

# Algorithms for Shaping a Particle Swarm with a Shared Input by Exploiting Non-Slip Wall Contacts\*

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**Abstract**— There are driving applications for large populations of tiny robots in robotics, biology, and chemistry. These robots often lack onboard computation, actuation, and communication. Instead, these “robots” are particles carrying some payload and the particle swarm is controlled by a shared control input such as a uniform magnetic gradient or electric field. In previous works, we showed that the 2D position of each particle in such a swarm is controllable if the workspace contains a single obstacle the size of one particle.

Requiring a small, rigid obstacle suspended in the middle of the workspace is a strong constraint, especially in 3D. This paper relaxes that constraint, and provides position control algorithms that only require non-slip wall contact in 2D. Both *in vivo* and artificial environments often have such boundaries. We assume that particles in contact with the boundaries have zero velocity if the shared control input pushes the particle into the wall. This paper provides a shortest-path algorithm for positioning a two-particle swarm, and a generalization to positioning an  $n$ -particle swarm. Results are validated with simulations and a hardware demonstration.

## I. INTRODUCTION

Particle swarms propelled by a uniform field, where each particle receives the same control input, are common in applied mathematics, biology, and computer graphics. As a current example, micro- and nano-robots can be manufactured in large numbers, see [1]–[7]. Someday large swarms of robots will be remotely guided to assemble structures in parallel and through the human body to cure disease, heal tissue, and prevent infection. For each task, large numbers of micro robots are required to deliver sufficient payloads, but the small size of these robots makes it difficult to perform onboard computation. Instead, these robots are often controlled by a broadcast signal. The tiny robots themselves are often just rigid bodies, and it may be more accurate to define the *system*, consisting of particles, a uniform control field, and sensing, as the robot. Such systems are severely underactuated, having 2 degrees of freedom in the shared control input, but  $2n$  degrees of freedom for the particle swarm. Techniques are needed that can handle this underactuation. In previous work, we showed that the 2D position of each particle in such a swarm is controllable if the workspace contains a single obstacle the size of one particle.

Positioning is a foundational capability for a robotic system, e.g. placement of brachytherapy seeds. However, requiring a single, small, rigid obstacle suspended in the

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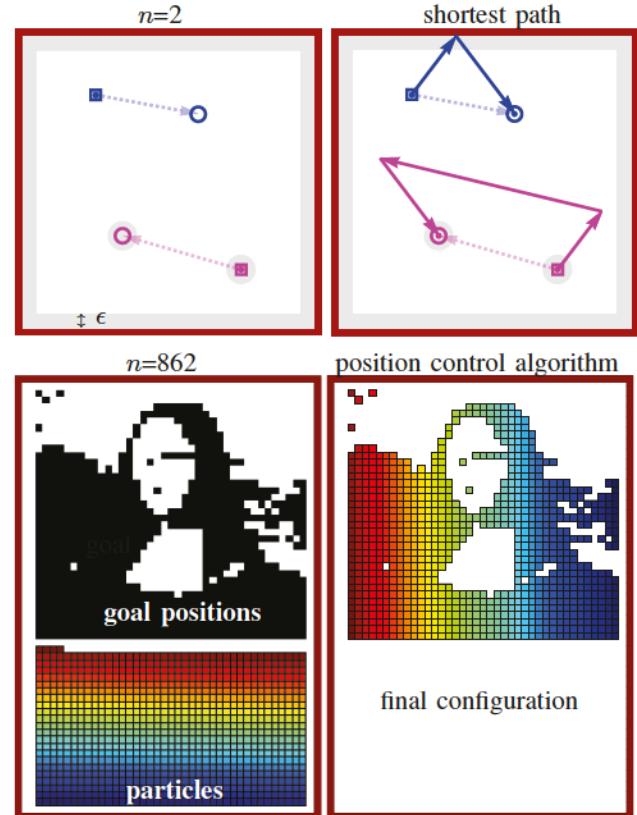


Fig. 1. Positioning particles that receive the same control inputs, but cannot move while a control input pushes them into a boundary. Top row shows the output of a best-first-search algorithm that finds the shortest path for two particles. Top left shows the initial position and goal position of the particles. The shortest path consists of moving at angle  $50^\circ$  until the blue robot contacts the top wall, then moving the magenta robot at angle  $165^\circ$  until the particles reach the desired relative spacing, then moving  $-60^\circ$  to the goal positions. The two bottom pictures show  $n$ -particle positioning using shared control inputs and boundary interaction.

middle of the workspace is often an unreasonable constraint, especially in 3D. This paper relaxes that constraint, and provides position control algorithms that only require non-slip wall contacts. We assume that particles in contact with the boundaries have zero velocity if the uniform control input pushes the particle into the wall.

The paper is arranged as follows. After a review of recent related work in Sec. II, Sec. III introduces a model for boundary interaction. We provide a shortest-path algorithm to arbitrarily position two robots in Sec. IV, and Sec. V extends this to prove a rectangular workspace with non-slip boundaries can position a swarm of  $n$  robots arbitrarily within a subset of the workspace. Sec. VI describes im-

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#### Using Boundaries: Friction and Boundary Layers

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$$x_i t \quad u t \quad F x_i t , u t , i \in , n .$$

$$F x_i t , u t \quad \begin{cases} -u t & x_i t \in \text{oundar and} \\ N \text{ oundar } x_i t \cdot u t \leq \\ e e & \end{cases}$$

ere N oundar  $x_i t$  i he nor a o he oundar a o i ion  $x_i t$  , and  $F x_i t , u t$  i he ric iona orce rovided he oundar .

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B A T A T

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eno e he curren o i ion o he ro o  $r_1$  and  $r_2$  . a ue .x and .y deno e he x and y coordina e , i.e.,  $r_1.x$  and  $r_1.y$  deno e he x and y oca ion o  $r_1$  . The a orih a i n a uni or con ro in u a ever in ance . The oa i o ero he error in o h coordina e , e .x, e.y

$g - r - g_2 - g_1 - r_2 - r_1$  u in a hared con ro in u . The a e ca e occur hen e , . n hi a e ca e , he hor e a h i a rai h i ne e een he curren o i ion o each ro o o i oa o i ion , a ho n in i . a . e / , , eco ue he o ove rea ch a e e enera ed r i o ii in one arice con ac in he a and hen re o i ion in he econd arice . There are our rea ch a e e , a ho n in i . . The hori on a and verica rea ch a e e are e uiva en in he con ura ion ace , o e can an in hi ace and choo e o i o i i e he arice co e o a a . ach non i ove chan e he con ura ion ace , a ho n in i .

he oa con ura ion can e reached in hree ove , hen  $m_1$  ake one arice hi a a ,  $m_2$  ad u he re a i ve acin error e o ero , and  $m_3$  ake he arice o heir na o i ion , a ho n in i . .  $m_2$  canno e horened , o o i i a ion de end on choo in he oca ion here he ro o hi he a . Since he hor e di ance e een o oin i a rai h i ne , re ec in he oa o i ion acro he oundar a and o in a rai h i ne i ve he o i a

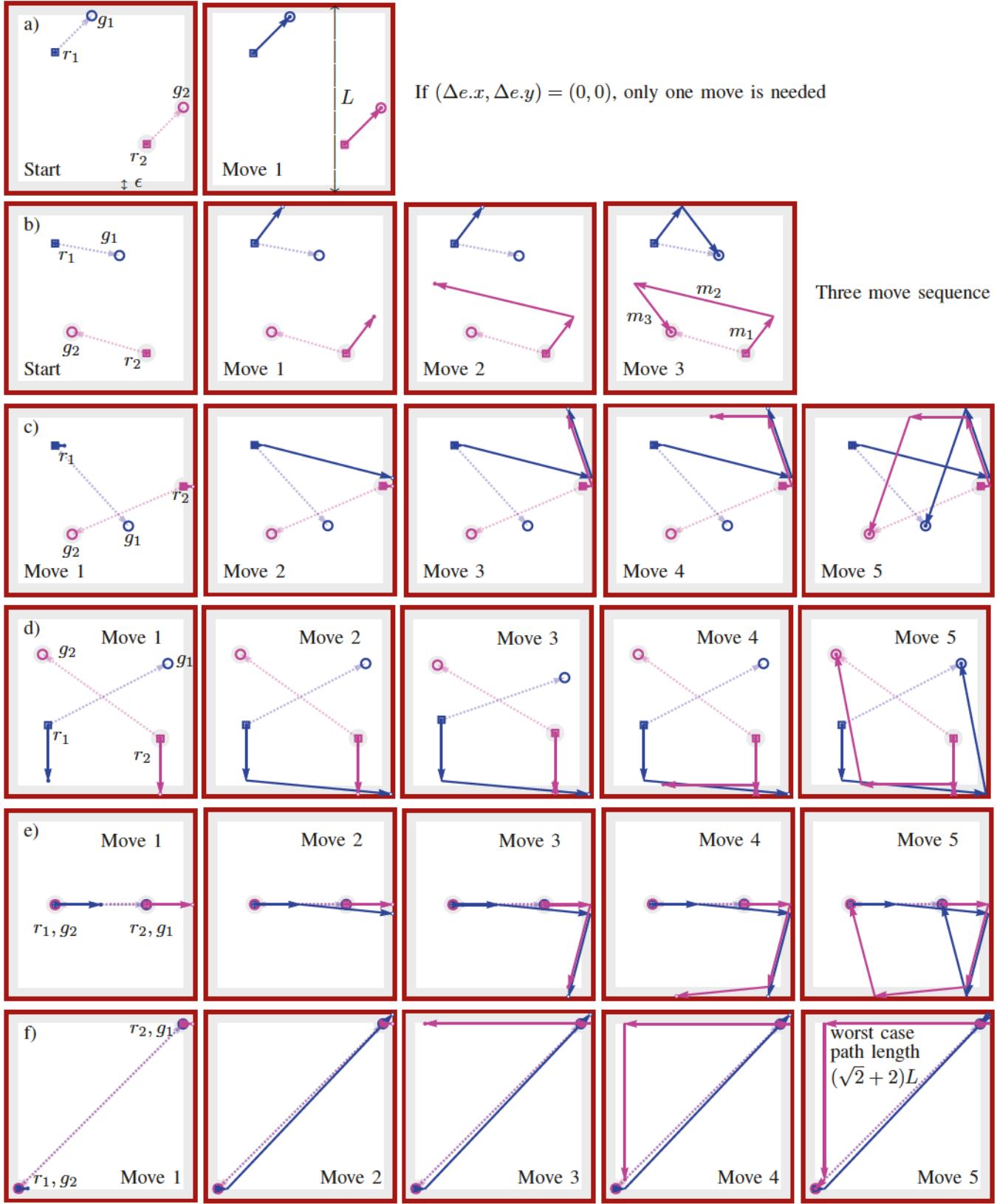


Fig. 2. Frames from an implementation of Alg. 1: two robot positioning using walls with non-slip contacts. Robot start positions are shown by a square, and goal positions by a circle. Dashed lines show the shortest route if robots could be controlled independently. Solid arrows show path given by Alg. 1. Online demonstration and source code at [20].

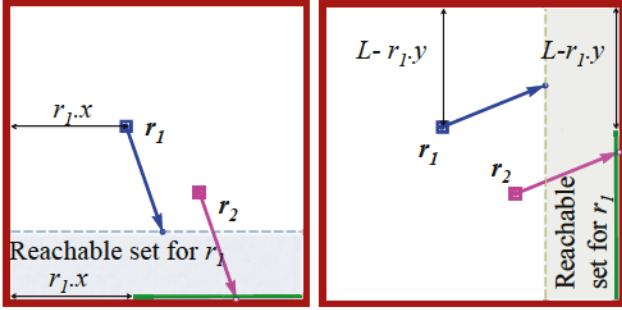


Fig. 3. Boundary interaction is used to change the relative positions of the robots. Each robot gets the same control input. (left) If robot 2 hits the bottom wall before robot 1 reaches a wall, robot 2 can reach anywhere along the green line, and robot 1 can move to anywhere in the shaded area. (right) Similarly, if robot 2 hits the right wall before robot 1 reaches a wall, robot 2 can reach anywhere along the green line, and robot 1 can move to anywhere in the shaded area.

hit location, as shown in Fig. 5. That point is selected when possible, but if this point would cause  $m_2$  to push the moving robot out of the workspace, the hit point is translated until the moving robot will not leave the workspace. If  $m_2$  causes the two particles to overlap, we add or subtract  $\epsilon$  to  $m_2.x$  to avoid collisions. This is shown in Fig. 6 with three different  $\epsilon$  values.

If  $\Delta g$  is not in the reachable set, we choose the nearest reachable  $\Delta x$  and  $\Delta y$  to  $\Delta g$ .

Alg. 1 uses an admissible heuristic that adds the current path length to the greatest distance from each robot to their goal. This heuristic directs exploration by expanding favorable routes first.

$$h(\text{moves}, r_1, r_2, g_1, g_2) = \sum_{i=1}^{|\text{moves}|} \|\text{moves}_i\| + \max(\|g_1 - r_1\|, \|g_2 - r_2\|) \quad (2)$$

We exploit symmetry in the solution by labeling the leftmost (or, if they have the same  $x$  coordinate, the topmost) robot  $r_1$ . If  $r_1$  is not also the topmost robot, we mirror the coordinate frame about the right boundary. As an example, consider the two starting positions,  $r_1 = (0.2, 0.2)$  and  $r_2 = (0.8, 0.8)$ . Because the leftmost robot is not the topmost robot, we mirror the coordinate frame about the right boundary giving  $r_1 = (0.2, 0.8)$  and  $r_2 = (0.8, 0.2)$ . After the path is found, we undo the mirroring to the output path. Similarly, we exploit rotational symmetry and assume the command pushes a robot to hit the top wall. If a different wall is selected, we rotate the coordinate frame by  $90^\circ$ ,  $180^\circ$ , or  $270^\circ$  counterclockwise and then push the robot to hit the top wall. After the path is found, we undo the rotation. This symmetry allows us to use a single function, Alg. 2, for collisions with all four walls.

## V. POSITION CONTROL OF $n$ ROBOTS USING BOUNDARY INTERACTION

The ideas from Alg. 1 can be extended to control the position of  $n$  particles using walls with non-slip contact. The solution is complete, but not optimal, and requires the

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### Algorithm 1 2-PARTICLEPATHFINDER( $r_1, r_2, g_1, g_2, L$ )

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**Require:** knowledge of current  $(r_1, r_2)$  and goal  $(g_1, g_2)$  positions of two robots.  $(0, 0)$  is bottom corner,  $L$  is length of the walls.  $\text{PathList}$  contains all the paths sorted by their path length plus an admissible heuristic.

- 1:  $\text{PathList} \leftarrow \{\}$
- 2:  $P \leftarrow \{h(\{\}, r_1, r_2, g_1, g_2), \{\}, r_1, r_2\}$  ▷  $P$  contains the admissible heuristic, the move sequence, and the current robot positions
- 3: **while**  $P.r_1 \neq g_1$  and  $P.r_2 \neq g_2$  **do**
- 4:   **for**  $\theta \in \{0^\circ, 90^\circ, 180^\circ, 270^\circ\}$  **do**
- 5:      $(r_1, r_2, g_1, g_2) \leftarrow \text{ROTATE}(P.r_1, P.r_2, g_1, g_2, \theta)$
- 6:      $\{d, \text{moves}, r_1, r_2\} \leftarrow \text{PLANMOVEUP}(r_1, r_2, g_1, g_2, L, P.\text{moves})$
- 7:      $(\text{moves}, r_1, r_2) \leftarrow \text{ROTATE}(\text{moves}, r_1, r_2, -\theta)$
- 8:     **PUSH**  $\{d, \text{moves}, r_1, r_2\}$  onto  $\text{PathList}$
- 9:   **end for**
- 10:   **SORT**( $\text{PathList}$ ) ▷ sort by admissible heuristic
- 11:    $P \leftarrow \text{POP}$  first element of  $\text{PathList}$
- 12: **end while**
- 13: **return**  $\text{moves}$

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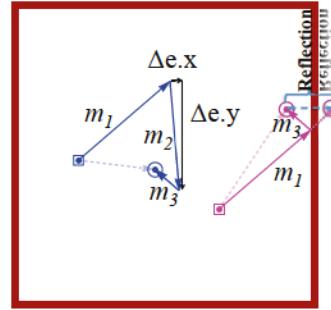


Fig. 5. If the goal configuration can be reached in three moves, the first move makes one particle hit a wall, the second move adjusts the relative spacing error  $\Delta e$  to zero, and the third move takes the particles to their final positions. The second move cannot be shortened, so optimization depends on choosing the location where the robot hits the wall. Since the shortest distance between two points is a straight line, reflecting the goal position across the boundary wall and plotting a straight line gives the optimal hit location.

starting and final configurations of particles to be disjoint. The solution described here is an iterative procedure with  $n$  loops. The  $k^{\text{th}}$  loop moves the  $k^{\text{th}}$  robot from a *staging zone* to the desired position in a *build zone*. All robots move according to the uniform input, but due to non-slip wall contacts, at the end of the  $k^{\text{th}}$  loop, robots 1 through  $k$  are in their desired final configuration in the build zone, and robots  $k+1$  to  $n$  are in the staging zone. See Fig. 8 for a schematic of the build and staging zones.

Assume an open workspace with four axis-aligned walls with non-slip contact. The axis-aligned build zone of dimension  $(w_b, h_b)$  containing the final configuration of  $n$  robots must be disjoint from the axis-aligned staging zone of dimension  $(w_s, h_s)$  containing the starting configuration of  $n$  robots. Without loss of generality, assume the build zone is above the staging zone. Let  $d$  be the diameter of the particles. Furthermore, there must be at least  $\epsilon$  space above

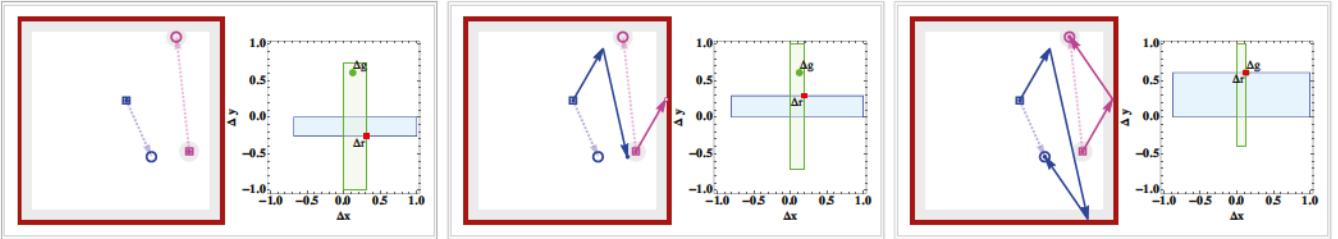


Fig. 4. Workspace and  $\Delta$  configuration space for three sets of robot configurations with the same final goal. The red square represents the starting  $\Delta x$  and  $\Delta y$  and the green circle represents the goal  $\Delta x$  and  $\Delta y$ . The green rectangle illustrates reachable  $\Delta x$  and  $\Delta y$  when one particle is in contact with a horizontal wall and the blue rectangle illustrates the reachable region when in contact with a vertical wall.

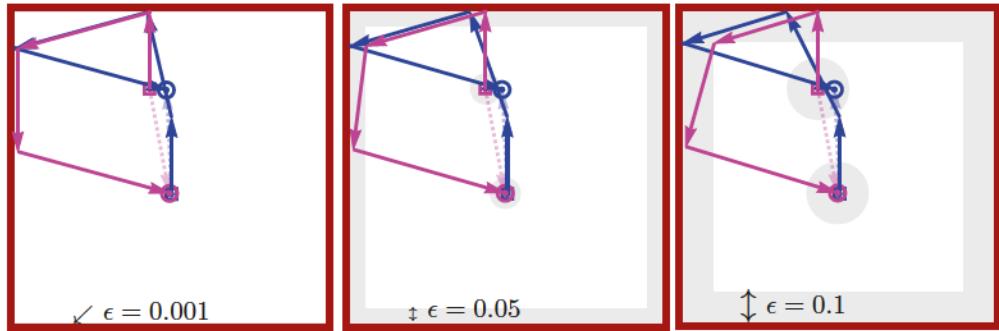


Fig. 6. Changing the minimum spacing  $\epsilon$  changes the path.  $\epsilon$  is the minimum spacing between two robots and the minimum separation from the boundaries.

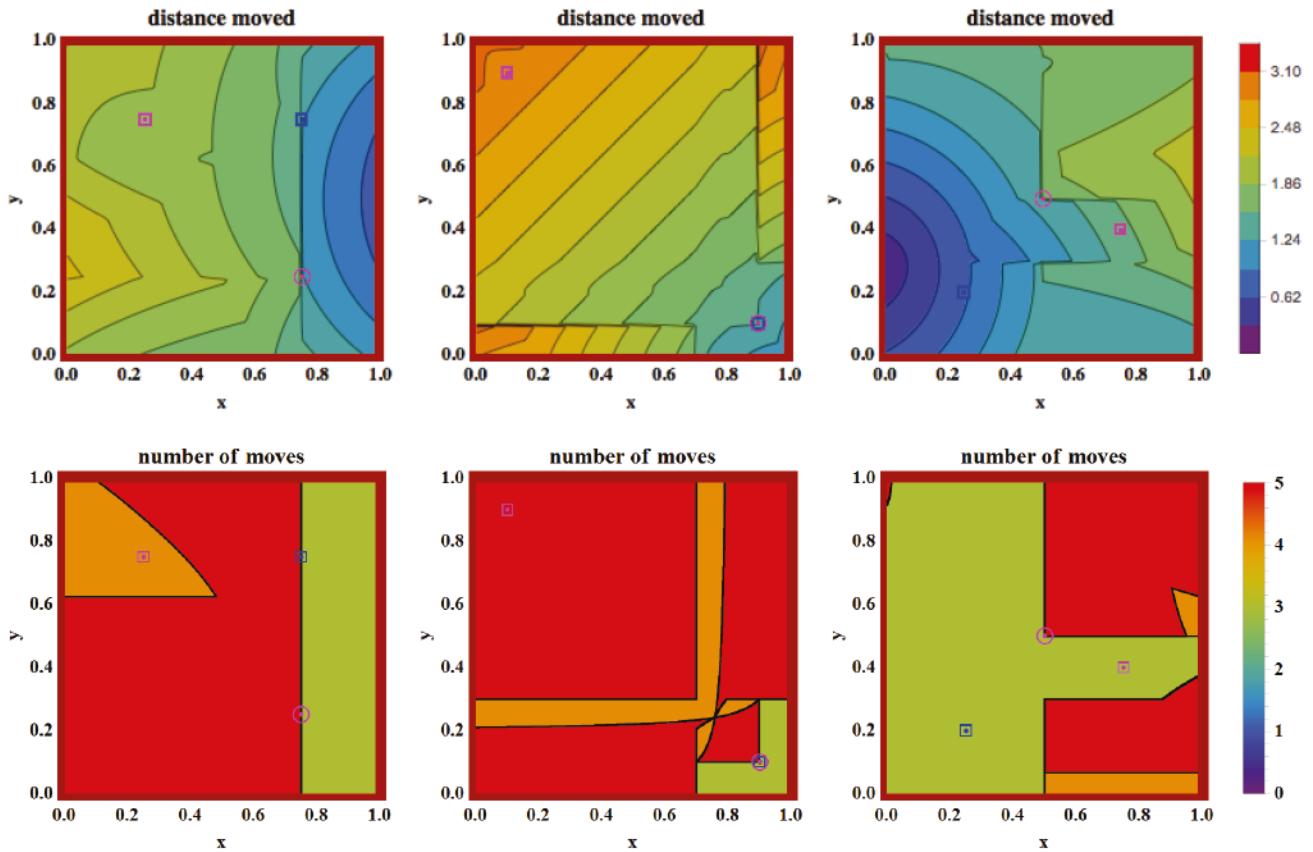


Fig. 7. Starting positions of robots 1 and 2 and goal position of robot 2 are fixed, and  $\epsilon = 0.001$ . The top row of contour plots show the distance if robot 1's goal position is varied in  $x$  and  $y$ . The bottom row shows the number of moves required for the same configurations.

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**Algorithm 2** PLANMOVEUP( $r_1, r_2, g_1, g_2, L, moves$ )

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**Require:** knowledge of current  $(r_1, r_2)$  and goal  $(g_1, g_2)$  positions of two robots.  $(0, 0)$  is bottom corner,  $L$  is length of the walls. The array  $moves$  is the current sequence of moves up to the current position. Assume  $r_1.x < r_2.x$  and  $r_1.y \geq r_2.y$ . If not, mirror the coordinate frame and swap the robots, then undo the mirroring before returning.  $\epsilon$  is a small, nonzero, user-specified value.

**Ensure:**  $(g_1, g_2), (r_1, r_2)$  all at least  $\epsilon$  distance from walls the goals and starting points have at least  $\epsilon$  distance from each other.  $m_1$  is the first move toward the wall or goal.  $m_2$  is the second move adjusting  $\Delta e$ .

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1:  $\Delta e \leftarrow (g_2 - g_1) - (r_2 - r_1)$ 
2: if  $\Delta e = (0, 0)$  then                                 $\triangleright$  base case
3:    $m_1 \leftarrow g_2 - r_2$ 
4:    $moves \leftarrow \{moves, m_1\}$ 
5:    $(r_1, r_2) \leftarrow \text{APPLYMOVE}(m_1, r_1, r_2)$ 
6:   return  $\{h(moves, r_1, r_2, g_1, g_2), moves, r_1, r_2\}$ 
7: end if
8: if  $r_2.x - r_1.x - 1 + 2\epsilon \leq \Delta g.x \leq 1$  and  $r_2.y - r_1.y \leq \Delta g.y \leq 0$  then           $\triangleright \Delta g \in \text{reachable region}$ 
9:    $m_1 \leftarrow \left( \frac{1-r_1.y}{2-g_1.y-r_1.y} (g_1.x - r_1.x), 1 - r_1.y \right)$ 
10:  if  $r_2.x + m_1.x > L$  then
11:     $m_1.x \leftarrow 1 - r_2.x$ 
12:  else if  $r_2.x + m_1.x < 0$  then
13:     $m_1.x \leftarrow -r_2.x$ 
14:  end if
15: else
16:    $m_1 = (0, 1 - r_1.y)$ 
17:    $\Delta g \leftarrow \text{closest reachable } (\Delta x, \Delta y)$ .
18: end if
19:  $moves \leftarrow \{moves, m_1\}$ 
20:  $(r_1, r_2) \leftarrow \text{APPLYMOVE}(m_1, r_1, r_2)$ 
21:  $m_2 \leftarrow \Delta g - (r_2 - r_1)$ 
22: if robots on each other or on the wall then
23:   Add  $\pm\epsilon$  to  $m_2.x$  to avoid collision
24: end if
25:  $moves \leftarrow \{moves, m_2\}$ 
26:  $(r_1, r_2) \leftarrow \text{APPLYMOVE}(m_2, r_1, r_2)$ 
27: return  $\{h(moves, r_1, r_2, g_1, g_2), moves, r_1, r_2\}$ 

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the build zone,  $\epsilon$  below the staging zone, and  $\epsilon + d$  to the left of the build and staging zone. The minimum workspace is then  $(\epsilon + d + \max(w_b, w_s), 2\epsilon + h_s, h_b)$ .

The  $n$  robots position control algorithm relies on a DRIFTMOVE( $\alpha, \beta, \epsilon, \theta$ ) control input, described in Alg. 4 and shown in Fig. 9. For  $\theta = 0^\circ$ , a drift move consists of repeating a triangular movement sequence  $\{(\beta/2, -\epsilon), (\beta/2, \epsilon), (-\alpha, 0)\}$ . Any particle touching a top wall moves right  $\beta$  units, while every particle not touching the top moves right  $\beta - \alpha$ .

Let  $(0, 0)$  be the lower left corner of the workspace,  $p_k$  the  $x, y$  position of the  $k^{\text{th}}$  robot, and  $f_k$  the final  $x, y$  position of the  $k^{\text{th}}$  robot. Label the robots in the staging zone from

Fig. 8. Illustration of Alg. 3,  $n$  robot position control using walls with non-slip contact.

Fig. 9. A DRIFTMOVE( $\alpha, \beta, \epsilon, 0^\circ$ ) repeats a triangular movement sequence  $\{(\beta/2, -\epsilon), (\beta/2, \epsilon), (-\alpha, 0)\}$ . At the sequence end, robot  $A$  has moved  $\beta$  units right, and robot  $B$  has moved  $\beta - \alpha$  units right.

left-to-right and bottom-to-top, and the  $f_k$  configurations top-to-bottom and right-to-left as shown in Fig. 10.

Alg. 3 proceeds as follows: First, the robots are moved left away from the right wall, and down so all robots in  $k^{\text{th}}$  row touch the bottom wall. Second, a set of DriftMoves are executed that move all robots in  $k^{\text{th}}$  row left until  $k$  touches the left wall, with no net movement of the other robots. Third, a set of DriftMoves are executed that move only robot  $k$  to its target height and return the other robots to their initial heights. Fourth, all robots except robot  $k$  are pushed left until robot  $k$  is in the correct relative  $x$  position compared to robots 1 to  $k - 1$ . Finally, all robots are moved right until robot  $k$  is in the desired target position. Running time is  $O(n(w + h))$ .

The hardware platform depicted in Fig. 10 is an assembled practical setup that assumes that  $\epsilon = 1$  cm. The workspace is a  $7 \times 7$  cm grid space. All particles are 3D-printed plastic whose top is a 1cm diameter cylinder with a narrower base

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**Algorithm 3** PositionControlnRobots( $k$ )

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- 1: Move( $-\epsilon, d/2 - p_{ky}$ )
- 2: **while**  $p_{kx} > d/2$  **do**
- 3:     DRIFTMOVE( $\epsilon, \min(p_{kx} - d/2, \epsilon), \epsilon, 180^\circ$ )
- 4: **end while**
- 5:  $m \leftarrow \text{ceil}(\frac{f_{ky} - d/2}{\epsilon})$
- 6:  $\beta \leftarrow \frac{f_{ky} - d/2}{m}$
- 7:  $\alpha \leftarrow \beta - \frac{d/2 - p_{ky} - \epsilon}{m}$
- 8: **for**  $m$  iterations **do**
- 9:     DRIFTMOVE( $\alpha, \beta, \epsilon, 90^\circ$ )
- 10: **end for**
- 11: Move( $d/2 + \epsilon - f_{kx}, 0$ )
- 12: Move( $f_{kx} - d/2, 0$ )

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**Algorithm 4** DRIFTMOVE( $\alpha, \beta, \epsilon, \theta$ )

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particles touching the wall move  $\beta$  units, while particles not touching the wall move  $\beta - \alpha$  units.

- 1:  $R = \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix}$
- 2: MOVE( $R \cdot [\beta/2, -\epsilon]^\top$ )
- 3: MOVE( $R \cdot [\beta/2, \epsilon]^\top$ )
- 4: MOVE( $R \cdot [-\alpha, 0]^\top$ )

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that encapsulates a steel bearing ball. Non-slip wall contact is generated by a toothed wall design to keep particles from moving out of place while implementing the drift move. The workspace boundary is mounted on top of a white sheet of cardboard. Underneath the cardboard, a grid of 3mm diameter magnets glued with 1 cm spacing to a thin board generates the uniform control input. A video attachment shows the algorithm at work. This discretized setup requires several modifications to Alg. 3. In this demonstration, all moves are 1 cm in length. All drift moves are a counterclockwise *square* move of size 1 cm  $\times$  1 cm. Once the  $k^{\text{th}}$  particle gets to its designated location in each loop, a correction step is implemented. This correction step increases by two the total number of moves required per particle. Fig. 8 shows there are only 6 stages per particle involved in Alg. 3. The fixed step algorithm requires 8 stages per particle as shown in Fig. 10.

A significant difference between Alg. 3 and the fixed move implementation of it is that Alg. 3 enables placing particles at arbitrary, non-overlapping locations, while the fixed move implementation requires goal locations at the center of grid cells.

## VI. SIMULATION

Two simulations were implemented using non-slip contact walls for position control. The first controls the position of two robots, the second controls the position of  $n$  robots.

### A. Position Control of Two Robots

Algorithm 1 was implemented in Mathematica using point robots (radius = 0). Fig. 2 shows an implementation of this algorithm with robot initial positions represented by hollow squares and final positions by circles. Dashed lines show the

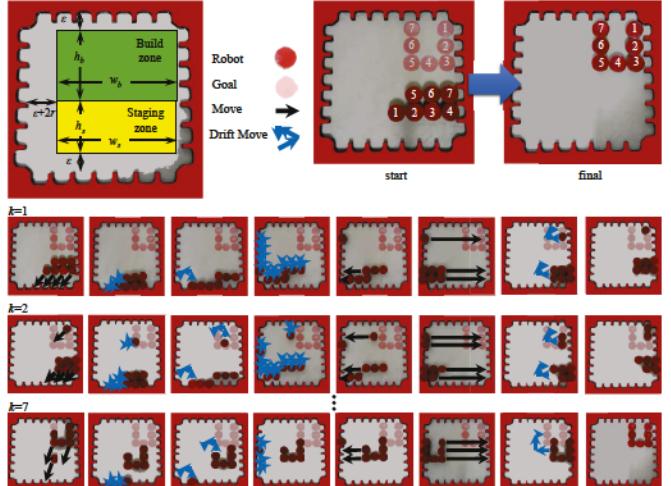


Fig. 10. Illustration of Alg. 3, discretized  $n$  robot position control using walls that enforce non-slip contact.

shortest route if robots could be controlled independently, while solid lines show the optimal shortest path using uniform inputs.

The contour plots in Fig. 7 top row show the length of the shortest path for given  $s_1, s_2, g_1$  with  $g_2$  ranging over all the workspace. This plot clearly shows the nonlinear nature of the path planning, with multiple isolated islands showing regions that are difficult to reach. If the length of each side of the square workspace is  $L$ , the worst case path length is  $(\sqrt{2} + 2)L$ .

The contour plots in Fig. 7 bottom row show the same configurations, but plot the required number of moves. There is never more than one  $g_2$  position reachable in one move. If  $g_2$  results in a contraction of  $(\delta x, \delta y)$ , there are many three move sequences. Four and five moves are sometimes required.

## B. Position Control of $n$ Robots

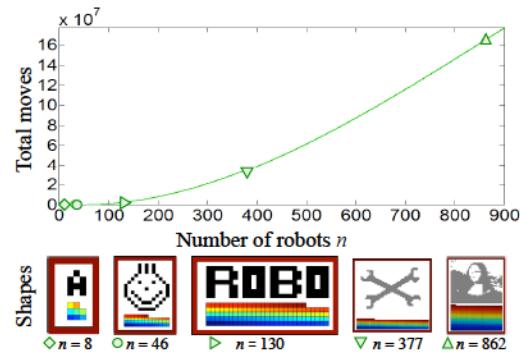
Alg. 3 was simulated in MATLAB using square block robots with unity width. Simulation results are shown in Fig. 11 for arrangements with an increasing number of robots,  $n = [8, 46, 130, 390, 862]$ . The distance moved grows quadratically with the number of robots  $n$ . A best-fit line  $210n^2 + 1200n - 10,000$  is overlaid by the data.

In Fig. 11, the amount of clearance is  $\epsilon = 1$ . Control performance is sensitive to the desired clearance. As  $\epsilon$  increases, the total distance decreases asymptotically, as shown in Fig. 12, because the robots have more room to maneuver and fewer DriftMoves are required.

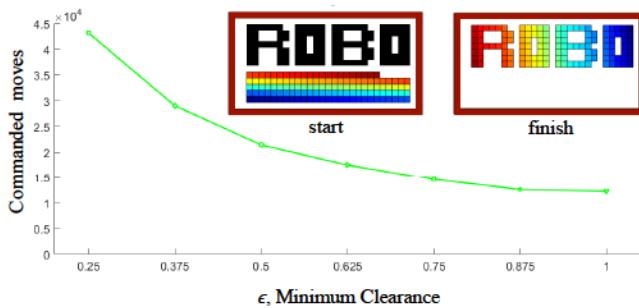
## VII. DEMONSTRATION

### Hardware Demonstration: Position Control of $n$ Robots

A hardware setup with a bounded platform, magnetic sliders, and a magnetic guide board was used to implement Alg. 3. The pink boundary is toothed with a white free space, as shown in Fig 10. Only discrete, 1 cm moves in the  $x$  and  $y$  directions are used. The goal configuration highlighted in the top right corner represents a 'U' made of seven particles. The



The required number of moves under A. in non i con ac a o rearren e n uare ha ed ro o ro uadra ica i h n. See code a .



on ro er or ance i en iive o he de ired clearance  $\epsilon$ . A  $\epsilon$  increa e, he o a di ance decrea e a o ica .

dark red con ura ion i he curren o iion o he aric e . ue o he di cre i ed ove en a o ed he oundar , dri ove o o a c uare. ree aric e reurn o heir ar o iion u aric e on he oundar o ove aera , enera in a ne idin o ion.

o o he o iion o he aric e hrou h ier a ion  $k$  , , and . A aric e receive he a e con ro in u , u oundar in eracion reak he con ro er . aric e reach heir oa o iion in a r in, r ou arran e en e innin ih he o o e aric e ro he a in one occu in he o ri h o iion o he uid one.

## S A T

This a er re ened echni ue or con ro in he o iion o a ar o ro o uin uni or in u and in eracion i h oundar ric ion orce . The a er rovided a ori h or recie o iion con ro, a e a rou and e cien covariance con ro . endin a ori h and o i rai h or ard u increa e he co e i . Addi ion a , hi a era u ed ric ion a u cien oco ee o aric e in con ac ih he oundar . The a ori h re uire re oo in o hande a ric ion coe cien . The a ori h a u ed a rec an uar ork ace . Thi i a rea ona e a u ion or aric a environ en , u in vivo environ en are curved . A e r each ro ra cou d i ork, u i canno ake advan a e o he o d ro a iona er a in a rec an uar environ en . u ure e or hou d e direc ed o ard i rovin he echno o and ai orin i o eci cro o a ica ion .

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