

# Online Algorithms



IT'S OKAY IF THE FUTURE IS  
UNCERTAIN.  
I THINK WE'RE JUST MEANT TO DO OUR  
BEST WITH WHAT'S IN FRONT  
OF US TODAY.

# Admin

Exam Review Session led by Daphne:  
Wednesday April 24th, 7-9pm, BBB 1670.

HW11 is due Tuesday 4/23 at 8pm.

Reminder: Filling out the course evaluations is 1% of your grade, which is otherwise covered by the final exam. Remember to submit receipt on Gradescope.

# It's the end of 376 as we know it

And I feel fine?



If you enjoyed this course:

Fall '24:

[**475:** Introduction to Cryptography  
**477:** Introduction to Algorithms  
**498:** Advanced Data Structures  
**498/598:** Algorithms for Machine Learning and Data Science  
**572:** Randomness and Computation  
**575:** Advanced Cryptography  
**598:** Machine Learning Theory

Winter '25:

[**475:** Introduction to Cryptography  
**477:** Introduction to Algorithms  
**574:** Computational Complexity Theory  
**598:** Machine Learning Theory  
**598:** Graph Algorithms (taught by me)

Also, consider applying to become an IA for 376

\*recently some courses were renamed “CSE” instead of “EECS” so try both names

\*\*When registering for grad-level courses, uncheck the box for “open courses only”

# It's the end of 376 as we know it

If you really enjoyed this course:

Consider trying **research** in theoretical computer science (TCS)!

## What is research?

Pushing the boundaries of human knowledge.

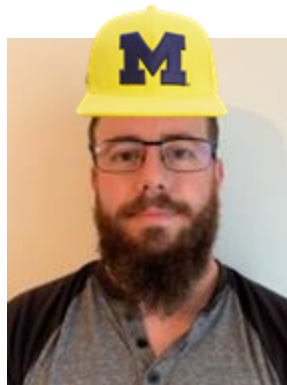
## What is TCS research?

Well, it's pretty similar to the experience of working on 376 HW  
...except you're working on the same problem for months/years and  
nobody knows the solution.

## How to get involved in research:

- Research programs at UM: [cse.engin.umich.edu/academics/undergraduate/undergraduate-resources-and-opportunities/undergraduate-research/](https://cse.engin.umich.edu/academics/undergraduate/undergraduate-resources-and-opportunities/undergraduate-research/)
- Apply to **REU** programs
- Look up the research of professors at UM and reach out to them

# A puzzle game made by UM alums!



# Online Algorithms

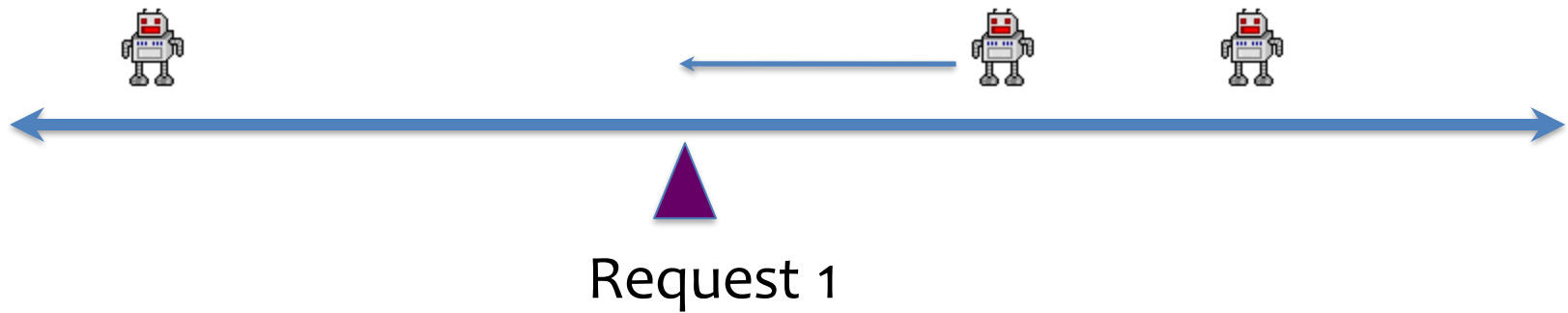
The input is revealed over time.

The algorithm needs to make decisions with partial information!

E.g. ridesharing, stock market, caching, ...

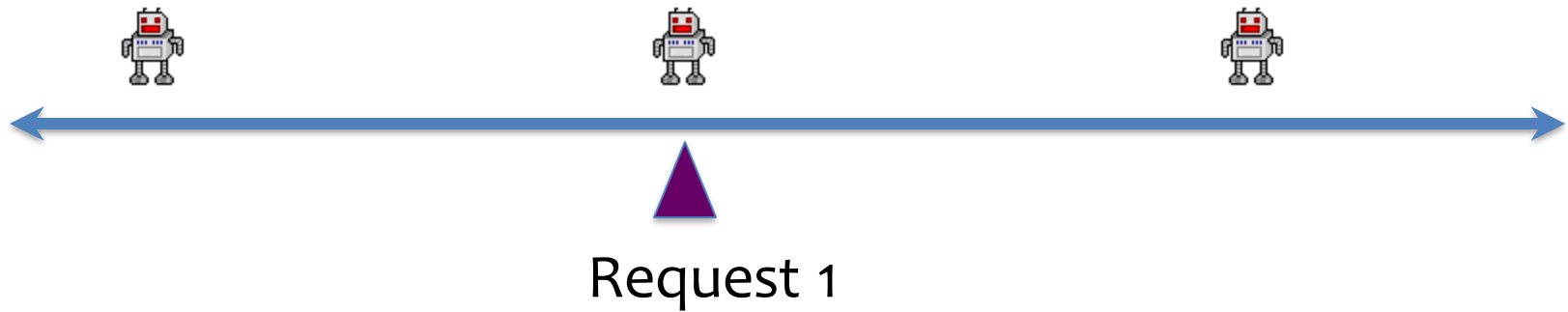
A classic problem in online algorithms:

# The k-Server Problem



A classic problem in online algorithms:

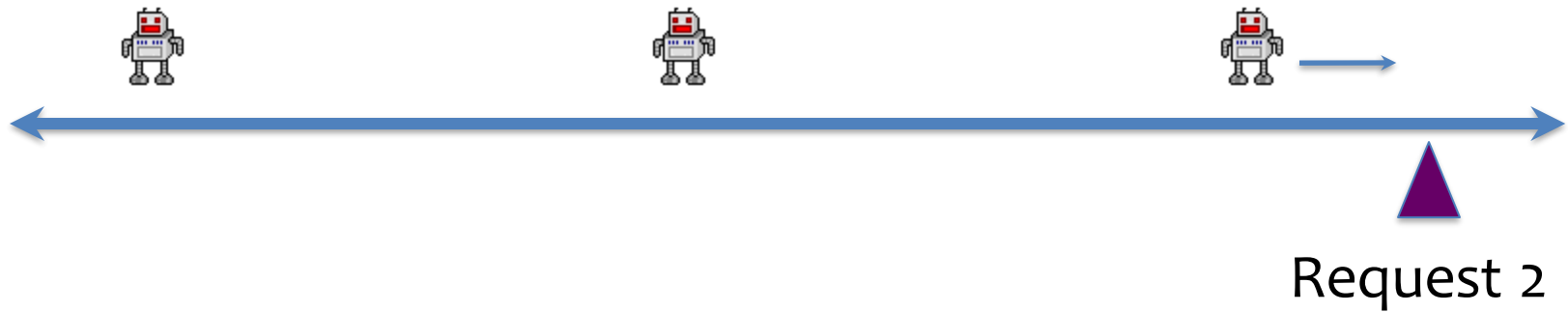
# The k-Server Problem





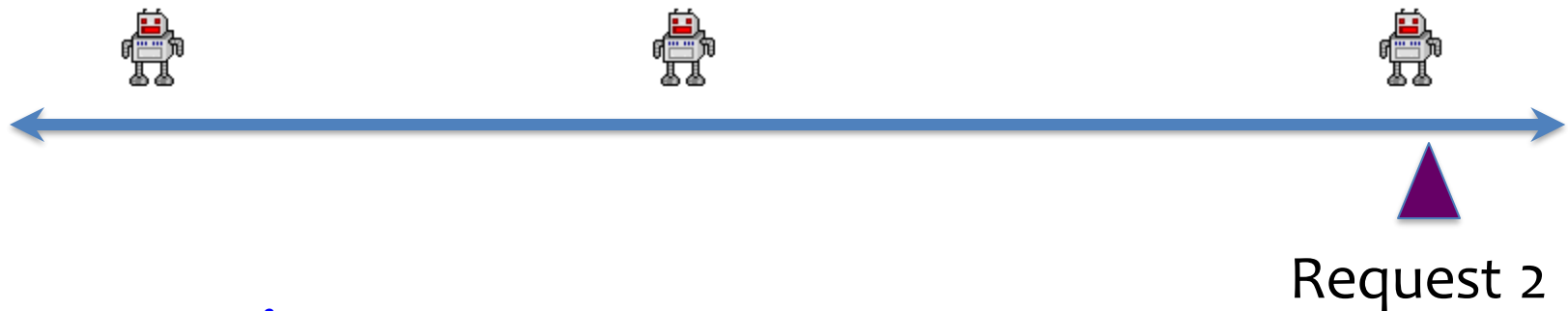
A classic problem in online algorithms:

# The k-Server Problem



A classic problem in online algorithms:

# The k-Server Problem



## Assumptions:

- $k$  identical robots. Any robot can service any request.
- Requests arrive one-at-a-time. A request must be serviced before the next request arrives.

**Goal:** Minimize total travel distance of all robots

# Competitive Ratio

We measure the quality of an online algorithm by its **competitive ratio**.

An online algorithm for the k-Server problem is **c-competitive** if its cost is at most **c** times the cost of an optimal **offline** algorithm i.e. an algorithm that *knows the entire request sequence in advance*.

$$\text{ALG} \leq c \cdot \text{OPT}, \quad c > 1$$

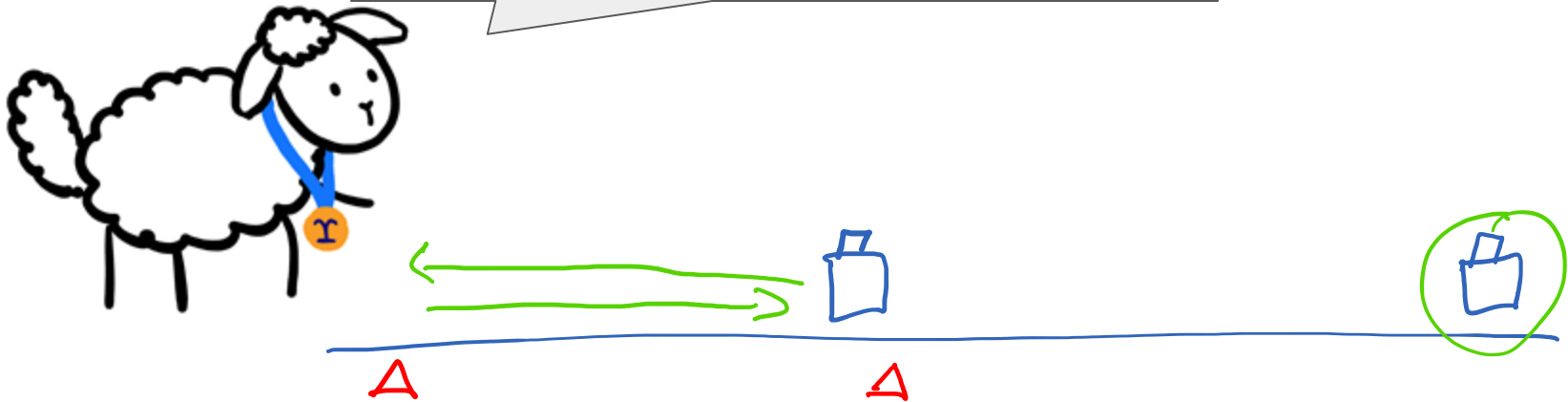
Cost of our algorithm      Cost of optimal **offline** algorithm (can see future)

**c** is called the **competitive ratio** (smaller is better here).

This is similar to approximation algorithms, but here OPT can see the future and ALG can't



Check out my greedy algorithm!  
Always dispatch the robot closest  
to the request point.



For your algorithm,  
there's a request  
sequence of length  $n$ ,  
whose competitive ratio  
goes to infinity as  $n$   
goes to infinity!



# The Double Coverage (DC) Algorithm

**The DC Algorithm** (Chrobak, Karloff, Payne, Vishwanathan, 1990):

- The closest robots on both sides of the request (if they exist) move towards the request at equal speed.
- Both stop moving once one of them reaches the request.
- If two or more robots are co-located, only one will move.



**We will show:** If there are  $k$  robots, DC is  $k$ -competitive!

It turns out  $k$  is the *best possible* competitive ratio for a deterministic algorithm.

But you can do better with randomization (we won't show).

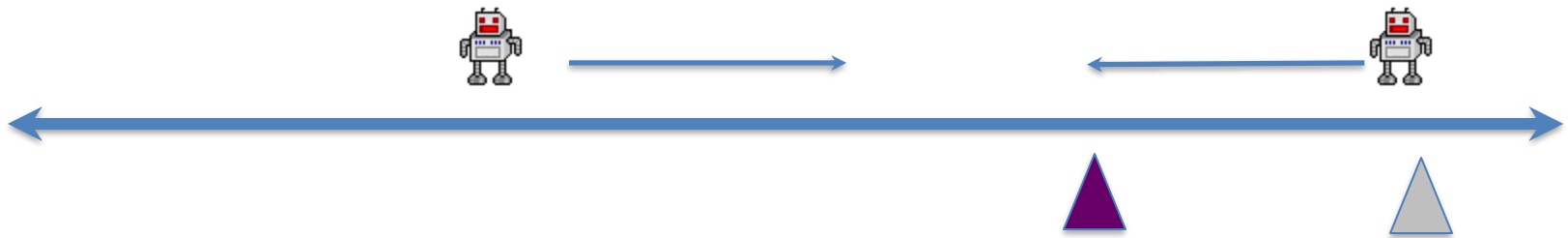
# The Double Coverage (DC) Algorithm



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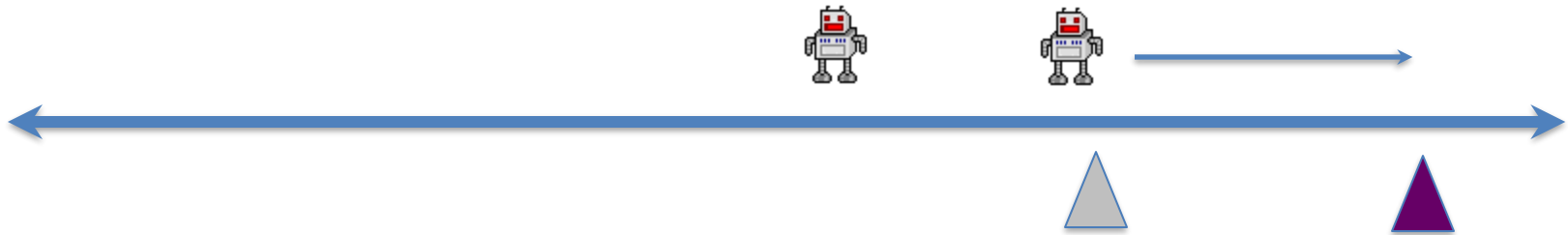


# The Double Coverage (DC) Algorithm

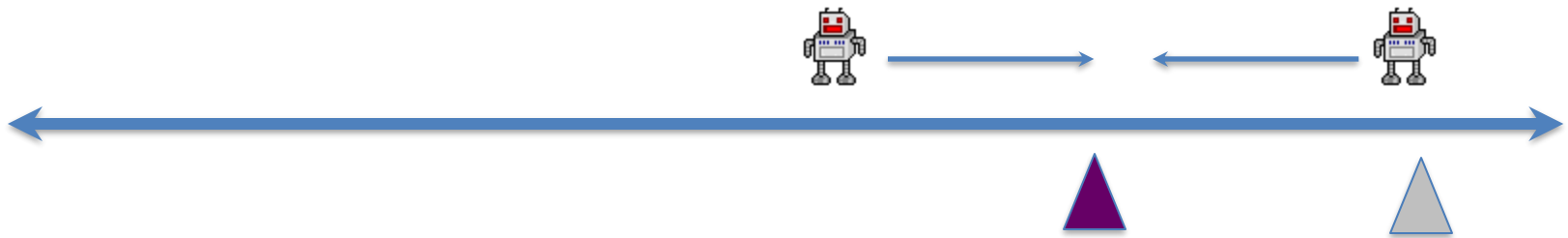




# The Double Coverage (DC) Algorithm



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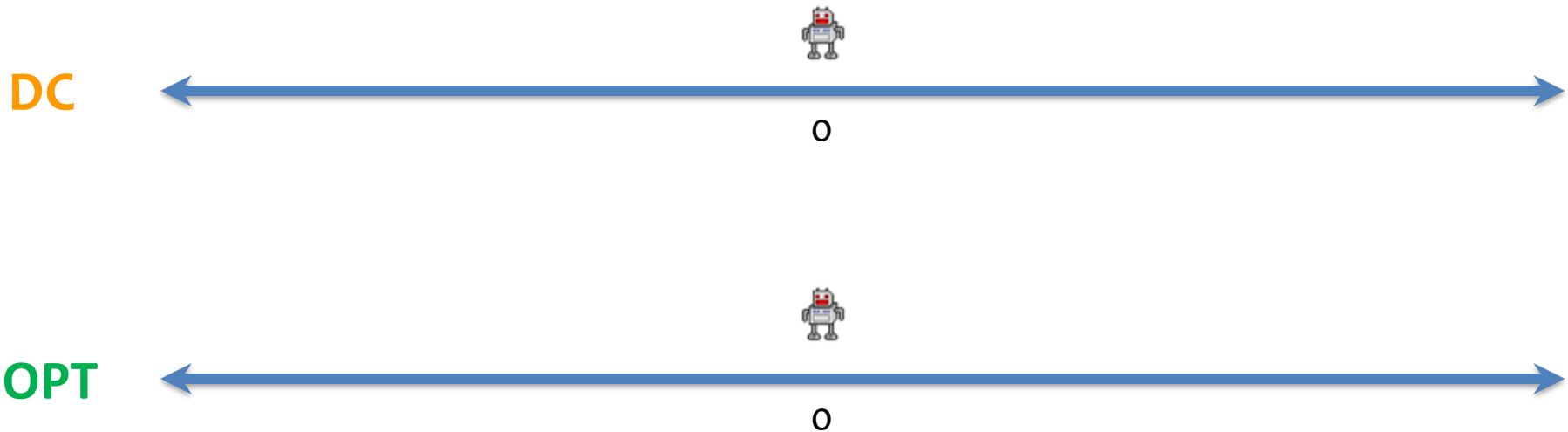
# The Double Coverage (DC) Algorithm



DC certainly looks better  
than the greedy algorithm

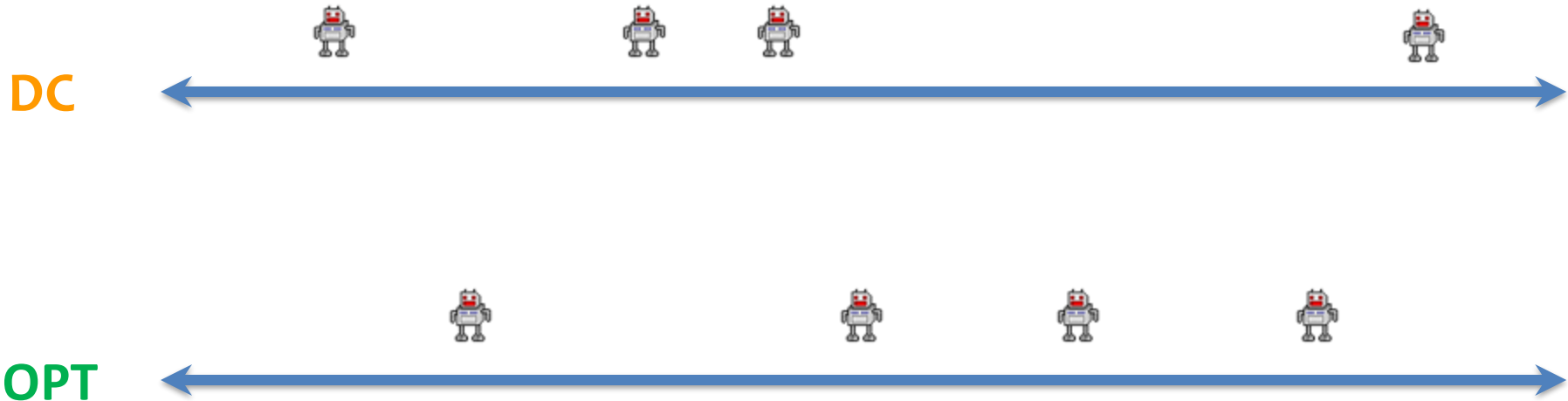


# Goal: Show DC is k-competitive



**Assumption:** Initially, **DC** and **OPT** have all of their robots at the origin...

# Goal: Show DC is $k$ -competitive



... but over time, as they service requests, the robots spread out and the configurations of **DC** and **OPT** may start looking different

**Claim:** Without loss of generality, we can assume **OPT** only moves 1 robot per request. **Why?** *Consider alg OPT' that's the same as OPT but only moves the 1 robot that services the request.*

# A Potential Function Argument

To prove that **DC** is  $k$ -competitive, we will use a **potential function**!

**Previously:** We used a potential function  $S_i$  to measure the *running time* of an algorithm.

⇒ Smaller  $S_i$  value means the algorithm is closer to termination.

**Now:** We will use a potential function  $S_i$  to measure the *competitive ratio* of the **DC** algorithm.

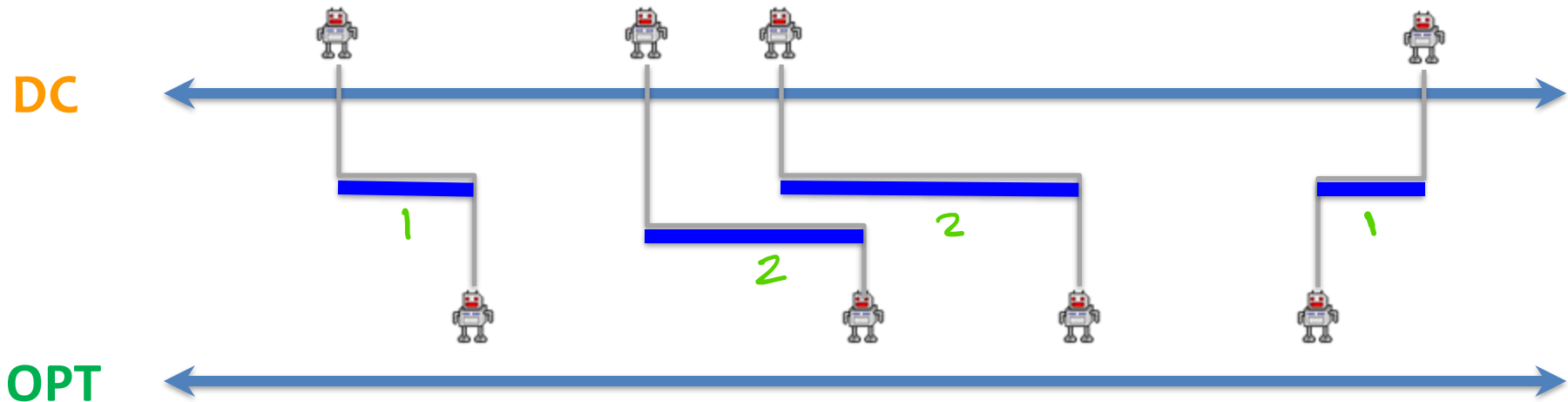
⇒ Smaller  $S_i$  value means **DC** is in a “better configuration”  $\approx$  closer to **OPT**.

(Here,  $S_i$  can go up and down over time.)

Potential  
functions are  
more versatile  
than I thought!



# Potential Function: Attempt #1



**M** = “match” the robots and take the total length of the “horizontal legs” 6

Potential function (attempt #1):  $S_i = M$   
(i.e. value of **M** right after  $i^{\text{th}}$  request has been serviced.)

This is a way of measuring how close DC is to OPT





# How will we use our potential function?

## Observations:

- $S_0 = 0$ .
- $S_i \geq 0$  for all  $i$ .

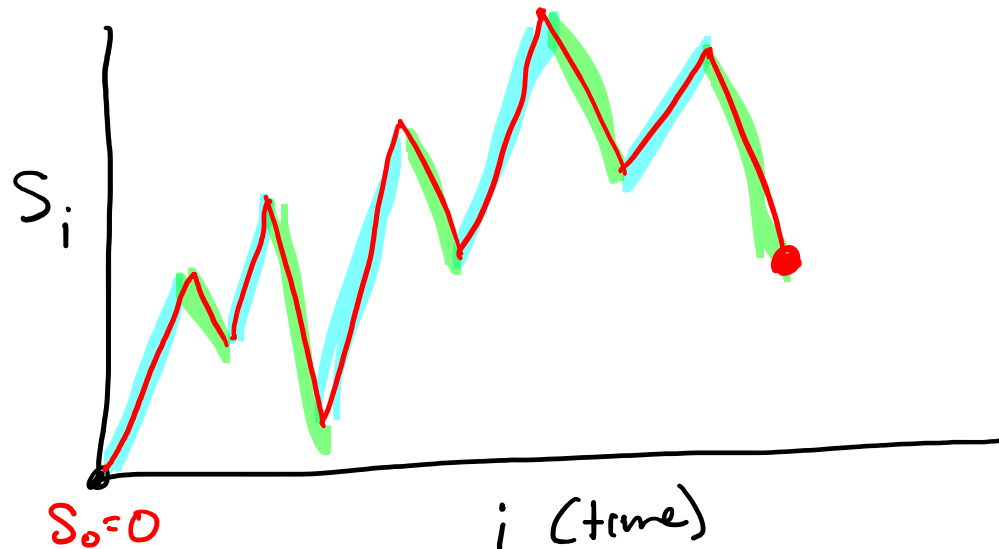
therefore



total amount  $S_i$  ever decreases

$\leq$

total amount  $S_i$  ever increases



$S_0$  OPT  $S_{0.5}$  DC  $S_i$  OPT  
mes mes mes.



For each request, we measure the change in potential right after OPT moves and then again after DC moves!



# How will we use our potential function?

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- $S_0 = 0$ .
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therefore



total amount  $S_i$  ever decreases

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total amount  $S_i$  ever increases

## Key properties we want to show:

- For any given request, if **OPT** moves its server a distance of **d** then  $S_i$  increases by  $\leq k \cdot d$  (or decreases).
- For any given request, if **DC** moves its server(s) a total distance **d** then  $S_i$  decreases by  $\geq d$ .



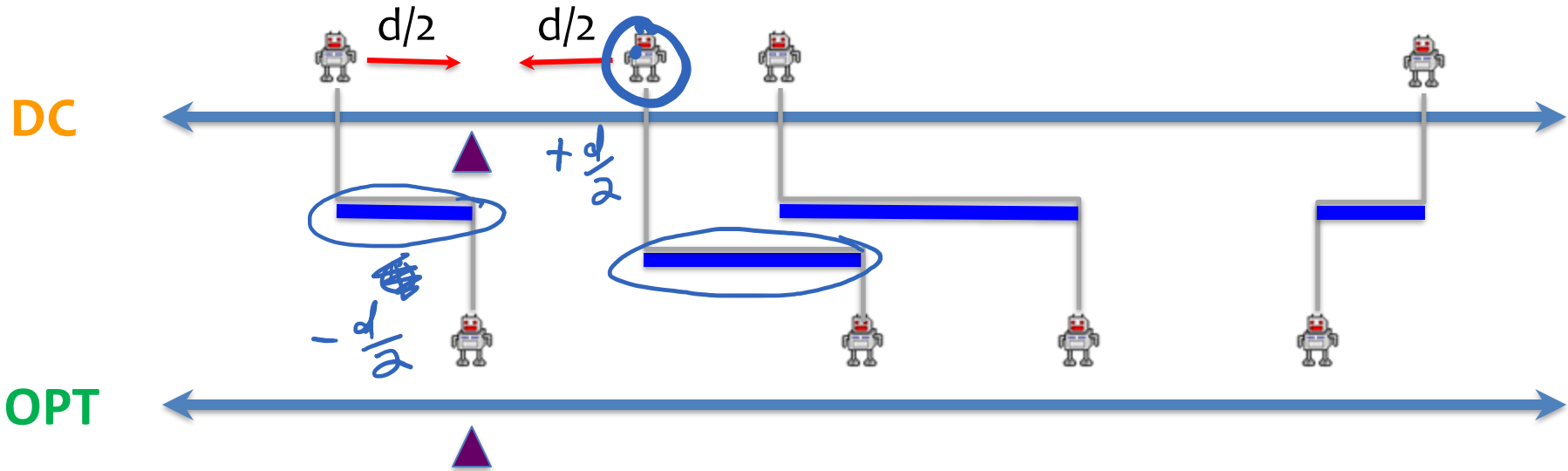
## Consequence of key properties:

- total amount  $S_i$  ever increases  $\leq k \cdot \text{OPT}$
- total amount  $S_i$  ever decreases  $\geq \text{DC}$   
 $\Rightarrow \text{DC} \leq k \cdot \text{OPT}$  (as desired)

For each request, we measure the change in potential right after OPT moves and then again after DC moves!



# Potential Function: Attempt #1



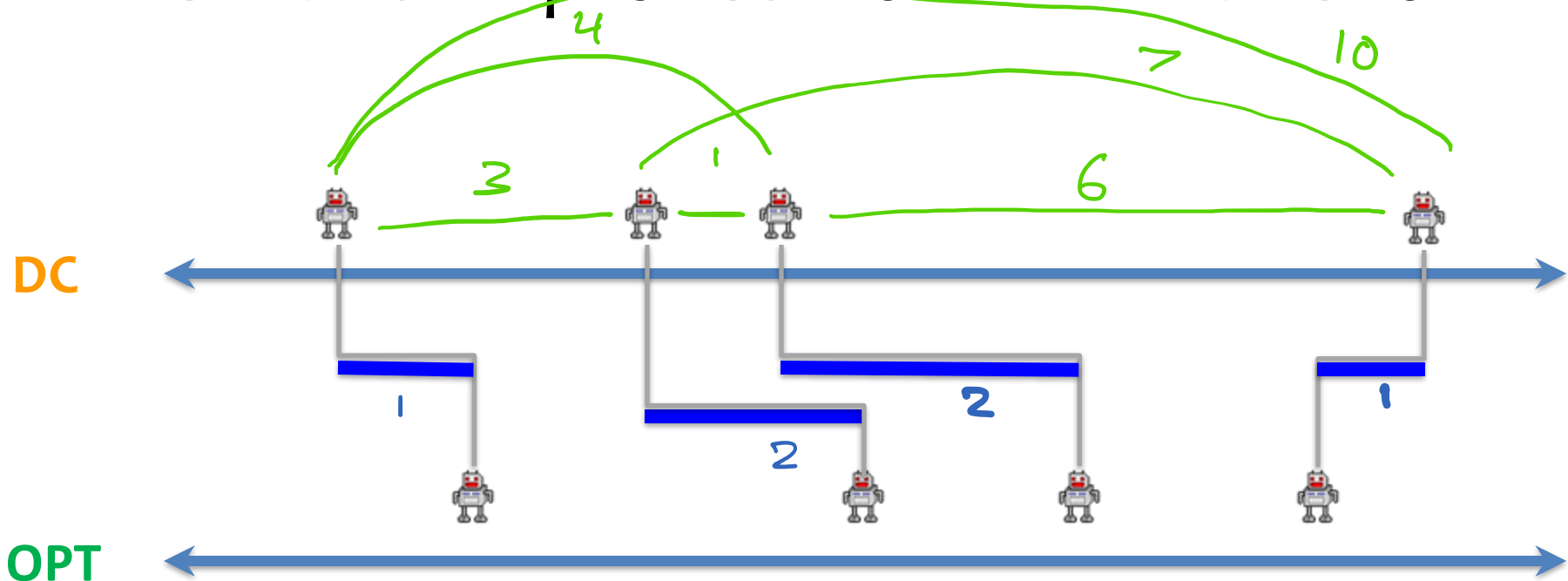
$S_i$  (attempt #1) =  $M$  (= total length of “horizontal legs”)

**Want:** if DC moves its server(s) a total distance  $d$  then  $S_i$  decreases by  $\geq d$ .

oops, we need a  
new potential  
function



# A New and Improved Potential Function



$M$  = total length of the “horizontal legs” 6

$\Sigma_{DC}$  = Sum of all pairwise distances between DC's servers

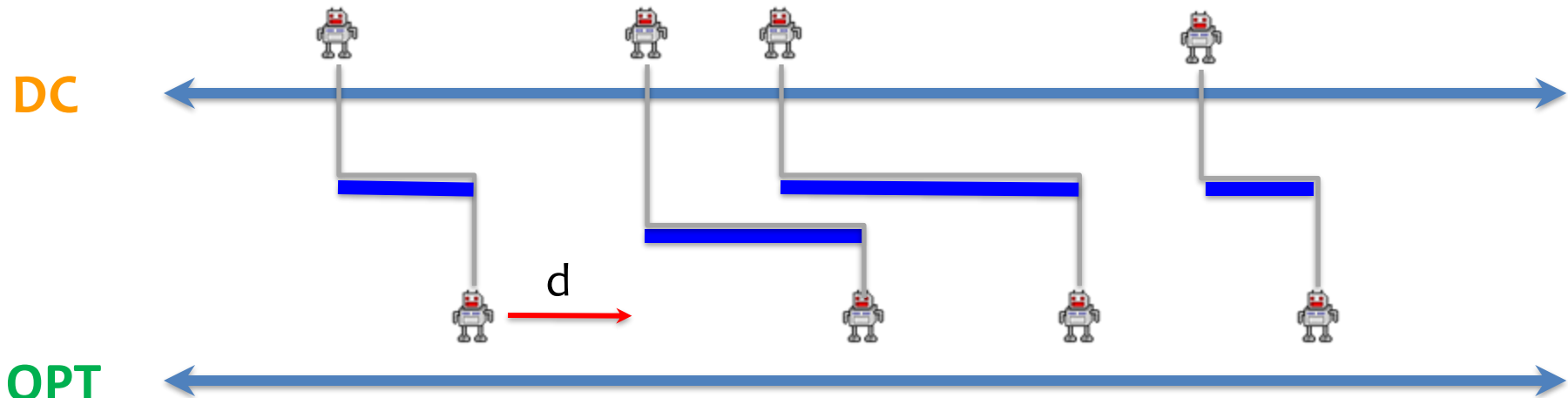
New potential function:  $S_i = \frac{k \cdot M}{4 \cdot 6} + \Sigma_{DC} = 55$

What planet did this potential function come from?



# Proving the Key Properties

Want to show: If **OPT** moves its server a distance of **d** then  **$S_i$**  increases by  $\leq k \cdot d$  (or decreases).



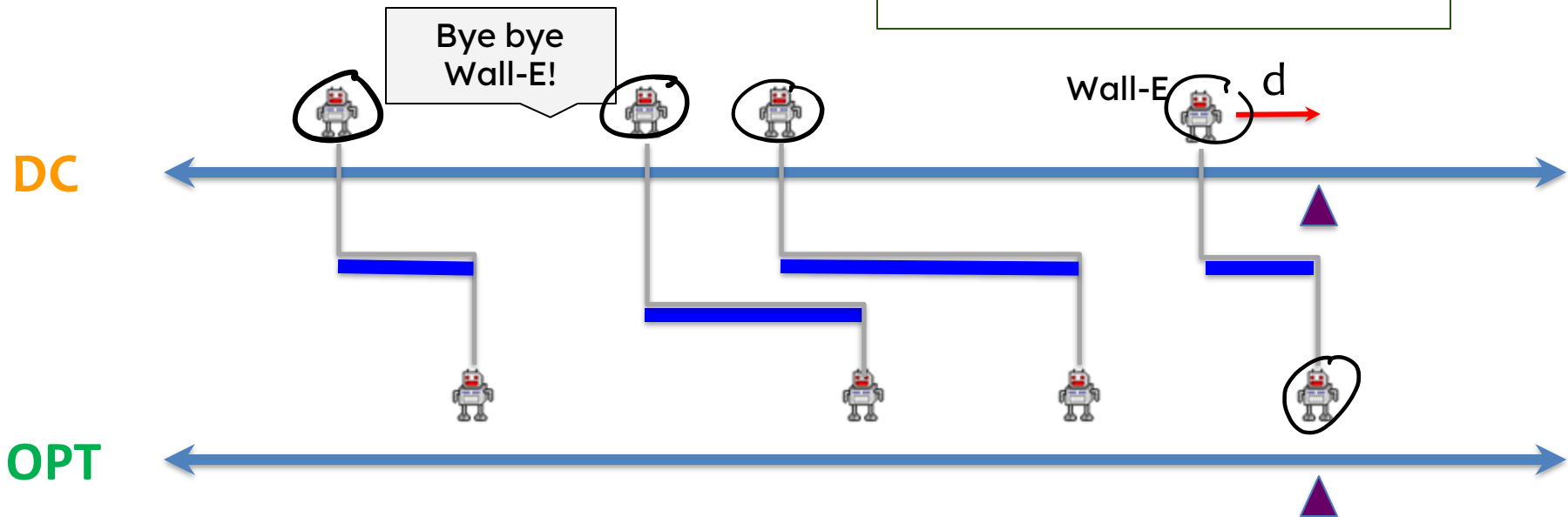
Potential function:  $S_i = k \cdot \frac{M}{\uparrow d} + \frac{\sum DC}{\uparrow 0}$

$$\leq kd$$

# Proving the Key Properties

Want to show: If **DC** moves its server(s) a total distance **d** then  **$S_i$**  decreases by  $\geq d$ .

Case 1: DC moves 1 robot



Potential function:  $S_i = \underbrace{k \cdot M}_{-d} + \underbrace{\sum_{DC}}_{+(k-i)d}$

$$-kd + (k-i)d$$

$$-d$$

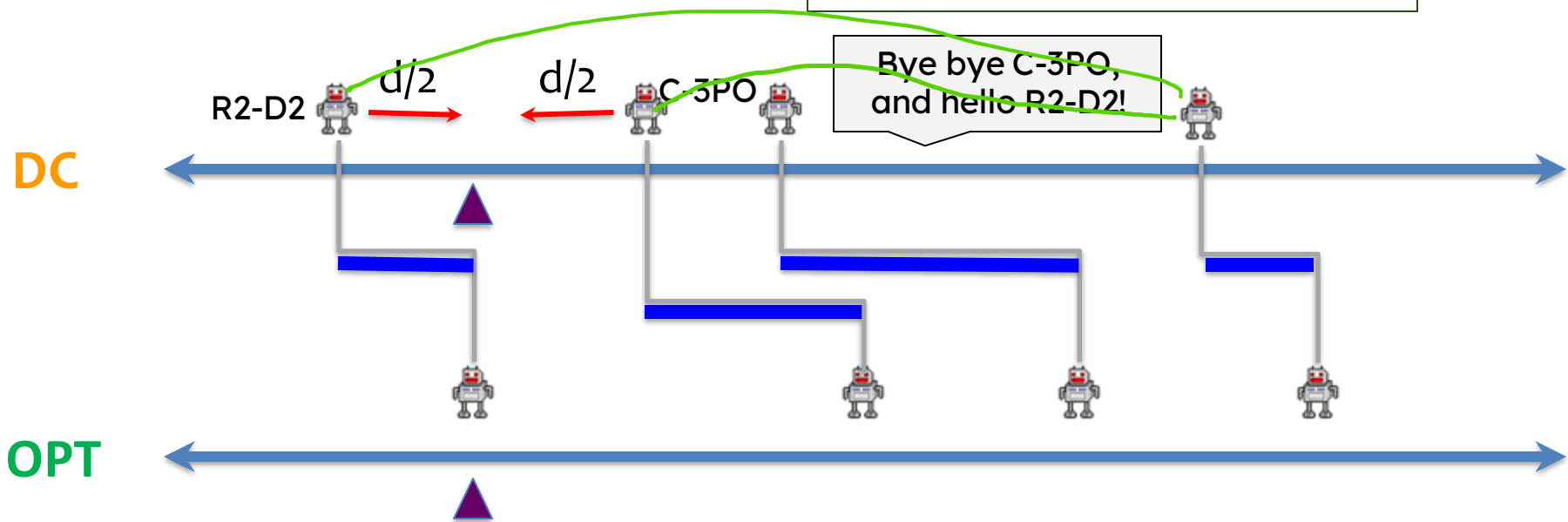


Note that OPT already has a robot at the request

# Proving the Key Properties

Want to show: If **DC** moves its server(s) a total distance **d** then  **$S_i$**  decreases by  $\geq d$ .

Case 2: DC moves 2 robots



First, how does  $\Sigma_{DC}$  change?

$-d$

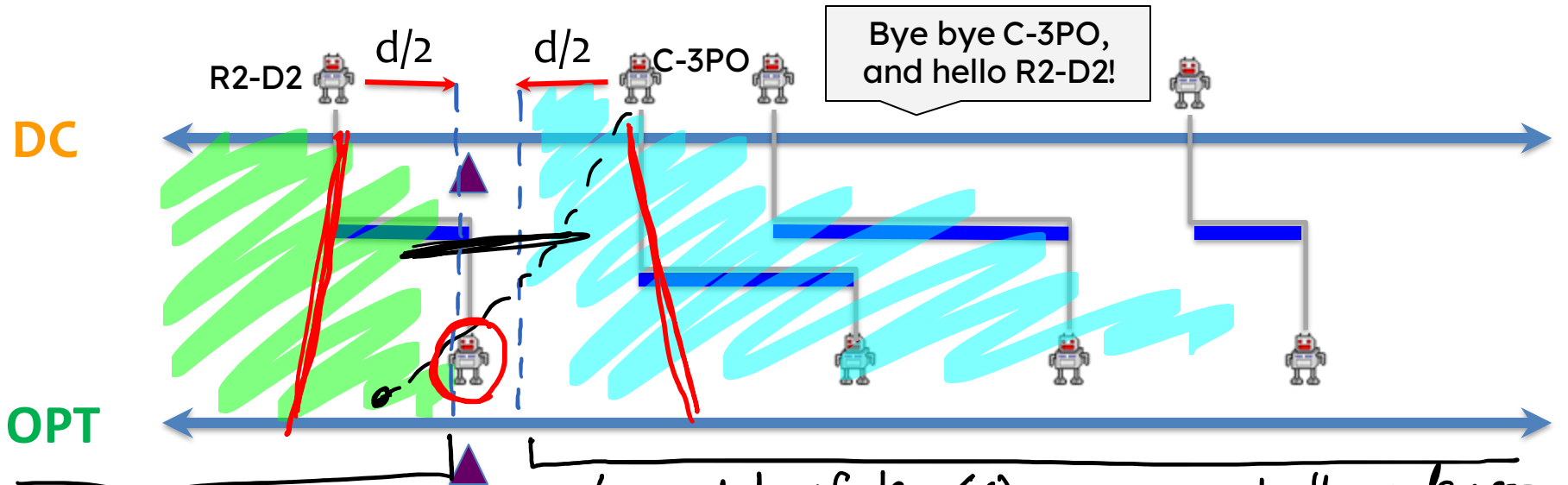


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# Proving the Key Properties

Want to show: If **DC** moves its server(s) a total distance **d** then **S<sub>i</sub>** decreases by  $\geq d$ .

**Case 2: DC moves 2 robots**



R's match if  $\log(R)$  increases at all or decreases by  $< \frac{d}{2}$

C's match if  $\log(C)$  increases at all or decreases by  $< \frac{d}{2}$   
Both can't happen!

At least one of  $\log(C)$  or  $\log(R)$  decrease by  $\geq \frac{d}{2}$



Note that OPT already has a robot at the request

Second, how does **M** change?  
**Claim: net same or dec**



# How we used our potential function

## Observations:

- $S_0 = 0$ .
- $S_i \geq 0$  for all  $i$ .

therefore  
 $\Rightarrow$

total amount  $S_i$  ever decreases  
 $\leq$   
total amount  $S_i$  ever increases

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## Consequence of key properties:

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For each request, we measure the change in potential right after OPT moves and then again after DC moves!

