

# Beachcombing on Strips and Islands

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# The Beachcombers' problem

## Setting

- Robots have to **search every point** of a given domain
- They have different maximum **searching speeds**  $s_i$  and maximum **walking speeds**  $w_i > s_i$



Motivation: Careful searching (or some other action) takes more time than casual traversal of the domain.

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# Applications

## Searching type

- Beachcombing (beach looking for things of value)
- Looking for leaks/corrosion on pipelines
- Searching the lost Teddy bear after a family excursion
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## Processing type

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- Snow removal and de-icing of roads
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# Basic model

## Input

- Set of robots with
  - A maximum **walking** speed  $w_i$
  - A maximum **searching** speed  $s_i < w_i$
- A **domain**: so far, the line segment

## Additional assumptions

- No communication
- No cost for changing speed, mode, direction

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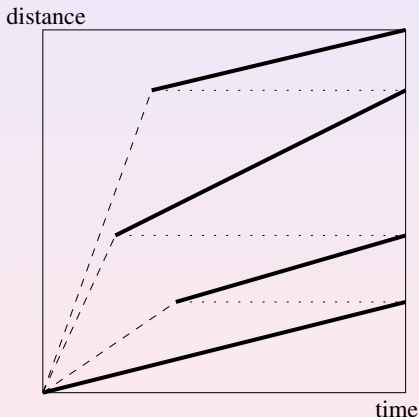
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# A schedule

Combining a segment of **known length** from one endpoint

**Optimal** solution:  
(computable in  
polynomial time)



Algorithm COMB [Cyzowicz et al., SIROCCO 2014]

# The online variants

Online version: the **length** of the segment is **not known**

Alternatively: the domain is a semi-infinite line

## Discrete Online Beachcombers' problem

- The segment length is an integer
- Goal: maximizing  $\inf_{\ell \in \mathbb{N}^+} \frac{\text{length } \ell}{\text{time of } A \text{ over segment } [0; \ell]}$

## Continuous Online Beachcombers' problem

- The segment length is a real (at least 1)
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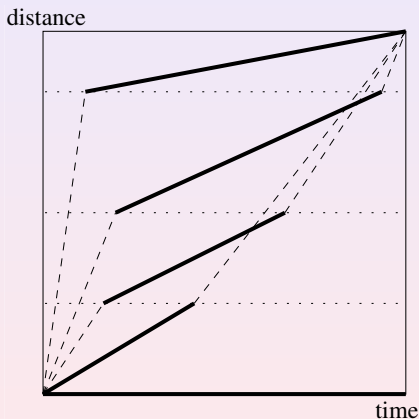
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# Another schedule

Combing a segment of **unknown length** from one endpoint

Algorithm LEAPFROG  
[Cyzowicz et al.,  
SIROCCO 2014]



**Conjectured to be optimal** in the discrete online setting



# The multi-source Beachcombers' problem

Introduced in [Czyzowicz et al., ALGOSENSORS 2014]

## $t$ -source Beachcombers' problem ( $t$ -SBP)

Main modifications:

- The robots **cannot change direction**
- Additional outputs
  - A partition of the robots in at most  $t$  **groups**
  - A **source** and a **direction of move** for each of the groups

Their hardness result

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# Previous work (summary)

Only studied on the segment

[Czyzowicz et al., SIROCCO 2014]

- Offline COMB algorithm: proved optimal
- Online LEAPFROG algorithm: 2-competitive

[Czyzowicz et al., ALGOSENSORS 2014]

- 2-SBP is NP-hard, even with uniform walking speeds
- 0.5 and 0.56 approximation algorithms for 2-SBP
- Poly-time algo for  $t$ -SBP with uniform searching speeds
- Randomized algorithm for  $t$ -SBP with expected approximation ratio  $1 - (1 - 1/t)^t$

# Our results

## Online Beachcombers' Problem on the segment

Algo **LEAPFROG** is optimal, even for the continuous variant

## Offline Beachcombers' Problem on the cycle

- Equivalent to 2-SBP, thus NP-hard
- Efficient algorithms  
(from [Czyzowicz et al., ALGOSENSORS 2014])

## Offline Multi-source Beachcombers' Problem with zigzags

- Zigzags do not help (both in segments and cycles)
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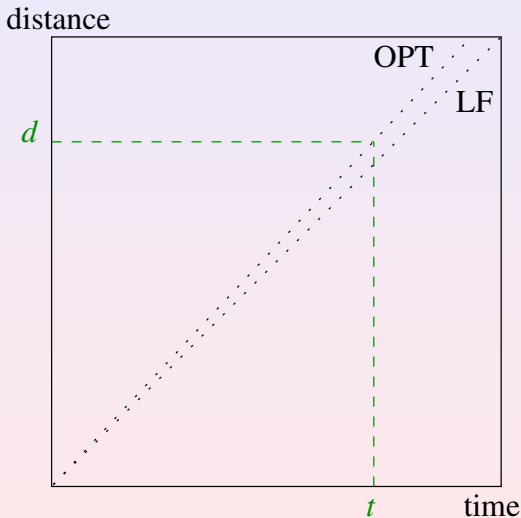
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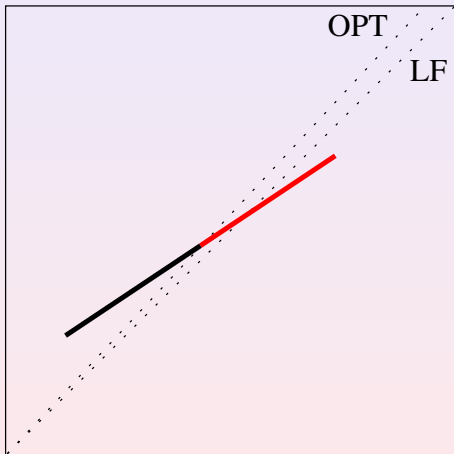
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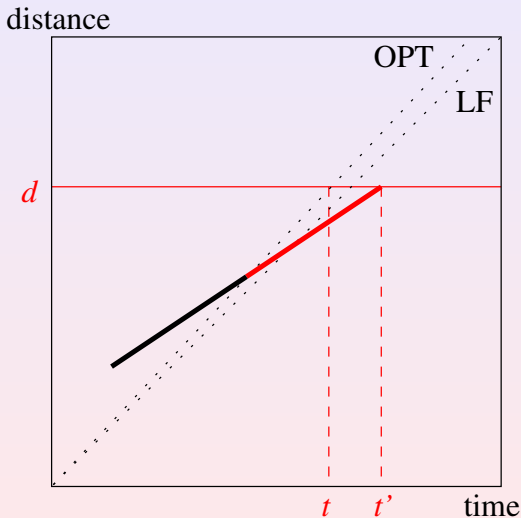
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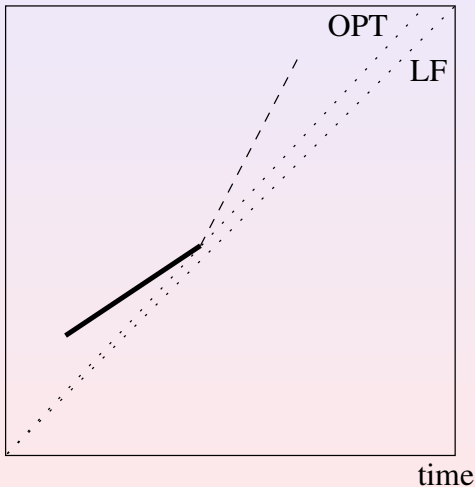


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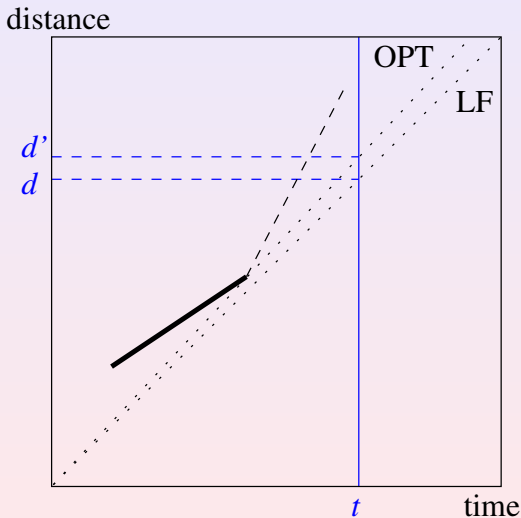


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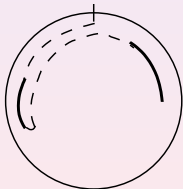
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# Combing a cycle of known circumference

**Assumption:** Initially collocated robots.

- If zig-zags were not allowed, this would be equivalent to combing a known segment from both endpoints.
- However, zig-zags **are** allowed:
  - either use them in order to obtain better solutions,
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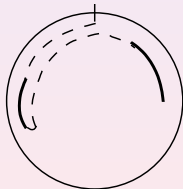


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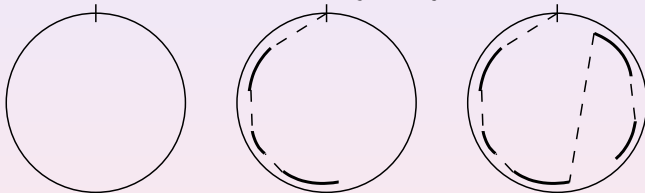
# Some simple observations

- It doesn't make sense to walk at a speed smaller than  $w_i$  or to search at a speed smaller than  $s_i$ .
- It doesn't make sense to have **overlapping** searched arcs.
- We can assume that each trajectory looks like this:

- In fact, if we look at all robots' searched arcs:

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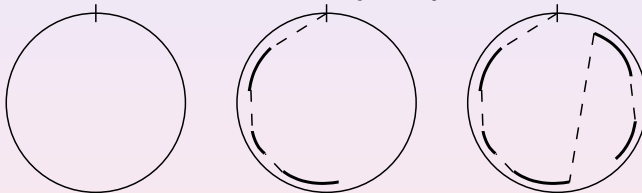
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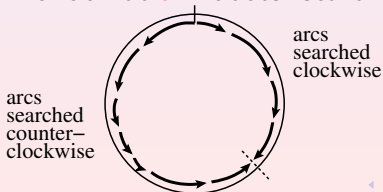
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# Normal schedules

## Definition

We call a schedule **normal** if it is as described in the previous slide.

## Lemma

We can transform every **non-normal** schedule into a **normal** schedule whose completion time is **not increased**, and which has **at least one** robot whose trajectory is **strictly decreased**.

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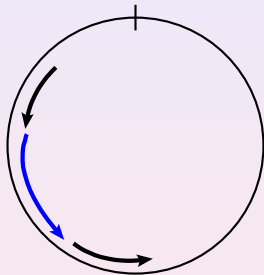
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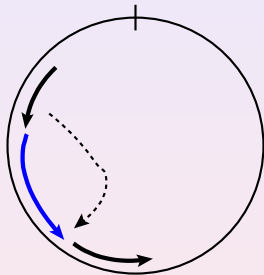
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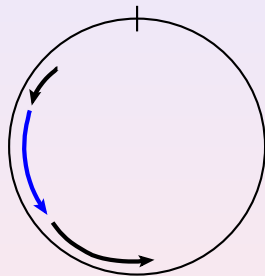
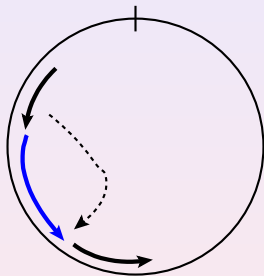
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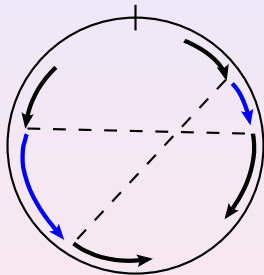


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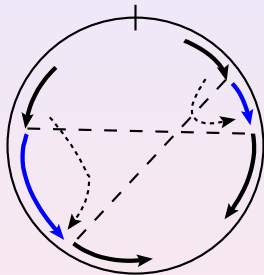


Robots searching the **blue part** finish their trajectories **earlier**.

# No crossing robots

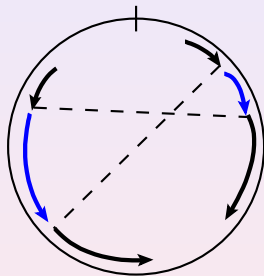
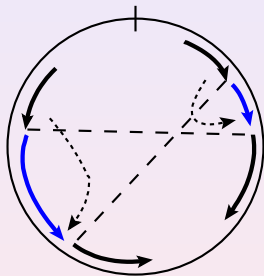


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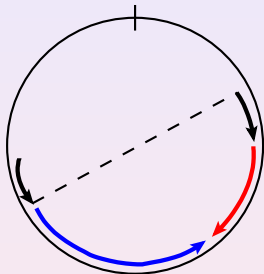


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All affected robots conclude their trajectories earlier.

# Finally: no zig-zags



**Idea:** We can “nudge” the two arcs of the zig-zagging robot so that **its completion time is reduced** while the completion time of robots covering the blue and red arcs remain the same.

Many natural extensions are worth being looked at:

- Different domains
- Robots have to return to their starting position
- Online version with communication (adaptive schedule)
- Fault-tolerance
-

Thank you  
for your attention