

Can we find a short path, without a map?

Can we parallelize maze-solving?

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Based on joint works with Laurent Massoulié & Laurent Viennot

PhD Seminar

The logo for Inria, featuring the word "Inria" in a stylized, red, cursive font.

**Keywords:** collaborative exploration, trees, competitive analysis

*20 slides / 20 mins*

# Online framework: the science of decision-making

Online problem = Information arrives over time. Examples :

- |                               |   |           |
|-------------------------------|---|-----------|
| • Linear Search               | (Where should I search for my lost wallet?)                 | Starter   |
| • Learning with expert advice | (Who should I trust?)                                       |           |
| • Metrical Task Systems       | (When should I move out to another city?)                   |           |
| • Secretary problem           | (When should I stop a hiring process?)                      |           |
| • Bandit problems             | (Should I be safe (exploitation) or be bold (exploration)?) |           |
| • List Update Problem         | (How should I organize my bookshelf?)                       |           |
| • Collective Tree Exploration | (How to solve a maze, with a team?)                         | Main Dish |
| • Layered Graph Traversal     | (How to find a short path, without a map?)                  |           |

# A simple example: Linear Search

*Where should I look for my lost wallet?*

- Introduced by Bellman in the 1950s
  - *you* start from  $0 \in \mathbb{Z}$  in an infinite street, your **Cost = #steps**
  - *wallet* is lost at  $x \in \mathbb{Z}$  at distance  $|x| = \text{OPT}$  (unknown)
  - Your strategy has competitive ratio  $\alpha$  if you always find your wallet with at most

$$\text{Cost} \leq \alpha \cdot \text{OPT}$$

*The doubling strategy is 9-competitive*



# Another example : Learning with Expert advice

*Who should I trust when there are many self-claimed « experts »?*

- Introduced independently in many fields in the late 20th century.
  - Each day  $t \in \{1, \dots, T\}$  there are  $n$  experts forecasting rain ( $y_{t,n} = 1$ ) or sun ( $y_{t,n} = 0$ )
  - **OPT** = number of mistakes made by the most accurate forecaster
  - **Cost** = the number of mistakes that you make
  - Your strategy has regret  $R(n, T)$  if it satisfies

$$\text{Cost} \leq \text{OPT} + R(n, T)$$

The multiplicative weights strategy has  $R(n, T) = \mathcal{O}(\sqrt{T \log n})$  regret



# Recap: on online problems

- **OPT** : the optimal possible cost, if you had all the information !
- **Cost** : the actual cost you pay,  $\text{Cost} \geq \text{OPT}$
- A strategy achieves :

- **Competitive ratio  $\alpha$**  if

$$\text{Cost} \leq \alpha \times \text{OPT}$$

- **Regret  $R$**  if

$$\text{Cost} \leq \text{OPT} + R$$

# Main dish !

- **Layered Graph Traversal**

- ❖ *Can you find the shortest path, when you don't have a map?*
- Papadimitriou and Yannakakis, 1991 (online algorithms)

- **Collective Tree Exploration**

- ❖ *Is maze-solving parallelizable?*
- Fraigniaud, Gasieniec, Kowalski and Pelc, 2004 (distributed algorithms)



# Layered Tree Traversal

Can you find the shortest path,  
Without a map?

# Can you find the shortest path, without a map?

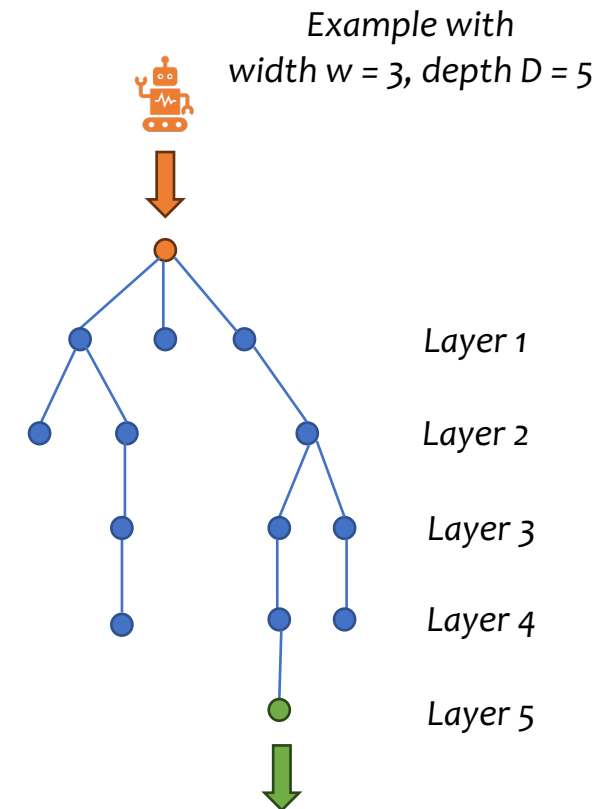
**Notation** :  $n = \#nodes = \#edges+1$  and  $D = depth$  and  $w = width$

- « **the width** » : max # nodes at given combinatorial depth
- « **a layer** » : set of nodes at the given combinatorial depth

**Online Problem** : Layers are revealed one after the other !

- **OPT = D** so an «  $\alpha(w)$ -competitive path » has length

$$Cost \leq \alpha(w)D$$



15 moves, i.e. 3-competitive



# Is layered-feedback realistic?



length  $D$



width  $w$

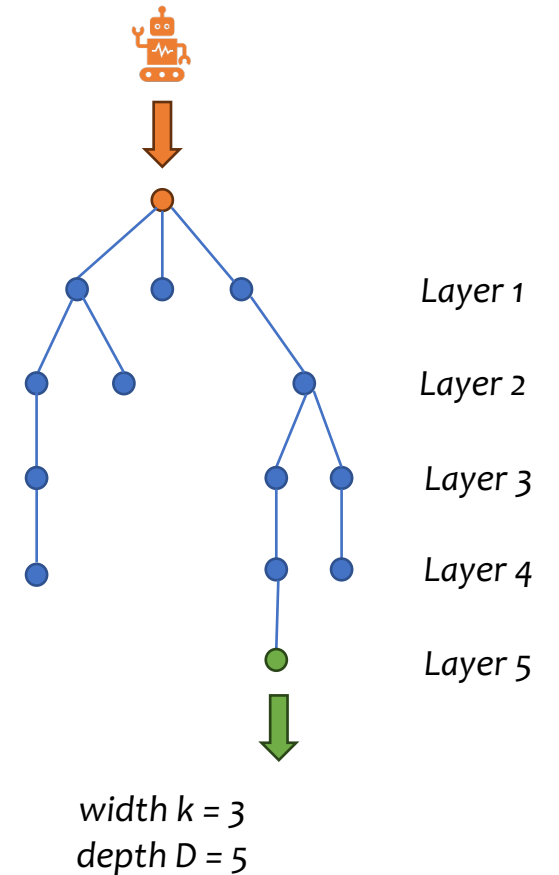
# The unweighted variant

- ✓ **Observation:** In unweighted tree  $n = wD$ 
  - ✓ Depth-First Search is thus  $O(w)$  - competitive

- ✓ **Question:** Can we do better than Depth-First Search?

## Our work [CM]

- ✓ **Yes!** There is a  $O(\sqrt{w})$ -competitive strategy
  - ✓ For a more general formulation of the problem
  - ✓ Uses random choices!





# Definition and Motivation

## *Collective Tree Exploration*

*Is maze-solving  
parallelizable?*

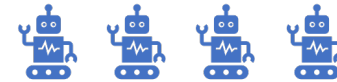
# Is solving a (tree) maze parallelizable?

**Notation :**  $n = \text{\#nodes} = \text{\#edges}+1$  and  $D = \text{depth}$

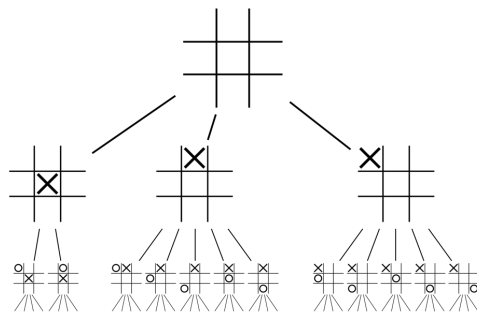
• **With 1 agent:** right-hand on wall (RHW, aka DFS)  $\leq 2n$  moves

• **With 2 agents:** right+left-hand on wall (RHW+LHW)  $\leq n$  moves each

• **What about  $k \geq 3$  agents?** Moving synchronously at each round



• Exploration  $\leftrightarrow$  Finding exit



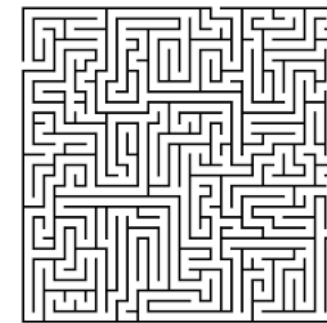
Tic-tac-toe tree game



Silver coin of Knossos (Crete, 400 BC)

Berlin state museum

Collective Tree Exploration



A computer-generated  
(tree) maze



**VLAM!**

**?!**

**VOUS NE  
SORTIREZ JAMAIS  
D'ICI, ÉTRANGERS! CE  
TOMBEAU SERA VOTRE  
TOMBEAU!**

# « Collective Tree Exploration »

[FGKP 2004] Fraigniaud, Gasieniec, Kowalski, Pelc [Collective Tree Exploration](#)

➤ **Goal:** Traverse all edges of unknown tree  $T = (V, E)$  with  $n$  nodes and depth  $D$

✓ With  $k \in \mathbb{N}$  agents moving synchronously at each round

➤ **Communication Models:**

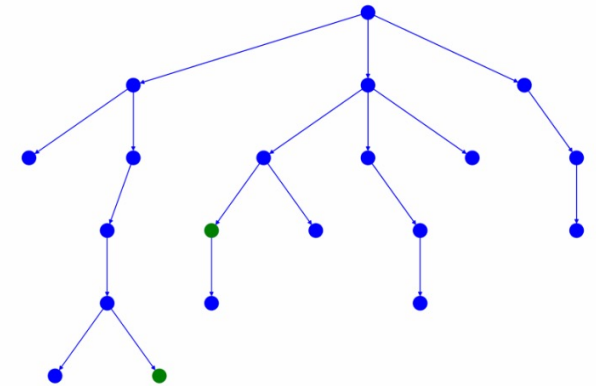
☎ ✓ Centralized (full « complete » communication)

🗺 ✓ Distributed (restricted « write-read » communication)

➤ **Main result [FGKP 2004]:**

✓ Distributed algorithm **SPLIT** in which explorers split evenly at intersections

✓  $\text{Runtime}_{\text{SPLIT}} \leq O\left(\frac{2n}{\log k} + D\right)$



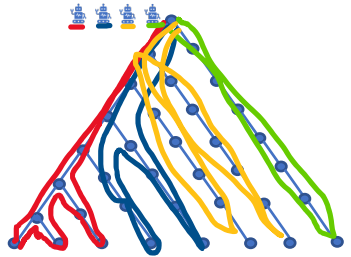
A tree with  $n = 20$  and  $D = 6$

# The Competitive Ratio approach

What runtime could we hope for, had we know the tree in advance?

➤ **Offline variant** :  $OPT$  = Exploration time if tree were known in advance

$$D \leq OPT \quad \text{and} \quad \frac{n}{k} \leq OPT \quad \text{in fact,} \quad \frac{n}{k} + D \approx OPT$$



➤ **Consequence** :  $SPLIT$  has a **competitive ratio** in  $O\left(\frac{k}{\log k}\right)$

$$\text{Runtime}_{SPLIT} = O\left(\frac{n}{\log k} + D\right) \leq O\left(\frac{k}{\log k}\right) OPT$$

# Recent results

- ✓ Continuous **analysis** of online algorithms (online convex optimization)
- ✓ New idea: the explorers behave like electrons in an **electric network**



## Latest results [C., Massoulié]

- Regret:  $O(kD)$
- Competitive Ratio:  $O(\sqrt{k})$



# Open Questions



# Some Open Questions



**Open Question 1:** Is there a competitive collective **graph** exploration algo

(competitive ratio)  $c(k) \left( \frac{m}{k} + D \right)$  or in  $\frac{2m}{k} + f(k, D)$  (regret)

where  $m$  is # of edges,  $D$  is **graph diameter**, and  $f(\cdot, \cdot)$  is some arbitrary function.

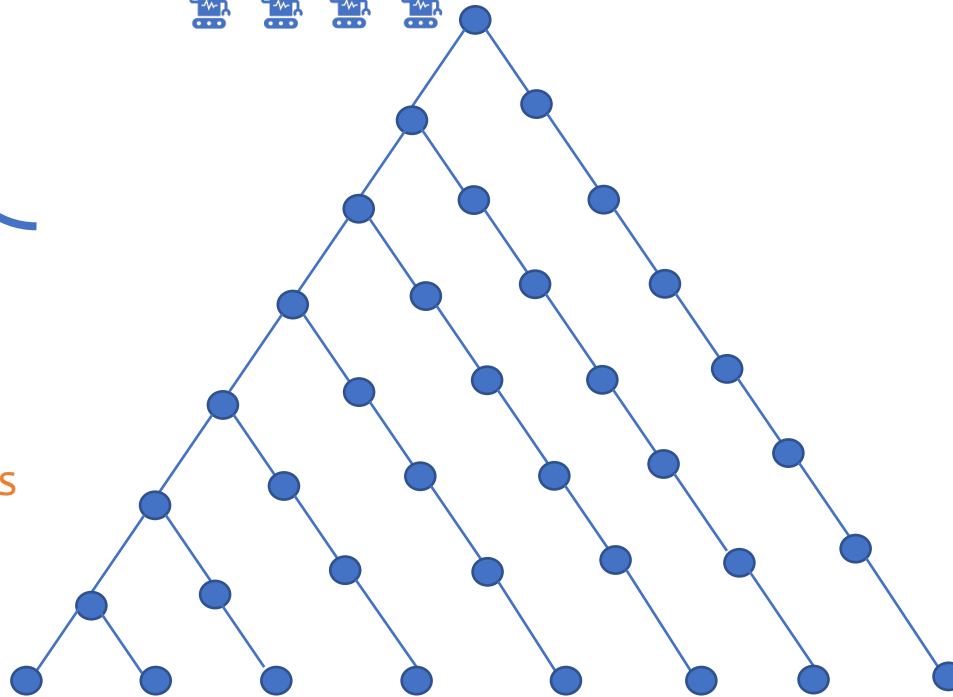
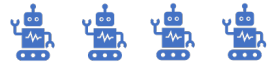
**Open Question 2:** Is there a (competitive) gap between **distributed** and **centralized** collective tree exploration?



Thank you!  
Feel free to reach out  
Office # C320

# Why $\log k$ appears in SPLIT?

$\log_2 k$



The « comb »

$$n \approx \frac{D^2}{2}$$

We need  $\frac{D}{\log_2(k)}$  such steps

This means  $\approx \frac{D^2}{2 \log_2 k}$  rounds

At least  $\Omega\left(\frac{n}{\log_2 k}\right)$  rounds

[HKLT 2014] This lower-bound applies to any « greedy » algorithm.

def. « Greedy » = a robot never goes upwards, if there is an unexplored edge below