
TuringMobile: A Turing Machine of Oblivious Mobile Robots with Limited Visibility and Its Applications

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Received: date / Accepted: date

Abstract In this paper we investigate the computational power of a set of mobile robots with limited visibility. At each iteration, a robot takes a snapshot of its surroundings, uses the snapshot to compute a destination point, and it moves toward its destination. Robots are punctiform and memoryless, they operate in \mathbb{R}^m , they have local reference systems independent of each other, and are activated asynchronously by an adversarial scheduler. Moreover, robots are non-rigid, in that they may be stopped by the scheduler at each move before reaching their destination (but are guaranteed to travel at least a fixed unknown distance before being stopped).

We show that despite these strong limitations, it is possible to arrange $3m + 3k$ of these weak entities in \mathbb{R}^m to simulate the behavior of a stronger robot that is rigid (i.e., it always reaches its destination) and is endowed with k registers of persistent memory, each of which can store a real number. We call this arrangement

a *TuringMobile*. In its simplest form, a TuringMobile consisting of only three robots can travel in the plane and store and update a single real number. We also prove that this task is impossible with fewer than three robots.

Among the applications of the TuringMobile, we focused on Near-Gathering (all robots have to gather in a small-enough disk) and Pattern Formation (of which Gathering is a special case) with limited visibility. Interestingly, our investigation implies that both problems are solvable in Euclidean spaces of any dimension, even if the visibility graph of the robots is initially disconnected, provided that a small amount of these robots are arranged to form a TuringMobile. In the special case of the plane, a basic TuringMobile of only three robots is sufficient.

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1 Introduction

Framework and background. The investigations of systems of autonomous mobile robots have long moved outside the boundaries of the engineering, control, and AI communities. Indeed, the computational and complexity issues arising in such systems are important research topics within theoretical computer science, especially in distributed computing. In these theoretical investigations, the robots are usually viewed as punctiform computational entities that live in a metric space, typically \mathbb{R}^2 or \mathbb{R}^3 , in which they can move. Each robot operates in “Look-Compute-Move” (LCM) cycles: it observes its surroundings, it computes a destination within the space based on what it sees, and it moves

toward the destination.¹ Each robot as its own local coordinate system. The only means of interaction between robots are observations and movements: that is, communication is *stigmatic*. The robots, identical and outwardly indistinguishable, are *oblivious*: when starting a new cycle, a robot has no memory of its activities (observations, computations, and moves) from previous cycles (“every time is the first time”). In other words, the robots have no persistent memory; for this reason, they are sometime said to be memoryless. Clearly obliviousness is a desirable property as it ensures a degree of self-stabilization and fault-tolerance into the system and its computations. Equally clear is that being memoryless severely constrains the computational capabilities of the robots.

There have been intensive research efforts on the computational issues arising with such robots, and an extensive literature has been produced in particular in regard to the important class of *Pattern Formation* problems [10, 20, 22, 23, 31, 33] as well as for *Gathering* [1, 2, 4, 9, 11–14, 16, 21, 28]; and *Scattering* [6, 24]; see also [8, 15, 34]. The goal of the research has been to understand the minimal assumptions needed for a team (or swarm) of such robots to solve a given problem, and to identify the impact that specific factors have on feasibility and hence computability.

The most important factor is the power of the adversarial scheduler that decides when each activity of each robot starts and when it ends. The main adversaries (or “environments”) considered in the literature are: *synchronous*, in which the computation cycles of all active robots are synchronized, and at each cycle either all (in the fully synchronous case) or a subset (in the semi-synchronous case) of the robots are activated, and *asynchronous*, where computation cycles are not synchronized, each activity can take a different and unpredictable amount of time, and each robot can be independently activated at each time instant. An important factor is whether a robot moving toward a computed destination is guaranteed to reach it (i.e., it is a *rigid* robot), or it can be stopped on the way (i.e., it is a *non-rigid* robot) at a point decided by an adversary. In all the above cases, the power of the adversaries is limited by some basic fairness assumption. All the existing investigations have concentrated on the study of (a-)synchrony, several on the impact of rigidity, some on other relevant factors such as agreement on local coordinate systems or on their orientation, etc.; for a review, see [19].

From a computational point of view, there is another crucial factor: the visibility range of the robots, that is, how much of the surrounding space they are able to observe in a Look operation. In this regard, two basic settings are considered: *unlimited visibility*, where the robots can see the entire space (and thus all other robots), and *limited visibility*, when the robots have a fixed visibility radius. In the latter case, it is customary to assume that all robots have the same visibility radius, which also implicitly provides a common unit distance. In Section 6 we will show how to drop this assumption.

While the investigations on (a-)synchrony and rigidity have concentrated on all aspects of those assumptions, this is not the case with respect to visibility. In fact, almost all research has assumed unlimited visibility; few exceptions are the algorithms for Convergence [4], Gathering [17, 18, 21], and Near-Gathering [28] when the visibility range of the robot is limited. The unlimited visibility assumption clearly greatly simplifies the computational universe under investigation; at the same time, it neglects the more general and realistic one, which is still largely unknown.

Let us also stress that, in the existing literature, all results on oblivious robots are for \mathbb{R}^1 and \mathbb{R}^2 ; the only exception is the recent result on plane formation in \mathbb{R}^3 by semi-synchronous rigid robots with unlimited visibility [34]. No results exist for robots in higher dimensions.

Contributions. In this paper we contribute several constructive insights on the computational universe of oblivious robots with limited visibility, especially asynchronous non-rigid ones, in any dimension.

The first and main contribution is a technique to construct a “moving Turing Machine” made solely of oblivious non-rigid robots in \mathbb{R}^m with limited visibility, for any $m \geq 2$. More precisely, we show how to arrange $3m + 3k$ identical non-rigid oblivious robots in \mathbb{R}^m with a visibility radius of $V + \varepsilon$ (for any $\varepsilon > 0$) and how to program them so that they can collectively behave as a single rigid robot in \mathbb{R}^m with k persistent registers and visibility radius V would. This team of identical robots is informally called a *TuringMobile*.

We obtain this result by using a fundamental construction called *basic component* of a TuringMobile, which is itself a TuringMobile that is able to move in \mathbb{R}^2 while storing and updating a single real number. Interestingly, we show that three robots are necessary and sufficient to build such a basic component. The full-fledged TuringMobile will then be built by arranging multiple copies of this basic component side by side.

¹ Another model of swarm robotics, alternative to the LCM one, is the “continuous” model. Some examples are found in [5, 7, 25–27].

We stress that the three robots forming a basic TuringMobile are *asynchronous*, that is, the scheduler makes them move at independent arbitrary speeds, and each robot takes the next snapshot an arbitrary amount of time after terminating each move; furthermore, they are *anonymous*, in that they are indistinguishable from each other, and they all execute the same program to compute their destination points. Notably, this program only performs arithmetic operations, square roots, and comparisons (hence no transcendental function has to be computed by the robots).

Even the basic component of a TuringMobile is a powerful construct that, once deployed in a swarm of robots, can act as a rigid leader with persistent memory, allowing the swarm to overcome many handicaps imposed by obliviousness, limited visibility, and asynchrony. As examples we present a variety of applications in \mathbb{R}^m , with $m \geq 2$.

First of all we show how a basic TuringMobile can explore and search the space. We then show how it can be employed to solve the long-standing open problem of (Near-)Gathering with limited visibility in spite of an asynchronous non-rigid scheduler and disagreement on the axes, a problem still open without a TuringMobile. Interestingly, the presence of the TuringMobile allows Gathering to be done even if the initial visibility graph is disconnected (this does not change the fact that there are cases in which Gathering is impossible, as remarked in [4,21]). Finally we show how the arbitrary Pattern Formation problem can be solved under the same conditions (asynchrony, limited visibility, possibly disconnected visibility graph, etc.).

There are limitations to the use of a TuringMobile when deployed in a swarm of robots: The TuringMobile must be always recognizable (e.g., by its unique shape) so that other robots cannot interfere by moving too close to the machine, disrupting its structure.

Also, when joining multiple basic components to form a TuringMobile that simulates a robot with more than one persistent register, we need to relax some of our asynchrony assumptions. Namely, we need a subset of the robots in the TuringMobile (one robot in each basic component) to always take snapshots at the same time, and thus have some form of synchrony.

The paper is organized as follows: In Section 2 we give formal definitions, introducing mobile robots with or without memory as *oracle semi-oblivious real RAMs*.² In Section 3 we illustrate our implementation of the TuringMobile, starting from the basic 3-robot compo-

nent, and then explaining how to join multiple components and how to work in higher dimensions. The correctness of the proposed construction is proved in Section 4. In Section 5 we show how to apply the TuringMobile to solve some fundamental problems for mobile robots with limited visibility. As a by-product of these applications, we also obtain a proof that two robots with limited visibility are insufficient to build a device that reliably stores and updates a real number and can freely move in the plane. This implies in particular that our basic three-robot TuringMobile design is optimal. In Section 6 we conclude the paper with some extra remarks and open problems.

2 Definitions and Preliminaries

2.1 Oracle Semi-Oblivious Real RAMs

Real random-access machines. A *real RAM* is a model of computation defining a machine that can operate on real numbers. That is, instead of just manipulating and storing integers, the machine can handle arbitrary real numbers and do infinite-precision operations on them. It has a finite set of internal *registers* and an infinite ordered sequence of *memory cells*; each register and each memory cell can hold a single real number, which the machine can modify by executing its program.³

A real RAM's instruction set contains at least the four arithmetic operations, but it may also contain k -th roots, trigonometric functions, exponentials, logarithms, and other analytic functions, depending on the application. The machine can also compare two real numbers and branch depending on which one is larger.

The initial contents of the memory cells are the *input* of the machine (we stipulate that only finitely many of them contain non-zero values), and their contents when the machine halts are its *output*. So, each program of a real RAM can be viewed as a partial function mapping tuples of reals into tuples of reals.

Remark 1 The real RAMs can at least compute the Turing-computable partial functions over the integers. Indeed, it is well known that all these functions can be computed by traditional RAMs whose programs only contain integer additions, subtractions, and comparisons. It is obvious that a real RAM running such a program on an integer-valued input behaves exactly as a traditional RAM, and therefore computes the same partial function.

² The *real RAM* was introduced by Shamos in [29, Chapter 2.3] (see also [30, Chapter 1.4]). It is a modified *RAM* (i.e., Random-Access Machine, see [3, Chapter 1.2]) that can natively operate on real numbers.

³ Nonetheless, the constant operands in a real RAM's program cannot be arbitrary real numbers, but have to be integers.

Oracles and semi-obliviousness. We introduce the *oracle semi-oblivious real RAM*, which is a real RAM with an additional “ASK” instruction. Whenever this instruction is executed, the contents of all the memory cells are replaced with new values, which are a function of the numbers stored in the registers.

In other words, the machine can query an external oracle by putting a question in its k registers in the form of k real numbers. The oracle then reads the question and writes the answer in the machine’s memory cells, erasing all pre-existing data. The term “semi-oblivious” comes from the fact that, every time the machine invokes the oracle, it “forgets” everything it knows, except for the contents of the registers, which are preserved.⁴

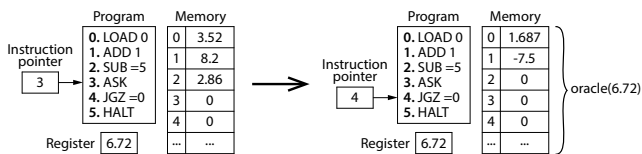


Fig. 1 An oracle semi-oblivious real RAM with a single register before and after executing an “ASK” instruction. The entire memory is overwritten by the oracle based on the number read from the register, which remains unaltered.

In spite of their semi-obliviousness, these real RAMs with oracles are at least as powerful as Turing Machines with oracles.

Theorem 1 *Given an oracle Turing Machine, there is an oracle semi-oblivious real RAM with one register that computes the same partial function.*

Proof Following van Melkebeek [32, Chapter 2.4.1], we define an oracle Turing Machine as a Turing Machine with an additional read-only tape containing the answers to all possible oracle queries. The i th cell of the oracle tape contains a symbol that is read by the machine whenever the head of the oracle tape is in position i .

Given such a machine M , we construct an oracle semi-oblivious real RAM with one register M' that “simulates” M step by step. As already observed, a real RAM can compute any Turing-computable partial function, and M' behaves as a real RAM as long as it does not invoke its oracle. So, M' can encode and decode the entire state of M , including the contents of its non-oracle tapes and the positions of its heads on the tapes,

⁴ Observe that, in general, the machine cannot salvage its memory by encoding its contents in the registers: since its instruction set has only analytic functions, it cannot injectively map a tuple of arbitrary real numbers into a single real number.

as a single integer: indeed, the functions that encode and decode a Turing Machine’s state are themselves Turing-computable.

To simulate one step of M , M' encodes the current state of M in its register and executes an “ASK” instruction. The oracle of M' reads the register, decodes the state of M , fetches the position of the head on the oracle tape, and answers with the symbol s that M would read on its oracle tape at that position. Next, M' finds s in the first cell of its own memory. So, M' decodes the contents of the register to retrieve the state of M , and uses it along with s to compute the next state of M . \square

2.2 Mobile Robots as Real RAMs

Mobile robots. Our oracle semi-oblivious real RAM model can be reinterpreted in the realm of *mobile robots*. A mobile robot is a computational entity, modeled as a geometric point, that lives in a metric space, typically \mathbb{R}^2 or \mathbb{R}^3 . It can observe its surroundings and move within the space based on what it sees. The same space may be populated by several mobile robots, each with its local coordinate system, and static objects.

To compute its next destination point, a mobile robot executes a real RAM program with input a representation of its local view of the space. After moving, its entire memory is erased, but the content of its k registers is preserved. Then it makes a new observation; from the observation data and the contents of the registers, it computes another destination point, and so on. If $k = 0$, the mobile robot is said to be *oblivious*. Note that robots have no notion of time or absolute positions.

The actual movement of a mobile robot is controlled by an external *scheduler*. The scheduler decides how fast the robot moves toward its destination point, and it may even interrupt its movement before the destination point is reached. If the movement is interrupted midway, the robot makes the next observation from there and computes a new destination point as usual. The robot is not notified that an interruption has occurred, but it may be able to infer it from its next observation and the contents of its registers. For fairness, the scheduler is only allowed to interrupt a robot after it has covered a distance of at least δ in the current movement, where δ is a positive constant unknown to the robots. This guarantees, for example, that if a robot keeps computing the same destination point, it will reach it in a finite number of iterations. If $\delta = \infty$, the robot always reaches its destination, and is said to be *rigid*.

Mobile robots, revisited. A mobile robot in \mathbb{R}^m with k registers can be modeled as an oracle semi-oblivious real RAM with $2m + k + 1$ registers as follows.

- m *position registers* hold the absolute coordinates of the robot in \mathbb{R}^m .
- m *destination registers* hold the destination point of the robot, expressed in its local coordinate system.
- 1 *timestamp register* contains the time of the robot’s last observation.
- k *true registers* correspond to the registers of the robot.

As the RAM’s execution starts, it ignores its input, erases all its registers, and executes an “ASK” instruction. The oracle then fills the RAM’s memory with the robot’s initial position p , the time t of its first observation, and a representation of the geometric entities and objects surrounding the robot, as seen from p at time t .

The RAM first copies p and t in its position registers and timestamp register, respectively. Then it executes the program of the mobile robot, using its true registers as the robot’s registers and adding $m + 1$ to all memory addresses. This effectively makes the RAM ignore the values of p and t , which indeed are not supposed to be known to the mobile robot.

When the robot’s program terminates, the RAM’s memory contains the output, which is the next destination point p' , expressed in the robot’s coordinate system. The RAM copies p' into its destination registers, and the execution jumps back to the initial “ASK” instruction.

Now the oracle reads p , p' , and t from the RAM’s registers (it ignores the true registers), converts p' in absolute coordinates (knowing p and the orientation of the local coordinate system of the robot) and replies with a new position p'' , a timestamp $t' > t$, and observation data representing a snapshot taken from p'' at time t' . To comply with the mobile robot model, p'' must be on the segment pp' , such that either $p'' = p'$ or $\overline{pp''} \geq \delta$. The execution then proceeds in the same fashion, indefinitely.

Note that in this setting the oracle represents the scheduler. The presence of a timestamp in the query allows the oracle to model dynamic environments in which several independent robots may be moving concurrently (without a timestamp, two observations from the same point of view would always be identical). Also note that in this formulation there are no actual robots moving through an environment in time, but only RAMs which query an oracle, which in turn provides a “virtual” environment and timeline by writing information in their memory.

Snapshots and limited visibility. In the mobile robot model we consider in this paper, an observation is simply an instantaneous *snapshot* of the environment taken from the robot’s position. In turn, each entity and object that the robot can see is modeled as a dimensionless point in \mathbb{R}^m . A mobile robot has a positive *visibility radius* V : it can see a point in \mathbb{R}^m if and only if it is located at distance at most V from its current position.⁵ If $V = \infty$, the robot is said to have *unlimited visibility*.

As we hinted at earlier in this section, a mobile robot has its own local reference system in which all the coordinates of the objects in its snapshots are expressed. The origin of a robot’s local coordinate system always coincides with the robot’s position (hence it follows the robot as it moves), and its orientation and handedness are decided by the scheduler (and remain fixed). Different mobile robots may have coordinate systems with a different orientation or handedness. (However, when two robots have the same visibility radius, they also implicitly have the same unit of distance.)

So, a snapshot is just a (finite) list of points, each of which is an m -tuple of real numbers.

Simulating memory and rigidity. The main contribution of this paper, loosely speaking, is a technique to turn non-rigid oblivious robots into rigid robots with persistent memory, under certain conditions. More precisely, if $3m + 3k$ identical non-rigid oblivious robots in \mathbb{R}^m with a visibility radius of $V + \varepsilon$ (for any $\varepsilon > 0$) are arranged in a specific pattern and execute a specific algorithm, they can collectively act in the same way as a single rigid robot in \mathbb{R}^m with $k > 0$ persistent registers and visibility radius V would. This team of identical robots is informally called a *TuringMobile*, from now on abbreviated as **tm**.

We stress that the robots of a **tm** are *asynchronous*, that is, the scheduler makes them move at independent arbitrary speeds, and each robot takes the next snapshot an arbitrary amount of time after terminating each move. The robots are also *anonymous*, in that they are indistinguishable from each other, and they all execute the same program.

Although our technique is fairly general and has a plethora of concrete applications (some are discussed in Section 5), a “perfect simulation” is achieved only under additional conditions on the scheduler or on the environment. These conditions will be discussed toward the end of Section 3.2.

⁵ Note, in particular, that robots do not obstruct each other’s line of sight: two robots are able to see each other whenever they are at distance at most V , even if there is a third robot between them.

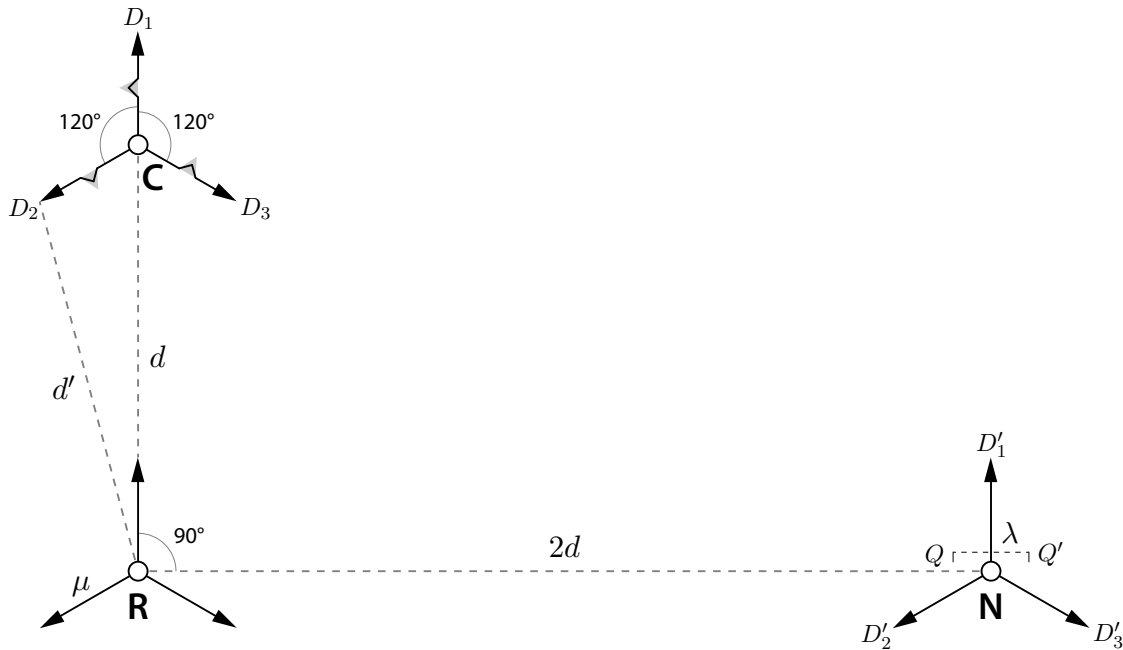


Fig. 2 Basic TuringMobile at rest, not drawn to scale (μ and λ should be smaller); here and in later figures, boldface letters indicate robots, normal uppercase letters indicate points, and lowercase letters indicate numbers or angles (with the exception of V in Figure 4, which is uppercase but indicates a visibility radius, i.e., a number)

3 Implementing the TuringMobile

3.1 Basic Implementation

We will first describe how to construct a basic version of the **tm** with just three oblivious non-rigid asynchronous robots in \mathbb{R}^2 . This **tm** can remember a single real number and rigidly move in the plane by fixed-length steps: its layout is sketched in Figure 2. In Section 3.2, we will show how to combine several copies of this basic machine to obtain a full-fledged **tm**.

It should be noted that in Section 5.2 we will prove that our basic **tm** design is minimal. That is, it is impossible for only two robots with limited visibility to implement a device that can remember a real number and rigidly move anywhere in the plane. Intuitively, this is because two anonymous robots always form a symmetric figure, and are unable to break symmetry. Thus, they can never agree on a direction of movement, and cannot get too far from their starting positions without losing vision of each other. By contrast, three robots can form a scalene triangle, which allows them to break symmetry and elect a leader, which in turn decides the direction of movement.

General layout. The elements of the basic **tm** are three robots: a *Commander* robot, a *Number* robot, and a *Reference* robot. We will denote as C , N , and R , respectively, the points where these robots are located. The

three robots have the same visibility radius of $V + \varepsilon$, where $\varepsilon \ll V$, and there is always a disk of radius ε containing all three of them. As a consequence, the three robots will always be able to see one another.

In the following, we will describe the behavior of the basic **tm** assuming it is “isolated”, i.e., no fourth robot is visible to any of its three members. In Sections 3.2 and 5, we will discuss when and how this assumption can be removed, and how the basic **tm** can be utilized in more general and interesting contexts.

Position at rest. When the machine is “at rest”, $\angle NRC$ is a right angle, the distance between C and R is some fixed value $d \ll \varepsilon$, and the distance between R and N is approximately $2d$. More precisely, N lies on a segment QQ' of length λ , where $\lambda \ll d$ is some fixed value, such that Q has distance $2d - \lambda/2$ from R and Q' has distance $2d + \lambda/2$ from R .

Representing numbers. The distance between the Reference robot and the Number robot when the **tm** is at rest is a representation of the real number r that the machine is currently storing. There are several possible ways of defining such a code: an easy one is to encode the number r as $\overline{RN} = \alpha(r) = 2d + \arctan(r) \cdot \lambda/\pi$ and to decode it as $r = \tan((\overline{RN} - 2d) \cdot \pi/\lambda)$. A different method that does not use transcendental functions is discussed in Section 6.

Algorithm 1 Basic TuringMobile in \mathbb{R}^2

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1: Identify Commander, Number, Reference (located in  $C$ ,  $N$ ,  $R$ , respectively)
2: if I am Commander then
3:   Compute Virtual Commander  $C'$  (based on  $R$  and  $N$ ) and points  $A_i$ ,  $S_i$ ,  $S'_i$ ,  $B_i$ ,  $D_i$ 
4:   if I am in  $C'$  then Choose final destination  $D_i$  and move to  $A_i$ 
5:   else if  $\exists i \in \{1, 2, 3\}$  s.t. I am on segment  $C'A_i$  but not in  $A_i$  then Move to  $A_i$ 
6:   else if  $\exists i \in \{1, 2, 3\}$  s.t. I am in  $A_i$  then
7:     Move to point  $P$  on segment  $S_iS'_i$  such that  $\overline{PS_i} = f(\overline{NQ})$ 
8:   else if  $\exists i \in \{1, 2, 3\}$  s.t. I am in triangle  $A_iS_iS'_i$  but not on segment  $S_iS'_i$  then
9:     Move to the intersection of segment  $S_iS'_i$  with the extension of line  $A_iC$ 
10:  else if  $\exists i \in \{1, 2, 3\}$  s.t. I am on  $S_iS'_i$  and  $\overline{NQ} = \overline{CS_i}$  then Move to  $B_i$ 
11:  else if  $\exists i \in \{1, 2, 3\}$  s.t. I am in triangle  $B_iS_iS'_i$  but not in  $B_i$  then Move to  $B_i$ 
12:  else if  $\exists i \in \{1, 2, 3\}$  s.t. I am on segment  $B_iD_i$  but not in  $D_i$  then Move to  $D_i$ 
13: else if I am Number then
14:   if  $\overline{CR} = d + \mu$  or  $\overline{CR} = d'$  then
15:     Compute Virtual Commander  $C'$  (based on  $C$  and  $R$ ) and points  $D'_i$ 
16:     if  $\overline{CR} = d + \mu$  and I am not in  $D'_1$  then Move to  $D'_1$ 
17:     else if  $\overline{CR} = d'$  and  $\angle NRC > 90^\circ$  and I am not in  $D'_2$  then Move to  $D'_2$ 
18:     else if  $\overline{CR} = d'$  and  $\angle NRC < 90^\circ$  and I am not in  $D'_3$  then Move to  $D'_3$ 
19:   else
20:     Compute Virtual Commander  $C'$  (based on  $R$  and  $N$ ) and points  $S_i$ ,  $S'_i$ 
21:     if  $\exists i \in \{1, 2, 3\}$  s.t.  $C$  is on segment  $S_iS'_i$  then
22:       Move to point  $P$  on segment  $QQ'$  such that  $\overline{PQ} = \overline{CS_i}$ 
23:   else if I am Reference then
24:     if Commander and Number are not tasked with moving (based on the above rules) then
25:        $\gamma$  = circle centered in  $C$  with radius  $d$ 
26:        $\gamma'$  = circle with diameter  $CN$ 
27:       Move to the intersection of  $\gamma$  and  $\gamma'$  closest to  $R$ 

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Movement directions. The Commander's role is to decide in which direction the machine should move next, and to initiate the movement. When the machine is at rest, the Commander may choose among three possible *final destinations*, labeled D_1 , D_2 , and D_3 in Figure 2. The segments CD_1 , CD_2 , and CD_3 all have the same length μ , with $\lambda \ll \mu \ll d$, and form angles of 120° with one another, in such a way that D_1 is collinear with R and C .

Around the center of each segment CD_i there is a *midway triangle* τ_i , drawn in gray in Figure 2. This is an isosceles triangle of height λ whose base lies on CD_i and has length λ as well. When the Commander decides that its final destination is D_i , it moves along the segment CD_i , but it takes a small detour in the midway triangle τ_i , as we will explain shortly.

Structure of the algorithm. Algorithm 1 is the program that each element of the basic **tm** executes every time it computes its next destination point.

Since the robots are anonymous, they first have to determine their roles, i.e., who is the Commander, etc. (line 1 of the algorithm). We make the assumption that there exists a disk of radius ε containing only the **tm** (close to its center) and no other robot. Using the fact that the two closest robots must be the Commander and the Reference robot and that the two farthest robots must be the Commander and the Number robot, it is

then easy to determine who is who (these properties will be preserved throughout the execution, as we will see in the next section).

Once it has determined its role, each robot executes a different branch of the algorithm (cf. lines 2, 13, and 23). The general idea is that, when the Commander realizes that the machine is in its rest position, it decides where to move next, i.e., it chooses a final destination D_i . This choice is based on the number r stored in the machine's "memory" (i.e., the number encoded by \overline{RN}), the relative positions of the visible robots external to the machine, and also on the application, i.e., the specific program that the **tm** is executing.

When the Commander has decided its final destination D_i , the entire machine moves by the vector $\overline{CD_i}$, and the Number robot also updates its distance from the Reference robot to represent a different real number r' . Again, this number is computed based on the number r the machine was previously representing, the relative positions of the visible robots external to the machine, and the specific program: in general, the new distance between N and Q is a function f of the old distance.

When this is done, the machine is in its rest position again, so the Commander chooses a new destination, and so on, indefinitely.

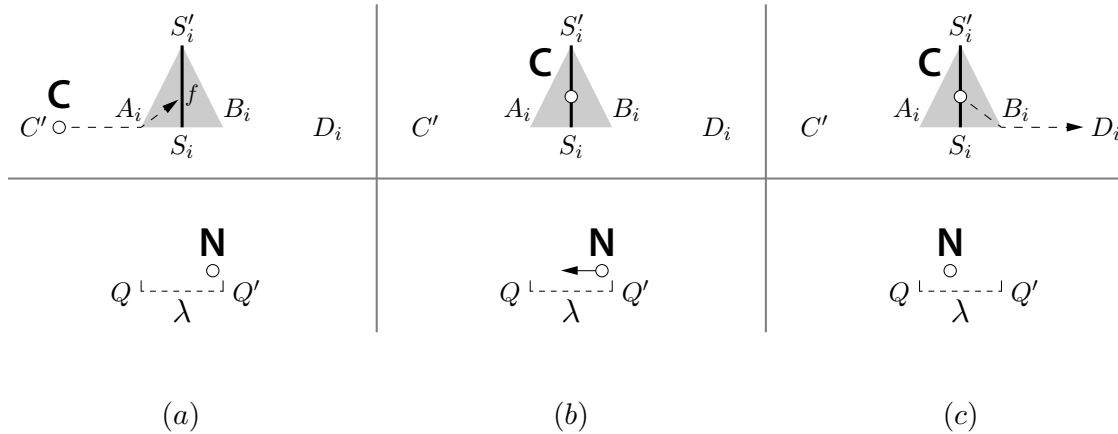


Fig. 3 Coordinated movement of the Commander and the Number robot, to cope with their asynchronous and non-rigid nature. (a) The Commander stops on $S_i S'_i$, recording the number that the machine is going to represent next (which is a function f of the number currently represented by the Number robot). (b) The Number robot moves within QQ' to match the Commander's position in $S_i S'_i$. (c) Finally, the Commander reaches D_i .

Coordinating movements. Note that it is not possible for all three robots to translate by $\overrightarrow{CD_i}$ at the same time, because they are non-rigid and asynchronous. If the scheduler stops them at arbitrary points during their movement, after the structure of the machine has been destroyed, they will be incapable of recovering all the information they need to resume their movement (recall that they are oblivious and they have to compute a destination point from scratch every time).

To prevent this, the robots employ various coordination techniques. First the Commander moves to the middle triangle τ_i , and precisely to its base vertex A_i , as shown in Figure 3(a) (cf. line 5 of Algorithm 1). Then it positions itself on the altitude $S_i S'_i$, in such a way as to indicate the new number r' that the machine is supposed to represent. That is, the Commander picks the point on $S_i S'_i$ at distance $f(\overline{NQ})$ from S_i (lines 6 and 7). Even if it is stopped by the scheduler before reaching such a point, it can recover its destination by simply drawing a ray from A_i to its current position and intersecting it with $S_i S'_i$ (lines 8 and 9).

When the Commander has reached $S_i S'_i$, it waits to let the Number robot adjust its position on the segment QQ' to match that of the Commander on $S_i S'_i$, as in Figure 3(b) (lines 21 and 22). This effectively makes the Number robot represent the new number r' . Note that the Number robot can do this even if it is stopped by the scheduler several times during its march, because the Commander keeps reminding it of the correct value of r' : since r' depends on the old number r , the Number robot would be unable to re-compute r' after it has forgotten r .

Once the Number robot has reached the correct position on QQ' , the Commander starts moving again

(line 10) and finally reaches D_i while the other robots wait, as in Figure 3(c) (lines 11 and 12).

When the Commander has reached D_i , the Number robot realizes it and makes the corresponding move (lines 14–18) while the other two robots wait. The destination point of the Number robot is D'_i , as shown in Figure 2. Finally, when the Number robot is in D'_i , the Reference robot realizes it and makes the final move to bring the **tm** back into a rest position (lines 23–27).

Note that the number r' stored in the machine is not erased after these final movements, because both the Number and Reference robot move by the same vector. As a consequence of this, the basic **tm** is able to retain memory of the real number it is storing, in spite of the obliviousness of the individual robots that compose it.

Computing the Virtual Commander. After the Commander has left its rest position and is on its journey to D_i , the **tm** loses its initial shape, and identifying the D_i 's and the midway triangles becomes a non-trivial task. To simplify this task, the robots try to guess where the Commander's original rest position may have been by computing a point called the *Virtual Commander* C' .

Assuming that the Reference and Number robots have not moved from their rest positions, the Virtual Commander is easily computed: draw a line ℓ through R perpendicular to RN ; then, C' is the point on ℓ at distance d from R that is closest to C . Once we have C' , we can construct the points D_i with respect to C' (in the same way as we did in Figure 2 with respect to C). This technique is used by Algorithm 1 at lines 3 and 20.

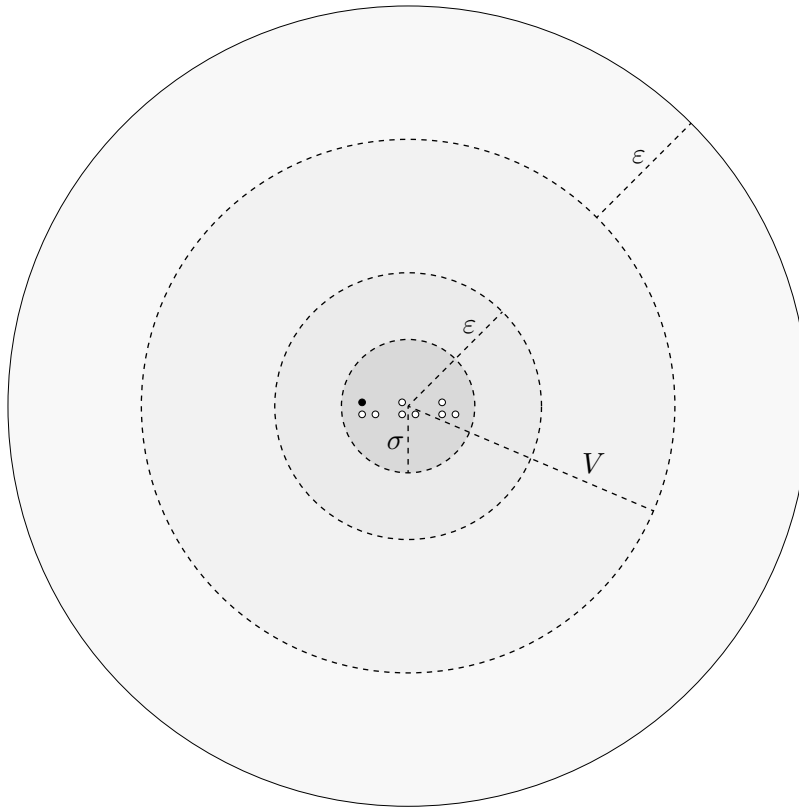


Fig. 4 Sketch of a complete TuringMobile, not drawn to scale (ϵ and σ should be smaller). All robots are in the central disk of radius σ ; the one drawn in black is the Leader

In the special case where the Commander has reached its final destination D_i and the Reference robot has not moved from its rest position (but perhaps the Number robot has moved), the Virtual Commander can also be computed. This situation is recognized because the distance between the Commander and the Reference robot is either maximum (i.e., $d + \mu$) or minimum (i.e., $d' = \sqrt{d^2 + \mu^2 - d\mu}$, by the law of cosines), as Figure 2 shows. If the distance is maximum, then C must coincide with D_1 ; otherwise, C coincides with D_2 (if the angle $\angle NRC$ is obtuse) or D_3 (if the angle $\angle NRC$ is acute). Since we know the position of R and one of the D_i 's, it is then easy to determine the other D_i 's. This technique is used at line 15.

The Reference robot's behavior. To know when it has to start moving, the Reference robot simply executes Algorithm 1 from the perspective of the Commander and the Number robot: if neither of them is supposed to move, then the Reference robot starts moving (line 24).

We have seen that the Number robot can determine its destination D_i' solely by looking at the positions of C and R , which remain fixed as it moves. For the Reference robot the destination point is not as easy to de-

termine, because the distance between C and N varies depending on what number is stored in the **tm**.

However, the Reference robot knows that its move must put the **tm** in a rest position. The condition for this to happen is that its destination point be at distance d from C (line 25) and form a right angle with C and N (line 26). There are exactly two such points in the plane, but one of them has distance much greater than μ from R , and hence the Reference robot will pick the other (line 27).

As the Reference robot moves toward such a point, all the above conditions must be preserved, due to the asynchronous and non-rigid nature of the robots. This is not a trivial requirement, and we will prove that it is indeed fulfilled in Section 4.

3.2 Complete Implementation

We have shown how to implement a basic component of the **tm** in \mathbb{R}^2 consisting of three robots: a Commander, a Number, and a Reference. The basic component is able to rigidly move by a fixed distance μ in three fixed directions, 120° apart from one another. It can also store and update a single real number.

Planar layout. We can obtain a full-fledged **tm** in \mathbb{R}^2 by putting several tiny copies of the basic component side by side as in Figure 4.

For the machine to work, we stipulate that there exists a disk of radius σ that contains all the robots constituting the **tm** and no extraneous robot, where $\sigma \ll \varepsilon$. The distance between two consecutive basic components of the **tm** is roughly s , where $d \ll s \ll \sigma$. This makes it easy for the robots to tell the basic components apart and determine the role of each robot within its basic component.

Since a basic component of the **tm** is a scalene triangle, which is chiral, all its members implicitly agree on a clockwise direction even if they have different handedness. Similarly, all robots can agree that the vector from the Number robot of any component to the Reference robot of the same component points in the “left” direction (in spite of the fact that robots, by themselves, do not have a common frame of reference). Hence, all robots in the **tm** can agree on a “leftmost” basic component, whose Commander is said to be the *Leader* of the whole machine.

Coordinated movements. All the basic components of the **tm** are always supposed to agree on their next move and proceed in a roughly synchronous way. To achieve this, when all the basic components are in a rest position, the Leader decides the next direction among the three possible, and executes line 4 of Algorithm 1. Then all the other Commanders see where the Leader is going, and copy its movement.

When all the Commanders are in their respective A_i 's, they execute line 7 of the algorithm, and so on. At any time, each robot executes a line of the algorithm only if all its homologous robots in the other basic components of the **tm** are ready to execute that line or have already executed it; otherwise, it waits.

When the last Reference robot has completed its movement, the machine is in a rest position again, and the coordinated execution repeats with the Leader choosing another direction, etc.

Simulating a non-oblivious rigid robot. Let a program for a rigid robot \mathcal{R} in \mathbb{R}^2 with k persistent registers and visibility radius V be given. We want the **tm** described above to act as \mathcal{R} , even though its constituting robots are non-rigid and oblivious.

Our **tm** consists of $2 + k$ basic components, each dedicated to memorizing and updating one real number. These $2 + k$ numbers are the x coordinate and the y coordinate of the destination point of \mathcal{R} and the contents of the k registers of \mathcal{R} . We will call the first two numbers the *x variable* and the *y variable*, respectively.

When the **tm** is in a rest position, its x and y variables represent the coordinates of the destination point of \mathcal{R} relative to the Leader of the machine. Whenever the **tm** moves by μ in some direction, these values are updated by subtracting the components of an appropriate vector of length μ from them. Of course, this update is computed by the Commanders of the first two basic components of the machine, which communicate it to their respective Number robots, as explained in Section 3.1.

Let P be the destination point of \mathcal{R} . Since the **tm** can only move by vectors of length μ in three possible directions, it may be unable to reach P exactly. So, the Leader always plans the next move trying to reduce its distance from P until this distance is at most 2σ (this is possible because $\mu \ll d \ll \sigma$).

When the Leader is close enough to P , it “pretends” to be in P , and the **tm** executes the program of \mathcal{R} to compute the next destination point. Recall that the visibility radius of \mathcal{R} is V , and that of the robots of the **tm** is $V + \varepsilon$. Since $\sigma \ll \varepsilon$, each member of the **tm** can therefore see everything that would be visible to \mathcal{R} if it were in P , and compute the output of the program of \mathcal{R} independently of the other members. The only thing it should do when it executes the program of \mathcal{R} is subtract the values of the x and y variables to everything it sees in its snapshot, discard whatever has distance greater than V from the center of the snapshot, and of course discard the robots of the **tm** and replace them with a single robot in the center of the snapshot (representing the robot itself). Then, the robots that are responsible for updating the x and y variables add the relative coordinates of the new destination point of \mathcal{R} to these variables. Similarly, the robots responsible for updating the k registers of \mathcal{R} do so.

Note that the above simulation works also in the special case where \mathcal{R} is supposed to update its registers without moving. The Leader will move by μ in any direction, followed by the entire machine (because this is the only way the **tm** can update its registers), and the x and y variables will be updated with the old position of the Leader.

Restrictions. The above **tm** correctly simulates \mathcal{R} under certain conditions. The first one is that, if all robots are indistinguishable, then no robot extraneous to the **tm** may get too close to it (say, within a distance of σ of any of its members). This kind of restriction cannot be dispensed with: whatever strategy a team of oblivious robots employs to simulate a single non-oblivious robot's behavior is bound to fail if extraneous robots join the team creating ambiguities between its members. Nevertheless, the restriction can be removed if we

stipulate that the members of a **tm** are distinguishable from all other robots.

Another difficulty comes from the fact that, if the **tm** is made of more than one basic component and its Commanders are all in their respective A_i 's and ready to update the values represented by the machine, they may get their snapshots at different times, due to asynchrony. If the environment moves in the meantime, the snapshots they get are different, and this may cause the machine to compute an incorrect destination point or put inconsistent values in its simulated registers.

There are several possible solutions to this problem: we will only mention two trivial ones. We could for instance assume the Commanders to be *synchronous*, that is, make the scheduler activate them in such a way that all of them take their snapshots at the same time. This way, all Commanders get compatible snapshots and compute consistent outputs. Another possible solution is to make the **tm** operate in an environment where everything else is static, i.e., no moving entities are present other than the **tm**'s members.

We stress that these restrictions make sense if a perfect simulation of \mathcal{R} is sought. However, as we will see in Section 5, there are several applications of the **tm** technique where no such restrictions are required.

Higher dimensions. Let us now generalize the above construction of a planar **tm** to \mathbb{R}^m , for any $m \geq 2$. We start with the same **tm** \mathcal{M} with $2+k$ basic components laid out on a plane $\gamma \subset \mathbb{R}^m$. Since \mathcal{M} has only two basic components for the x and y variables, we will add $m-2$ basic components to it, positioned as follows.

Let vectors v_1 and v_2 be two orthonormal generators of γ , and let us complete $\{v_1, v_2\}$ to an orthonormal basis $\{v_1, v_2, \dots, v_m\}$ of \mathbb{R}^m . Now, for all $i \in \{3, 4, \dots, m\}$, we make a copy of the basic component of \mathcal{M} containing the Leader, we translate it by $s \cdot v_i \cdot i/m$, and we add it to the **tm** (s is the same value used in the construction of the planar **tm** at the beginning of Section 3.2). Note that the Leader of this new **tm** \mathcal{M}' is still easy to identify, as well as the plane γ when \mathcal{M}' is at rest.

Clearly, m basic components allow the machine to record a destination point in \mathbb{R}^m , as opposed to \mathbb{R}^2 . Additionally, the positions of the basic components with respect to γ give the machine an m -dimensional sense of direction.

For instance, say that $m = 3$, γ is a horizontal plane, and v_3 points upward. Then, when the Leader decides to move up, it moves by μ in the direction of the basic component of the **tm** not lying on γ (first stopping in a midway triangle, as per Algorithm 1). The rest of \mathcal{M}' can reconstruct the direction of v_3 , for instance by inspecting the relative positions of the Reference robots,

and move as required when the time comes. In the subsequent moves, the Leader still retains a consistent notion of up and down, and can therefore lead \mathcal{M}' close enough to the destination point.

The same restrictions that apply to the planar **tm** as a simulator of course extend to its higher-dimensional versions. The next section will be devoted to proving the following theorem, which summarizes the results obtained so far.

Theorem 2 *Under the aforementioned restrictions, a rigid robot in \mathbb{R}^m with k persistent registers and visibility radius V can be simulated by a team of $3m + 3k$ non-rigid oblivious robots in \mathbb{R}^m with visibility radius $V + \varepsilon$.*

4 Correctness

This section is devoted to the proof of Theorem 2. The crux of the proof is the following lemma, which states that a single basic component of the TuringMobile, as described in Section 3.1, works as intended.

The fundamental lemma. To state the next lemma, the following definition will be useful: A robot is *frozen* at time t if it has already reached the last destination point that it has computed before time t , and it has not taken the next snapshot before time t (although it may be taking the snapshot exactly at time t).

Lemma 1 *Let a **tm** in \mathbb{R}^2 consisting of a single basic component execute Algorithm 1, and assume that throughout the execution no object extraneous to the machine approaches any of its members by less than σ . If at some point in time t the **tm** is in a rest position and all of its members are frozen, then, at a point in time $t' > t$, the **tm** is in a rest position again, its Commander and Reference robot have translated by a vector of length μ in one of three predefined directions (as in Figure 2), its Number robot has correctly updated its distance from the Reference robot (according to some function f of the previous distance and the **tm**'s surrounding environment as observed by the Commander in a single snapshot taken between times t and t'), and all members of the **tm** are frozen.*

Note that for Lemma 1 we do not make all the restrictions of Theorem 2, because we do not have to synchronize several basic components of the machine.

If Lemma 1 holds, then a **tm** in \mathbb{R}^2 with only one basic component correctly performs a single step of the execution, rigidly moving by μ and updating the real number that it is storing. By repeatedly applying this

lemma, we have the correctness of the entire execution of a basic component.

Then, the proof of Theorem 2 follows as expected, because the coordination of several basic components in \mathbb{R}^m for $m \geq 2$ is done as described in Section 3.2.

Proof structure. Let us prove Lemma 1. The intended behavior of the machine is for the execution to go through the following five phases in chronological order:

1. The Commander moves to $S_i S'_i$ (lines 4–9 of Algorithm 1);
2. The Number robot moves within QQ' (lines 21 and 22);
3. The Commander moves to D_i (lines 10–12);
4. The Number robot moves to D'_i (lines 14–18);
5. The Reference robot moves, bringing the machine in a rest position (lines 24–27).

During each phase, only one robot is supposed to move, while the other two wait. If we can ensure this behavior, then Lemma 1 follows.

Recall that the robots constituting the **tm** are asynchronous and non-rigid. This means that we have to guarantee two things for each of the above phases:

- If a robot moves as per phase i and another robot sees it at any time before it has finished (due to asynchrony), the second robot correctly recognizes the current phase as being i , and hence it waits.
- If a robot moves as per phase i and the scheduler stops it before it has reached its destination (due to non-rigidity), the robot takes another snapshot, and correctly resumes phase i .

Phase 1. If the assumptions of Lemma 1 are satisfied, the first robot to take a snapshot after time t (or exactly at time t) will see a **tm** in a rest position. As a consequence, the Virtual Commander coincides with the Commander, and therefore only the Commander is allowed to move toward some A_i .

While the Commander moves, its distance from the Reference robot never gets as small as d' or as large as $d + \mu$ (cf. Figure 2), hence the conditions of line 14 of Algorithm 1 are never satisfied. Also, the Virtual Commander computed with respect to R and N always coincides with the starting position C' of the Commander, which means that the Commander will be seen on the segment $C'A_i$, implying that only the Commander will be allowed to move.

Since the Commander approaches A_i by at least δ at every movement (cf. Section 2.2), it eventually reaches it. When it reaches A_i , it chooses a destination point on $S_i S'_i$ based on a single snapshot of the environment (as required by Lemma 1): once a destination point has been chosen, it never changes even if the Commander is

stopped before reaching it, due to lines 8 and 9. Since the Number robot and the Reference robot have not moved yet, the number stored in the machine is still the same as it was a time t , and therefore the point on $S_i S'_i$ chosen by the Commander is correctly computed by applying function f to \overline{QN} . Again, only the Commander is allowed to move, and it eventually reaches $S_i S'_i$, for the same reasons as before.

Phase 2. When the Commander is on $S_i S'_i$, it waits until the Number robot has a distance from Q of $\overline{CS_i}$. Observe that, as the Number robot moves within QQ' , the slope of the line RN does not change, and therefore the Virtual Commander C' computed with respect to R and N is always the position that C occupied at time t . So, the point S_i is always the same, and the Number robot keeps consistently moving toward the same destination point on QQ' .

As for phase 1, \overline{CR} never becomes as small as d' or as large as $d + \mu$, and therefore the Number robot always executes line 21 until it reaches the correct point on QQ' .

Phase 3. When $\overline{NQ} = \overline{CS_i}$, the Commander knows it has to start moving again, first to B_i and then to D_i , due to lines 10–12. Again, while this happens the Virtual Commander computed based on R and N is always the same point C' , so the positions of S_i , B_i , D_i , etc. remain consistent, and the distance between R and C never gets as small as d' or as large as $d + \mu$ until the Commander has reached D_i . In particular the Number robot never sees the Commander on $S_i S'_i$ after it has left it, and so it does not move. Eventually, the Commander reaches D_i .

Phase 4. When the Commander reaches D_i , its distance from R finally becomes $d + \mu$ (if $i = 1$) or d' (if $i = 2$ or $i = 3$), and so the Number robot executes lines 15–18 and starts moving. While the Number robot moves, the Commander does not: indeed, as long as the Number robot is tasked with moving, the Reference robot never moves (cf. line 24), and hence \overline{CR} remains the same. Therefore, if the Commander computes a Virtual Commander \tilde{C} based on N and R , and then computes the points D_i and the midway triangles τ_i with respect to \tilde{C} , it will never believe to be in \tilde{C} or in the interior of the segment $\tilde{C}D_i$ or in τ_i , no matter where N is. This is because all such points have distance greater than d' and smaller than $d + \mu$ from R (cf. Figure 2). So, the conditions of lines 5–12 are never satisfied.

Suppose that the Commander is in D_1 . This configuration is correctly identified by the Number robot no

matter how it moves, because it is the only one in which $\overline{CR} = d + \mu$. So the Number robot computes the point D'_1 correctly (it does so only based on C and R) and keeps moving toward D'_1 until it reaches it (line 16).

Suppose now that the Commander is in D_2 . The Number robot recognizes this configuration because $\overline{CR} = d'$ and $\angle NRC > 90^\circ$. Again, it correctly computes D'_2 and moves toward it (line 17). As the Number robot moves, the angle $\angle NRC$ grows (cf. Figure 2), and so the condition of line 17 keeps being satisfied. Eventually, the Number robot reaches D'_2 .

Finally, suppose that the Commander robot is in D_3 . Now $\overline{CR} = d'$ and $\angle NRC < 90^\circ$, and so the Number robot starts moving toward D'_3 (cf. Figure 2). We have to prove that, if it stops on its way to D'_3 and gets a new snapshot, the inequality $\angle NRC < 90^\circ$ keeps being satisfied, and so the Number robot re-computes D'_3 as its destination point, until it reaches it. This is not trivial, since the angle $\angle NRC$ grows as N approaches D'_3 .

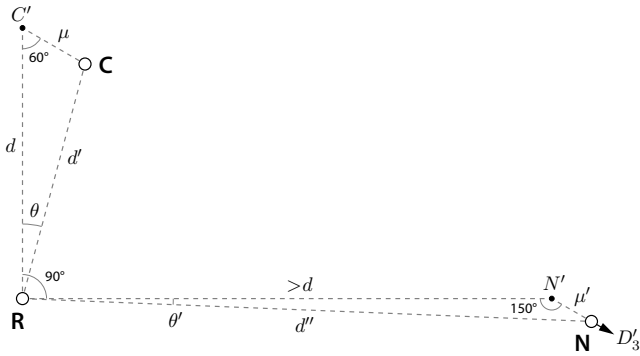


Fig. 5 As N moves toward D'_3 , we have $\theta > \theta'$, and hence $\angle NRC < 90^\circ$

The situation is illustrated in Figure 5, where C' represents the starting position of the Commander and N' the starting position of the Number robot. Since $\angle N'RC' = 90^\circ$, proving that $\angle NRC < 90^\circ$ is equivalent to proving that $\theta > \theta'$ (where $\theta' = \angle NRN'$).

By the law of sines applied to triangle RCC' ,

$$\frac{d'}{\sin 60^\circ} = \frac{\mu}{\sin \theta},$$

implying that

$$\sin \theta = \frac{\sqrt{3}\mu}{2d'}. \quad (1)$$

Again for the law of sines applied to triangle RNN' ,

$$\frac{d''}{\sin 150^\circ} = \frac{\mu'}{\sin \theta'},$$

where $\mu' = \overline{N'N}$. Hence

$$\sin \theta' = \frac{\mu'}{2d''}. \quad (2)$$

Observe that $\mu > \mu'$, because N lies on $N'D'_3$, and therefore $\sqrt{3}\mu > \mu'$ (i.e., the numerator of (1) is greater than that of (2)). Recall that $\mu \ll d$, and so $d' < d$. Moreover, since $\overline{RN'} \geq 2d - \lambda/2$ (cf. Figure 2) and $\lambda \ll d$, it follows that $d < \overline{RN'}$. Also observe that $\overline{RN'} < d''$, from which we obtain that $2d' < 2d''$ (i.e., the denominator of (1) is smaller than that of (2)). As a consequence, $\sin \theta > \sin \theta'$. Since $\mu \ll d$, both θ and θ' are acute, which means that $\theta > \theta'$ (the function $\sin x$ increases monotonically when $x \in [0, 90^\circ]$).

Phase 5. Since in the previous phases either the Commander or the Number robot was always tasked with moving, the condition of line 24 was never satisfied, and hence the Reference robot never moved. Now that the Commander is in D_i and the Number robot is in D'_i , they are no longer tasked with moving, and so the Reference robot executes lines 25–27.

Ideally, the Reference robot should complete the translation of the **tm** in order to put it in a rest position again. This is achieved by moving by vector $\overline{C''C}$, where C'' is the initial position of the Commander (i.e., its position when phase 1 starts). Instead of trying to reconstruct C'' , the Reference robot constructs two circumferences γ and γ' and moves to their nearest intersection point. Note that γ' passes through the center of γ , and hence it has at most two intersection points with it. At least one intersection point exists: this is the point $P = R' + \overline{C''C}$, where R' is the initial position of the Reference robot (which coincides with its position when phase 5 starts). If there is another intersection point P' between the two circles, it must be symmetric to P with respect to line CN , because such line passes through the centers of both circles (recall that the segment CN is a diameter of γ'). So, assuming that the Commander and the Number robot do not move in this phase, P remains the destination point of the Reference robot as long as the robot never crosses the line CN . But this is impossible, since the segment $R'P$ has length $\mu \ll d$, and therefore it cannot cross the line CN , whose distance from R' is roughly $2d/\sqrt{5} \gg \mu$.

It remains to prove that, as the Reference robot moves toward P , the Commander and the Number robot remain still. Recall that, when phase 5 starts, either $\overline{CR} = d + \mu > d$ or $\overline{CR} = d' < d$. As R approaches P (and C does not move), \overline{CR} converges monotonically to d . It follows that \overline{CR} never becomes $d + \mu$ or d' again, and so the condition of line 14 is never satisfied.

Consider now the Virtual Commander C' computed with respect to R and N when R is strictly between R' and P , and construct the three segments $C'D_i$ and the three midway triangles τ_i around C' . If we can prove that, no matter where R is located in the interior of the segment $R'P$, C never lies on any of these segments and triangles, we are finished: indeed, this would mean that the conditions of lines 5–12 and line 21 are never satisfied.

Suppose that the Number robot has moved to D'_1 during phase 4, which means that at the beginning of phase 5 we have $\overline{CR} = d + \mu$: this case is illustrated in Figure 6. We have to show that C does not lie on any of the solid gray lines and triangles around the Virtual Commander C' . It is obvious that the lines $C''C$ and RC' are not parallel and intersect each other at R . Also, C' and N are always on opposite sides of RC' . This already implies that C cannot be on the segments $C'D_1$ and $C'D_2$ or on their respective midway triangles.

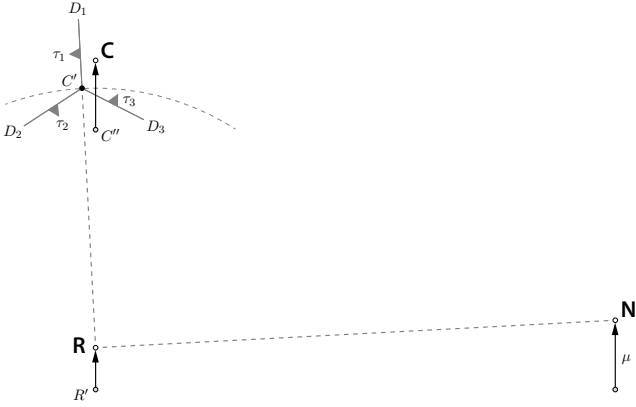


Fig. 6 Correctness of phase 5 when the Number robot has moved to D'_1

To show that C does not lie on $C'D_3$ or τ_3 , consider the circle through C' centered at R . Note that C is always outside the circle, because its radius is d , but $d < \overline{RC} < d + \mu$. Since μ can be arbitrarily small compared to d , the angle between $C''C$ and RC' can be made arbitrarily small, as well (cf. Figure 6). If we take a small-enough μ , the segment $C'D_3$ is entirely contained in the circle, and hence it cannot contain C . Moreover, since τ_3 has height $\lambda \ll \mu$, by taking a small-enough λ we ensure that τ_3 is contained in the circle, too.

Suppose now that the Number robot has moved to D'_2 during phase 4: then, at the beginning of phase 5, $\overline{CR} = d'$ and $\angle NRC > 90^\circ$. On the other hand, when R reaches P , we have $\overline{CR} = d$ and $\angle NRC = 90^\circ$. It follows that, when R is strictly between R' and P , $d' < \overline{CR} < d$ and $\angle NRC > 90^\circ$ (because both quan-

ties change monotonically), as Figure 7 shows. Since $\angle NRC > 90^\circ$, C cannot be located on $C'D_1$ or $C'D_3$ or τ_3 , because all their points X satisfy $\angle NRX \leq 90^\circ$. Also, all the points in τ_1 have distance at least $d + \mu/2 - \lambda/2 > d$ from R (recall that $\lambda \ll \mu$), and so C cannot lie in τ_1 , because $\overline{CR} < d$.

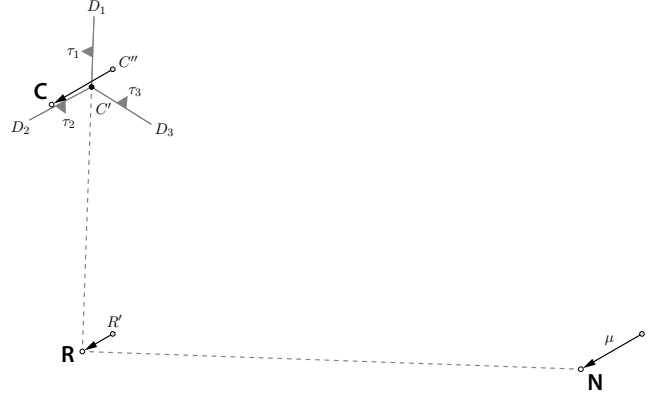


Fig. 7 Correctness of phase 5 when the Number robot has moved to D'_2

Let us show that C does not lie on $C'D_2$ or τ_2 . Observe that $\overline{R'C''} = \overline{RC'} = d$, that $\angle RC'D_2 = \angle R'C''C = 60^\circ$, and that $R'R$ and $C''C$ are parallel (cf. Figure 7). It follows that the line $C''C$ is obtained by rotating line $C'D_2$ about R by some angle $\theta > 0$. These two lines are not parallel, and hence they intersect in a single point K . As R approaches P , θ tends monotonically to 0, and K approaches the foot U of the altitude from R to the line $C''C$. So, if we take μ small enough with respect to d , we can keep K as close as we want to U . The distance between U and C'' is obviously minimum when $R = R'$, in which case $\overline{C''U} = d/2$. It follows that, for small-enough values of μ , $\overline{C''K}$ is always as close as we want to $d/2$. Hence we have $\overline{C''K} > \mu = \overline{C''C}$, proving that $C \neq K$, and so C cannot be on $C'D_2$. Also, since $\angle NRC > 90^\circ$, C and τ_2 are always on opposite sides of $C'D_2$ (cf. Figure 7), and so C cannot be in τ_2 .

Lastly, suppose that the Number robot has moved to D'_3 during phase 4: then, at the beginning of phase 5, $\overline{CR} = d'$ and $\angle NRC < 90^\circ$, as shown in Figure 8. Similarly to the previous case, we can prove that C cannot lie on $C'D_3$ or τ_3 because the line $C''C$ is obtained by rotating line $C'D_3$ about R by some angle $\theta > 0$ that can be made arbitrarily small by just decreasing μ . Again, this implies that the intersection point between the lines $C''C$ and $C'D_3$ can be kept at a distance from C'' arbitrarily close to $d/2$, and can therefore never coincide with C , which is only μ away from C'' . Also, because $\angle NRC < 90^\circ$ ($\angle NRC$ increases monotonically

and converges to 90° as R converges to P), C and τ_3 are always on opposite sides of $C'D_3$, and so C cannot be in τ_3 .

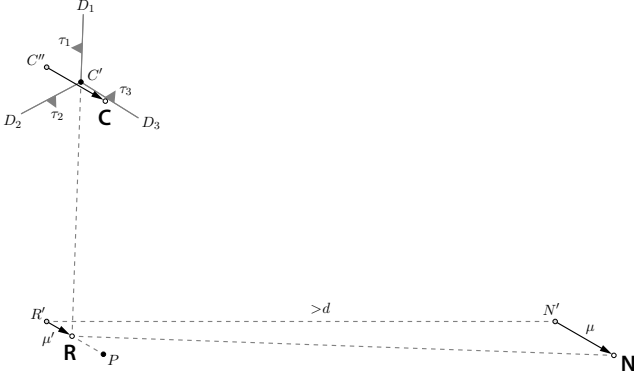


Fig. 8 Correctness of phase 5 when the Number robot has moved to D_3'

To conclude the proof, it suffices to show that C and R' lie on strictly opposite sides of line RC' : indeed, this would imply that C is not on the segments $C'D_1$ and $C'D_2$ or in their respective midway triangles, because these lie on the same side of RC' as R' or on the line RC' itself (cf. Figure 8). To prove this claim, consider a Cartesian coordinate system with origin in R and x axis oriented as $\overrightarrow{RN'}$. Let $0 < \mu' = \overline{R'R} < \mu$. Since the line RR' forms an angle of 60° with the y axis, the coordinates of R' are

$$R' = \left(-\frac{\sqrt{3}\mu'}{2}, \frac{\mu'}{2} \right).$$

We therefore have

$$C'' = R' + (0, d) = \left(-\frac{\sqrt{3}\mu'}{2}, \frac{\mu'}{2} + d \right)$$

and

$$C = C'' + \left(\frac{\sqrt{3}\mu}{2}, -\frac{\mu}{2} \right) = \left(\frac{\sqrt{3}(\mu - \mu')}{2}, d - \frac{\mu - \mu'}{2} \right).$$

We also have

$$N' = R' + (\overline{R'N'}, 0) = \left(\overline{R'N'} - \frac{\sqrt{3}\mu'}{2}, \frac{\mu'}{2} \right)$$

and

$$\begin{aligned} N &= N' + \left(\frac{\sqrt{3}\mu}{2}, -\frac{\mu}{2} \right) = \\ &= \left(\overline{R'N'} + \frac{\sqrt{3}(\mu - \mu')}{2}, \frac{\mu' - \mu}{2} \right). \end{aligned} \quad (3)$$

It follows that the line RN has equation

$$y = \frac{\mu' - \mu}{2 \cdot \overline{R'N'} + \sqrt{3}(\mu - \mu')} x.$$

Since the line RC' is orthogonal to RN , it has equation

$$y = \left(\frac{2 \cdot \overline{R'N'}}{\mu - \mu'} + \sqrt{3} \right) x. \quad (4)$$

Observe that RC' passes through the origin and its slope is positive. Hence R' lies above this line, because its x coordinate is negative and its y coordinate is positive.

Let us now plug the x coordinate of C in (4):

$$\begin{aligned} y &= \left(\frac{2 \cdot \overline{R'N'}}{\mu - \mu'} + \sqrt{3} \right) \cdot \frac{\sqrt{3}(\mu - \mu')}{2} = \\ &= \sqrt{3} \cdot \overline{R'N'} + \frac{3(\mu - \mu')}{2}. \end{aligned} \quad (5)$$

Recall from the discussion on phase 4 that $\overline{R'N'} > d$ (it corresponds to $\overline{RN'}$ in Figure 5), and therefore the y in (5) is abundantly greater than d . On the other hand, the y coordinate of C is $d - (\mu - \mu')/2$, which is smaller than d , implying that C lies below the line RC' . We conclude that C and R' lie on opposite sides of RC' .

We have just proved that the Reference robot keeps moving until it reaches P , thus bringing the \mathbf{tm} in a rest position again, say at time t' . We ultimately observe that the real number stored in the machine at time t' is the same the one the Commander computed in phase 1 and that the Number robot copied during phase 2. This is because the Number robot and the Reference robot, during phases 4 and 5 respectively, have moved by μ in the same direction: so, at the end of phase 5, they have the same distance they had at the end of phase 2. Also, all members of the \mathbf{tm} are clearly frozen at time t' . \square

5 Applications

In this section we discuss some applications of the \mathbf{tm} . We also prove that the basic \mathbf{tm} constructed in Section 3.1 is minimal, in the sense that no smaller team of oblivious robots can accomplish the same tasks (Theorem 4).

5.1 Exploring the Plane

The first elementary task a basic \mathbf{tm} in \mathbb{R}^2 can fulfill is that of *exploring* the plane. The task consists in making all the robots in the \mathbf{tm} see every point in the plane in the course of an infinite execution. We first assume that the three members of the \mathbf{tm} are the only robots

in the plane. Later in this section, we will extend our technique to other types of scenarios and more complex tasks.

Theorem 3 *A basic TuringMobile consisting of three robots in \mathbb{R}^2 can explore the plane.*

Proof Recall that a basic **tm** can store a single real number r and update it at every move as a result of executing a real RAM program with input r . In particular, the **tm** can count how many times it has moved by simply starting its execution with $r = 0$ and computing $r := r + 1$ at each move.

Moreover, the Commander chooses the direction of the next move (in the form of a point D_i , see Figure 2) by executing another real RAM program with input r . If r is an integer, the Commander can therefore compute any Turing-computable function on r , and so it can decide to move to D_1 the first time, then to D_2 twice, then to D_3 three times, to D_1 four times, and so on. This pattern of moves is illustrated in Figure 9, and indeed it results in the exploration of the plane, because the visibility radius of the robots is much greater than the step μ . □

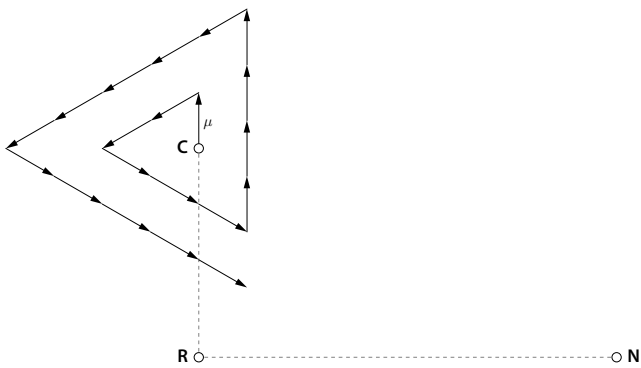


Fig. 9 Exploration of the plane by a basic TuringMobile

5.2 Minimality of the Basic TuringMobile

We can use the previous result to prove indirectly that our basic **tm** design is minimal, because no team of fewer than three oblivious robots in \mathbb{R}^2 can explore the plane.

Theorem 4 *If only one or two oblivious identical robots with limited visibility are present in \mathbb{R}^2 , they cannot explore the plane, even if the scheduler lets them move synchronously and rigidly.*

Proof Assume that a single oblivious robot is given in \mathbb{R}^2 (hence no other entities or obstacles are present).

Since the robot always gets the same snapshot, it always computes the same destination point in its local coordinate system, and so it always translates by the same vector. As a consequence, it just moves along a straight ray, and therefore it cannot explore the plane.

Let two oblivious robots be given, and suppose that their local coordinate systems are oriented symmetrically. Whether the robots see each other or not, if they take their snapshots simultaneously, they get identical views, and so they compute destination points that are symmetric with respect to their midpoint O . If they keep moving synchronously and rigidly, O remains their midpoint. So, if the robots have visibility radius V , they see each other if and only if they are in the circle γ of radius $V/2$ centered in O .

Now consider a Cartesian coordinate system with origin O . Without loss of generality, when the robots do not see each other, they move by vectors $(1, 0)$ and $(-1, 0)$, respectively. Let ξ be the half-plane $y \geq V$, and observe that ξ lies completely outside γ .

It is obvious that the robots cannot explore the entire plane if neither of them ever stops in ξ . The first time one of them stops in ξ , it takes a snapshot from there, and starts moving parallel to the x axis, thus never seeing the other robot again, and never leaving ξ . Of course, following a straight line through ξ is not enough to explore all of it. □

5.3 Near-Gathering with Limited Visibility

The exploration technique can be applied to several more complex problems. The first we describe is the *Near-Gathering* problem, in which all robots in the plane must get in the same disk of a given radius ϵ (without colliding) and remain there forever. It does not matter if the robots keep moving, as long as there is a disk of radius ϵ that contains them all.

It is clear that solving this problem from every initial configuration is not possible, and hence some restrictive assumptions have to be made. The usual assumption is that the initial visibility graph of the robots be connected [21, 28]. Here we make a different assumption: there are three robots that form a basic **tm** somewhere in the plane, and each robot not in the **tm** has distance at least ϵ from all other robots. (Actually we could weaken this assumption much more, but this simple example is good enough to showcase our technique.) Also, in the existing literature on the Near-Gathering problem it is always assumed that the robots agree on at least one coordinate axis, but here we do not need this assumption.

Theorem 5 *Let a swarm of asynchronous, anonymous, oblivious robots with limited visibility be in an initial configuration in \mathbb{R}^2 where three robots form a basic TuringMobile at rest, and each robot not in the TuringMobile is at distance at least ε from all other robots. Then, the swarm can solve the Near-Gathering problem. (Note that we make no assumption on the visibility graph and its connectedness.)*

Proof Say that all robots in the plane have a visibility radius of $V \gg \varepsilon$, and that the **tm** moves by $\mu \ll \varepsilon$ at each step. The **tm** starts exploring the plane as above, and it stops in a rest position as soon as it finds a robot whose distance from the Commander is smaller than $V/2$ and greater than ε . On the other hand, if a robot is not part of the **tm**, it waits until it sees a **tm** in a rest position at distance smaller than $V/2$. When it does, it moves to a designated area \mathcal{A} in the proximity of the Commander. Such an area has distance at least $3d$ from the Commander, so no confusion can arise in the identification of the members of the **tm**. If several robots are eligible to move to \mathcal{A} , only one at a time does so: note that the layout of the **tm** itself gives an implicit total order to the robots around it. Observe that the robots cannot form a second **tm** while they move to \mathcal{A} : in order to do so, two of them would have to move to \mathcal{A} at the same time and get close enough to a third robot. Once they enter \mathcal{A} , the robots position themselves on a segment much shorter than d , so they cannot possibly be mistaken for a **tm**.

Once the eligible robots have positioned themselves in \mathcal{A} , the **tm** resumes its exploration of the plane, and the robots in \mathcal{A} copy all its movements. Of course, at each step the **tm** waits for all the robots in \mathcal{A} to catch up before carrying on with the exploration. Now, if the total number of robots in the plane is known, the **tm** can stop as soon as all of them have joined it. Otherwise, the machine simply keeps exploring the plane forever, eventually collecting all robots. In both cases, the Near-Gathering problem is solved. \square

5.4 Pattern Formation with Limited Visibility

Suppose n robots are tasked with forming a given *pattern* consisting of a multiset of n points: this is the *Pattern Formation* problem, which becomes the *Gathering* problem in the special case where the points are all coincident. For this problem, it does not matter where the pattern is formed, nor does its orientation or scale.

Again, the Pattern Formation problem is unsolvable from some initial configurations, so we make the same assumptions as with the Near-Gathering problem.

Theorem 6 *Under the assumptions of Theorem 5, the swarm can solve the Pattern Formation problem, as well. (Again, we make no assumption on the visibility graph and its connectedness.)*

Proof The algorithm starts by solving the Near-Gathering problem as in Theorem 5. The only difference is that now there is a second tiny area \mathcal{B} , attached to \mathcal{A} (and still far enough from the **tm**), which the robots avoid when they join \mathcal{A} . This is because this second area will later be used to form the pattern.

Since n is known (as it is the size of the pattern), the **tm** knows when it has to interrupt the exploration of the plane because all robots have already been found. At this point, the robots switch algorithms: one by one, they move to \mathcal{B} and form the pattern. This task is made possible by the presence of the **tm**, which gives an implicit order to all robots, and also unambiguously defines an embedding of the pattern in \mathcal{B} . So, each robot is implicitly assigned one point in \mathcal{B} , and it moves there when its turn comes.

If $n = 3$ or $n = 4$, there are uninteresting ad-hoc algorithms to do this; so, let us assume that $n \geq 5$. The first to move are the robots in \mathcal{A} : this part is easy, because they all lie on a small segment, which already gives them a total order, and allows them to move one by one. The robots only have to be careful enough not to collide with other robots before reaching their final positions. Again, this is trivial, because only one robot is allowed to move at a time.

When this part is done, there are at least two robots in \mathcal{B} , all of which have distance much smaller than d from each other. Then the members of the **tm** join \mathcal{B} as well, in order from the closest to the farthest. Each of them chooses a position in \mathcal{B} based on the robots already there and the remnants of the **tm**. Moreover, the members of the **tm** that have not started moving to \mathcal{B} yet cannot be mistaken for robots in \mathcal{B} , because they are at a greater distance from all others (and vice versa).

Note that, when the last robot leaves the **tm** and joins \mathcal{B} , it is able to find its final location because there are already at least four robots there, which provide a reference frame for the pattern to be formed. When this last robot has taken position in \mathcal{B} , the pattern is formed. \square

5.5 Higher Dimensions

Everything we said in this section pertained to robots in the plane. However, we can generalize all our results to robots in \mathbb{R}^m , for $m \geq 2$. Recall that, at the end of Section 3.2, we have described a **tm** for robots in \mathbb{R}^m ,

which can move within a specific plane γ exactly as a bidimensional **tm**, but can also move back and forth by μ in all other directions orthogonal to γ .

Now, extending our results to \mathbb{R}^m actually boils down to exploring the space with a **tm**: once we can do this, we can easily adapt our techniques for the Near-Gathering and the Pattern Formation problem, with negligible changes.

There are several ways a **tm** can explore \mathbb{R}^m : we will only give an example. Consider the exploration of the plane described at the beginning of this section, and let P_i be the point reached by the Commander after its i th move along the spiral-like path depicted in Figure 9 (P_0 is the initial position of the Commander).

Our m -dimensional **tm** starts exploring γ as if it were \mathbb{R}^2 . Whenever it visits a P_i for the first time, it goes back to P_0 . From P_0 , it keeps making moves orthogonal to γ until it has seen all points in \mathbb{R}^m whose projection on γ is P_0 and whose distance from P_0 is at most i . Then it goes back to P_0 , moves to P_1 , and repeats the same pattern of moves in the section of \mathbb{R}^m whose projection on γ is P_1 . It then does the same thing with P_2 , etc. When it reaches P_{i+1} (for the first time), it goes back to P_0 , and proceeds in the same fashion. By doing so, it explores the entire space \mathbb{R}^m .

Note that this algorithm only requires the **tm** to count how many moves it has made since the beginning of the execution: thus, the machine only has to memorize a single integer. The direction of the next move according to the above pattern is then obviously Turing-computable given the move counter.

6 Conclusions

We have introduced the TuringMobile (**tm**) as a special configuration of oblivious non-rigid robots that can simulate a rigid robot with memory. We have also applied the **tm** to some typical robot problems in the context of limited visibility, showing that the assumption of connectedness of the initial visibility graph can be dropped if a unique **tm** is present in the system. Our results hold not only in the plane, but also in Euclidean spaces of higher dimensions.

The simplest version of the **tm** (Section 3.1) consists of only three robots, and is the smallest possible configuration with these characteristics (Theorems 3 and 4). Our generalized **tm** (Section 3.2), which works in \mathbb{R}^m and simulates k registers of memory, consists of $3m+3k$ robots (Theorem 2). We believe we can decrease this number to $m+k+3$ by putting all the Number robots in the same basic component and adopting a more complicated technique to move them. However, minimizing

the number of robots in a general **tm** is left as an open problem.

Our basic **tm** design works if all robots have the same radius of visibility, because that allows them to implicitly agree on a unit of distance. We could remove this assumption and let each of them have a different visibility radius, but we would have to add a fourth robot to the **tm** for it to work (as well as keep the **tm** small compared to *all* these radii). The fourth robot is placed directly below the Reference robot at distance $d/2$ from it, and defines the unit distance. This robot is allowed to move, following the other three, only when the **tm** is at rest.

Recall that, in order to encode and decode arbitrary real numbers we used the α function and its inverse, which in turn are computed using the arctan and the tan functions. However, using transcendental functions is not essential: we could achieve a similar result by using only comparisons and arithmetic operations. The only downside would be that such a real RAM program would not run in a constant number of machine steps, but in a number of steps proportional to the value of the number to encode or decode. With this technique, we would be able to dispense with the trigonometric functions altogether, and have our robots use only arithmetic operations and square roots to compute their destination points.

Acknowledgments. The authors are grateful to the anonymous reviewers for greatly improving the quality of this paper with helpful comments and suggestions.

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