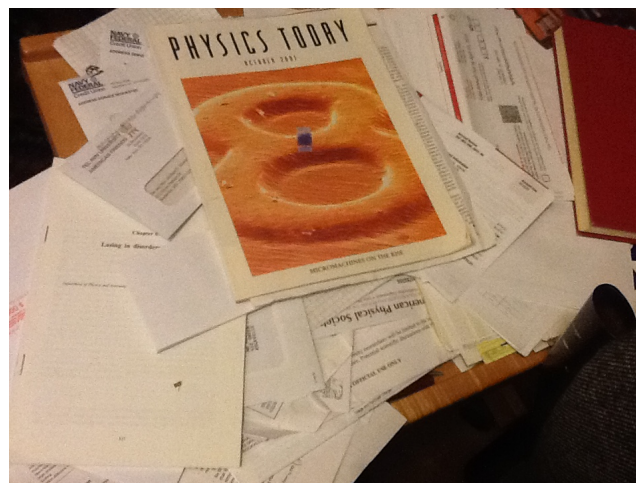
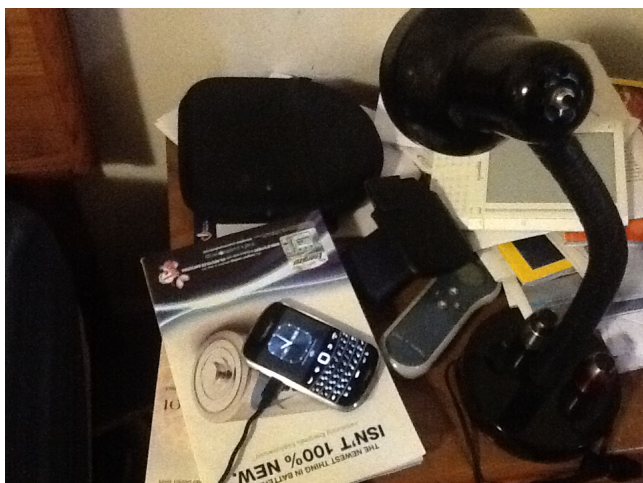


# Is There an Optimal Search Strategy, When Time is Limited?



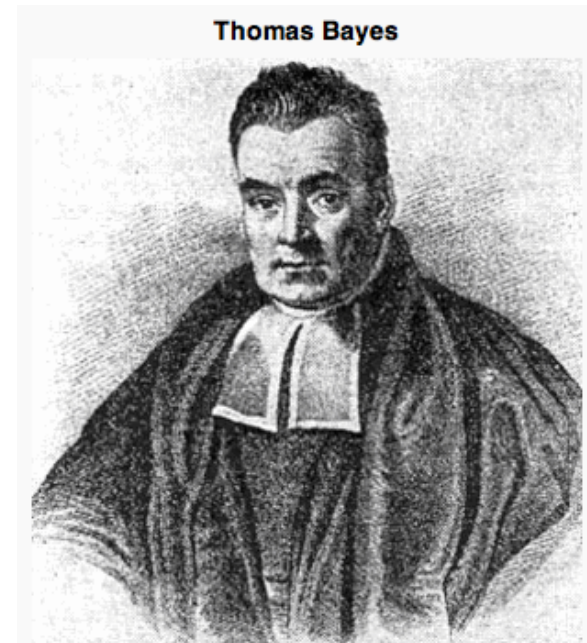
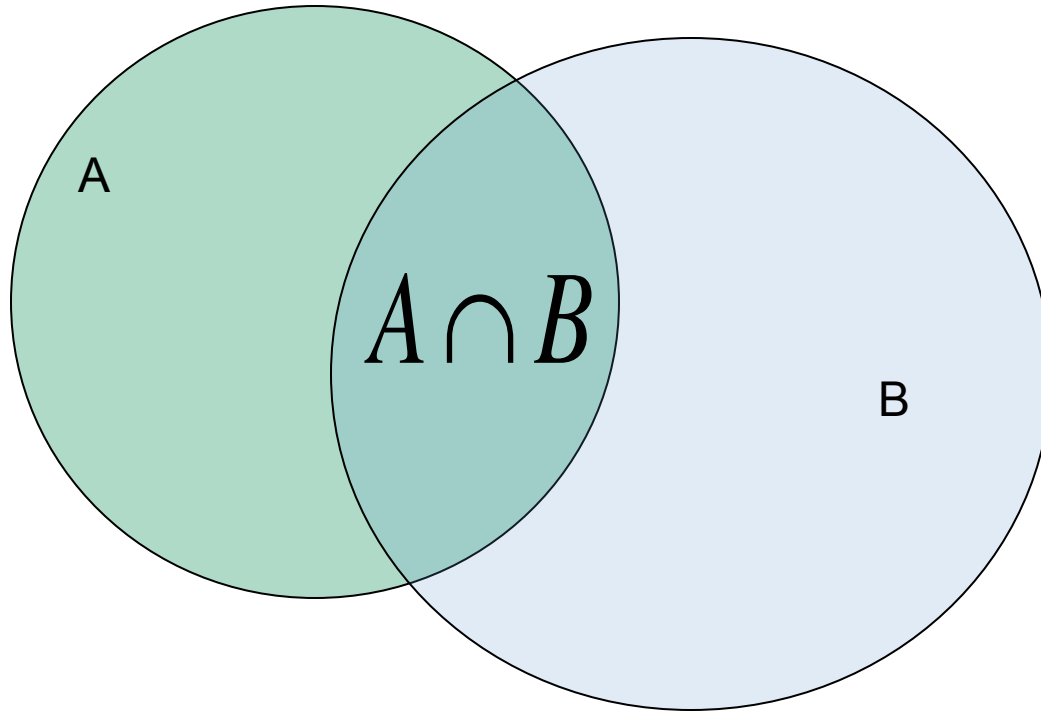
Michael Shlesinger  
mike.shlesinger@navy.mil

Unsolved Problems of Noise  
Barcelona July 13-17, 2015

Reverend Bayes (1701-1761)

Bayes' Theorem(1763)

$$p(A \cap B) = p(A|B)p(B) = p(B|A)p(A)$$



# The Sceptical Bayesian's Unfair Coin: STATISTICS VS. PROBABILITY

(when to stop searching)

**heads = found target**

Flip a coin n times and get n TAILS

What can you infer about the fairness of the coin?

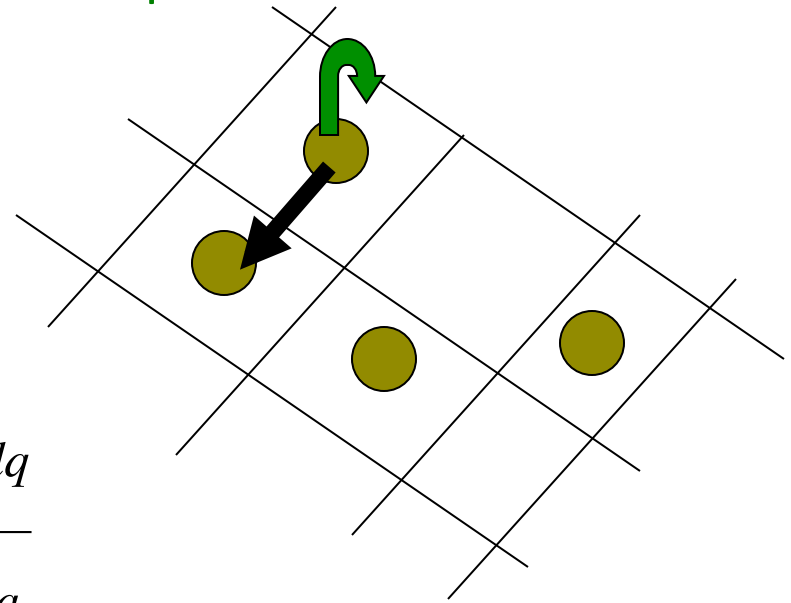
Use data to estimate q

$$p(nT) = \left(\frac{1}{2}\right)^n \text{ for a fair coin}$$

$$p(q|nT) = \frac{p(nT|q)p(q)}{\int_0^1 p(nT|q)p(q)dq}$$

$$\langle q \rangle = \frac{\int_0^1 qp(nT|q)p(q)dq}{\int_0^1 p(nT|q)p(q)dq} = \frac{\int_0^1 qq^n p(q)dq}{\int_0^1 q^n p(q)dq}$$

$$\langle q \rangle = \frac{n+1}{n+2}$$



$$\langle q(n=0) \rangle = 1/2$$

$$\langle q(n \rightarrow \infty) \rangle \rightarrow 1$$

## When to switch coins: Stick with same coin or switch coins?

<i>outcome</i>	<i>estimate</i> <T>	<i>estimate</i> <H>	<i>switch if</i>	<i>same as</i>
no coin toss	1/2	1/2		
T	2/3	1/3	$1/2 - \text{cost} > 1/3$	$\text{cost} < 1/6$
TT	3/4	1/4	$1/2 - \text{cost} > 1/4$	$\text{cost} < 1/4$
TTT	4/5	1/5	$1/2 - \text{cost} > 1/5$	$\text{cost} < 3/10$
TTTT	5/6	1/6	$1/2 - \text{cost} > 1/6$	$\text{cost} < 1/3$

Flipping a coin and getting a HEAD, then win one coin.

$w = \text{prob}(\text{next space has a coin})$

$N = \text{number of TAILS in a row}$

$C = \text{cost to switch coins}$

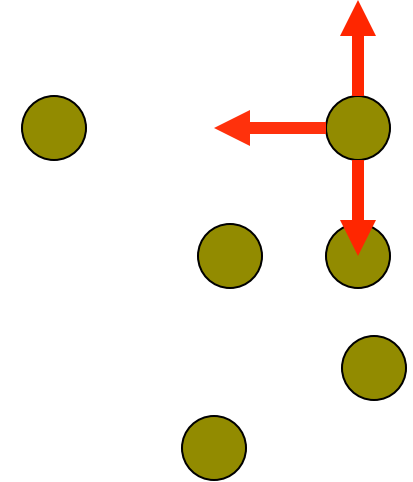
One flip T case estimated gain by switching  $= (1/2)w - C$

Switch coins if  $w/2 - C > 1/3$

$N$  tails TTT...T switch if  $(1/2)w - C > 1/(N+2)$

switch if  $\text{cost} < (w/2) - (1/(N+2))$

What if  $w = w(t) = \exp(-bt)$ ?



# Finite Time T Searching Two Regions

$\lambda$ =rate of success to find target in a search if a target is present

$K$ = rate of success to retrieve/capture the target

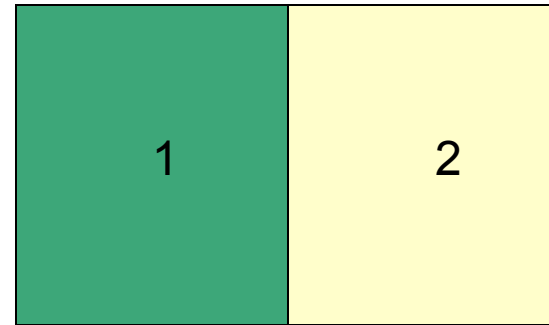
$G$ = gain (payoff of search)

$C$ =cost of search/unit time

$p$ =prob that a target is present

$N$ = net return

$\leftarrow (T - \tau) \rightarrow \quad \leftarrow \tau \rightarrow$



$$N(T, \tau) = G_1 K_1 (T - \tau) p_1 (1 - \exp(-\lambda_1 (T - \tau))) + G_2 K_2 (\tau) p_2 (1 - \exp(-\lambda_2 \tau)) - C_1 (T - \tau) - C_2 \tau$$

Choose  $\tau$  to maximize  $N$

$$\frac{\partial N(T, \tau)}{\partial \tau} = 0$$

## Rate Limited or Reaction Limited



Getting to the right location does not mean the target is found

**For Brownian motion with diffusion constant  $D$  and average target spacing  $d$**

$$\lambda \propto \frac{D}{\langle d \rangle^2}$$

**For ballistic motion with velocity  $v$  and sighting range  $R$**

$$\lambda \propto \frac{vR}{\langle d \rangle^2} \quad (2D)$$

$$\text{cost} \propto mv^2 t$$

$$\lambda \propto \frac{vR^2}{\langle d \rangle^3} \quad (3D)$$

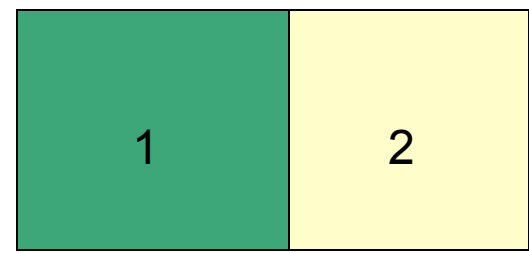
$$\text{gain } G = \begin{cases} \text{constant} \\ 1/\lambda \\ f(\lambda) \end{cases}$$

$$d \rightarrow d^{1/\alpha}$$

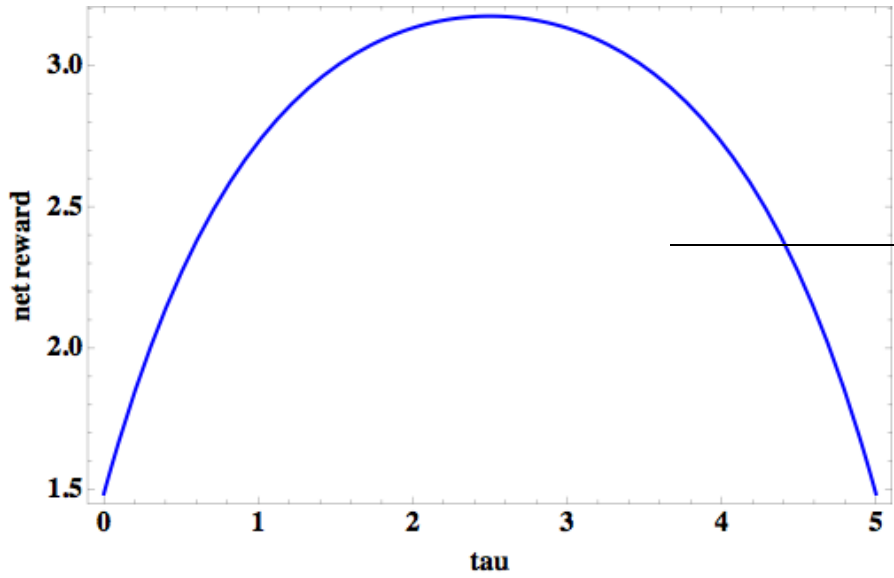
**For fractal set of targets with dimension  $\alpha$**

In the figure on the left both regions are the same;

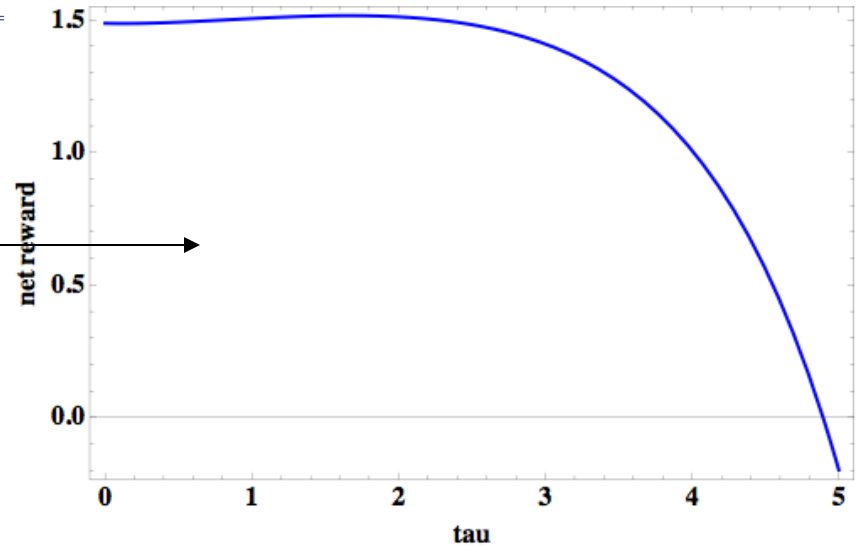
$$\leftarrow (T - \tau) \rightarrow \quad \leftarrow \tau \rightarrow$$



Out[64]=



Out[52]=

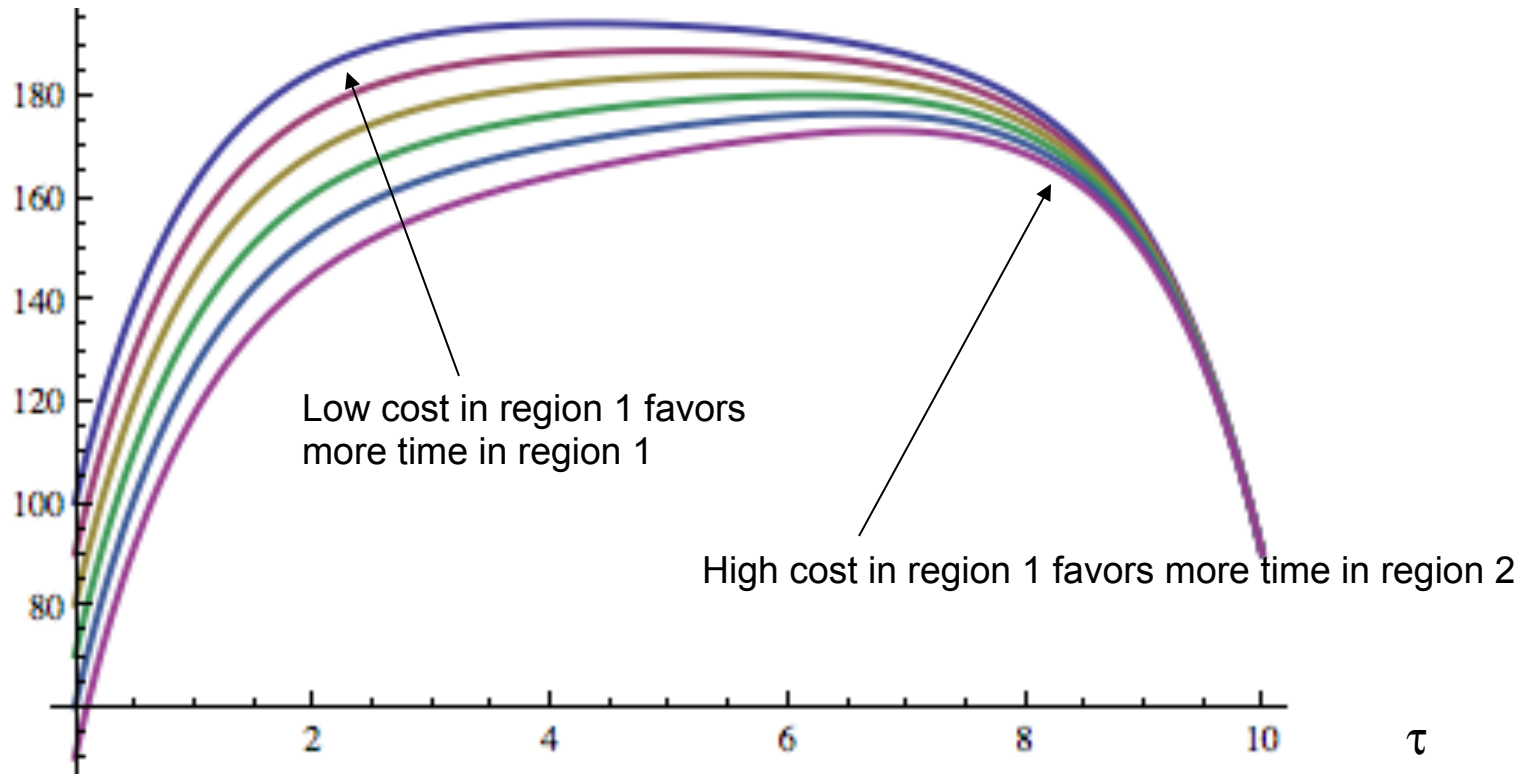


in the figure on the right a time penalty was imposed to move from region 1 to region 2 so the max net return shifted to spending more time in region 1



# Higher cost in region 1 shows better to spend more time in region 2

Payoff versus time in region 2



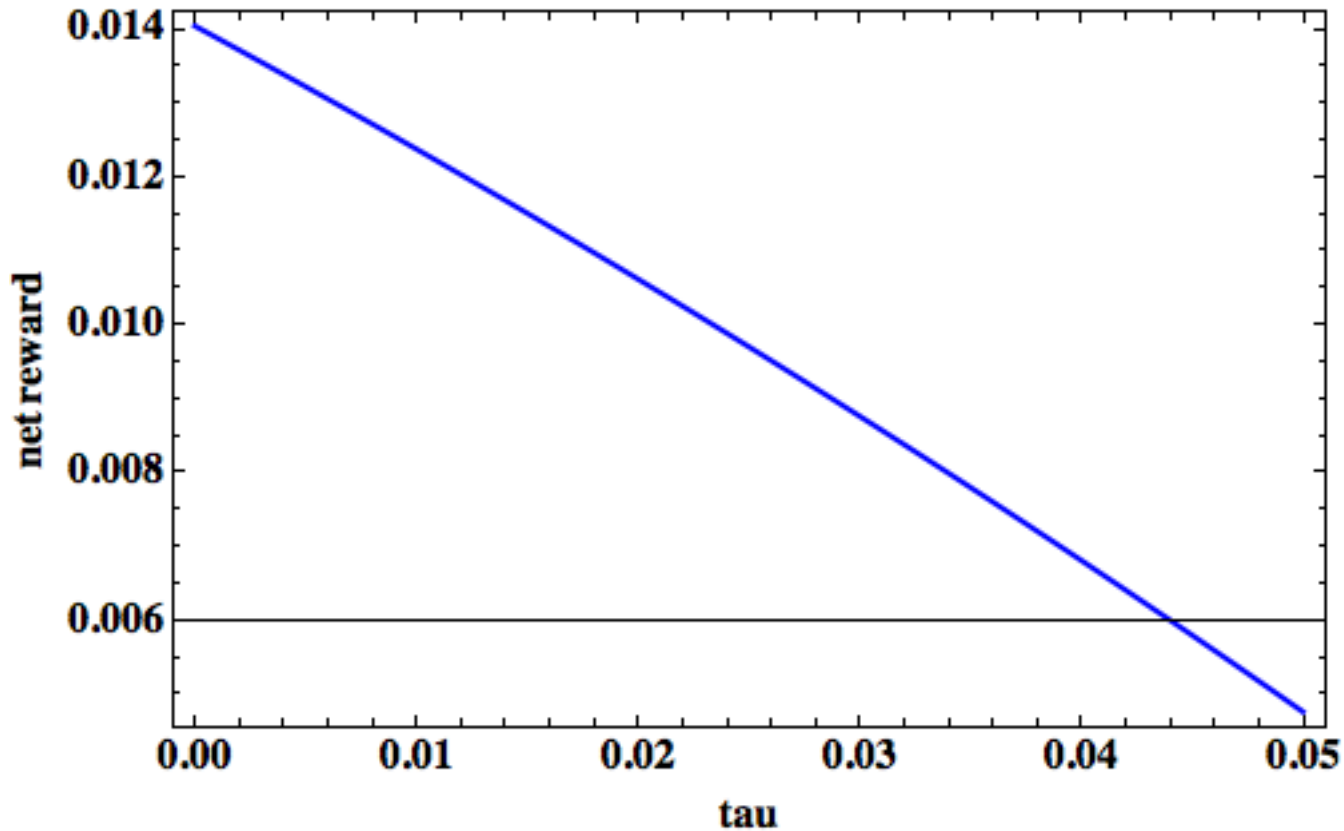
Cost in region 1 highest in bottom curve lowest in top curve



Hunter chooses  
between going to the  
plains (gain  $G_1$ ) or to  
the river (gain  $G_2$ )



Search time  $T=0.05$   
success rate in region 1 = 2.0  
success rate in region 2 = 1.0

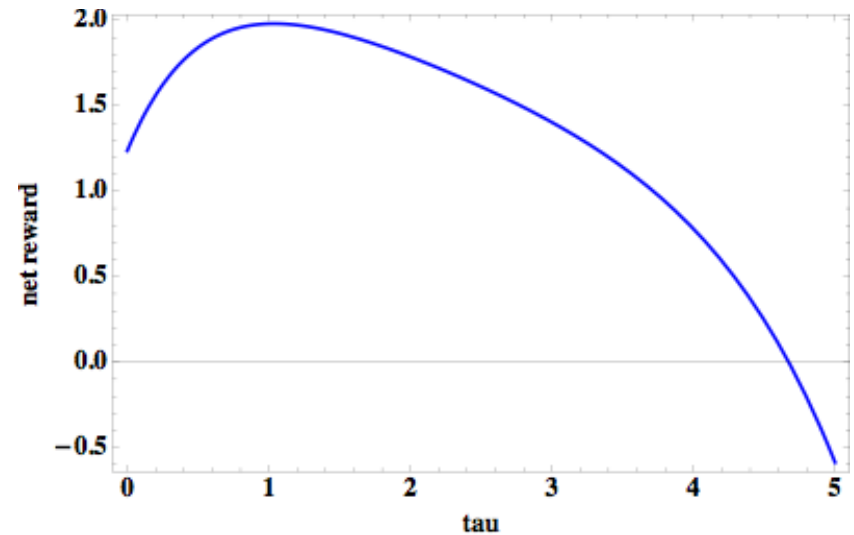
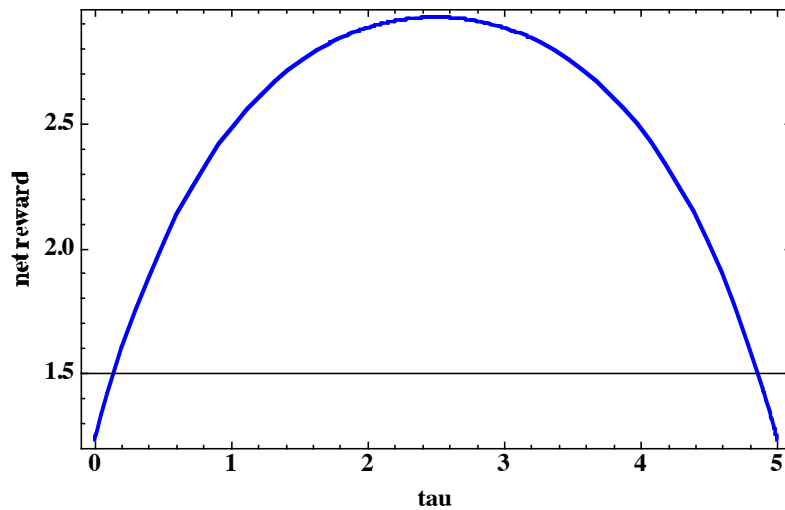


Better to stay in region 1 as search time is very short

## Generalizations: scare away the target in region 2

(the longer the search the more likely the target has left)

$$N(T, \tau) = G_1 p_1 (1 - \exp(-\lambda_1 (T - \tau))) + G_2 p_2 (\tau) (1 - \exp(-\lambda_2 \tau)) - C_1 (T - \tau) - C_2 \tau$$

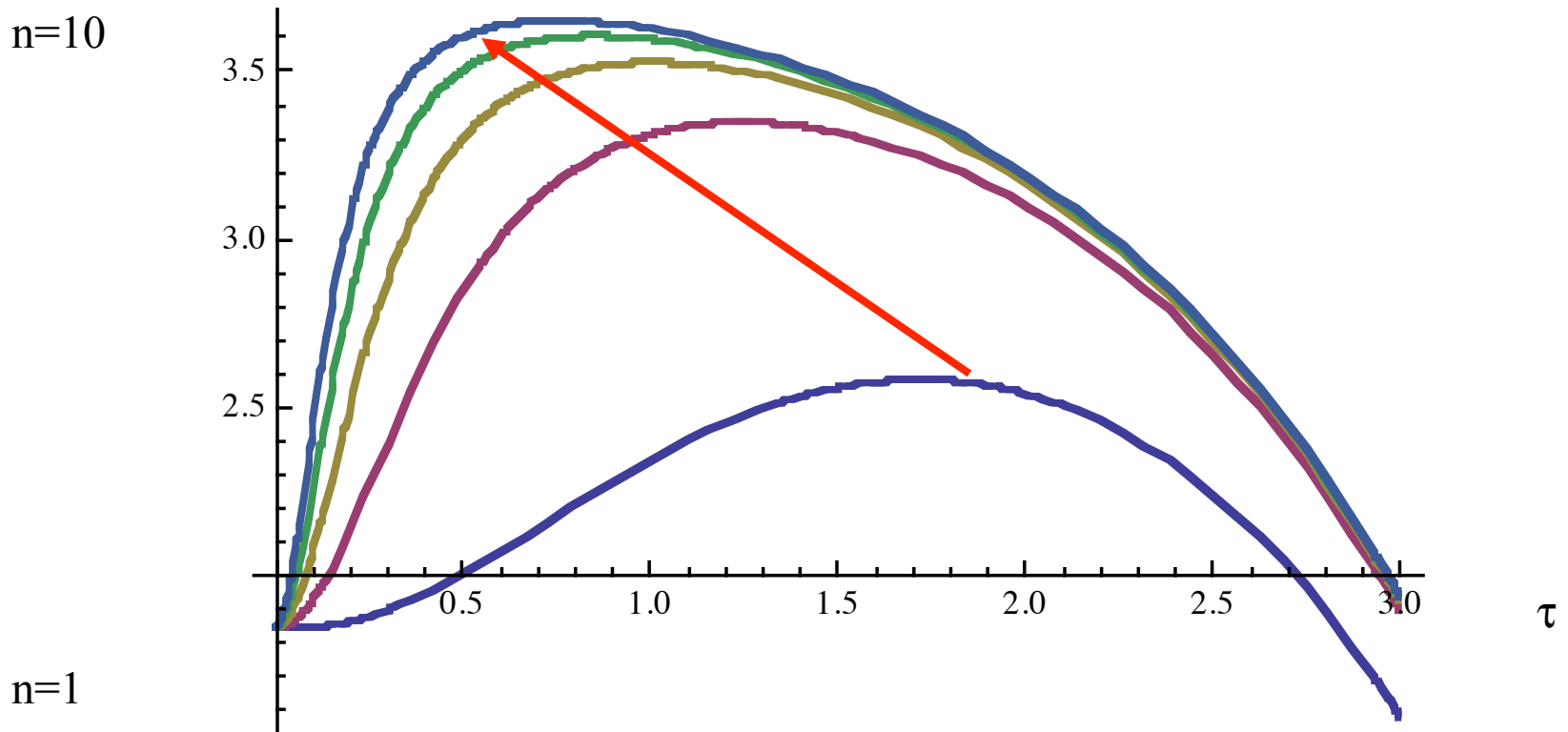


in the figure on the right, the probability  $p$  that the target is in region 2 decreases exponentially with time.

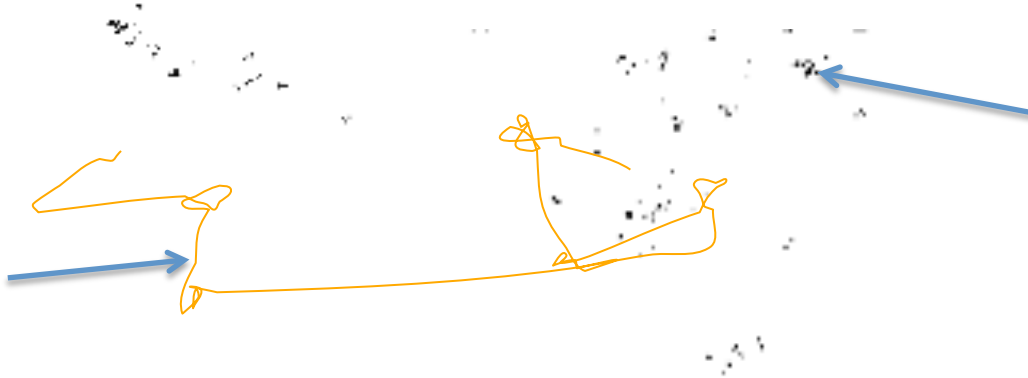
**Small tau means search maximized by spending shorter time in region 2.**

Slower than exponential search success rate in region 2

$$\lambda \exp(-\lambda t) \Rightarrow \frac{\tau_0}{\tau^2_0 + t^2 / n}$$



# Levy Flights vs Drives



$$\Psi(\ell, t) = \psi(t|\ell) p(\ell)$$

$$\psi(t|\ell) = \delta\left(t - \frac{\ell}{V(\ell)}\right)$$



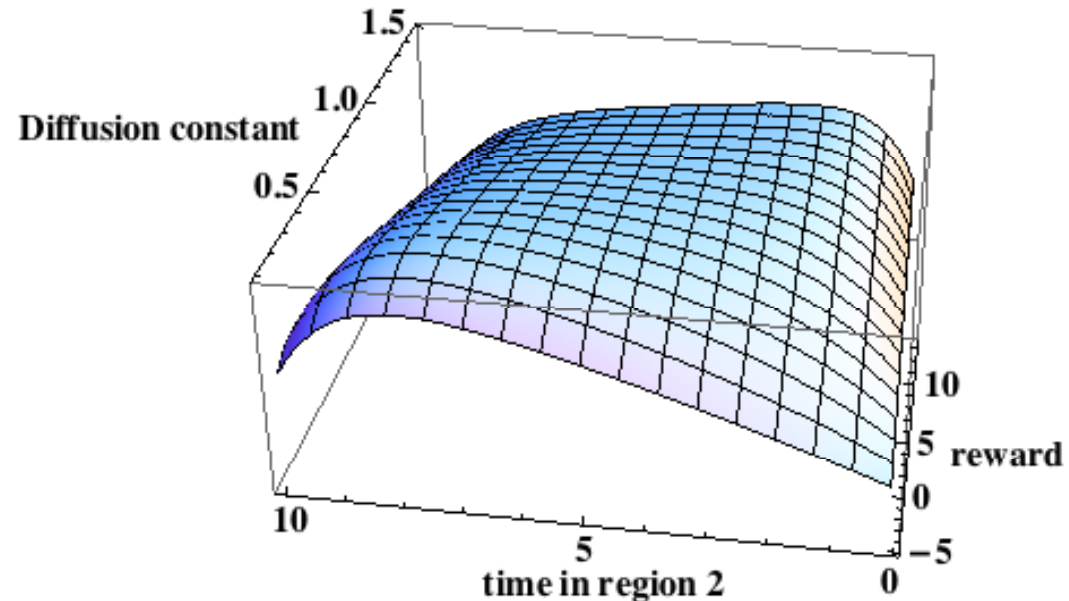
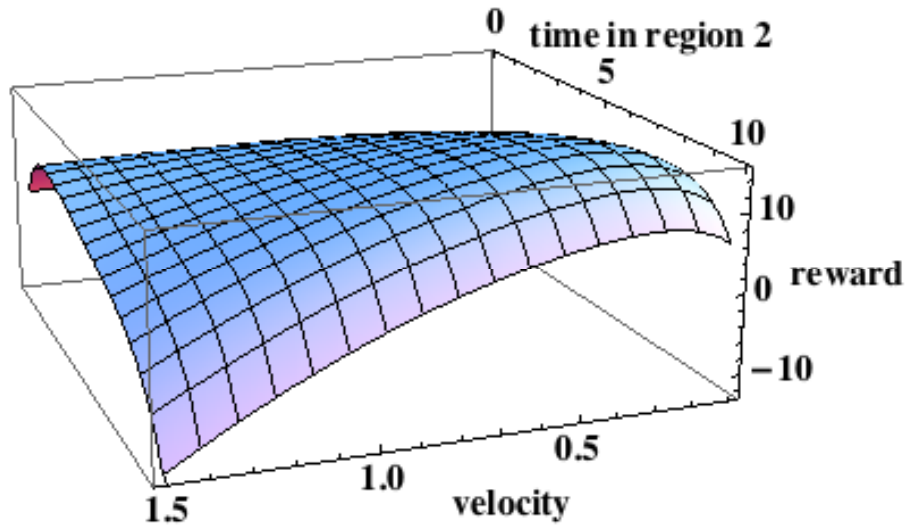
diffusion

What if target sites look like this, a fractal set?



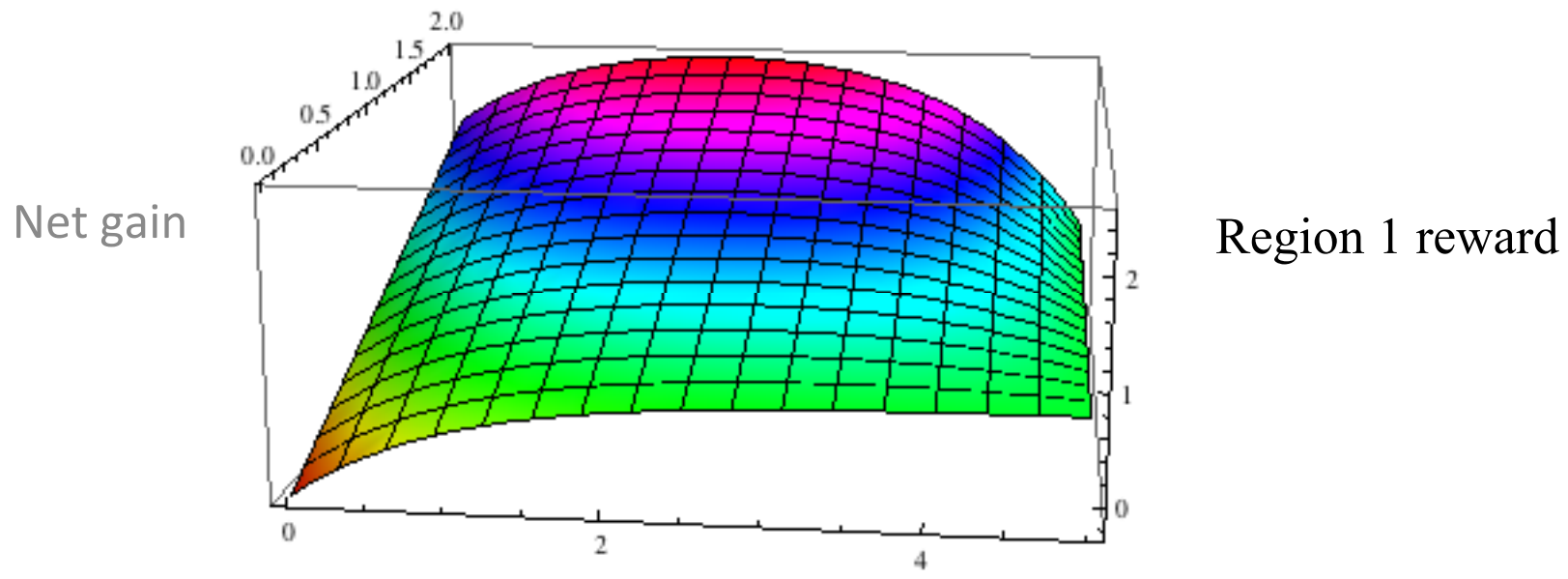
Sometimes a long flight pays off: C. Columbus (1492)

# Success rate $\lambda$ depending on velocity $v$ or diffusion constant $D$





Gain as a function of time in region 2 and reward value in region 1

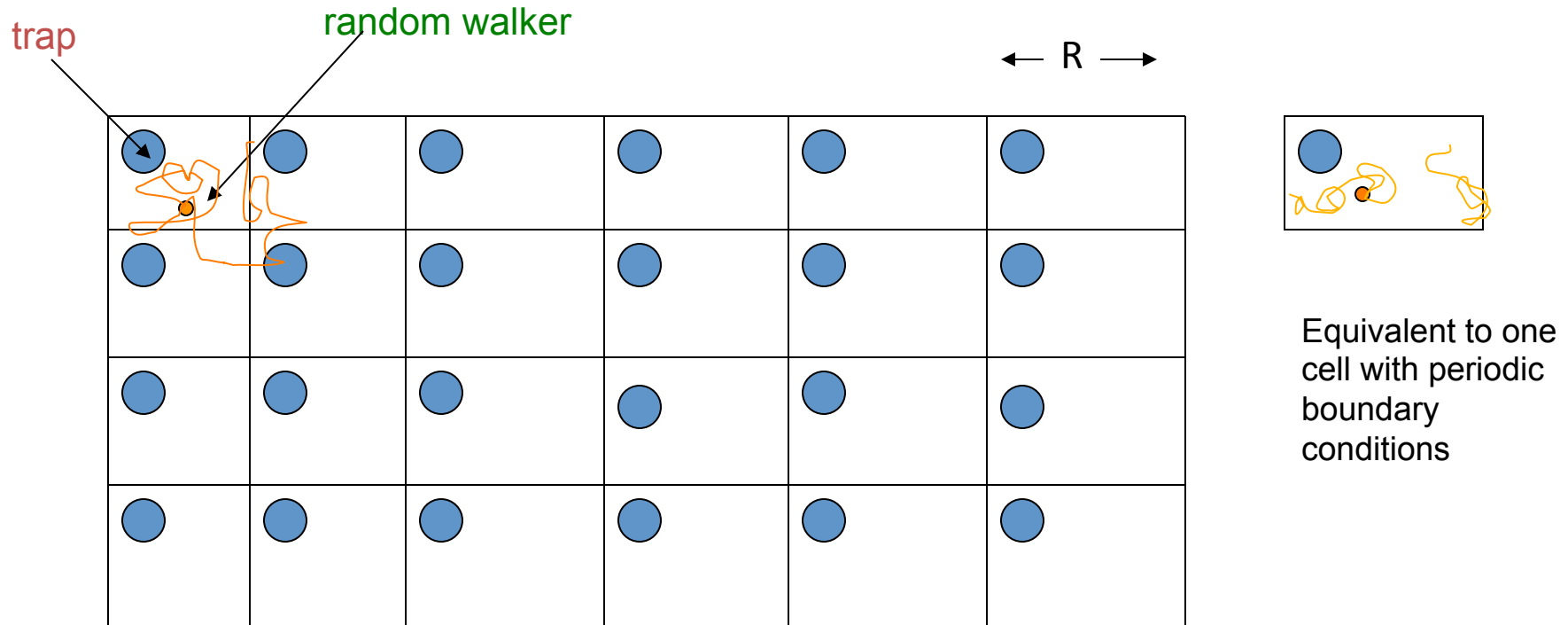


# Brownian Random Walker Lifetime Probability

## *the trapping problem*

$$\Phi(t) \propto \exp\left(-\frac{4Dt}{R^2}\right)$$

B. Ya Balagorov and V. G. Vaks, Sov. Phys. JETP **38** 968 (1974)



Mean number of contacts  
in time  $t$

$$\langle n(t) \rangle \propto \frac{4D}{R^2} t$$

## Donsker-Varadhan Brownian trapping problem

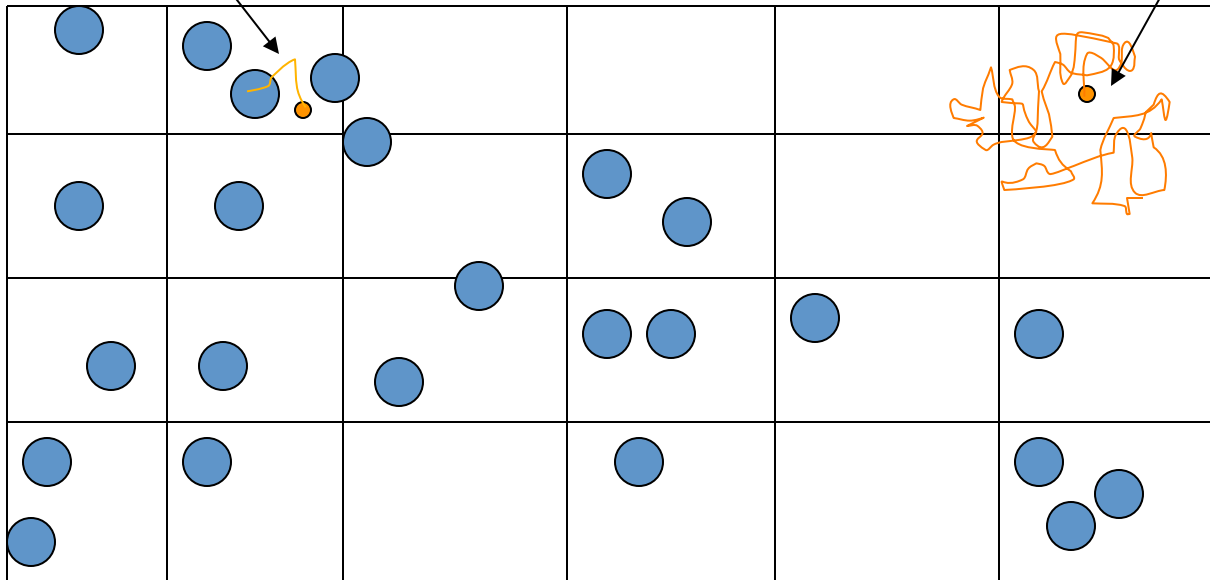
$$\lim_{t \rightarrow \infty} \Phi(t) \propto \lim_{t \rightarrow \infty} \int^R \exp\left(-\frac{Dt}{R^2}\right) \frac{1}{V_0} \exp\left(-\frac{V}{V_0}\right) dR$$

$$\propto \exp\left(-t^{\frac{d}{d+2}}\right), \quad t \gg 0$$

$$V \propto R^d$$

short lifetime

long lifetime

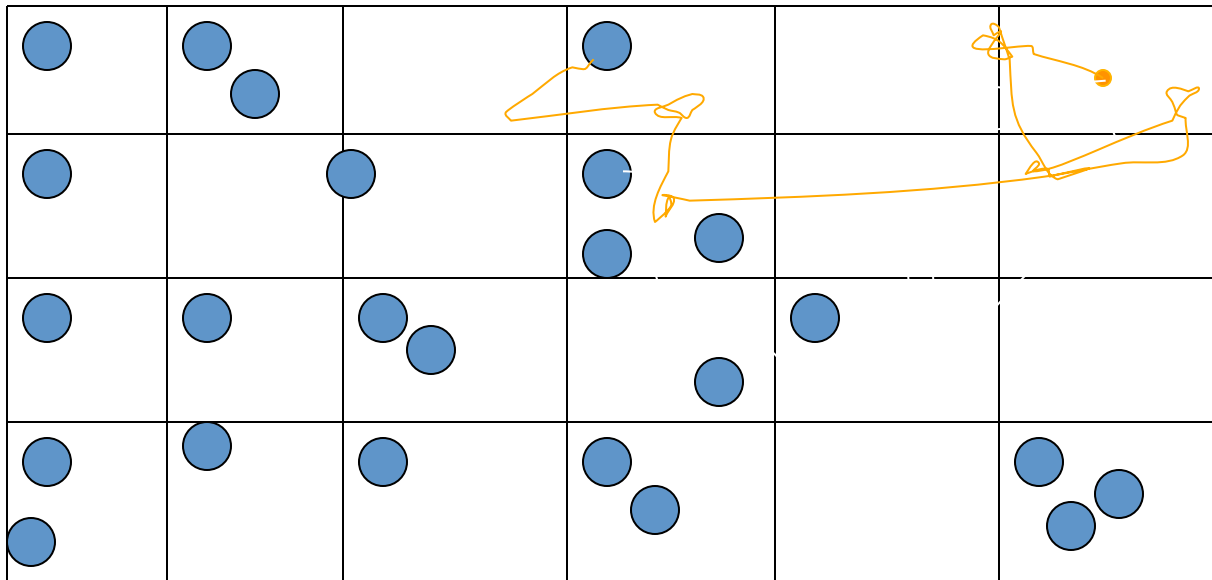


Poisson distribution of trap sites

## Donsker-Varadhan Levy trapping problem

$$\lim_{t \rightarrow \infty} \Phi(t) \propto \lim_{t \rightarrow \infty} \int^R \exp\left(-\frac{Dt}{R}\right) \frac{1}{V_0} \exp\left(-\frac{V}{V_0}\right) dR \propto \exp\left(-t^{\frac{d}{d+1}}\right)$$

shallower tail than  $\exp\left(-t^{\frac{d}{d+2}}\right)$



When should Brownian search give up and take a Levy flight?

## Estimation of success $p^*$ decreases in time with continued failure

$$\begin{aligned} p^* &= P(T \text{ present} | \text{search fails}, T \text{ not found}) \\ &= \frac{P(T \text{ not found} | T \text{ present})P(T \text{ present})}{P(T \text{ not found} | T \text{ present})P(T \text{ present}) + P(T \text{ not found} | T \text{ not present})P(T \text{ not present})} \\ &= \frac{(1-f)p}{(1-f)p + 1^*(1-p)} \\ &= \frac{p - fp}{1 - fp} < p \end{aligned}$$

$$p^*(t) = \frac{p\phi(t)}{p\phi(t) + (1-p)} = \frac{pe^{-\lambda t}}{pe^{-\lambda t} + (1-p)} < p$$