

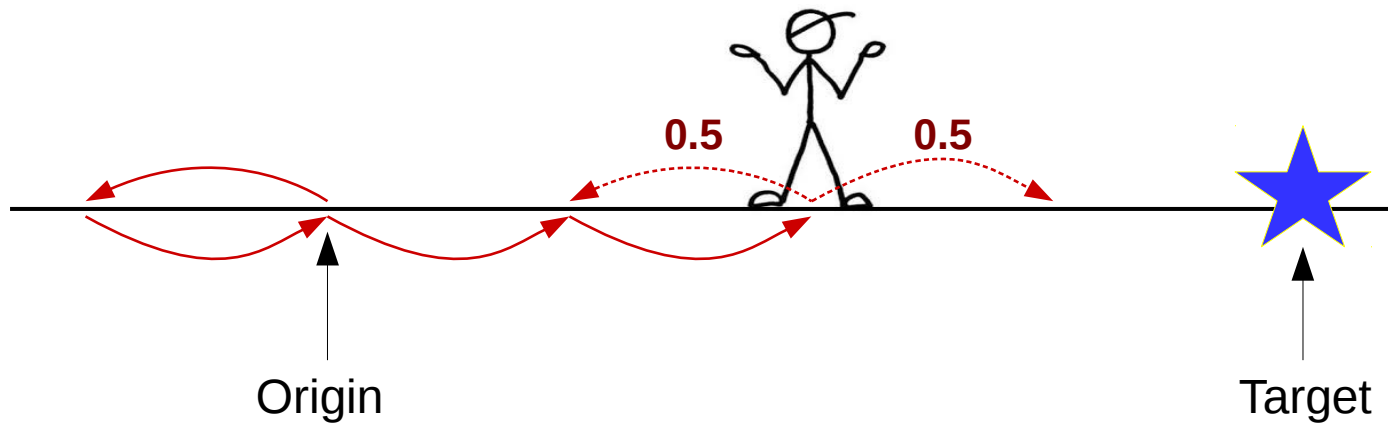
# Efficiency of Stochastic Searches with Costly and Punctual Restarts



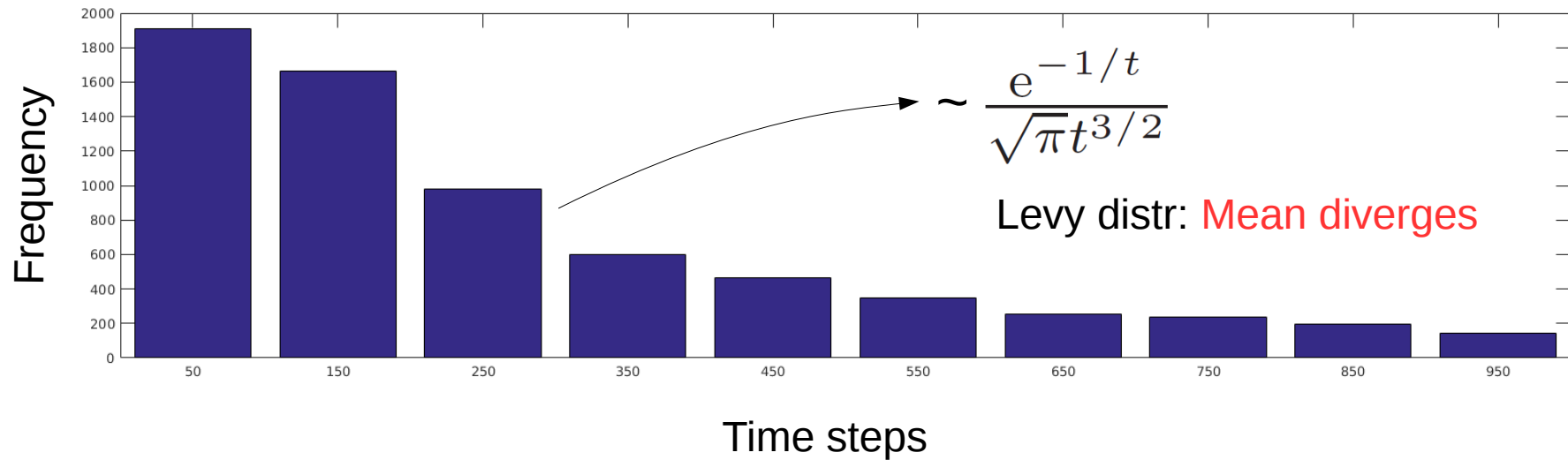
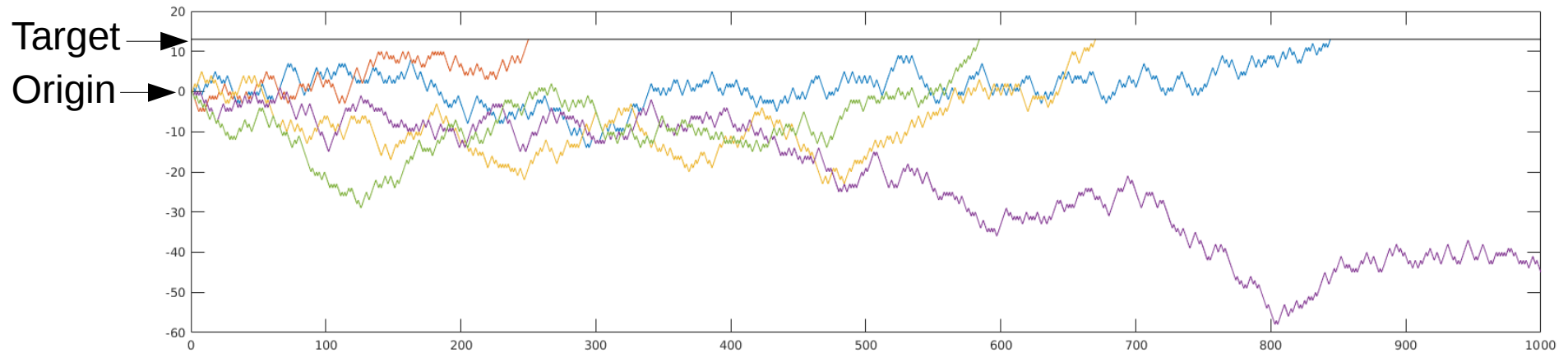
Kabir Husain, Sandeep Krishna  
National Centre for Biological Sciences, Bangalore

More details:  
Kabir's poster  
[arXiv:1609.03754](https://arxiv.org/abs/1609.03754)

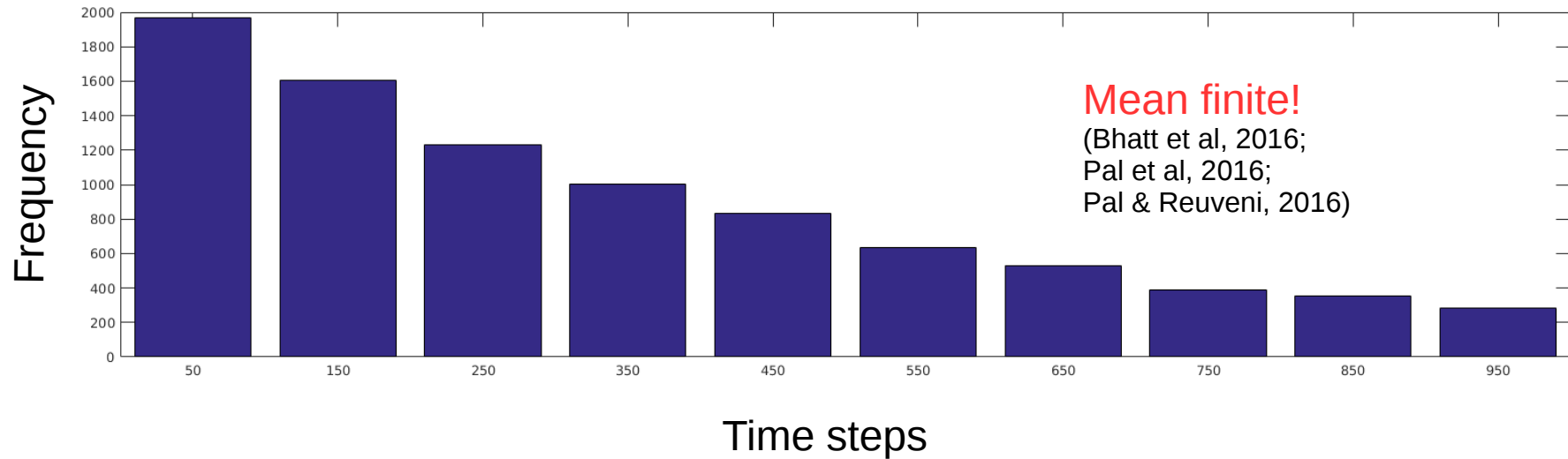
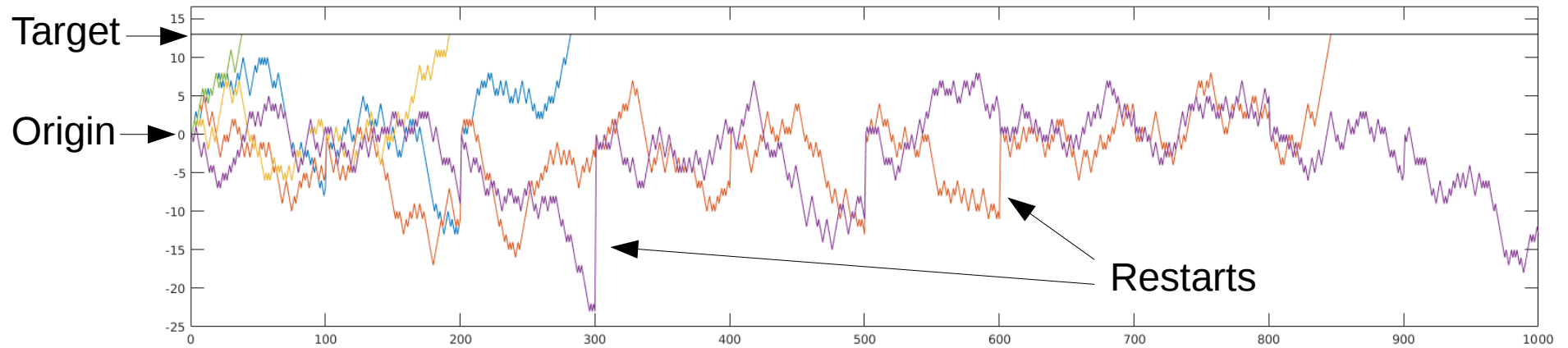
# 1D Diffusive Search for a Fixed Target



# 1D Diffusive Search for a Fixed Target



# 1D Diffusive Search for a Fixed Target with Restarts



# Stochastic Searches with Restarts in ...

## Computer science:

### Satisfiability (SAT) problems

GSAT Algorithm [Selman, Mitchell, Levesque, 1992]

WalkSAT algorithm family [McAllester et al., 1997]

Hoos 1999

### Travelling Salesman problem

Johnson & McGeoch 1997

Stutzle & Hoos 1997–2000

### Graph partitioning

Martin, Houdayer, Schreiber, 1999

### Quadratic Assignment Problem

Stutzle 1997-1999; Taillard 1995

### Combinatorial Auctions

Hoos & Boutilier 2000

## Biology (the role of restarts is speculation):

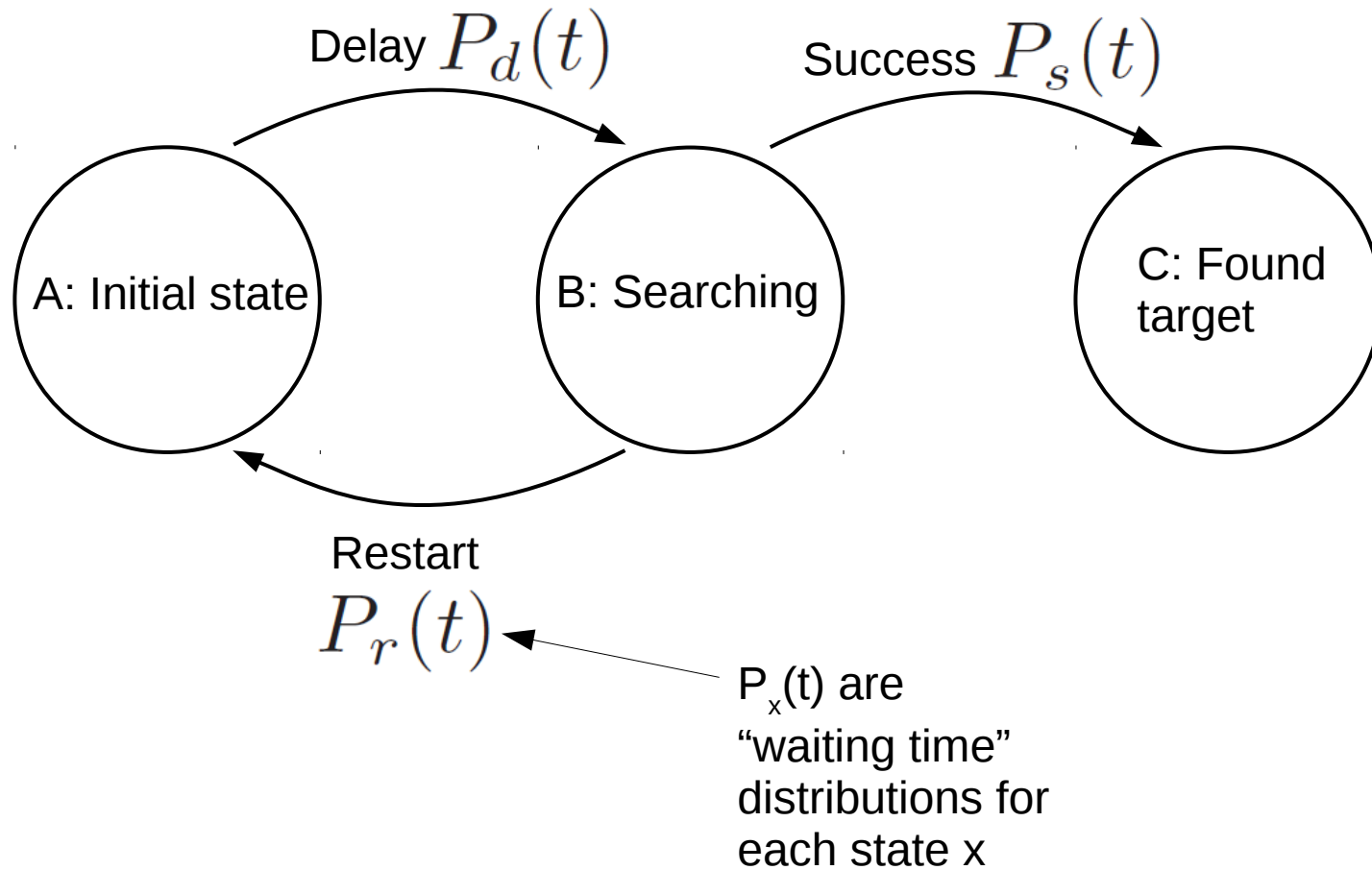
### Animal foraging

Immune system recognition of self vs. non-self

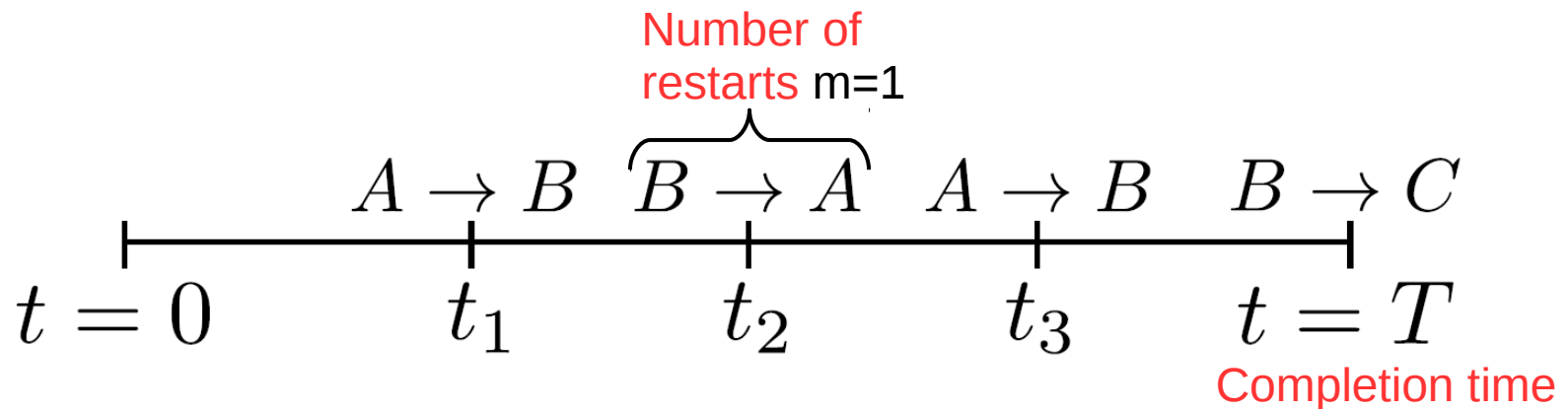
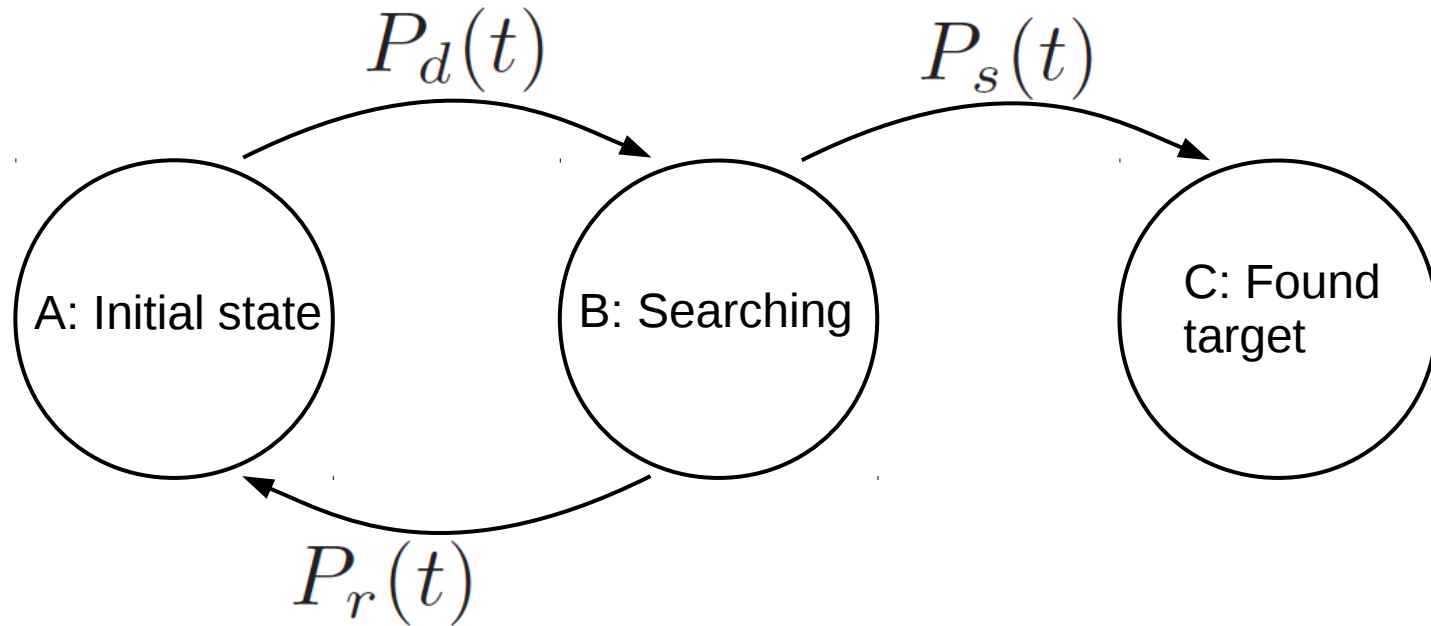
Protein folding (do some chaperones implement restarts?)

Binding of regulatory proteins to DNA

# General Framework

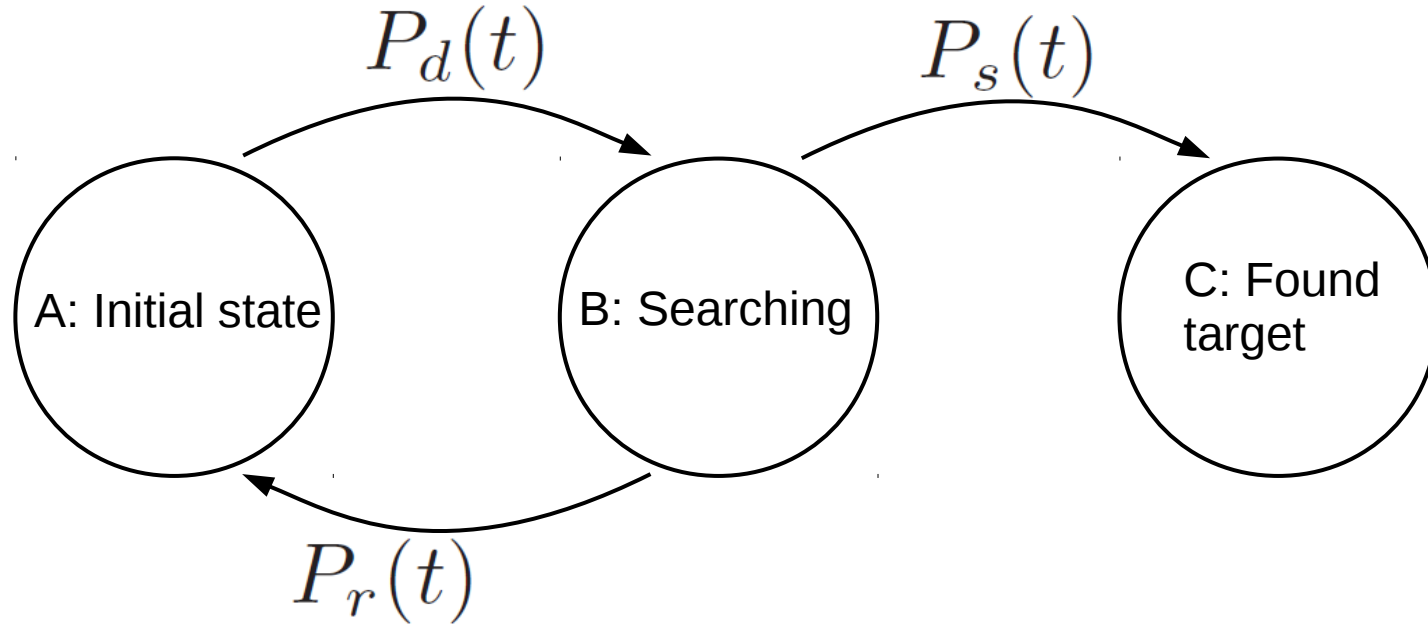


# General Framework



Joint probability distribution:  $P(m, T)$

# General Framework



$$\mathcal{L}(P(m, T)) \equiv \int_0^\infty dT P(m, T) e^{-sT} = \hat{G}_f(s) \left( \hat{G}_i(s) \right)^m$$

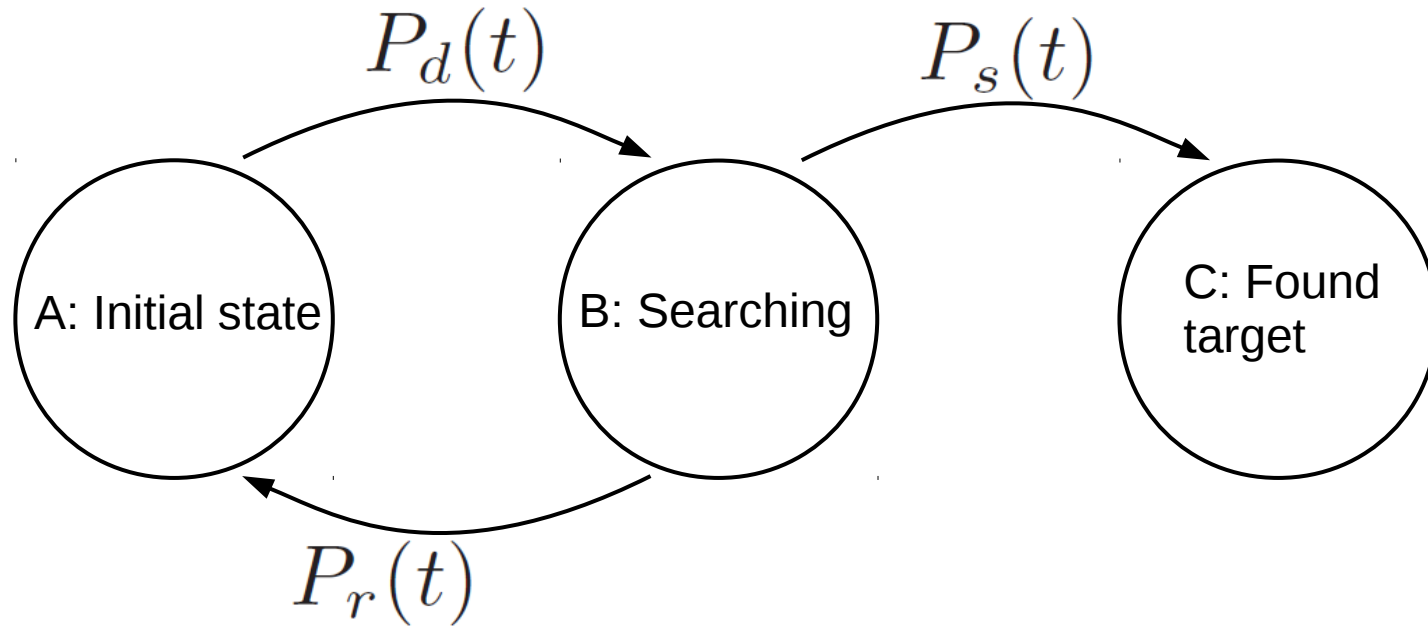
where  $\hat{G}_i(s) = \mathcal{L}[P_d(t)] \times \mathcal{L}[P_r(t)S_s(t)],$

$$\hat{G}_f(s) = \mathcal{L}[P_d(t)] \times \mathcal{L}[P_s(t)S_r(t)], \quad S_x(t) = 1 - \int_0^t dt' P_x(t')$$

$$\mathcal{L}(P(T)) = \hat{G}_f \sum_{m=0}^{\infty} \hat{G}_i^m = \frac{\hat{G}_f(s)}{1 - \hat{G}_i(s)}$$



# General Framework



Computation of  
all moments  
reduced to  
quadratures

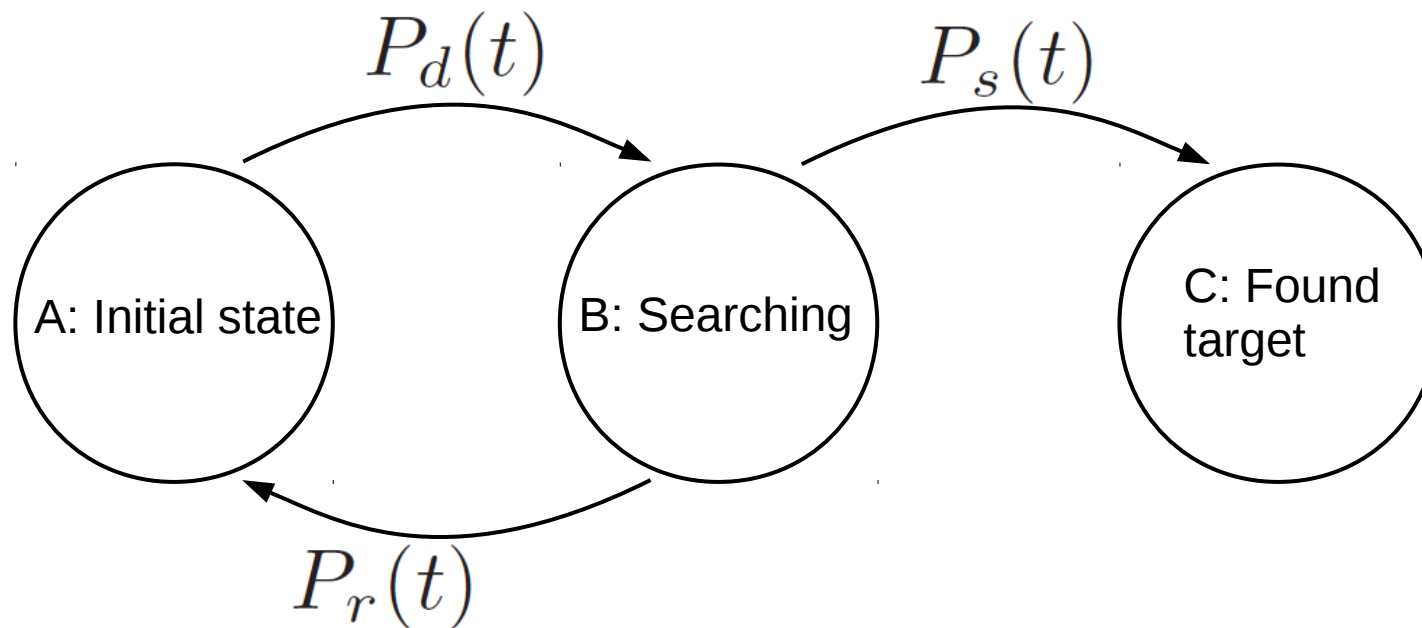
The case of zero delay:  $P_d(t) = \delta(t)$

$$\langle T \rangle = \frac{G_f^{(1)} + G_i^{(1)}}{G_f^{(0)}}, \quad \begin{aligned} G_i(t) &= P_r(t) S_s(t) \\ G_f(t) &= P_s(t) S_r(t) \end{aligned} \quad G_x^{(n)} = \int dt t^n G_x(t)$$

$$\langle T^2 \rangle = \frac{2G_i^{(1)} (G_f^{(1)} + G_i^{(1)}) + G_f^{(0)} (G_f^{(2)} + G_i^{(2)})}{(G_f^{(0)})^2}$$

$$\mathcal{L}(P(T)) = \hat{G}_f \sum_{m=0}^{\infty} \hat{G}_i^m = \frac{\hat{G}_f(s)}{1 - \hat{G}_i(s)}$$

# General Framework



## Re-deriving previously known results:

(1) No delay, deterministic restart:  $P_r(t) = \delta(t - \tau)$ ;  $P_d(t) = \delta(t)$

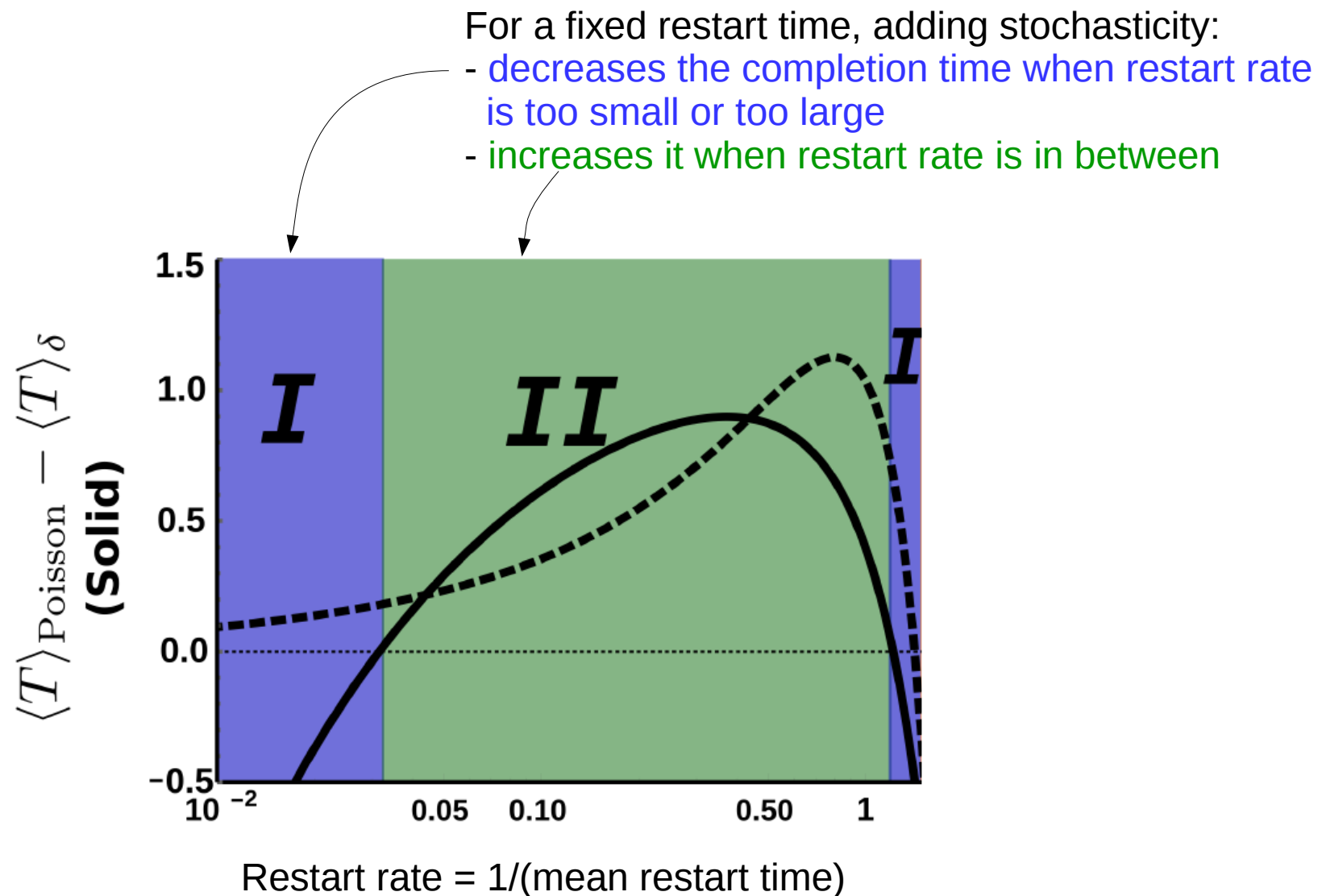
$$\langle T \rangle_\delta = \frac{\int_0^\tau dt S_s(t)}{1 - S_s(\tau)} \quad (\text{Bhatt et al, 2016; Pal et al, 2016; Pal \& Reuveni, 2016})$$

Consequence: **any deterministic restart renders  $\langle T \rangle$  finite**, as long as the denominator is non-zero

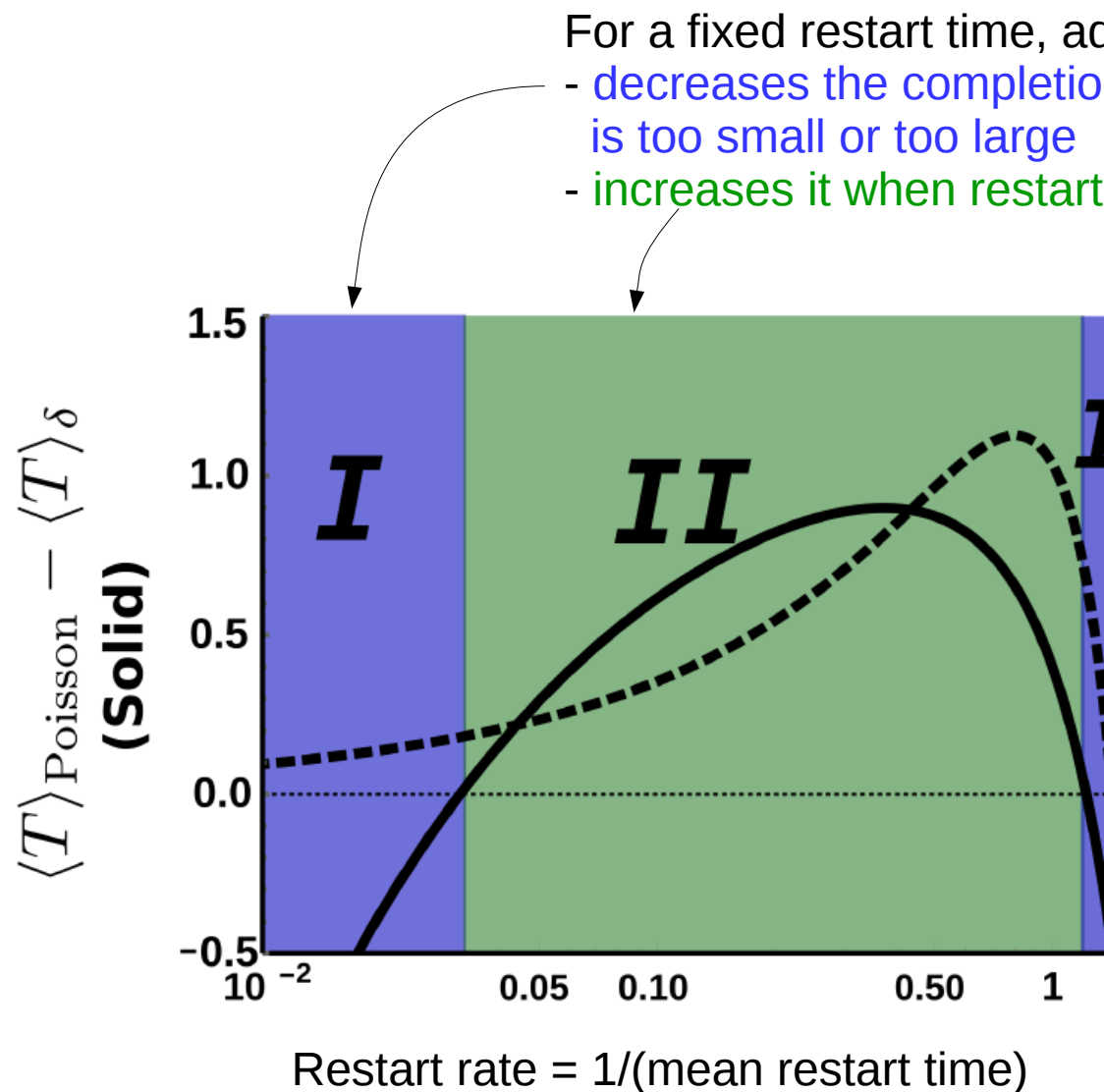
(2) No delay, Poisson restarts, 1D diffusive search:  $P_s(t) = \frac{e^{-1/t}}{\sqrt{\pi t^{3/2}}}$ ;  $P_r(t) = k \exp(-kt)$   
 $\langle T \rangle = (\exp(2\sqrt{k}) - 1) / k$  (Evans & Majumdar 2011)

(3) No delay, power law restarts, 1D diffusive search (Nagar & Gupta 2016): e.g.  $P_r(t) = \frac{\eta}{(\eta + t)^2}$   
 **$\langle T \rangle$  can be finite even if search process *and* restarts have diverging mean waiting times**

## New Results: When does stochasticity in restart distribution help in reducing mean completion time?



## New Results: When does stochasticity in restart distribution help in reducing mean completion time?




However, the *optimal* restart distribution is the deterministic one (Evans & Majumdar 2011; Pal & Reuveni 2016)

What if there are unavoidable sources of noise in the restarts?


**New Results:** What if restarts are 'punctual', i.e., almost-deterministic?

$$P_r(t) = \delta(t - \tau) + \sum_{n=2}^{\infty} \frac{(-1)^n}{n!} \mu_n(\tau) \delta^{(n)}(t - \tau)$$

$$\langle T \rangle = \langle T \rangle_{\delta}(\tau) + R_{\sigma^2}(\tau) \sigma_r^2 + \mathcal{O}(\mu_3(\tau))$$


$$\langle T \rangle_{\delta} = \frac{\int_0^{\tau} dt S_s(t)}{1 - S_s(\tau)}$$

Deterministic restarts

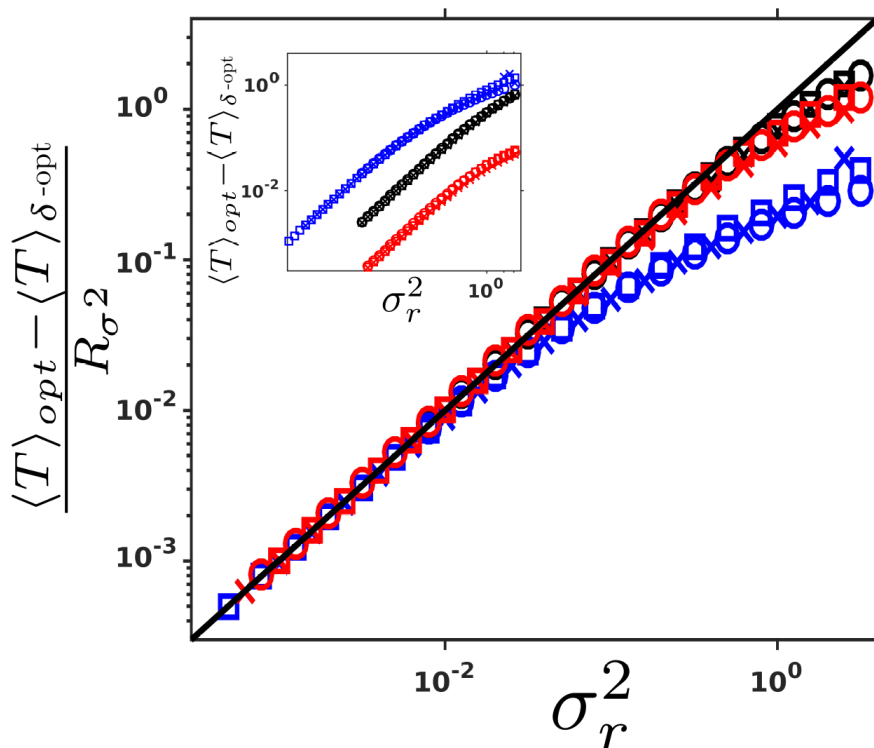


Variance of the  
restart distribution

**New Results:** What if restarts are 'punctual', i.e., almost-deterministic?

$$P_r(t) = \delta(t - \tau) + \sum_{n=2}^{\infty} \frac{(-1)^n}{n!} \mu_n(\tau) \delta^{(n)}(t - \tau)$$

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Variance of the  
restart distribution

Simple scaling of optimal completion time  
with variance of restart time distribution

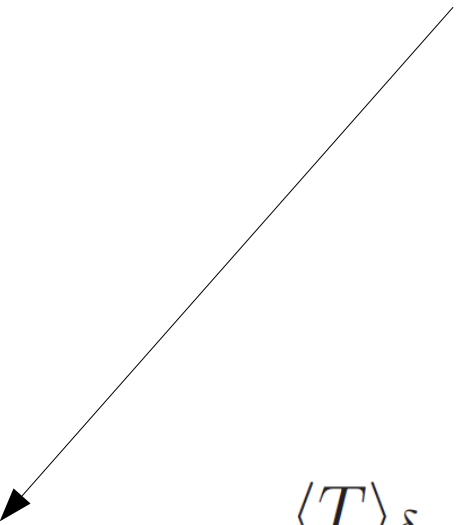
Search distr:  
Levy, log-normal, Frechet

Restart distr:  
Gamma (X), Weibull (O),  
truncated normal (□)

**New Results:** What if restarts are 'punctual', i.e., almost-deterministic?

$$P_r(t) = \delta(t - \tau) + \sum_{n=2}^{\infty} \frac{(-1)^n}{n!} \mu_n(\tau) \delta^{(n)}(t - \tau)$$

$$\langle T \rangle = \langle T \rangle_{\delta}(\tau) + R_{\sigma^2}(\tau) \sigma_r^2 + \mathcal{O}(\mu_3(\tau))$$



$$R_{\sigma^2} = -\frac{\langle T \rangle_{\delta}}{2(1 - S_s(\tau))} \left[ \frac{P_s(\tau)}{\langle T \rangle_{\delta}} + \frac{\partial P_s}{\partial t} \Big|_{t=\tau} \right]$$

Simple bounds on optimal restart time:  $\tau_{\text{opt}} > t_{\text{mode}}$

Optimal deterministic restart time

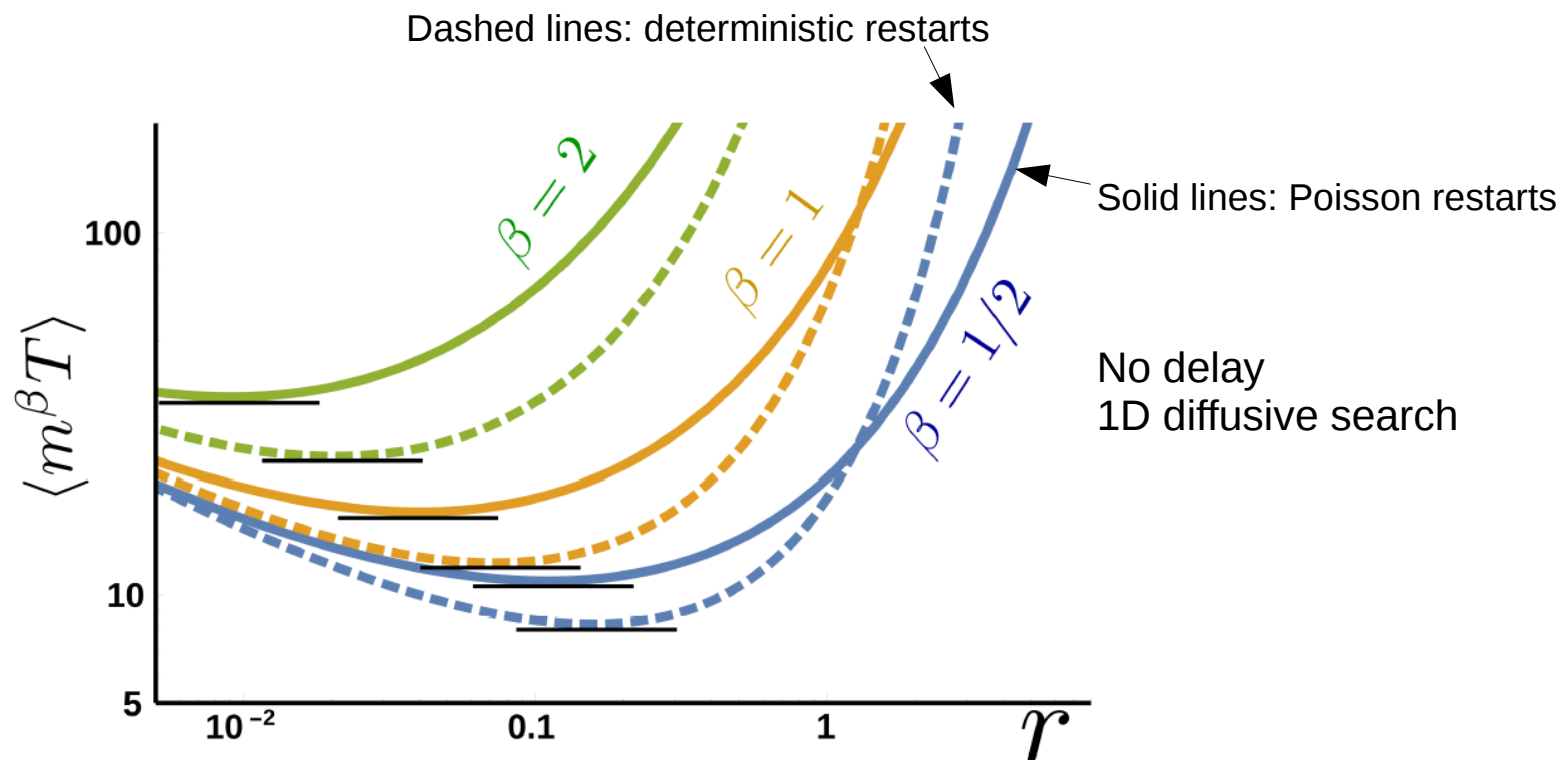
Mode of the search distribution time

**New Results:** What if restarts are 'costly', i.e., one wants to minimise some combination of completion time and number of restarts?

Moments of  $m$  and cross-moments are also easy to compute

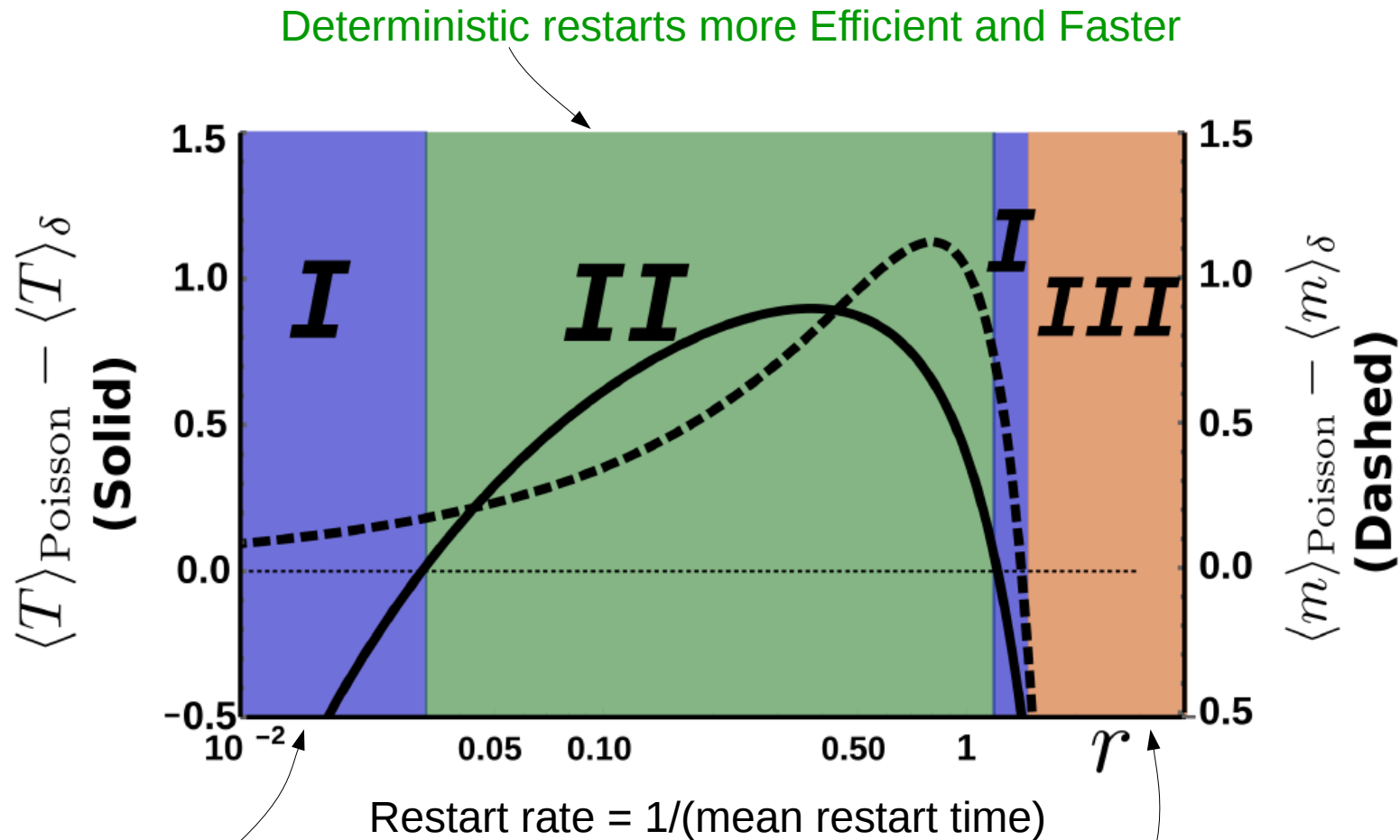
$$\langle m^\beta T \rangle = \frac{1}{\langle m \rangle} G_f^{(0)} G_i^{(1)} \Phi \left( G_i^{(0)}, -\beta - 1, 0 \right) + G_f^{(1)} \Phi \left( G_i^{(0)}, -\beta, 0 \right)$$

Lerch transcendent





**New Results:** What if restarts are 'costly', i.e., one wants to minimise some combination of completion time and number of restarts?

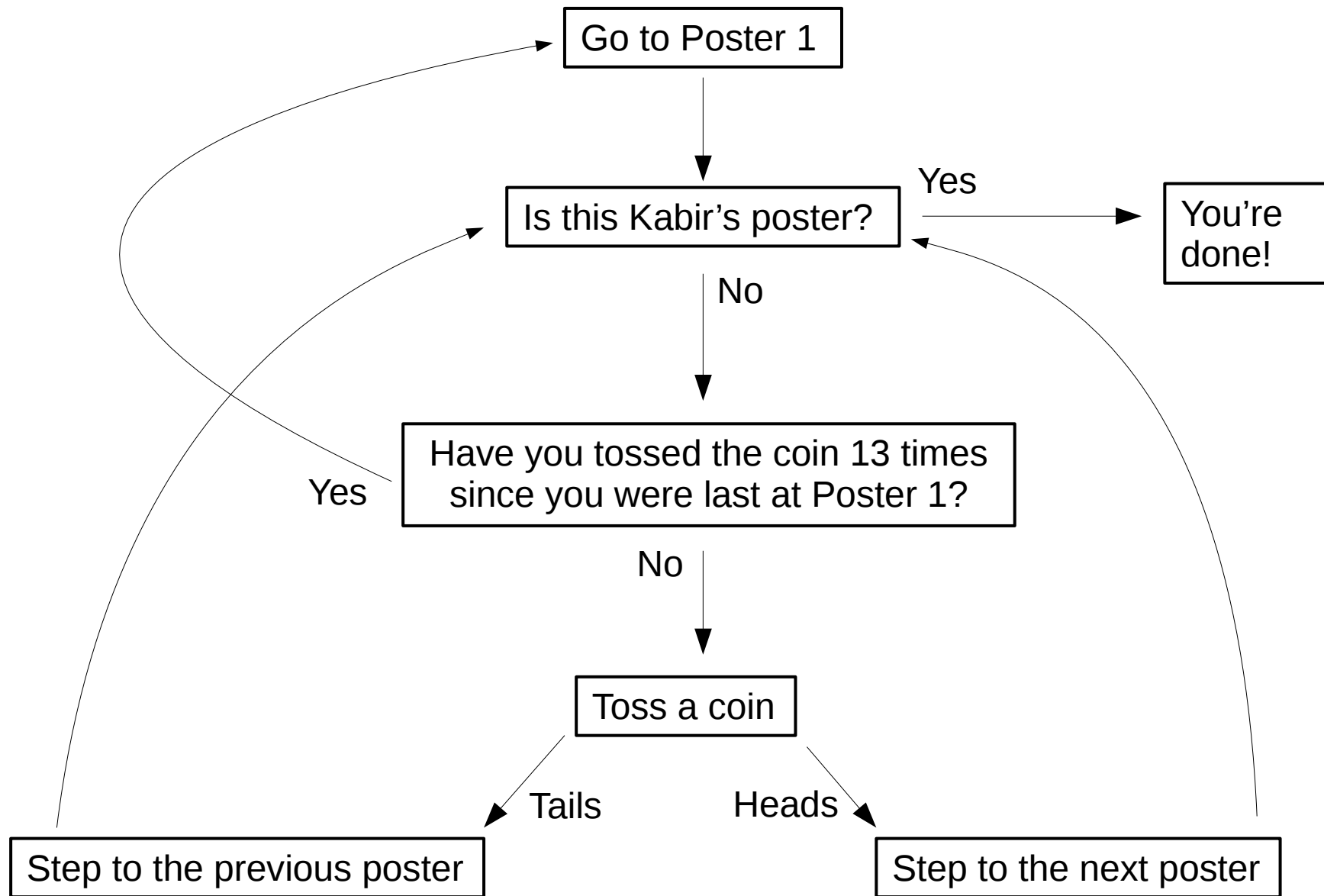


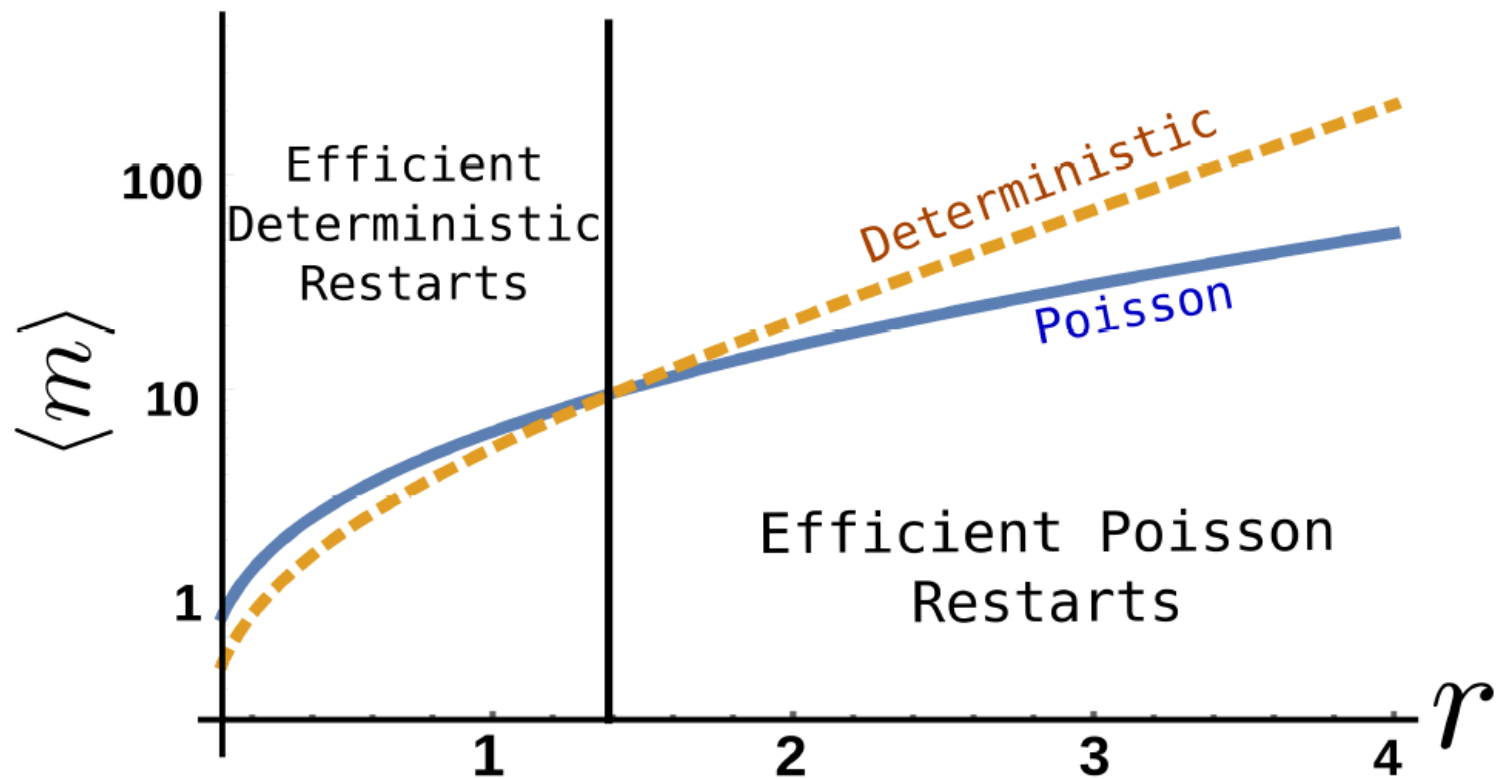
Deterministic restarts more Efficient  
but Stochastic restarts Faster

Stochastic restarts more Efficient and Faster

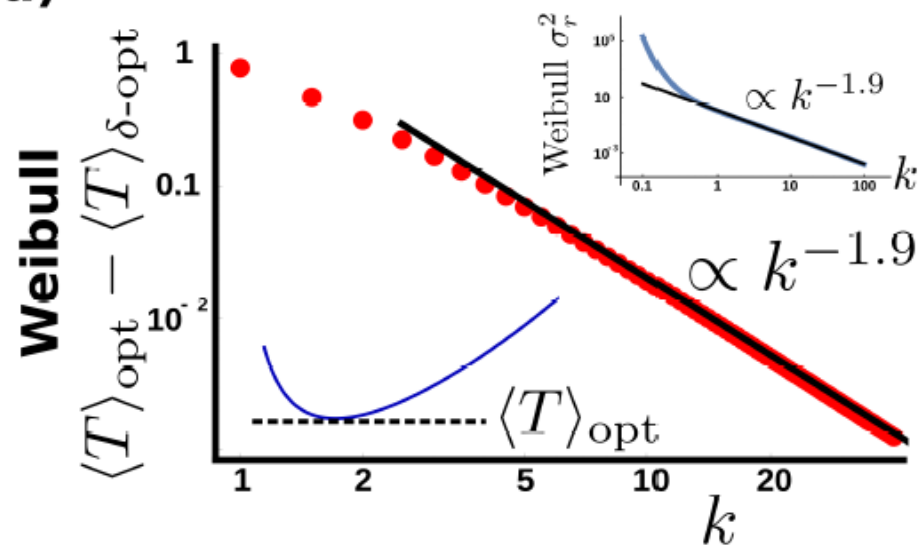
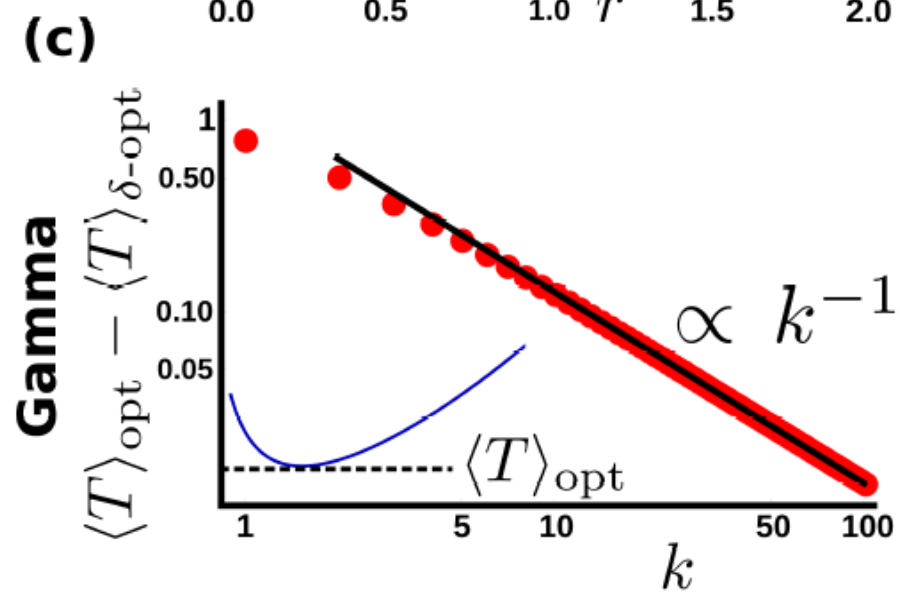
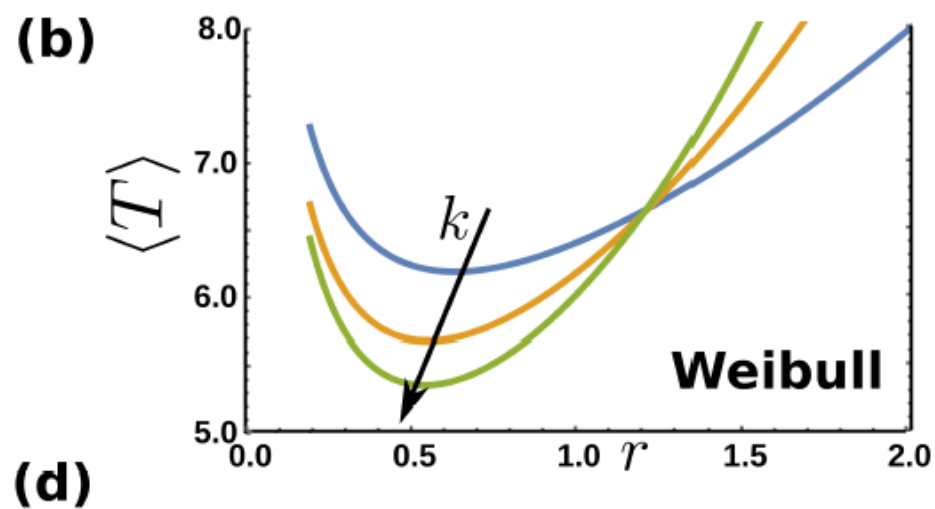
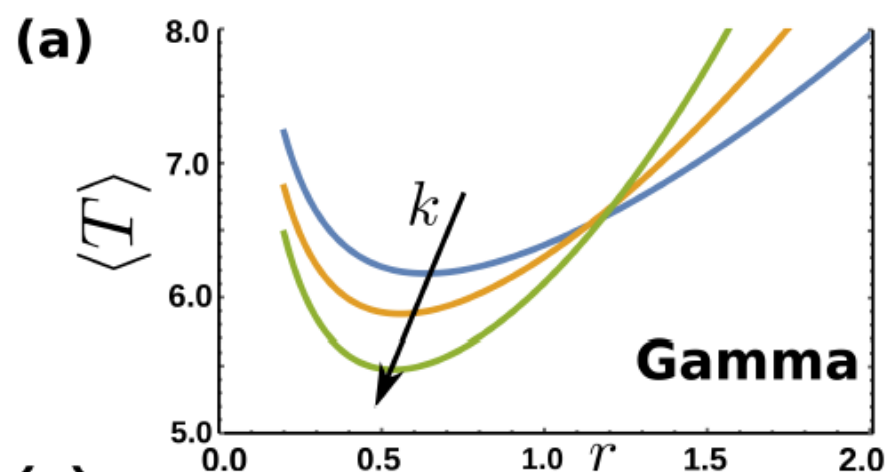
( 'more Efficient' => has smaller  $\langle m \rangle$ ; 'Faster' => has smaller  $\langle T \rangle$  )

# How to find Kabir's poster in a finite time

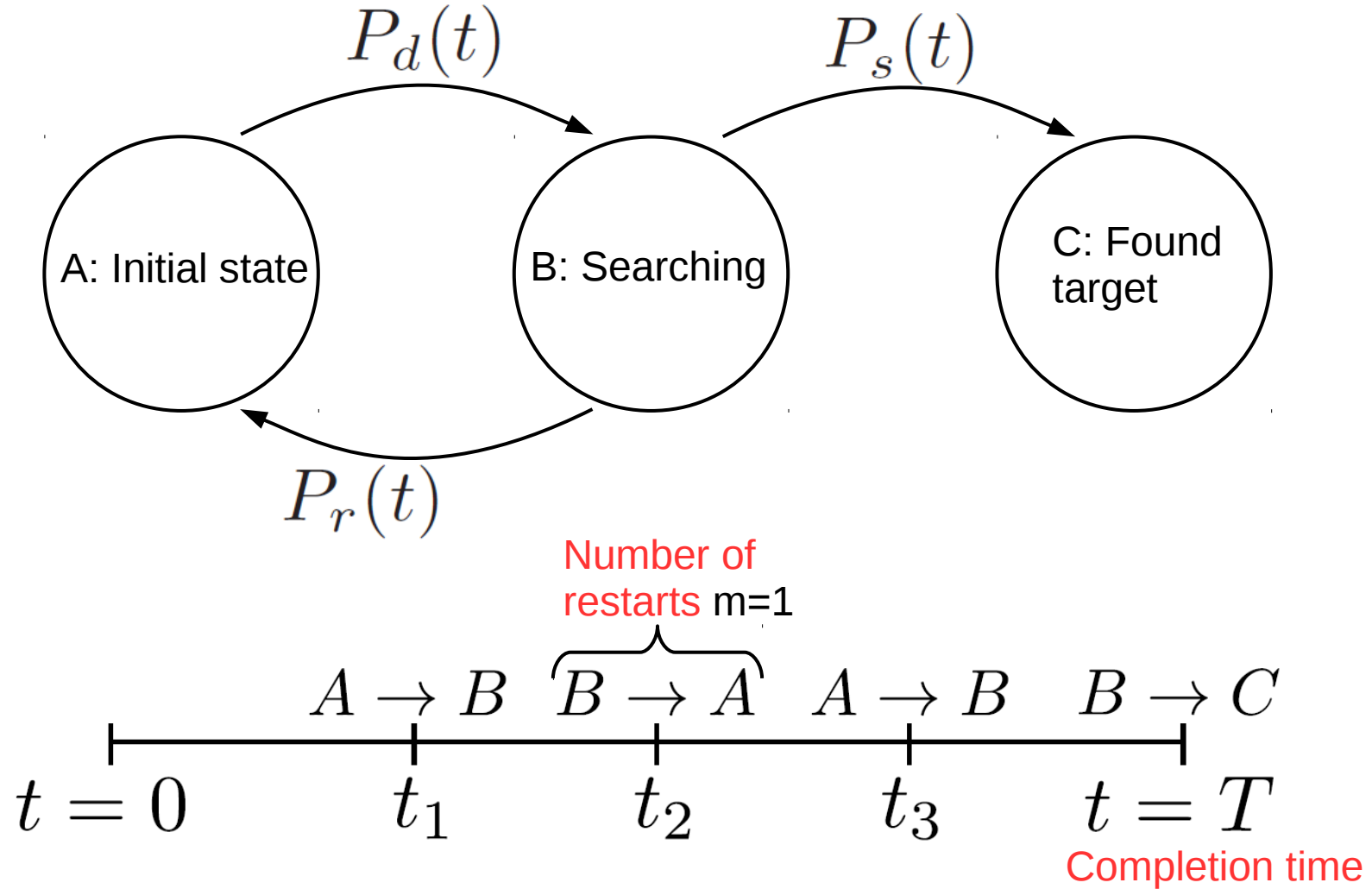




$$\langle m \rangle \approx \langle m \rangle_{\delta} - \frac{\sigma_r^2}{2 [1 - S_s(\tau)]^2} \left. \frac{\partial P_s}{\partial t} \right|_{t=\tau}$$

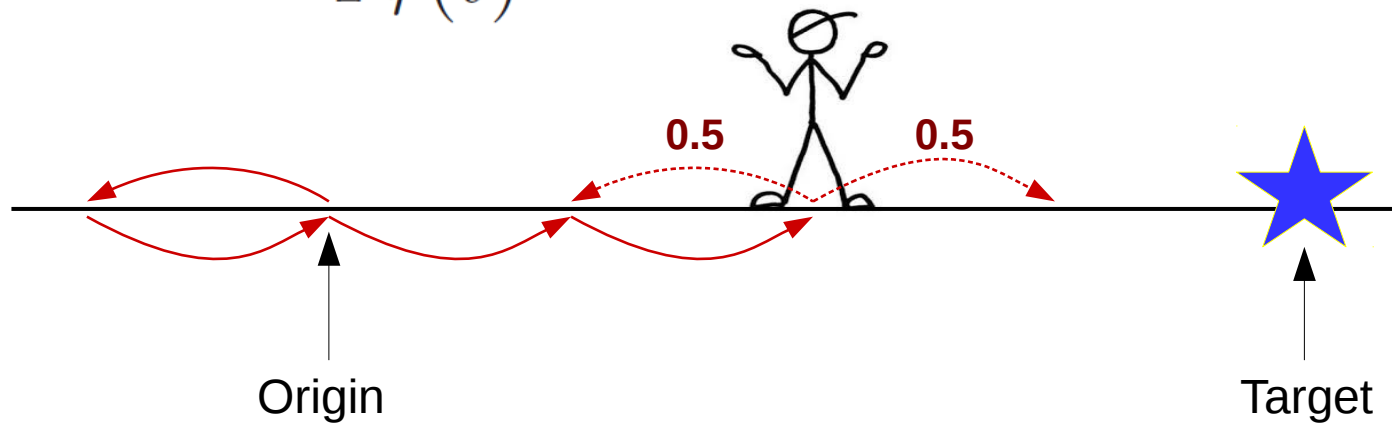
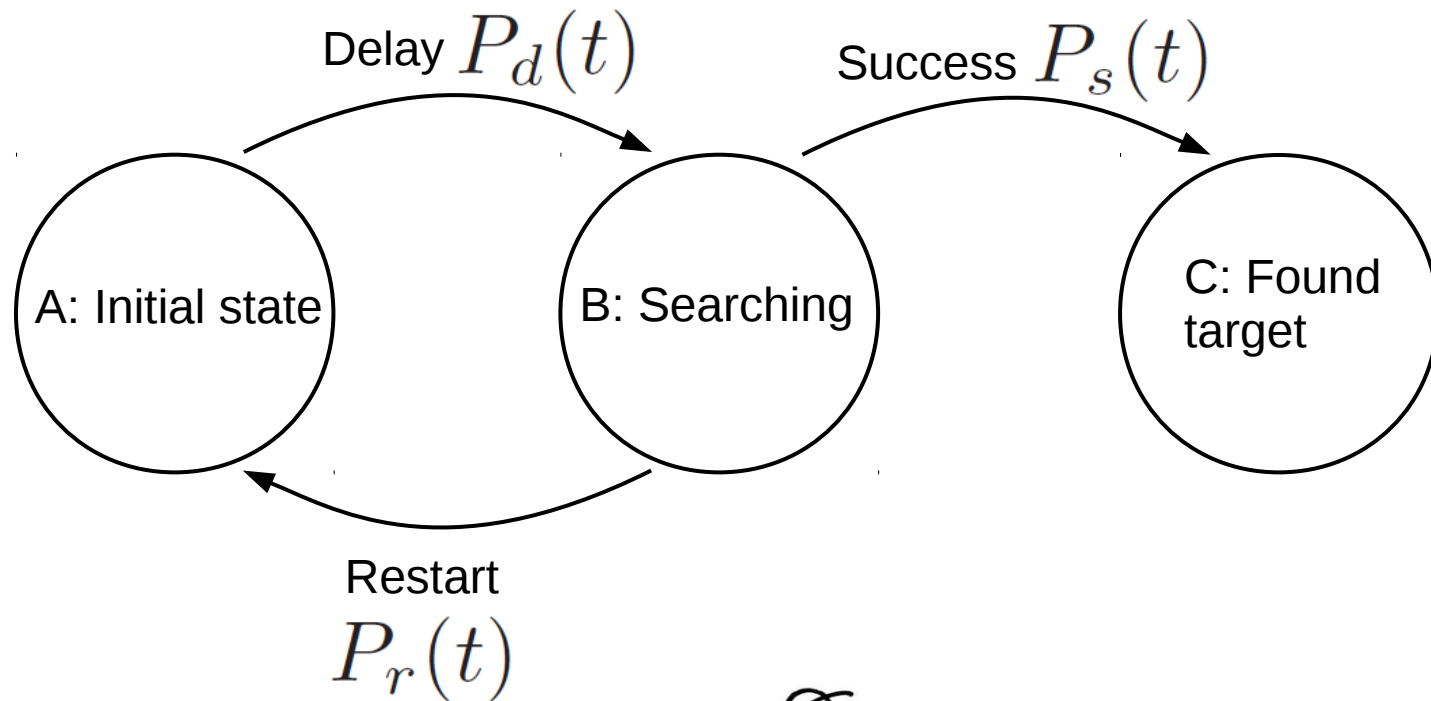


# General Framework



Joint probability distribution: 
$$P(m, T) = \int_0^T dt_{N-1} \dots \int_0^{t_2} dt_1 P(m, T, \{t_i\})$$

where 
$$P(m, T, \{t_i\}) = P_d(t_1) [P_r(t_2 - t_1) S_s(t_2 - t_1)] \dots \times [P_s(T - t_{N-1}) S_r(T - t_{N-1})]$$



$$\mathcal{L}(P(m, T)) \equiv \int_0^\infty dT P(m, T) e^{-st} = \hat{G}_f(s) \left( \hat{G}_i(s) \right)^m$$

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Stutzle & Hoos 1997–2000

Graph partitioning [Martin, Houdayer, Schreiber, 1999]

SAT [e.g., Selman et al. 1992–1997, Hoos et al. 1994–2000]

TSP [e.g., Johnson & McGeoch 1997, Stutzle & Hoos “  
1997–2000]

Quadratic Assignment Problem [e.g., Stutzle 1997-1999; “  
Taillard 1995]

Scheduling [e.g., den Besten, Stutzle, Dorigo 2000] “ Planning [e.g., Kautz et al.  
1996–1999, Brafman & Hoos  
1998–2000]

Combinatorial Auctions [Hoos & Boutilier 2000]

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