln u}{(1-u)^{1/2}}$$

$$= (Dr)^{1/2} 4(1 - \ln 2)$$

$$\text{so } Q_{\text{av}} \sim \frac{1}{t} \quad Q_{\text{typ}} \sim \exp\left[-8\rho(Dr)^{1/2} [1 - \ln 2] t\right]$$

Explanation: • average dominated by rare i.c. where searchers far from target

• strong memory of i.c. through resetting

•  $Q_{\text{typ}}$  decays faster than <sup>purely</sup> diffusive case

# Lecture III : Generalisations of resetting and Optimization Problems

## III A Resetting & Target distributions

- on resetting  $x \rightarrow x_T$  ~~chosen~~ <sup>drawn</sup> from distribution  $P_r(x_T)$
- for simplicity choose distribution of i.c.  $x_0$  to also be  $P_r(x_0)$
- Having known target position is artificial, instead let  $P_T(x_T)$  be the distribution of the target position.

Renewal equation  $\text{I} \text{ (with absorption at fixed target } x_T)$  generalizes to, when we integrate over  $P_r(x_T)$ ,

$$P(x, t | x_0, x_T) = e^{-\gamma t} G_D(x, t | x_0, x_T) + \gamma \int_0^\infty dt e^{-\gamma(t-t')} \int dx_T P_r(x_T) G_D(x, t-t' | x_T, x_T) \times \int dx P(x, t' | x_0, x_T) \quad (1)$$

Now integrate over i.c.  $x_0$  with dist<sup>bn</sup>  $P_r(x_0)$

$$P(x, t | x_T) = e^{-\gamma t} G_D(x, t | x_T) + \gamma \int_0^\infty dt e^{-\gamma(t-t')} \int dx G_D(x, t-t' | x_T) \int dx P(x, t' | x_T)$$

where now  $G_D(x, t | x_T) = \int dx_0 P_r(x_0) G_D(x, t | x_0, x_T)$   $P(x, t | x_T) = \int dx_0 P_r(x_0) P(x, t | x_0, x_T)$

Finally integrate over  $x$  to get survival prob  $q(t | x_T)$

$$q(t | x_T) = e^{-\gamma t} q_D(t | x_T) + \gamma \int_0^\infty dt e^{-\gamma(t-t')} q_D(t-t' | x_T) q(t' | x_T) \quad (2)$$

as before solve for L.T.  $\tilde{q}(s | x_T)$

$$\tilde{q}(s | x_T) = \frac{\tilde{q}_D(s + \gamma | x_T)}{1 - \gamma \tilde{q}_D(s + \gamma | x_T)}$$

③ # main result

$$q_D(t | x_T) = \int dx G_D(x, t | x_T)$$

$$\tilde{q}_D(s | x_T) = \int dx_0 P_r(x_0) \frac{1 - e^{-\gamma(x_0 - x_T)}}{\gamma - \gamma e^{-\gamma(x_0 - x_T)}}$$

$$T(x_T) = \frac{\tilde{q}_D(\gamma | x_T)}{1 - \gamma \tilde{q}_D(\gamma | x_T)} \quad (4)$$

$$= \frac{1}{\gamma} \left[ \frac{1}{\int dx_0 P_r(x_0) e^{-\gamma(x_0 - x_T)}} - 1 \right]$$

$$= \frac{1}{\gamma} - \frac{1}{\gamma} \int dx_0 P_r(x_0) e^{-\gamma(x_0 - x_T)}$$



Now the stationary distribution in the absence of an absorbing target is

$$p^*(x) = \int dx_r P_r(x_r) p^*(x|x_r)$$

$$= \frac{\kappa_0}{2} \int dx_r P_r(x_r) \exp(-\kappa_0 |x-x_r|)$$

using I (1) (5)

so  $T(x_T) = \frac{1}{\Gamma} \left[ \frac{\kappa_0}{2p^*(x_T)} - 1 \right]$

Average over target distribution  $P_T(x_T)$

$$\bar{T}(x_T) = \int dx_T P_T(x_T) \frac{1}{\Gamma} \left[ \frac{\kappa_0}{2p^*(x_T)} - 1 \right]$$

(5)

### III B Optimisation of MTA

Extremise  $\bar{T}(x_T)$  for given  $P_T(x_T)$  wrt  $P_r(x_r)$

assume  $P_T(x_T)$  symmetric about 0

minimise  $\int dx_T \frac{P_T(x_T)}{p^*(x_T)}$  subject to constraint

$$\delta \int dx \left[ \frac{P_T(x)}{p^*(x)} + \lambda \int dx P_r(x) \right] = 0$$

Lagrange Multiplier

→ E-L eqn  $\int dx \frac{P_T(x)}{[p_{opt}^*(x)]^2} e^{-\kappa_0 |x_T - x|} = \frac{2\lambda}{\kappa_0}$

holds  $\forall x_T \Rightarrow \frac{P_T(x)}{[p_{opt}^*(x)]^2} = \lambda$

$$\Rightarrow p_{opt}^*(x) = \frac{P_T^{1/2}(x)}{\lambda^{1/2}} = \frac{P_T^{1/2}(x)}{\int dx_T P_T^{1/2}(x_T)}$$

(6) (7)

so  $p^*(x)$  should be narrower than target dist<sup>n</sup>.

Solution of (5) for  $P_r(x_r)$

$$P_r(x) = p_{opt}^*(x) - \frac{1}{\kappa_0^2} \frac{d^2 p_{opt}^*(x)}{dx^2}$$

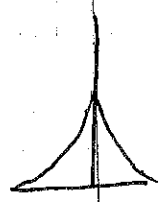
(8)

but need  $P_r(x_r) \geq 0$  for 'realisable'  $p_{opt}^*$

Example: Exponential target dist<sup>bn</sup>  $P_T(x) = \frac{\beta}{2} e^{-\beta|x|}$  small  $\beta$  broad  
 $\beta \rightarrow \infty \rightarrow \delta(x)$

if  $P_T(x_1) = \delta(x_1 - x_0)$  MTA diverges if  $\alpha_0 > \beta$

Small  $\beta < 2\alpha_0$   $\textcircled{7} \rightarrow P_{opt}^*(x) = \frac{\beta}{4} e^{-\beta|x|/2}$   
 $\textcircled{8} \rightarrow P_r^{opt}(x) = \frac{\beta}{4} e^{-\beta|x|/2} \left[ 1 - \frac{\beta^2}{4\alpha_0^2} \right] + \frac{\beta^2}{4\alpha_0^2} \delta(x)$



$\Rightarrow$  transition at  $\beta = 2\alpha_0$  pure  $\delta$  function is (locally) optimal  
 -ve when  $\beta > 2\alpha_0$

### III C Time dependent resetting

Generalisation  $\Gamma \rightarrow \Gamma(x)$  space dependent; use Master equation  
 $\Gamma(t)$  time dependent

Let  $\tau = t - t_e$   $t_e =$  time of last reset.  
 Prob of no resets up to  $t = e^{-\int_0^t \Gamma(\tau) d\tau} \equiv e^{-R(t)}$

prob of next reset  $t \rightarrow t+dt = \left( -\frac{d}{dt} e^{-R(t)} \right) dt$  cf. cts time r.w. "waiting time"

No absorption

Last renewal  $P(x,t|x_0) = e^{-R(t)} G_D(x,t|x_0) + \int_0^t dt' \Gamma(t') e^{-R(t-t')} P(x,t-t'|x_0)$   
 $+ \int_0^t dt' e^{-R(t-t')} \psi(t') G_D(x,t-t'|x_0)$

First renewal  $P(x,t|x_0) = e^{-R(t)} G_D(x,t|x_0) + \int_0^t dt' \Gamma(t') e^{-R(t-t')} P(x,t-t'|x_0)$   
 prob of 1<sup>st</sup> reset at  $t'$

solve by L.T  $\tilde{P}(x,s) = \frac{1}{sH} \tilde{G}_D$

$\tilde{G}_D = \int_0^\infty dt e^{-st - R(t)} G_D(x,t)$   
 $H = \int_0^\infty dt e^{-st - R(t)}$

st. st  $\lim_{s \rightarrow 0} [s\tilde{P}]$  non-zero

requires  $\int_0^\infty e^{-R(t)} dt < \infty$

$\Rightarrow \Gamma(t)$  decays more slowly than  $\frac{1}{t}$

e.g.  $\Gamma(t) \sim \frac{A}{t}$   $R \approx A \ln t$   $e^{-R(t)} \sim t^{-A}$   
 $\int_0^\infty e^{-R(t)} dt$  converges if  $A > 1$

Survival target  $x_0 = x_T$

survival prob  $q(t|x_0) = e^{-R(t)} q_D(t|x_0) + \int_0^t dt' r(t') e^{-R(t')} q_D(t-t'|x_0) q(t-t'|x_0)$

$$\hat{q}(s|x_0) = \frac{\hat{q}_D(s|x_0)}{s \hat{q}_D(s|x_0) - \hat{r}(s|x_0)}$$

$$\hat{q}_D = \int_0^\infty dt e^{-st - R(t)} q_D(t|x_0)$$

$$\hat{r}_D = \int_0^\infty dt e^{-st - R(t)} \frac{d}{dt} q_D(t|x_0)$$

$$s \rightarrow 0 \quad T(x_0) = - \frac{\int_0^\infty dt e^{-R(t)} q_D(x_0, t)}{\int_0^\infty dt e^{-R(t)} \left[ \frac{d}{dt} q_D(x_0, t) \right]} \quad (9) \equiv$$

### III) Optimal resetting rate

Extremize MTA wrt  $r(t)$

$$E-L. \frac{\delta T}{\delta r(t)} = \frac{\int_t^\infty dt' e^{-R(t')} q_D(t'|x_0)}{\hat{r}(x_0|x_0)} - \frac{q(0|x_0)}{[\hat{r}(0|x_0)]^2} \int_t^\infty dt' e^{-R(t')} \frac{\partial q_D}{\partial t}$$

= 0 no solution which holds  $\forall t$

Conjecture is global optimum  $r(t) = \begin{cases} 0 & t < t^* \\ \kappa & t > t^* \end{cases}$  (piecewise) (10)

$$(9) \rightarrow T(x_0) = \frac{\int_0^{t^*} dt q(t^*|x_0)}{1 - q(t^*|x_0)} \quad (11) \text{ otherwise}$$

$$t^* = \frac{x_0^2}{2D} \cdot 3.668$$

$$T = 2.671 \dots \frac{x_0^2}{2D}$$

better than constant rate  $r_c$

$$T = 3.0882 \frac{x_0^2}{2D}$$

(10) is local optimum of MTA subject to constraint  $r(t) > 0$

$$i.e. \quad \frac{\delta T}{\delta r(t)} \begin{cases} > 0 & t < t^* \\ = 0 & t > t^* \end{cases}$$