Randomized Strategy for Walking in Streets for a Simple Robot

Azadeh Tabatabaei*

Mohammad Ghodsi[†]

Abstract

We consider the problem of walking in an unknown street, for a robot that has a minimal sensing capability. The robot is equipped with a sensor that only detects the discontinuities in depth information (gaps) and can locate the target point as enters in its visibility region. First, we propose an online deterministic search strategy that generates an optimal search path for the simple robot to reach the target t, starting from s. In contrast with previously known research, the path is designed without memorizing any portion of the scene has seen so far. Then, we present a randomized search strategy, similar to the deterministic strategy. We prove that the expected distance traveled by the robot is at most a 5.33 times longer than the shortest path to reach the target.

1 Introduction

Path planning is a basic problem to almost all scopes of computer science; such as computational geometry, online algorithms, robotics and artificial intelligence [12]. Especially, path planning in an unknown environment for which there is no geometric map of the scene is interesting in many real life cases. Robot sensors is the only tool for gathering information in an unknown street. Amount of the information achieved from the environment depends on the capability of the robot. Due to the importance of using simple robot, including low cost, less sensitive to failure, robust against sensing errors and noise, many types of path planning for simple robot have been studied [1, 3, 6].

In this paper, we consider the problem of walking a simple robot in an unknown street. A simple polygon P with two separated vertices s and t is called a street if the left boundary chain L_{chain} and the right boundary chain R_{chain} constructed on the polygon from s to t are mutually weakly visible. In other words each point on the left chain can see at least one point on the right chain and vice versa [9], see Figure 1.

A point robot which its sensor has a minimal capability that can only detect discontinuities in depth information (gaps) and the target point t, starts searching the street. The robot can locate the target as soon



Figure 1: (a) A street polygon. The colored region is the visibility polygon of the point robot q in the street. (b) The position of discontinuities in the depth information (gaps) reported by the sensor.

as it enters in its visibility region. Also, the robot cannot measure any angles or distances or infer its position, see Figure 1. The goal is to reach the target t using the information gathered through its sensor, starting from s such that the traveled path by the robot is as short as possible.

In order to evaluate the efficiency of a search strategy for the robot, we use the notation of the com-The competitive analysis for a petitive analysis. strategy that leads the robot is the ratio of the distance traversed by the robot over the shortest distance from s to t, in the worst case. In previous research, Tabatabaei and Ghodsi gave a deterministic algorithm for the simple robot to reach the target tin the street, starting from s. The robot using some pebbles and memorizing some portion of the street has seen explores the street. The target t is achieved such that the traversed path is at most 11 times longer than the shortest path from s to t. Also they showed that 9 is a lower bound for the competitive ratio of each deterministic algorithm [15, 16]

In this paper, first, we present a deterministic strategy using the location of two special gaps which are updated during the walking, for the simple robot. The search path is optimal; length of the generated path is at most 9 times longer than shortest path. Then, we present a randomized strategy that generate a search path similar to the deterministic one, but the worst case ratio of the expected distances traveled by the robot to the shortest path is 5.33

Related Works: Klein proposed the first competitive algorithm for walking in streets problem for a robot that was equipped with a 360 degrees vision

^{*}Department of Computer Engineering, Sharif University of Technology.

[†]Sharif University of Technology and School of Computer Science, Institute for Research in Fundamental Sciences (IPM).

system [9]. Also, Icking et al. presented an optimal search strategy for the problem with the competitive factor of $\sqrt{2}$ [7]. Many online strategies for patrolling unknown environment such as street, generalized street, and star polygon are presented in [5, 10].

The limited sensing model (gap sensor) that our robot is equipped with, in this research, was first introduced by Tovar et al. [18]. They offered Gap Navigation Tree (GNT) to maintain and update the gaps seen along a navigating path. Other researcher presented some strategies, using GNT, for exploring unknown environments [4, 11, 13]. An optimal search strategy with minimum number of turns, for the simple robot equipped with the gap sensor, is presented in [17].

Another minimal sensing model was presented by Suri et al. [14]. They assumed that the simple robot can only sense the combinatorial (non-metric) properties of the environment. The robot can locate the vertices of the polygon in its visibility region, and can report if there is a polygonal edge between them. Despite of the minimal ability, they showed that the robot can accomplish many non-trivial tasks. Then, Disser et al. empowered the robot with a compass to solve the mapping problem in polygons with holes [2].

2 preliminaries

2.1 The Sensing Model and motion primitives

At the start point, the point robot reports a cyclically ordered of discontinuities in the depth information (gaps) in its visibility region. Each gap has a label of L or R which displays the direction of the part of the scene that is hidden behind the gap, see Figure 2.

The robot can orient its heading to each gap and moves towards the gap in an arbitrary number of steps. Also the robot moves towards the target as they enter in its visibility region.

While the robot moves, combinatorial changes occur in the visibility region of the robot that they are called critical events. There are four types of critical events: appearances, disappearances, merges, and splits of gaps. Appearance and disappearance events occur when the robot crosses inflection rays. An appeared gap, during the movement, corresponds to a portion of the environment that was already visible, but now is not visible. such the gaps are called primitive gaps and the other gaps are non-primitive gaps. Merge and split events occur when the robot crosses bitangent, as illustrated in Figure 2.

2.2 Known Properties

At each point of the search path, if the target is not visible, the robot reports a set of left and right gaps (*l*-gap and *r*-gap for abbreviation). Let g_l be the most



Figure 2: The dynamically changes of the gaps as the robot walks towards a gap. The dark circle is the location of the robot, and squares and other circles denote primitive and non-primitive gaps respectively. (a) Existing gaps at the start point. (b) A split event. (c) A disappearance event. (d) An appearance event. (e) Another split event. (f) A merge event.

advanced non-primitive left gap (l-gap) and g_r be the most advanced non-primitive right gap (r-gap) [16], see Figure 3. The two gaps have a fundamental role in path planning for the simple robot.

Theorem 1 [7, 16] While the target is not visible, it is hidden behind one of the two gaps, g_l or g_r .

From Theorem 1, if there exist only one of the two gaps $(g_r \text{ and } g_l)$ then the goal is hidden behind of the gap. Thus, there is no ambiguity and the robot moves towards the gap, see Figure 3(a). When both of g_r and g_l exist, a funnel case arises, see Figure 3(b). At each funnel case, usually, a detour from the shortest path is unavoidable.

As the robot moves in the street, the critical events that change the structure of the robot's visibility region may dynamically change g_l and g_r . Also, by the robot movement, a funnel case may end or a new funnel may start. We refer to the point, in which a funnel ends a *critical point* of the funnel.



Figure 3: g_r and g_l are the most advanced gaps at the start point s. (a) There is only one most advanced gap, at start point s. (b) Sequences of the most advanced gaps may occur, as the robot moves.

The following events update the location of g_l and g_r as well as a funnel situation.

- 1. When g_r/g_l splits into g_r/g_l and another r-gap/l-gap, then g_r/g_l will be replaced by the r-gap/l-gap, (point 1 in Figure 3(b)).
- 2. When g_r/g_l splits into g_r/g_l and another l-gap/r-gap, then l-gap/r-gap will be set as g_l/g_r . This point is a critical point in which a funnel situation ends, (point 1 in Figure 4(a).
- 3. When g_l or g_r disappears, the robot may achieve a critical point in which a funnel situation ends, (point 1 in Figure 4(b)).

Note that the split and disappearance events may occur concurrently, (point 1 in Figure 4(b)).

3 Algorithm

Now, we present our strategy for searching the street, from s to t. Since the target is constantly behind one of g_r and g_l , during the searching, the location of them is maintained and dynamically updated as explained in the previous section.

3.1 A deterministic strategy

At each point of the search path, especially at the start point s, there are two cases:

- If only one of the two gaps $(g_r \text{ and } g_l)$ exists, then the goal is hidden behind the gap. The robot moves towards the gap until the target is achieved or a funnel situation arises, see Figure 3(a).
- If there is a funnel case, in order to bound the detour, the robot moves towards g_r and g_l alternatively. At each stage $i \in \{1, 3, 5, ...\}$, the

robot moves a_i steps towards g_r , and at each stage $i \in \{2, 4, 6, ...\}$, the robot moves a_i steps towards g_l such that:

 $a_1 = 1, a_2 = 3, a_i = 2a_{i-1}$ for $i \in \{3, 4, 5, ...\}$. The robot continues moving towards g_r or g_l alternatively until a critical point of the funnel is achieved. At the point one of g_r or g_l disappears, or g_r and g_l are collinear. So, the robot moves along the existing gap direction until the target is achieved or a new funnel situation arises, as illustrated in Figure 3.

3.1.1 The randomized strategy

At each point of the search path that only one of g_r or g_l exists, or the two gaps are collinear, the robot moves along the existing direction, similar the deterministic strategy. In the funnel case, first, the robot chooses a random real uniformly variable from [0,1)and sets length of its step by 2^{ε} . Then, it chooses a uniform random variable from $\{0,1\}$ to select the direction towards g_r or g_l . If the random variable is 1(0), at each stage $i \in \{1,3,5,\ldots\}$, the robot moves a_i steps towards $g_r(g_l)$, and at each stage $i \in \{2,4,6,\ldots\}$, the robot moves a_i steps towards $g_l(g_r)$ such that: $a_1 = 1, a_2 = 3, a_i = 2a_{i-1}$ for $i \in \{3,4,5,\ldots\}$. Similar to the deterministic strategy, the robot continues moving towards g_r and g_l in the number of steps alternatively until the funnel case ends.

At each funnel, the actual randomization occurs only at first step for specifying length of steps, and for determining the direction of the movement. In the next section, we show the expected performance of our randomization algorithm is better than the performance of our deterministic algorithm.

3.2 Correctness and Analysis

Throughout the searching, the robot path coincides with the shortest path unless a funnel case arises. Then, in order to prove the competitive ratio of our strategy, we compare length of the path and shortest path in a funnel case.

For analyzing the strategy, we inspire from the doubling strategy by Baeza-Yates et al. [1]. In the strategy a robot moves back and forth on a line such that the distance to the start point doubles at each stage until the target is reached.

Theorem 2 [1] The doubling strategy for searching a point on a line has a competitive factor of 9, and this is optimal.

Opening angle, the angle between g_r and g_l , is always smaller than π [7]. The simple robot walks towards within the opening angle. An important attribute of the angle is characterized in the following lemma.



Figure 4: Illustration of the algorithm for two opening angles, small and large angles respectively in (a) and (b). The bold path is the robot search path, and the dotted path is shortest path.

Lemma 3 By our strategy, the detour from shortest path for small opening angle, in the funnel case, is shorter than detour for large opening angle.

Proof. In each funnel case, the robot moves some steps towards g_r or g_l , then changes its direction and moves some steps towards the other. In the alternative movement, one of the directions is correct and the other is a deviation. Clearly for large opening angle the deviation is greater, as shown in Figure 3.

Now, we can prove, the competitive factor of our deterministic strategy.

Theorem 4 Our deterministic strategy guarantee a path at most nine times as long as the shortest path in the street from s to t.

Proof. From Lemma 3, there is further deviation from shortest path for large opening angles. The angle never exceeds π . Then, for computing a competitive factor, we consider it equals π . Starting from s, the robot moves $a_1 = 1$ step towards g_r , then moves $a_2 = 1 + 2$ steps towards g_l , and again moves forth $a_3 = 2a_2 = 2 + 2^2$ steps towards g_l , moves back $a_4 = 2a_3 = 2^2 + 2^3$ steps towards g_l , and so on. In other words the robot moves back and forth on the line that contain g_l and g_r such that the distance to the start point *s* doubles at each stage until the critical point is reached. By Theorem 2 competitive factor for the search strategy is 9.

Kao, Reif, and Tate [8] offer a competitive randomized algorithm for searching on a line. The procedure is similar to the doubling strategy; the first step is characterized as a random number, and lengths of subsequent steps is multiplied by r. The competitive factor of their randomized strategy is $1 + (1+r)/\ln r$. A similar argument to the proof of theorem 4 shows that when the opening angle is π , our randomized search strategy coincides with Kao et al. strategy to search a point on a line with r = 2. So, the theorem below is immediately satisfied.

Theorem 5 The randomized strategy generates a search path to achieve target t in the street, starting from s, with a competitive ratio of 5.33.

4 Conclusions

In this paper we have developed two similar search strategies for walking in streets problem for a simple robot. The point robot can only detect the gaps and the target in the environment. Also the robot can only moves towards the gaps. Our deterministic strategy achieves optimal competitive factors of 9, and is simpler than previous known result. The other strategy is a randomized search strategy that has better performance with a competitive ratio of 5.33. It would be absorbing if there is an optimal randomized search strategy.

References

- Baezayates, R. A., Culberson, J. C., Rawlins, G. J. Searching in the plane. Information and Computation 106(2): 234–252, 1993.
- [2] Disser, Y., Ghosh, S. K., Mihalk, M., Widmayer, P. Mapping a polygon with holes using a compass. Theoretical Computer Science, 2013.
- [3] Fekete, Sandor P., Joseph SB Mitchell, and Christiane Schmidt. *Minimum Covering with Travel Cost.* Journal of Combinatorial Optimization, 24: 32-51, 2003.
- [4] Guilamo, L., Tovar, B., LaValle, S. M. Pursuitevasion in an unknown environment using gap navigation trees. In Intelligent Robot's and Systems Proceedings Vol. 4, (pp 3456–3462), 2004.
- [5] Ghosh, Subir Kumar, and Rolf Klein Online algorithms for searching and exploration in the plane. Computer Science Review 4.4 (2010): 189-201.

- [6] Hammar, Mikael, Bengt J. Nilsson, and Mia Persson. Competitive exploration of rectilinear polygons. Theoretical computer science 354(3): 367–378, 2006.
- [7] Icking, C., Klein, R., Langetepe, E. An optimal competitive strategy for walking in streets. In STACS 99, Springer Berlin Heidelberg, (pp. 110–120), 1999.
- [8] Kao, Ming-Yang, John H. Reif, and Stephen R. Tate. Searching in an unknown environment: An optimal randomized algorithm for the cow-path problem. Information and Computation 131(1): 63–79,1996.
- Klein, R. Walking an unknown street with bounded detour. Computational Geometry, 1(6): 325–351, 1992.
- [10] Lpez-Ortiz, Alejandro, and Sven Schuierer. Lower bounds for streets and generalized streets. International Journal of Computational Geometry and Applications 11.04 (2001): 401-421.
- [11] Lopez-Padilla, R., Murrieta-Cid, R., LaValle, S. M. Optimal Gap Navigation for a Disc Robot. In Algorithmic Foundations of Robotics, Springer Berlin Heidelberg, (pp 123–138), 2012.
- [12] Mitchell, Joseph SB. Geometric shortest paths and network optimization, Handbook of Computational Geometry. Elsevier Science Publishers, 1998.
- [13] Sachs, S., LaValle, S. M., Rajko, S. Visibilitybased pursuit-evasion in an unknown planar environment. The International Journal of Robotics Research 23(1): 3–26, 2004.
- [14] Suri, S., Vicari, E., Widmayer, P. Simple robots with minimal sensing: From local visibility to global geometry. The International Journal of Robotics Research, 27(9): 1055–1067, 2008.
- [15] Tabatabaei, Azadeh, and Mohammad Ghodsi. Walking in Streets with Minimal Sensing. Journal of Combinatorial Optimization http://link.springer.com/article/10.1007, 2014
- [16] Tabatabaei, Azadeh, and Mohammad Ghodsi. Walking in Streets with Minimal Sensing. In Combinatorial Optimization and Applications. Springer International Publishing, (pp. 361–372), 2013.
- [17] Tabatabaei, Azadeh, and Mohammad Ghodsi. Optimal Strategy for Walking in Streets with Minimum Number of Turns for a Simple Robot. Combinatorial Optimization and Applications. Springer International Publishing, (pp. 101–112), 2014.
- [18] Tovar, B., Murrieta-Cid, R., LaValle, S. M. Distanceoptimal navigation in an unknown environment without sensing distances. Robotics IEEE Transactions 23(3): 506–518, 2007.