



# From Ants to Query Complexity

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Based on joint works with

Biologists: Ofer Feinerman, Ehud (Udi) Fonio, Yael Heyman, Aviram Gelblum  
CS researchers: Lucas Boczkowski, Adrian Kosowski, Yoav Rodeh

# Collective transport by Crazy ants



*P. longicornis* ants

# Method:

## Scent mark detection





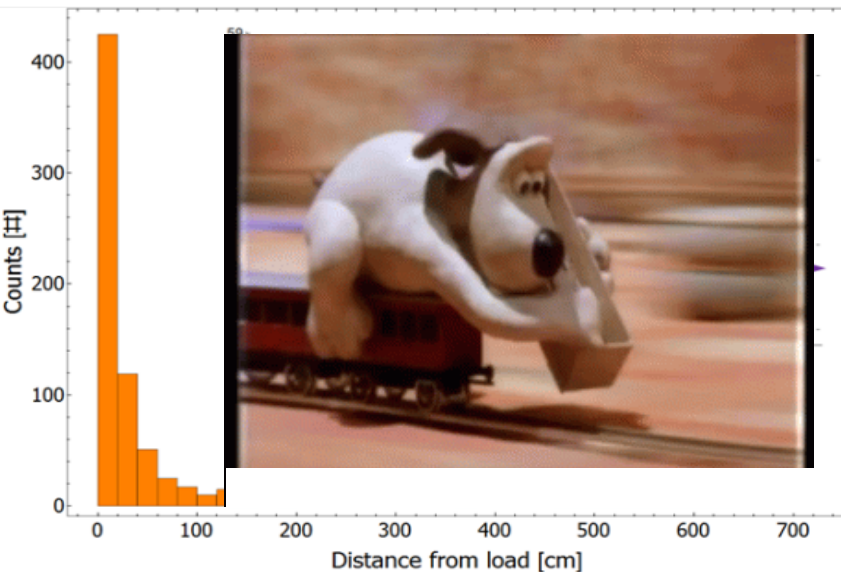
# A new type of ant trail

The known trail -  
Very long and steady

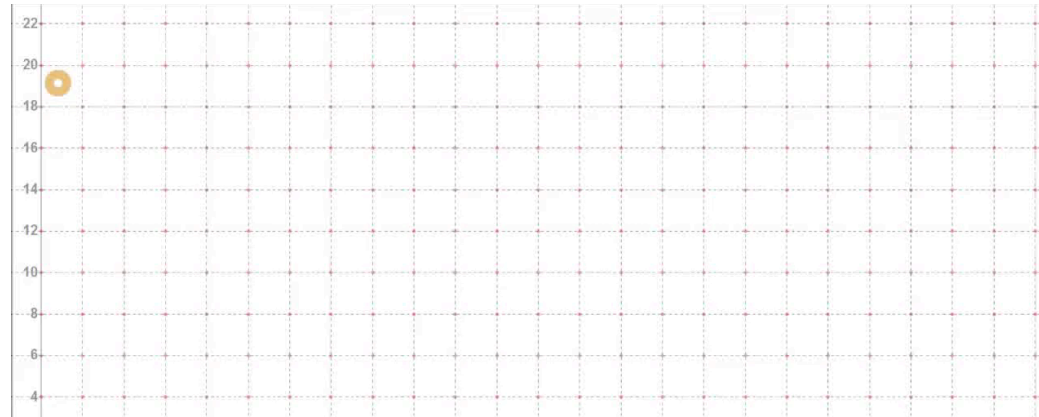


↑  
100 meters  
↓

## 1. Individuals lay local trails



## 2. Group follows them, but not religiously



What is it good for?



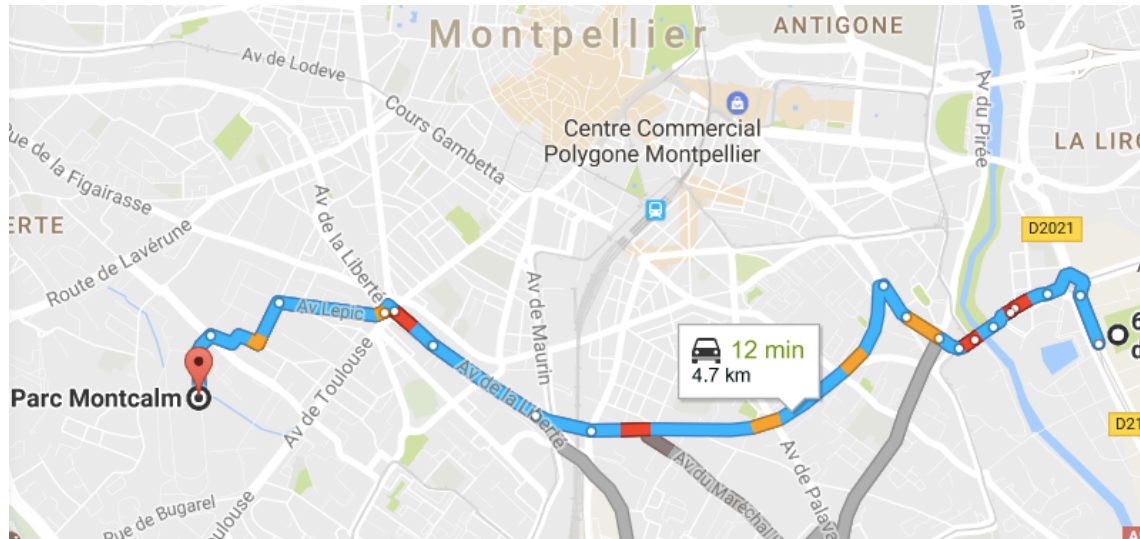


To nest

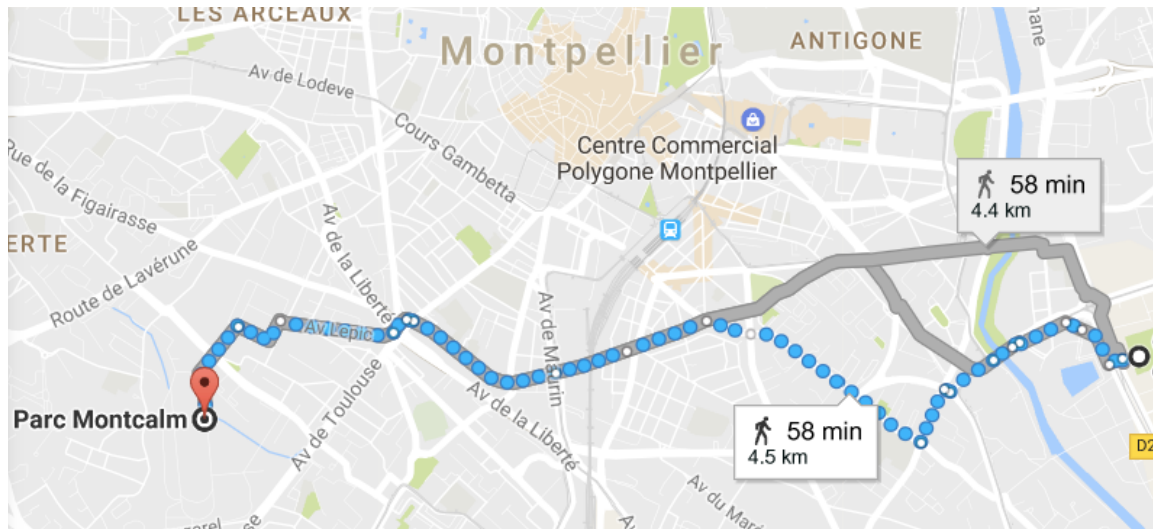




# You need to drive a car



# You get the map for walking



# Navigating with unreliable roadsigns

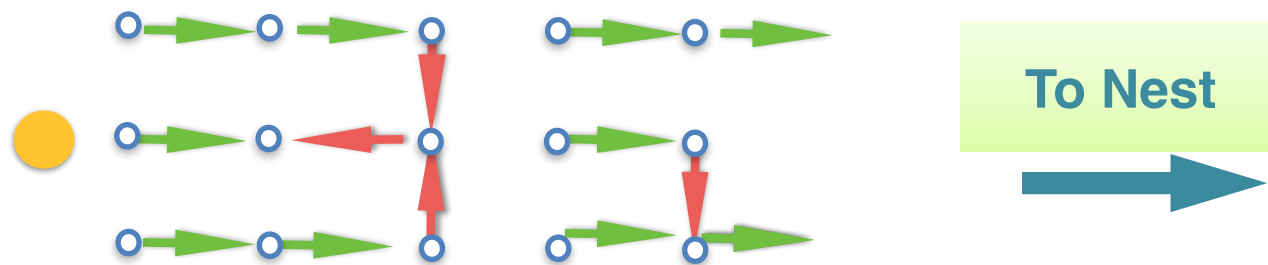
Imagine driving in a foreign country after a hurricane  
Most road signs are intact but some have been turned

How can you still get to your destination fast?



# The Noisy Advice model

Each node has an advice pointer, which is **correct** → with probability  $p$ .  
Otherwise, w.p.  $q=1-p$  it points at an arbitrary direction ←



How can we reach the nest fast?

Advice per node is fixed!

Very different from advice that can be resampled at any query (allows for busting)

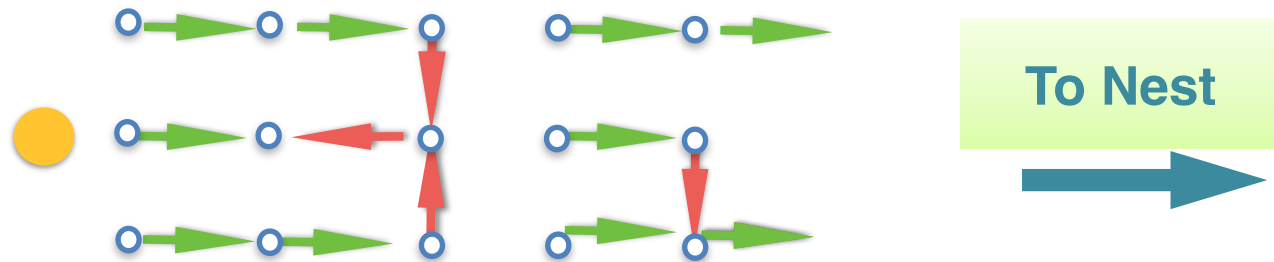
[Feige et al. SICOMP'94] [Ben-Or and Hassidim FOCS'08]

[Emamjomeh-Zadeh et al. STOC'16] [Karp and Kleinberg SODA'07]

Very different from worst case placement of wrong advice

[Hanusse et al. PODC'10, TCS'08]

# Random Listening (RL)

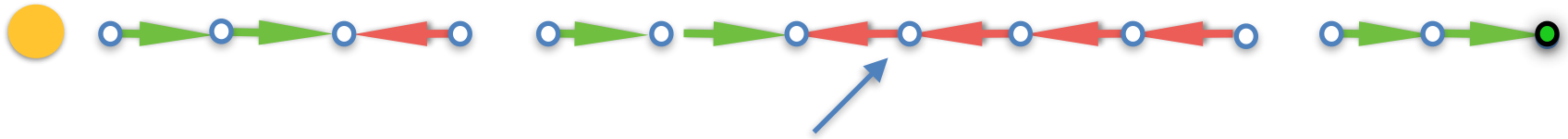


Random Listening:

- follow advice pointers with a fixed probability  $\lambda$  , otherwise:
- do a random walk

“Random Walks in Random Environments (RWRE)”

# Example: Line graph

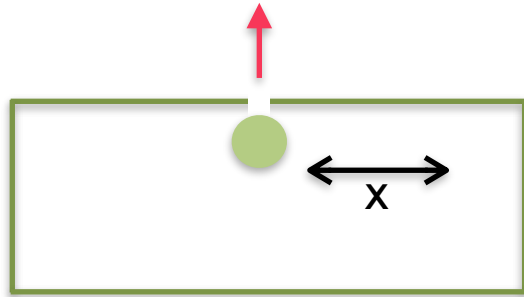


Time to pass a trap is exponential in its size  
But traps are exponentially rare

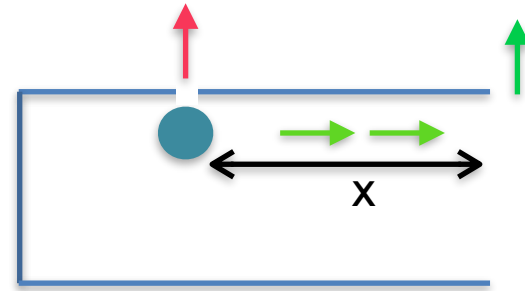
**Theorem:** If  $p > 0$  then any listening probability  $0 < \lambda < 1 - p$  allows for RL to reach distance  $d$  in  $O(d)$  time



# Predictions

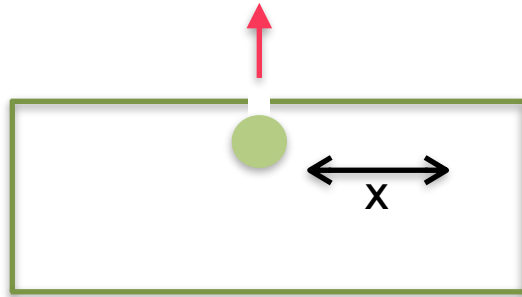


all advice is wrong  $\Rightarrow$   
time to pass obstacle of size  $x$  is  $\exp(x)$

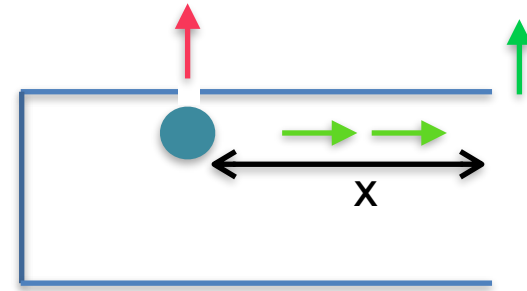


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# Predictions

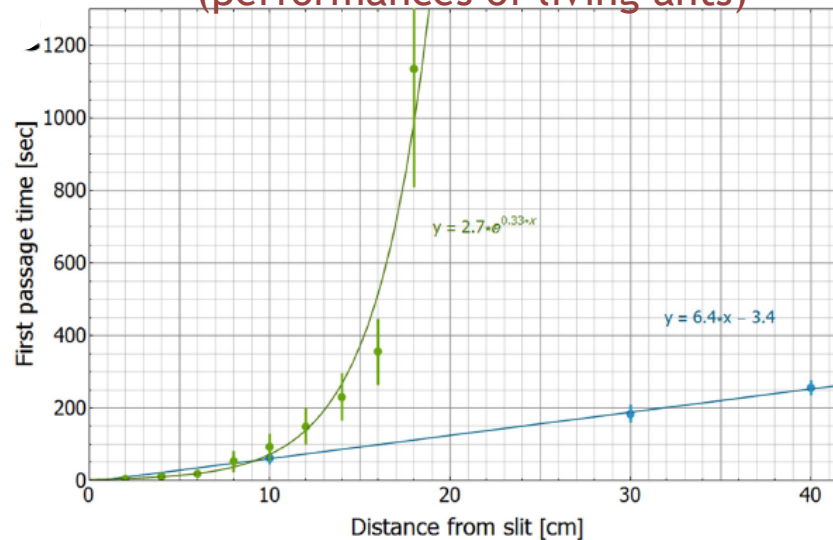


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## Experimental validations (performances of living ants)

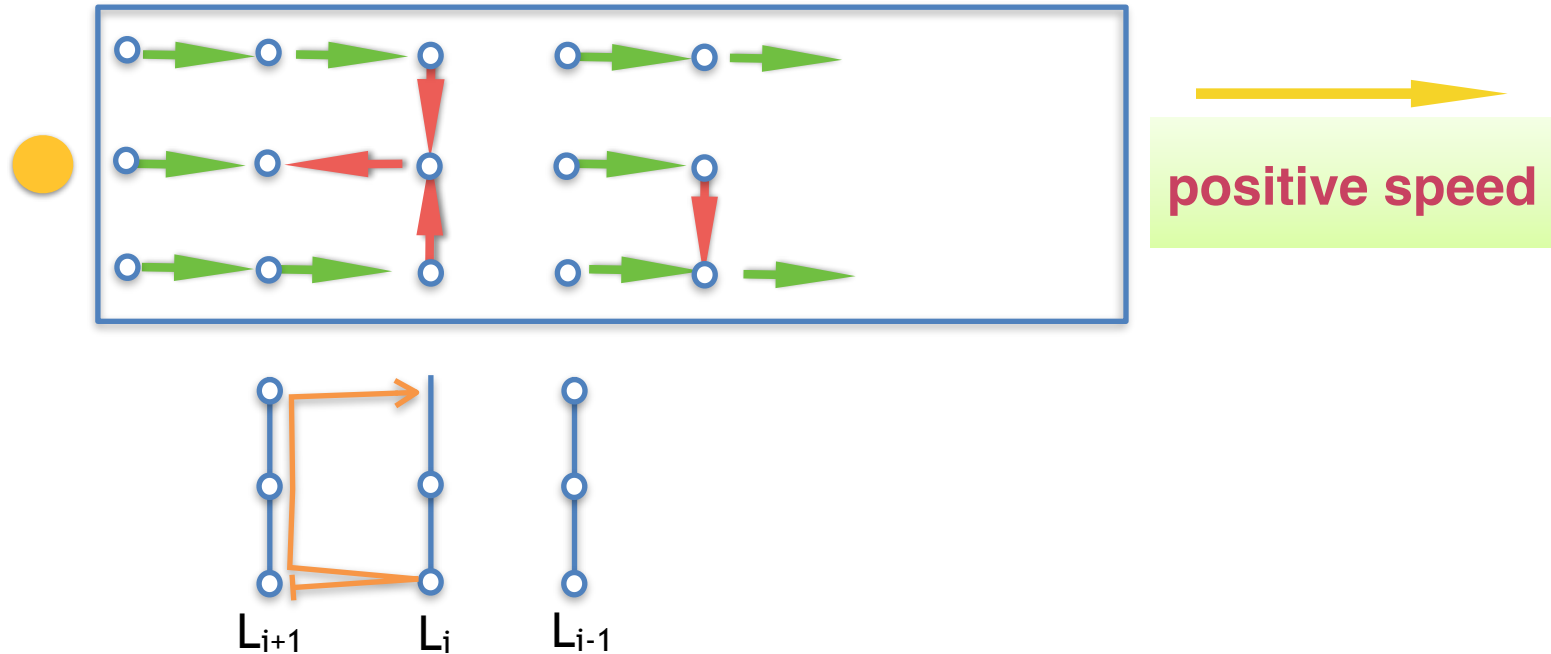


מכון ויצמן למדע  
WEIZMANN INSTITUTE OF SCIENCE

E. Fonio  
O. Feinerman

Consistent with step length of 10cm and probability of listening  $\lambda = 0.8$

# RL in grids



Situation is “in between” a line and a refreshed advice

Adapting results from RWRE [Snitzman, 2002], we prove:

## Theorem:

In grids and line graph if  $q$  is small enough then there exists a range of listening probabilities to allows for positive speed



## Conjecture:

There exists a constant  $c$  and a listening probability, s.t.  
for any graph, RL achieves linear hitting time if  
mistake  $q < c / \Delta$  at every node.



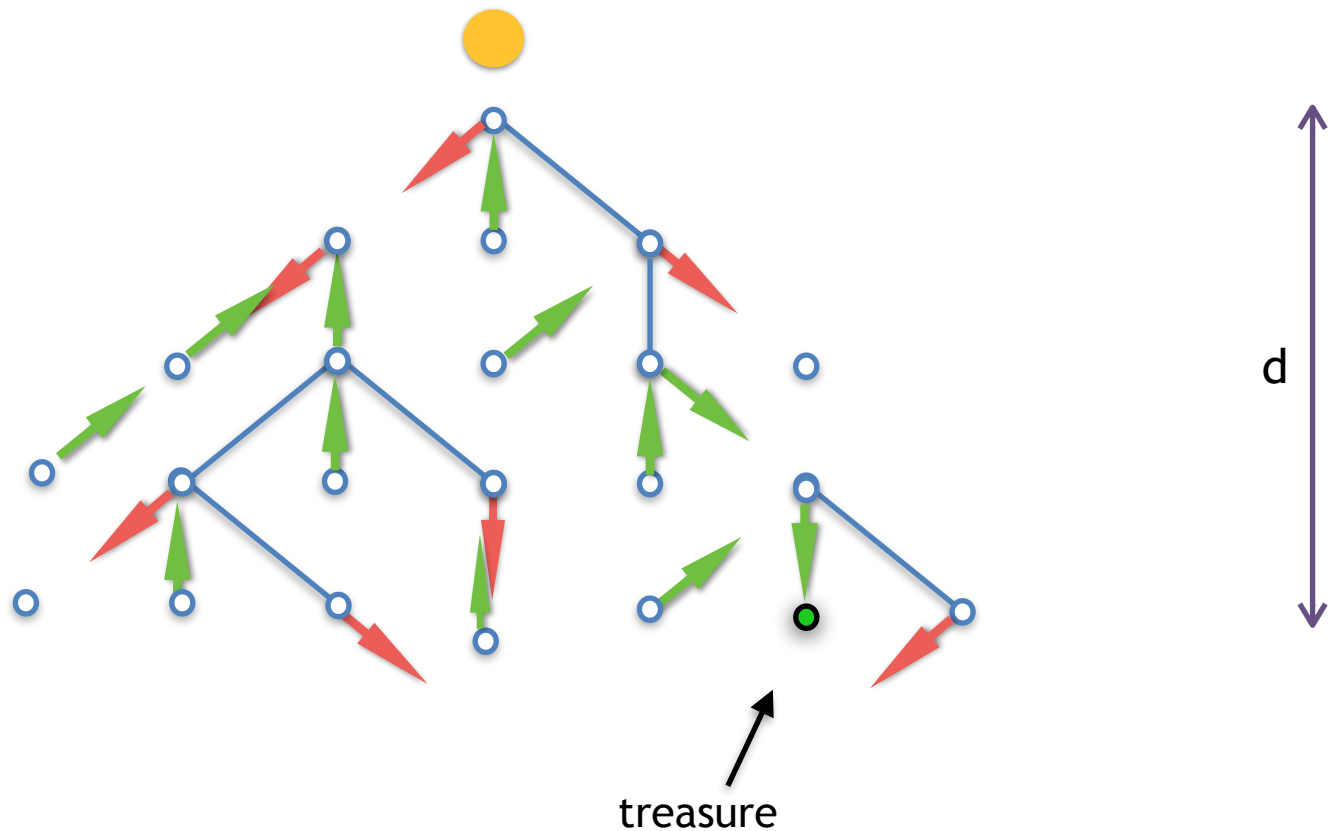
*eLife* 2016

“Le Monde” Jan. 2017

# Navigating on Noisy Trees

L. Boczkowski, A. Korman, Y. Rodeh

To be submitted





# Conjecture holds for Trees

$1 / \Delta$  is a threshold for the noise  $q$  in order for random listening strategies to be efficient

## Theorem

- There exists a constant  $c$  and a listening probability, s.t. for any tree, RL achieves linear hitting time  $O(d)$  if mistake  $q < c / \Delta$  at every node.
- Consider the complete  $\Delta$ -regular tree.  
There exists a constant  $c'$ , s.t. any RL achieves exponential hitting time (in  $d$ ) if mistake  $q > c' / \Delta$ .

# Conjecture holds for Trees

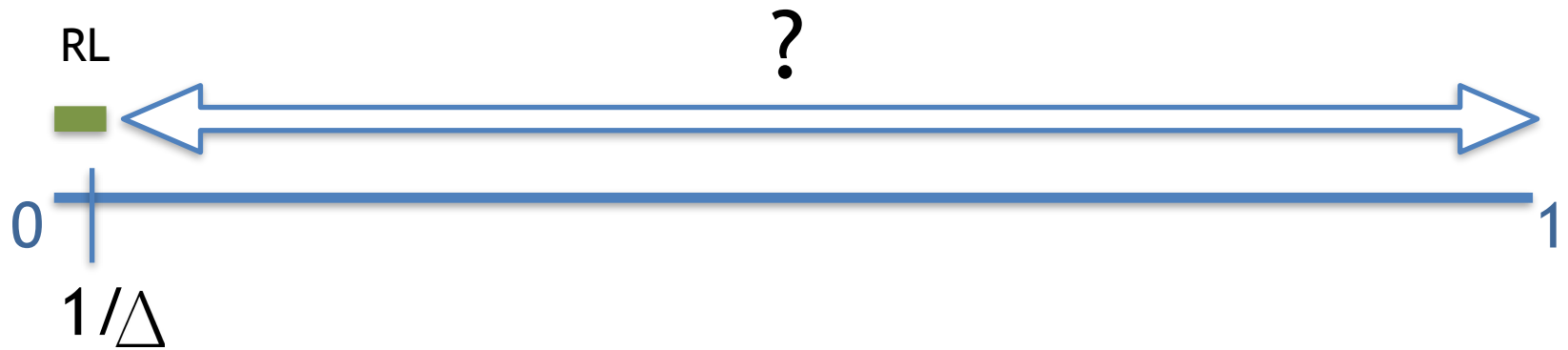
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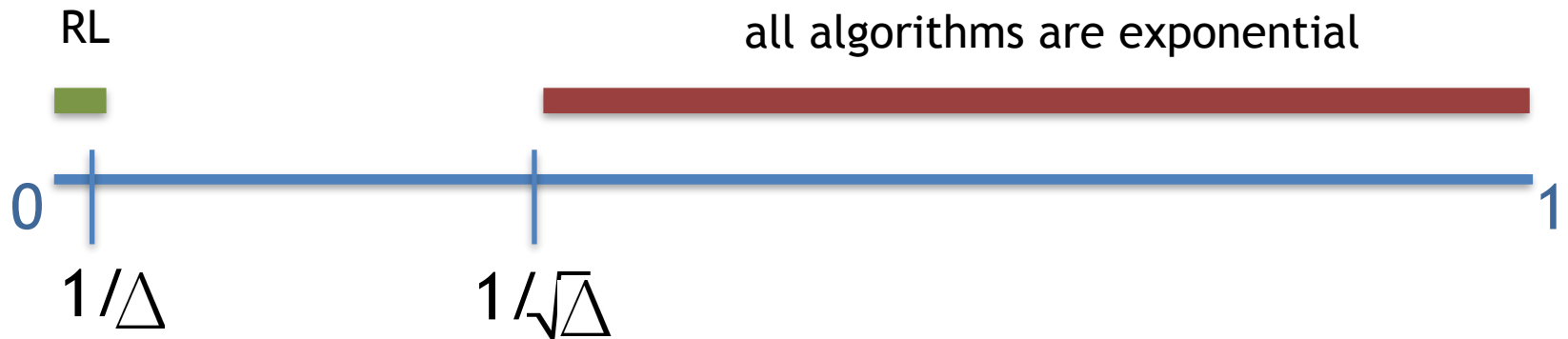
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**What about other algorithms?**

# Noise regimes



# $1/\sqrt{\Delta}$ is a lower bound on noise

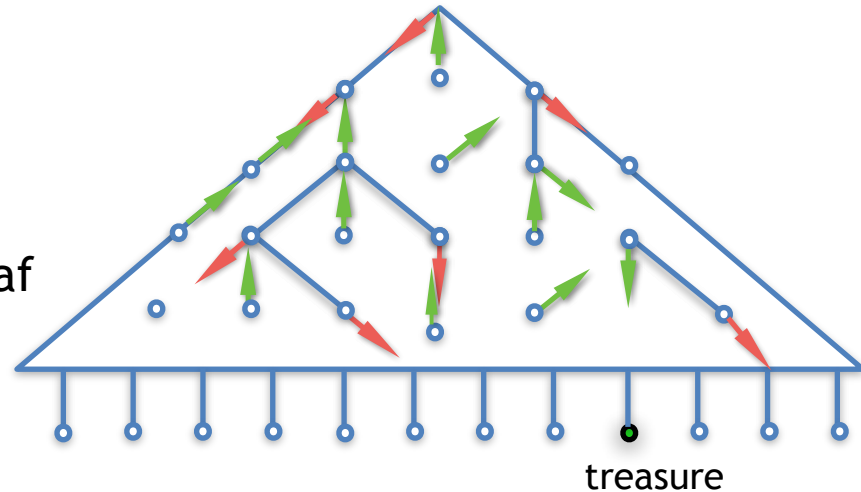


# Proof sketch

Consider the complete  $\Delta$ -ary tree

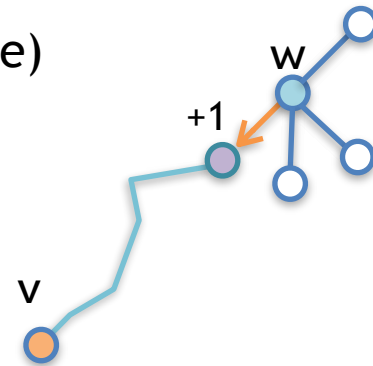
Assume:

- (1) full advice is given to alg, and
- (2) treasure is chosen u.a.r at a leaf



**Claim:**

**Best algorithm:** counts # of pointers (in the whole tree) pointing at each leaf and checks the leaves in order

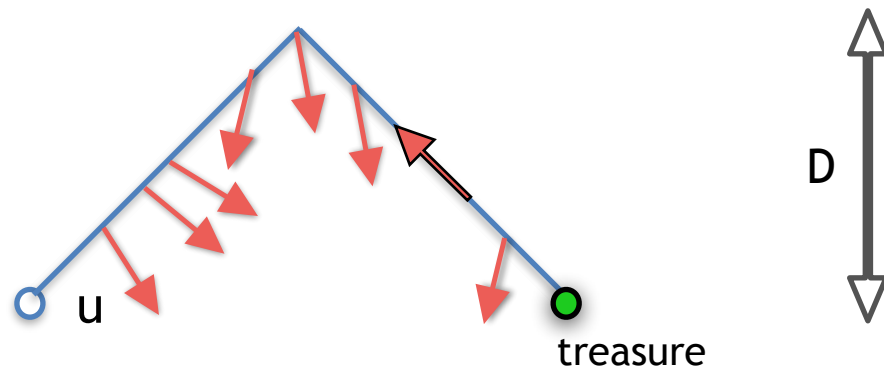


Given the claim, time is  $>$  the expected number of leaves that “beat” the treasure

# Expected number of competitors

Prob that a given leaf is “better looking” than  $t$  is small, but there are many leaves!

Prob that  $u$  is “better” than  $t$  is at least  $\frac{q}{\Delta} \cdot q^{d(u,\tau)} \approx q^{2D}$

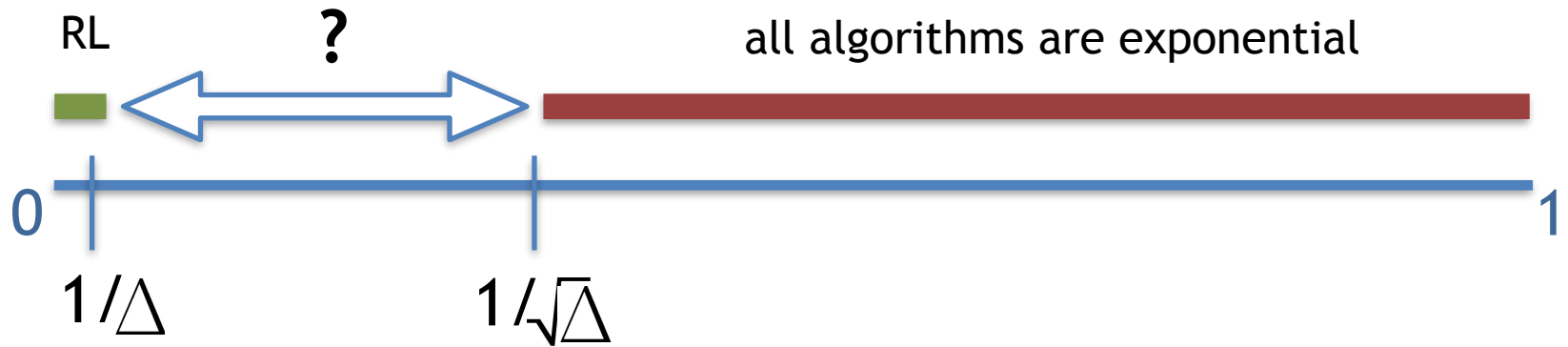


There are roughly  $\Delta^D$  leaves whose distance from  $t$  is  $2D$

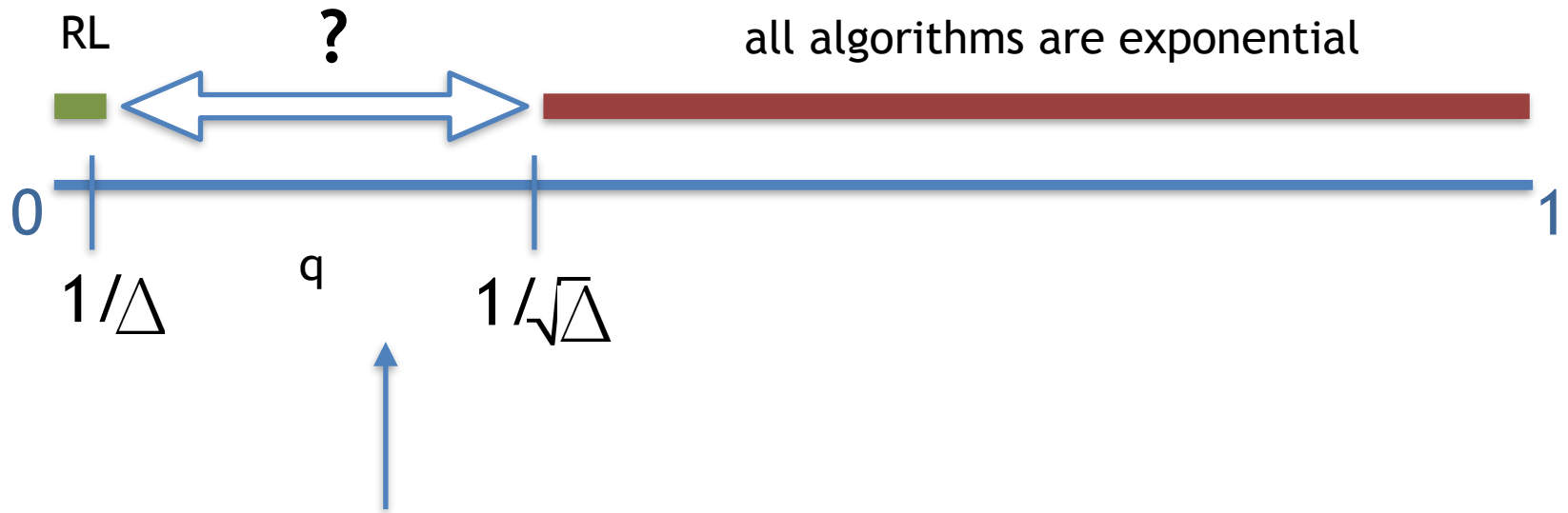
Therefore, the expected #leaves that beat  $t$  is at least:  $(q^2 \Delta)^D$



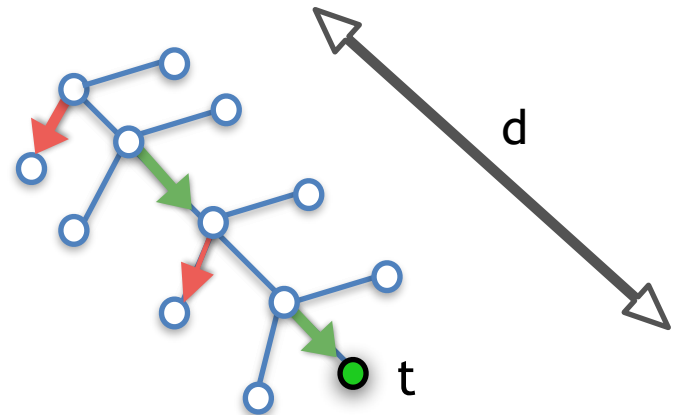
# Noise regimes



# Noise regimes



Time lower bound of  $\Omega(dq\Delta) = \Omega(d\sqrt{\Delta})$



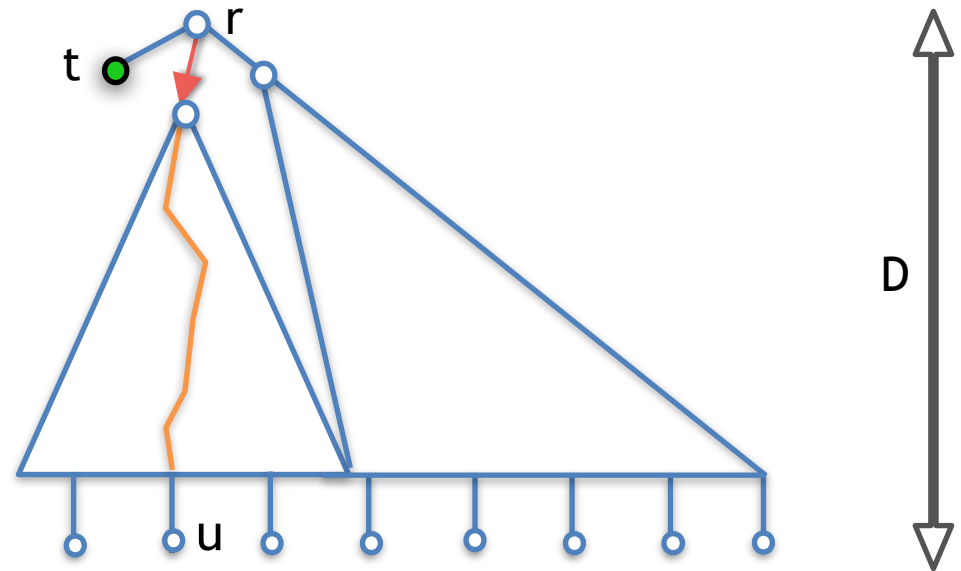
# A simple greedy strategy that fails

Walk to the currently most promising node -  
the one with most pointers pointing to it

# A simple greedy strategy that fails

Walk to the currently most promising node -  
the one with most pointers pointing to it

Consider the complete  $\Delta$ -ary tree  
with an extra child to  $r$



There are  $(\Delta - 1)^D$  leaves  $u$ . For any of them, prob that  $r$  points to  $u$ ,  
and nobody on the path to  $u$  points at  $r$  is  $> \frac{q}{\Delta} \cdot q^{D-1} \approx q^{D-1}$

So the expected number of nodes visited before the treasure is at least roughly:

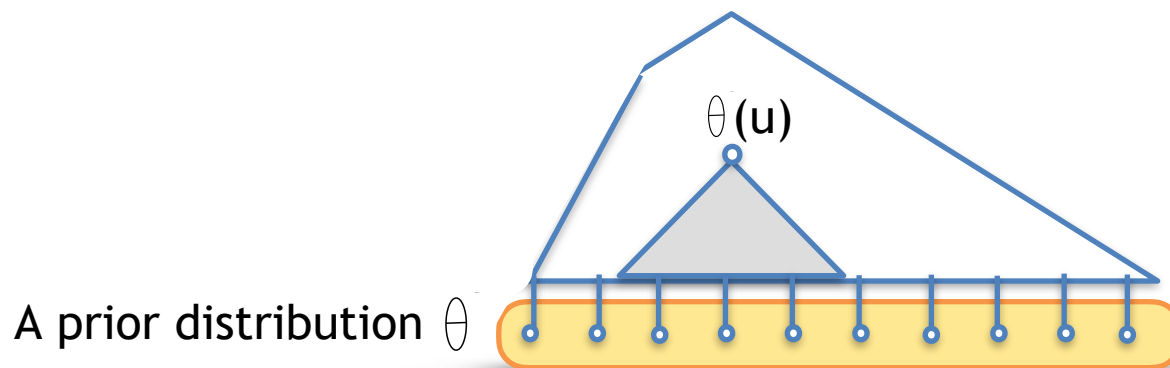
$$(q\Delta)^D$$

**Theorem:** if  $q < 1/\sqrt{\Delta}$  then there exists a walking algorithm that runs in optimal  $O(d\sqrt{\Delta})$  time

Intuition for the construction: Based on a Bayesian approach

Let us make our life easier, and assume:

1. tree structure is known to the algorithm
2. treasure is restricted to leaves
3.  $t$  is chosen at random according to a known dist  $\theta$



**Algorithm:** go to the node on the border of what you saw that maximizes the prob that the treasure is a descendant of that node

We want to choose  $\theta$  s.t. the corresponding algorithm will be good against an adversary

So what would be the good choice of  $\theta$ ?

The most natural choice is the uniform over all leaves or over all nodes.

This works for complete  $\Delta$ -ary trees, but fails for general trees :-)



# Choosing $\theta$

We define  $\theta$  according to a random walking down process:

Starting at the root, walk down to a child u.a.r.  
until reaching a leaf

For a leaf  $v$ , define  $\theta(v)$  as the probability that this process eventually reaches  $v$ . Our extension of  $\theta$  can be interpreted as  $\theta(v)$  being the probability that this process passes through  $v$ . Formally,

$$\theta(\sigma) = 1, \text{ and } \theta(u) = 1 / \prod_{w \in [\sigma, u)} \Delta_w$$

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$$\theta(\sigma) = 1, \text{ and } \theta(u) = 1 / \prod_{w \in [\sigma, u)} \Delta_w$$

This works! it gives an algorithm that runs in  $O(d\sqrt{\Delta})$  time

# The optimal walking algorithm

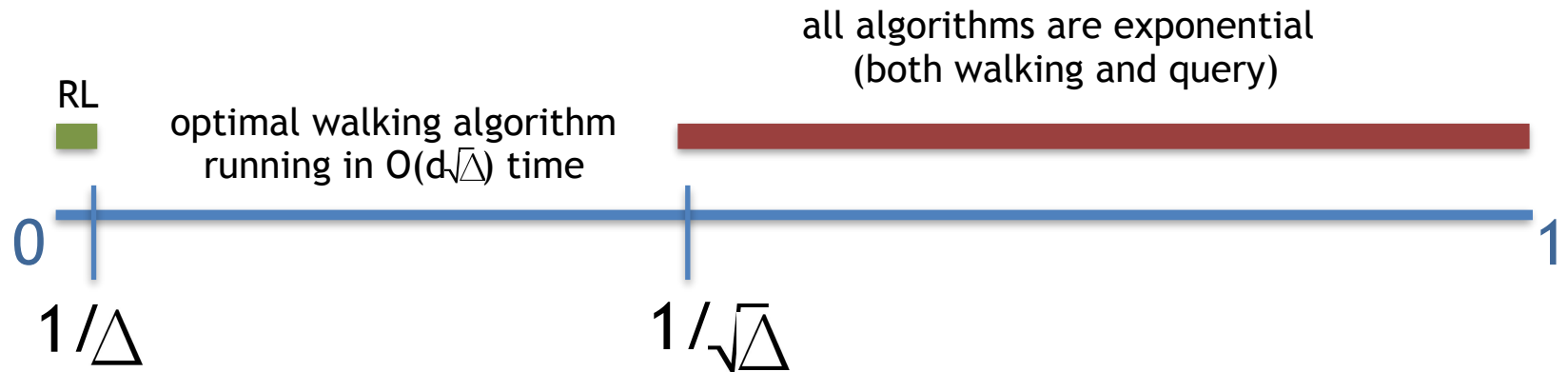
$$\theta(\sigma) = 1, \text{ and } \theta(u) = 1 / \prod_{w \in [\sigma, u)} \Delta_w$$

$$\text{score}(u) = \frac{2}{3} \log(\theta(u)) - \sum_{w \in \overleftarrow{\text{adv}}(u)} \log(\Delta_w)$$

Consider all advice discovered so far, and go to a node on the border with highest score

Note, the algorithm does not need any a priori knowledge of the structure of the tree!

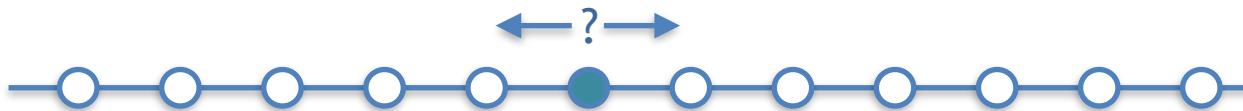
# Query Algorithms



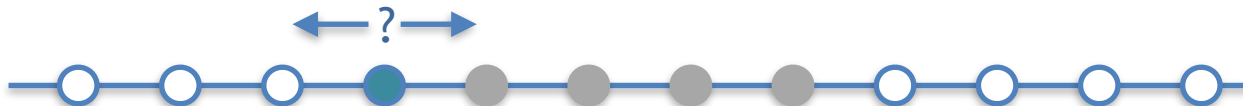
What about query algorithms?

# $O(\log n)$ query algorithm for the line

In the case of refreshable noise, there exists an  $O(\log n)$  query algorithm  
[Feige et al. SICOMP 1994]

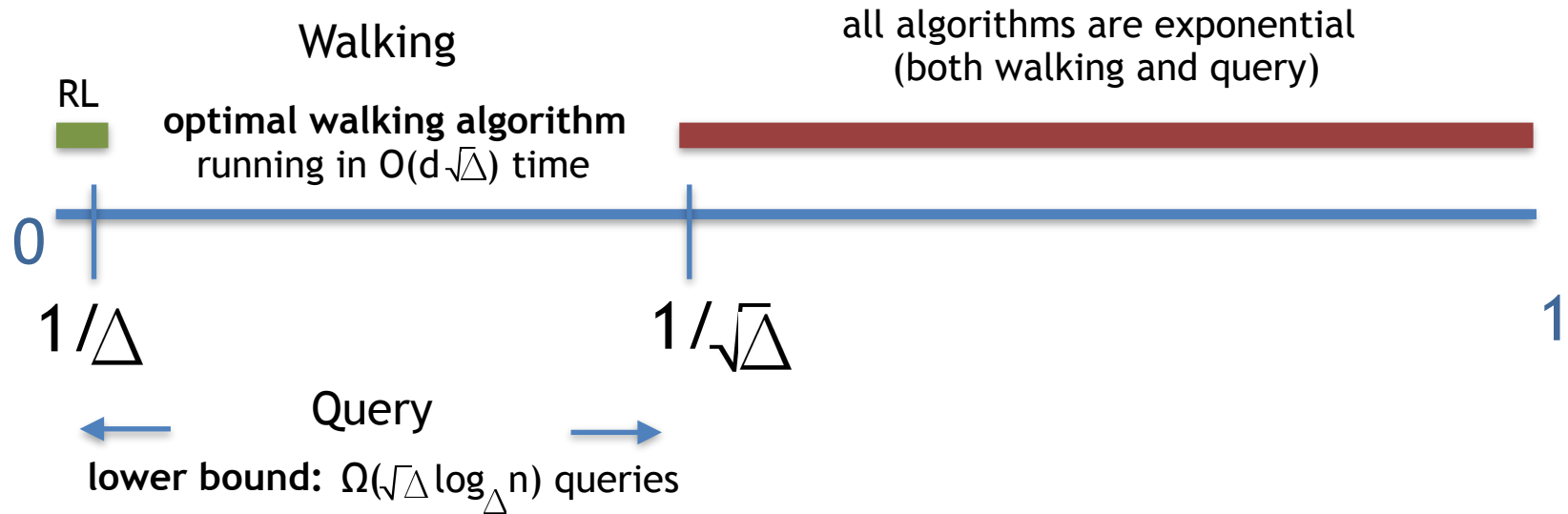


It is easy to immolate any protocol on the line by simply querying one of the neighbors of endnotes of the corresponding visited subpath





# Theorem: Lower Bound of $\Omega(\sqrt{\Delta} \cdot \log_{\Delta} n)$ for complete $\Delta$ -ary trees



# Query algorithms

**Theorem:** There is a query algorithm with # of queries  $O(\sqrt{\Delta} \cdot \log^2 n)$  on expectation, when  $q < 1/\sqrt{\Delta}$

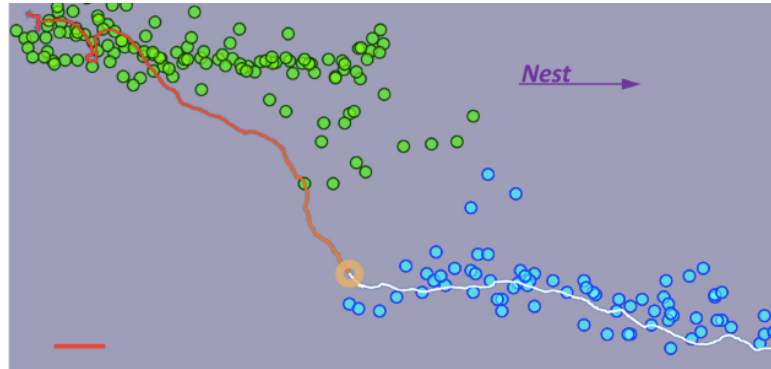
## Basic strategy

- Do a separator decomposition.
- For each junction  $v$  of the sep tree, apply the walking alg on the subtree  $T_v$  of depth  $O(\log n)$ , until finding a leaf  $w$  of  $T_v$ , for which 80% of the arrows point to it. W.h.p., this will happen by time  $O(\sqrt{\Delta} \log n)$ .
- Once  $w$  is point, the neighbor of  $v$  in the sep tree that contains  $w$  in its tree is w.h.p the correct separator to continue.
- W.h.p. this finds the treasure within  $O(\sqrt{\Delta} \log^2 n)$ . If the treasure is not found by this time, do an exhaustive search.

**Theorem:** There is a query algorithm with # of queries  $O(\sqrt{\Delta} \cdot \log n \cdot \log \log n)$  for  $\Delta$ -regular trees

# Summary

A new kind of ant trail

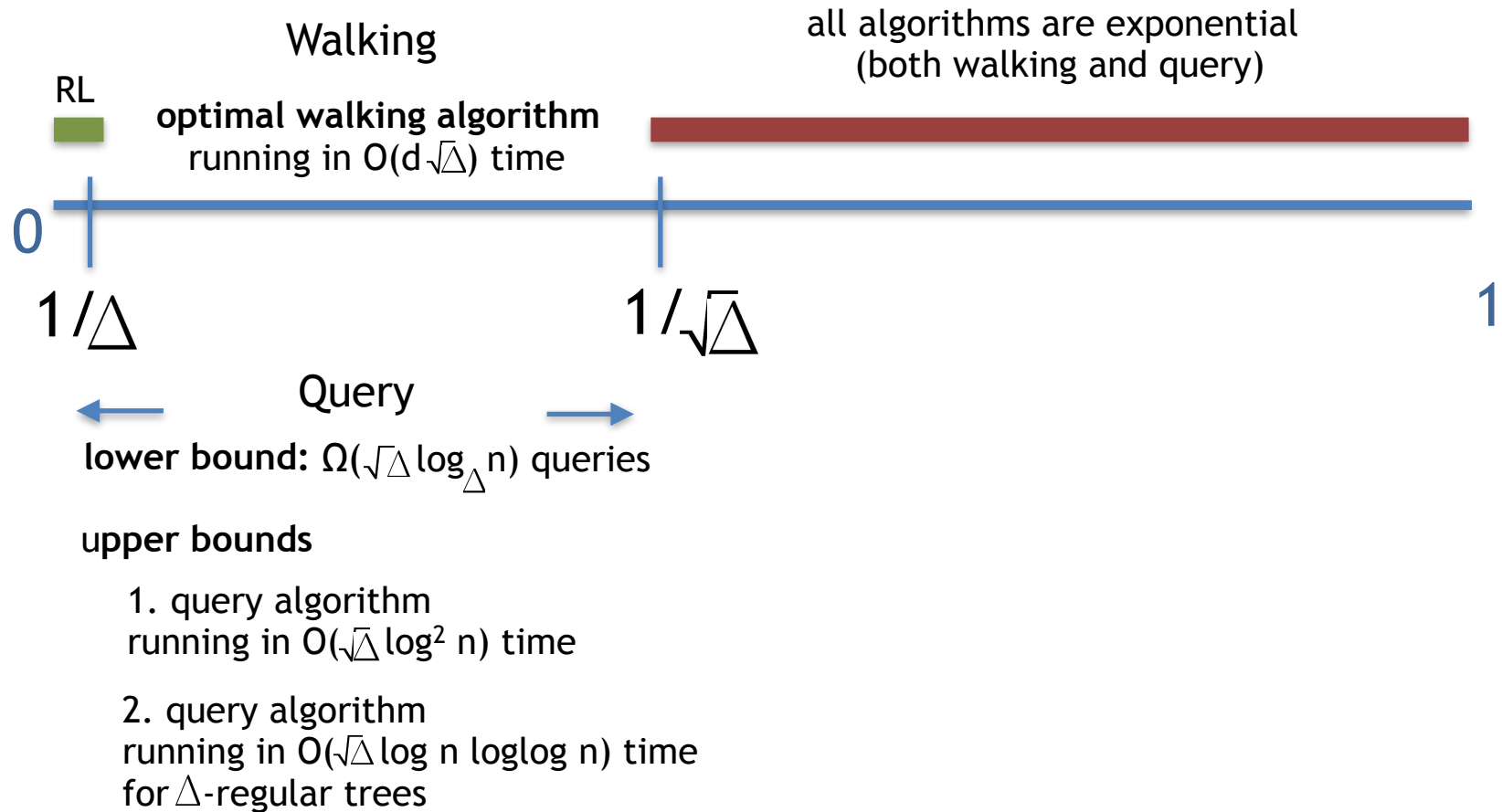


A new kind of model for search in  
unreliable conditions

RL - A memoryless strategy that is good for  
grids and trees as long as  $q < c/\Delta$

Conjecture: RL is good for any graph as long as  $q < c/\Delta$

# Summary - on noisy trees



# Open problems

- Solve the random listening conjecture for general graphs
- Find optimal algorithms for other graph families (e.g. expanders?)



**Merci!**