

RHEINISCHE FRIEDRICH-WILHELMS-UNIVERSITÄT BONN  
INSTITUT FÜR INFORMATIK I



---

Elmar Langetepe  
**Online Motion Planning**

MA INF 1314

---

---

Sommersemester 2016  
Manuscript: Elmar Langetepe

slightly misses the goal while visiting ray  $i$  up to distance  $x_k$ . Instead, it finds the goal at step  $x_{J_k}$  on ray  $i$  arbitrarily close to  $\beta_k x_k$ . Either we have  $x_{J_k} > \beta_k x_k$ ; that is, the searcher discovers the goal in distance  $x_{J_k}$  on ray  $i$  and moves  $x_{J_k} - \beta_k x_k$  to the goal, or we have  $x_{J_k} < \beta_k x_k$ . In the latter case, the searcher moves  $\beta_k x_k - x_{J_k}$  from  $x_{J_k}$  and finds the goal by accident. In both cases, the searcher moves  $|x_{J_k} - \beta_k x_k|$  in the last step. Altogether, the competitive factor,  $C(S)$ , is bigger than

$$\frac{|x_{J_k} - \beta_k x_k| + \sum_{i=1}^{J_k-1} \beta_i x_i - x_i + \sqrt{(\beta_i x_i)^2 - 2\beta_i x_i x_{i+1} \cos \gamma_{i,i+1} + x_{i+1}^2}}{\beta_k x_k}.$$

By simple trigonometry, the shortest distance from  $\beta_i x_i$  to a neighboring ray is given by  $\beta_i x_i \sin \frac{2\pi}{n}$ . Fortunately, this distance is smaller than the distance

$$\sqrt{(\beta_i x_i)^2 - 2\beta_i x_i x_{i+1} \cos \gamma_{i,i+1} + x_{i+1}^2}$$

to any other ray. Thus, we have

$$C(S) > \frac{\sum_{i=1}^{J_k-1} \beta_i x_i}{\beta_k x_k} \sin \frac{2\pi}{n}.$$

Altogether, we have to find a lower bound for  $\frac{\sum_{i=1}^{J_k-1} f_i}{f_k}$ , where  $J_k$  denotes the index of the next visit of the ray of  $x_k$  and  $f_i = \beta_i x_i$  denotes the search depth in step  $i$ . Fortunately, this problem is the same problem as in the competitive analysis for the usual  $m$ -ray problem where the searcher can move only along the rays. It was shown in Lemma 3.3 (see also Gal [Gal80] and Baeza-Yates et al. [BYCR93]) that for this problem there is an optimal strategy that visits the rays with increasing depth and in a periodic order; that is,  $J_k = k + n$  and  $i = k$ . Applying Theorem 3.2 the best achievable strategy is given by  $f_i = (n/(n-1))^i$ . Altogether, this results in a function

$$(n-1) \left( \frac{n}{n-1} \right)^n \sin \frac{2\pi}{n}$$

for  $n$  rays. We can make  $n$  arbitrarily big because our construction is valid for every  $n$ . Note that we also have a lower bound for the problem of searching a point in the plane; this lower bound is close to the factor that is achieved by a spiral search.

**Theorem 3.11** *For the ray search problem there is no strategy that achieves a better factor than*

$$\lim_{n \rightarrow \infty} (n-1) \left( \frac{n}{n-1} \right)^n \sin \frac{2\pi}{n} = 17.079\dots$$

*Additionally, every strategy for searching a point in the plane achieves a competitive factor bigger than 17.079\dots (the optimal spiral achieves a factor of 17.289\dots [Gal80]).*

### 3.3 Searching in street polygons

Now we consider a special class of polygons, such that a competitive search still can be performed. By the  $m$ -ray search problem we already know that a constant competitive strategy for searching a point in arbitrary polygons does not exist.

The following polygons resembles streets or rivers where the path to the endpoint is not arbitrary although the path can make many windings and there are many caves where the goal might be located. Formally, we define a street polygon as follows:

**Definition 3.12** Let  $P$  be a simple polygon with two points  $s$  and  $t$  on the boundary.  $P$  is denoted as a **street** (polygon), if the two boundary chains  $P_L$  and  $P_R$  of  $P$  between  $s$  and  $t$  are weakly visible, i.e., any point from  $P_L$  sees at least one point from  $P_R$  and vice versa. [Kle91]

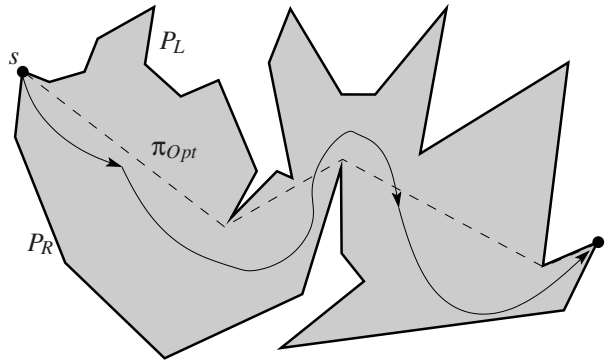
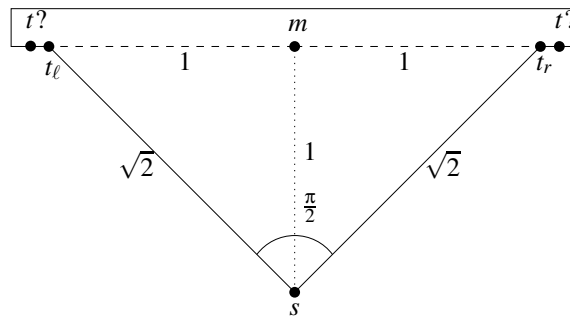


Figure 3.17: A street polygon.

Figure 3.17 shows an example. The main idea is that the shortest path from  $s$  to  $t$  also sees the boundary chains. Intuitively, if you use a street efficiently, you will always see the boundary chains.

Many structural properties have been proved for street polygons. For example, for a given polygon  $P$  one is interested on all possible pair of points  $(s, t)$  such that  $P$  is a street polygon. Surprisingly, this problem can be computed in linear time; see [THL98, DHN97]. In this section we consider the searching problem. That is, the start point  $s$  is given, the agent is equipped with a vision system and we are searching for a target  $t$ . The only information is, that  $P$  is a street for  $s$  and  $t$ . Against the shortest path to  $t$  we are searching for a competitive strategy with small ratio.

Figure 3.18: Lower bound for searching the target  $t$ .

A lower bound for the ratio in our problem can be constructed as follows.

**Theorem 3.13** (Klein, 1991)

There is no strategy that finds the target  $t$  in a street with a path of length smaller than  $\sqrt{2} \cdot \pi_{\text{Opt}}$ . The competitive ratio is at least  $\sqrt{2}$ . [Kle91]

**Proof.** Consider Figure 3.18. The agent is located at  $s$  and sees  $t_\ell$  and  $t_r$ . The target  $t$  lies behind one of them but the agent can only detect  $t$  if the line between  $t_\ell$  and  $t_r$  is visited. Then the agent can move to  $t$ . If the agent visits the segment between  $t_\ell$  and  $t_r$  to the left (right) of the midpoint  $m$ , the target is positioned at the right (left). Thus the best the agent can achieve is moving directly to  $m$ . Thus we have (where  $\varepsilon \rightarrow 0$ ):

$$|\pi_{\text{Rob}}| = 2 \quad \text{und} \quad \frac{|\pi_{\text{Rob}}|}{|\pi_{\text{Opt}}|} = \frac{2}{\sqrt{2}} = \sqrt{2}.$$

□

In search of  $t$  we can make use of some structural properties. Consider Figure 3.19(i). The agent is located at  $s$  and does not see the caves (the shaded parts). A cave is generated by a corresponding reflex vertex<sup>2</sup> of the polygon. We can subdivide the current cave generating reflex vertices into the set of left

<sup>2</sup>Vertices, with inner angle  $> \pi$ .

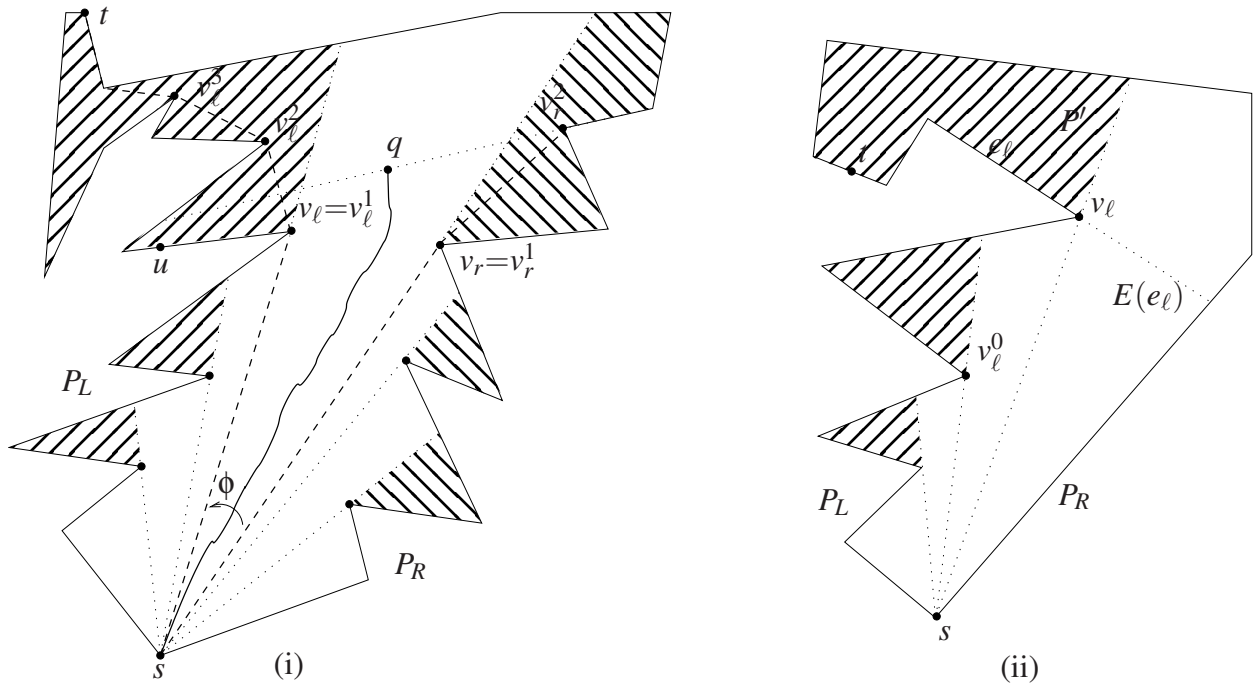


Figure 3.19: Typical situations for the task of searching the target in a street polygon.

reflex vertices (the cave is to the left) and right reflex vertices (the cave lies to the right). We call the vertices left or right reflex vertices, respectively.

Furthermore, we can consider the left reflex vertices in clockwise and the right reflex vertices in counter-clockwise order. One of these sequences can also be empty; in Figure 3.19(ii) there are no right reflex vertices.

We would like to argue that the unknown target  $t$  can only be located behind the rightmost left reflex vertex, say  $v_l$ , or the leftmost right reflex vertex, say  $v_r$ . The target cannot be located in one of the other caves. Assume that this is not the case. Assume that for example in Figure 3.19(i) the target is in the cave below  $v_l$ . In this case there is a point  $u$  on the right chain closely after  $v_l$  that does only see points on the right chain. This means that any reasonable strategy can concentrate on the current triangle of  $c$ ,  $v_l$  and  $v_r$ , where  $c$  is the current location of the agent. It only makes sense to run into this triangle and let the opening angle at  $c$  increase.

If there is only one vertex  $v_l$  or  $v_r$ , it is clear that the target can only lie behind this remaining vertex and any reasonable strategy move directly to this vertex. It is also clear the the shortest path to the target has to run over this vertex. The same holds, when the target gets visible. The agent directly moves toward it.

Formally, we consider the following cases or events while the agent moves into the triangle of  $c$ ,  $v_l$  and  $v_r$ .

- The target becomes visible. The agent moves toward it.
- The cave behind  $v_l$  or  $v_r$  becomes visible and does not contain the target; as in point  $q$  in Figure 3.19(i). The goal has to be behind the remaining vertex, the agent directly moves toward it.
- Behind the current vertex  $v_l$  or  $v_r$  another left or right reflex vertex becomes visible. For example  $v_l^2$  appears behind  $v_l$ . In this case the current left reflex vertex changes from  $v_l$  to  $v_l^2$ . The agent runs into the triangle of  $c$ ,  $v_l^2$  and  $v_r$

The last event successively builds segments of convex chain constructed from reflex vertices  $v_l^1, v_l^2, v_l^3, \dots, v_l^i$  and  $v_r^1, v_r^2, v_r^3, \dots, v_r^j$  to the left and to the right starting from  $s$ . The agent only moves inside these two

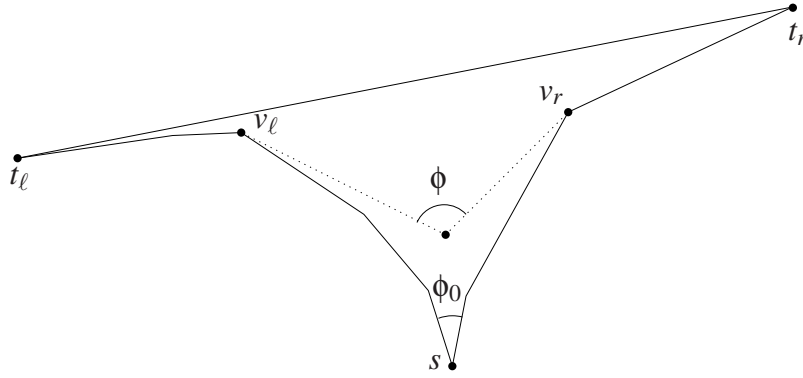


Figure 3.20: A funnel polygon.

chains. Therefore for simplicity we simply forget the original caves and only consider such funnel situations or so called funnel polygons. Beginning from  $s$  we have two convex chains that are finally closed by a segment  $t_l$  and  $t_r$  as shown in Figure 3.20. We assume that the current goal is either behind  $t_l$  or  $t_r$ . Actually there are two also caves behind  $t_l$  and  $t_r$ . Altogether the funnel polygons will invoke the same path as in the original polygon with caves.

These funnel situations are the only situations that can provoke a detour. If one such situation is resolved, either the goal is reached or the agent is located at a point on the shortest path to the goal. This means that we can consider this situation as the main challenge. If we can guarantee a competitive ratio of  $C$  for any single funnel, we can combine the path to a  $C$ -competitive strategy in total.

Therefore we concentrate on such polygons.

**Definition 3.14** A simple polygon is constructed by two convex chains  $P_L$  and  $P_R$  starting at a convex vertex  $s$ . The polygon can be closed by the segment  $\overline{t_l t_r}$  of the endpoints of the chains; see Figure 3.20. such a polygon is denoted as a **funnel (polygon)**,

Another important observation for the exploration of the funnel is, that the opening angle  $\phi$  for the current position  $c$  and the current active reflex vertices  $v_l$  and  $v_r$  will increase monotonically for any reasonable strategy. The agent starts with a opening angle  $\phi_0$  at  $s$  and finally we will reach  $\overline{t_l t_r}$  with opening angle  $180^\circ$ . Therefore it is quite natural to describe or parameterise a strategy by the opening angle  $\phi$ .

First, we define a more general lower bound dedicated to the opening angle  $\phi$ . We can generalize Theorem 3.13 as follows:

**Lemma 3.15** For a funnel polygon with opening angle  $\phi \leq \pi$  there is no strategy that has smaller path length than  $K_\phi \cdot |\pi_{\text{Opt}}|$  against the shortest path to the goal, where

$$K_\phi := \sqrt{1 + \sin \phi}.$$

Any strategy is at least  $K_\phi$  competitive.

**Proof.** Consider Figure 3.21. By the same argument as in the proof of Theorem 3.13 the best an agent can do is moving directly to the midpoint  $m$ . Any other movement results in a larger detour since we can place the target afterwards. Now the agent sees the target and moves toward it.<sup>3</sup> For  $\phi \leq \pi$  we have

$$\frac{|\pi_S|}{|\pi_{\text{Opt}}|} = \frac{\ell \cos \frac{\phi}{2} + \ell \sin \frac{\phi}{2}}{\ell} = \sqrt{1 + \sin \phi}.$$

□

<sup>3</sup>The path of length  $\epsilon$  from  $v_l$  or  $v_r$  to  $t$  need not be considered

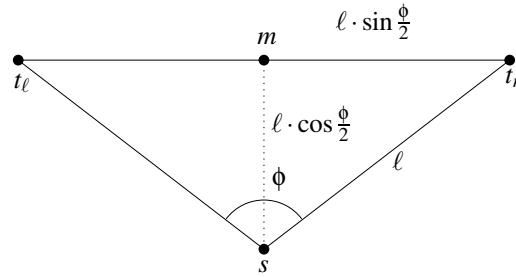


Figure 3.21: Generalized lower bound.

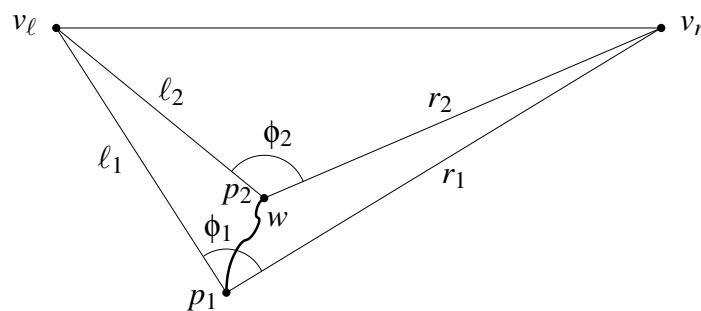
Note that for the final opening angle  $\phi = \pi$  and  $K_\phi = 1$  the agent will always move correctly, since the target is visible now. For  $\phi = \frac{\pi}{2}$  we have the ratio  $K_\phi = \sqrt{2}$  as in Theorem 3.13. For  $0 \leq \phi \leq \pi$  the function  $K_\phi$  gives a curve that starts at 1 rises up monotonically to  $\sqrt{2}$  at  $\frac{\pi}{2}$  and decreases monotonically toward 1 at  $\pi$ .

Assume that the agent explores a funnel starting from  $s$  with opening angle  $\phi_0$  and follows a path with monotonically increasing opening angles until  $\overline{t_l t_r}$  is visited and  $\phi = \pi$  holds.

For  $\frac{\pi}{2} \leq \phi_1 < \phi_2$  we have  $K_{\phi_1} > K_{\phi_2}$ , and the competitive ratio for the overall exploration is dominated by the smaller angle. For  $\phi_1 < \phi_2 \leq \frac{\pi}{2}$  we have  $K_{\phi_1} < K_{\phi_2} \leq \sqrt{2}$ , the ratio is dominated by the larger opening angle. If the agent starts from an opening angle  $\phi_0 < \frac{\pi}{2}$  along a path to angle  $\phi = \pi$  there will always be a point such that the opening angle  $\phi = \frac{\pi}{2}$  is attained. Therefore the worst case ratio  $\sqrt{2}$  is always included.

It seems to make sense to consider the case  $\phi_0 < \frac{\pi}{2}$  and  $\phi_0 \geq \frac{\pi}{2}$  separately. We start with  $\phi_0 \geq \frac{\pi}{2}$ . We already have a successful strategy for  $\phi = \pi$ . The following idea is that we apply a backward analysis that tells us how to prolong a successful strategy for opening angle  $\phi_2$  to a successful strategy for opening angle  $\phi_1 < \phi_2$ . By the following lemma we design a requirement for any path  $w$  from angle  $\phi_1$  to  $\phi_2$ .

**Lemma 3.16** *Let  $\Pi$  be a strategy that can reach the target of any funnel polygon with opening angle  $\phi_2 \geq \frac{\pi}{2}$  by competitive ratio  $K_{\phi_2}$ . We can extend this strategy to a  $K_{\phi_1}$  competitive strategy for funnel polygons with opening angle  $\phi_1$  with  $\phi_2 > \phi_1 \geq \frac{\pi}{2}$ , if the path  $w$  between the two corresponding points fulfils the length condition Equation 3.9 for the current situation as depicted in Figure 3.22.*

Figure 3.22: A path  $w$  from  $p_1$  with angle  $\phi_1$  to  $p_2$  with angle  $\phi_2$ .

**Proof.** We consider a triangle with opening angle  $\phi_1$ , start point  $p_1$  and a path  $w$  to a point  $p_2$  with opening angle  $\phi_2$ ; see Figure 3.22. From  $p_2$  the agent can use the strategy  $\Pi$  for the angle  $\phi_2$  which is known by assumption.  $\Pi$  is  $K_{\phi_2}$  competitive. Let us assume that during the movement  $w$  the vertices  $v_l$  and  $v_r$  do not change.

Let  $\ell_1$  and  $\ell_2$  denote the distances from  $p_1$  and  $p_2$  to  $v_l$ , as depicted in Figure 3.22,  $r_1$  and  $r_2$  are defined analogously. If the goal lies behind  $v_l$  we can assume that the overall path length for  $\pi_{p_1}^t$  from  $p_1$  to  $t$  is:

$$|\pi_{p_1}^t| \leq |w| + K_{\phi_2} \cdot \ell_2.$$

We would like to guarantee that the overall strategy is  $K_{\phi_1}$ -competitive, therefore we require:  $K_{\phi_1} = \frac{|\pi'_{p_1}|}{|\pi_{opt}|} \geq \frac{|w| + K_{\phi_2} \cdot \ell_2}{\ell_1}$ , also

$$K_{\phi_1} \cdot \ell_1 \geq |w| + K_{\phi_2} \cdot \ell_2.$$

Analogously, if the goal is behind  $v_r$ , we require  $K_{\phi_1} \cdot r_1 \geq |w| + K_{\phi_2} \cdot r_2$ .

If we can guarantee that the path  $w$  from  $p_1$  to  $p_2$  fulfils the length condition

$$|w| \leq \min\{K_{\phi_1} \ell_1 - K_{\phi_2} \ell_2, K_{\phi_1} r_1 - K_{\phi_2} r_2\}, \quad (3.9)$$

we conclude that the overall strategy starting at  $p_1$  attains a competitive ratio of  $K_{\phi_1}$  for the funnel with opening angle  $\phi_1$ .

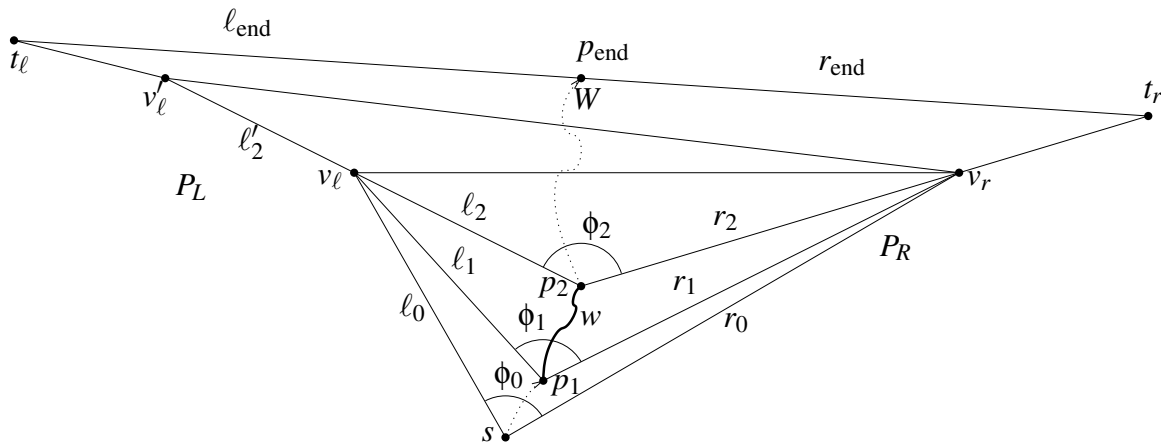


Figure 3.23: At  $p_2$  a new left reflex vertex is detected.

Now it is clear that from time to time the reflex vertices in the funnel will change. The path  $w$  and the condition Equation 3.9 should still guarantee the above conclusion. Therefore we consider the situation that condition Equation 3.9 is fulfilled but precisely at  $p_2$  there is a change of the reflex vertices as shown in Figure 3.23. In  $p_2$  behind  $v_l$  a new left reflex vertex  $v'_l$  appears. Since Equation 3.9 holds we can conclude:

$$\begin{aligned} |w| &\leq K_{\phi_1} \ell_1 - K_{\phi_2} \ell_2 \\ &= K_{\phi_1} \ell_1 - K_{\phi_2} \ell_2 + K_{\phi_2} \ell'_2 - K_{\phi_2} \ell'_2 \\ &\leq K_{\phi_1} (\ell_1 + \ell'_2) - K_{\phi_2} (\ell_2 + \ell'_2) \end{aligned} \quad (3.10)$$

The last inequality is true, since from Lemma 3.15 for  $\phi_2 > \phi_1 \geq \frac{\pi}{2}$  we have  $K_{\phi_2} < K_{\phi_1}$ . Note that  $\ell_1 + \ell'_2$  respectively  $\ell_2 + \ell'_2$  denote the lengths of the shortest paths from  $p_1$  respectively  $p_2$  to  $v'_l$ . Equation 3.10 says that the condition Equation 3.9 takes care that also for changes of the reflex vertices, we have obtained a  $K_{\phi_1}$  competitive strategy at  $p_1$   $\square$

Assume that Equation 3.9 holds for all small changes of opening angles for the overall path  $W$  from  $s$  to  $p_{end}$ , we conclude

$$|W| \leq \min\{K_{\phi_0} \cdot |P_L| - K_{\pi} \ell_{End}, K_{\phi_0} \cdot |P_R| - K_{\pi} r_{End}\}.$$

Altogether we have a  $K_{\phi_0}$  competitive strategy in this case.

Now it is sufficient to guarantee that the agent fulfils Equation 3.9 during the movements. The idea of fulfilling this requirement is as follows: The portions  $K_{\phi_1} \ell_1 - K_{\phi_2} \ell_2$  and  $K_{\phi_1} r_1 - K_{\phi_2} r_2$  somehow express how many path length  $w$  we can use in the next step for the left or the right location of the goal,

respectively. Since we do not know where the target will be at the end, we do not want to let one side have an advantage at this stage.

Therefore we would like to guarantee that both values are the same. This gives

$$K_{\phi_2}(\ell_2 - r_2) = K_{\phi_1}(\ell_1 - r_1).$$

Fortunately, by this requirement we indeed define a special curve for any starting situation with angle  $\phi_0$  and length  $l_0$  and  $r_0$ . Let  $A = K_{\phi_0}(\ell_0 - r_0)$ . The curve that fulfils the above equation all the time is given by

$$\begin{aligned} X(\phi) &= \frac{A}{2} \cdot \frac{\cot \frac{\phi}{2}}{1 + \sin \phi} \cdot \sqrt{\left(1 + \tan \frac{\phi}{2}\right)^2 - A^2} \\ Y(\phi) &= \frac{1}{2} \cdot \cot \frac{\phi}{2} \cdot \left(\frac{A^2}{1 + \sin \phi} - 1\right). \end{aligned}$$

We will now explain how we have developed the formulas above. We choose a coordinate system with axis parallel to  $v_l v_r$ , the midpoint of  $v_l v_r$  is the origin. We scale such that  $|v_l v_r| = 1$ . Let  $p$  be the point on the curve with opening angle  $\phi$ ; see Figure 3.24. We have starting values  $\phi_0$ ,  $l_0$  and  $r_0$  and set  $A := K_{\phi_0}(\ell_0 - r_0)$ .

In order to find  $p$  we have to fulfil two conditions. First, the difference  $l(p) - r(p)$  of the distances from  $p$  to  $v_l$  and  $v_r$  has to equal  $\frac{A}{K_\phi}$ . The locus of all such point is a hyperbola. Second the angle at  $p$  with respect to  $v_l$  and  $v_r$  has to be  $\phi$ . The locus of all such points is a circle; see Figure 3.24. This holds because of the Thales' circle property.

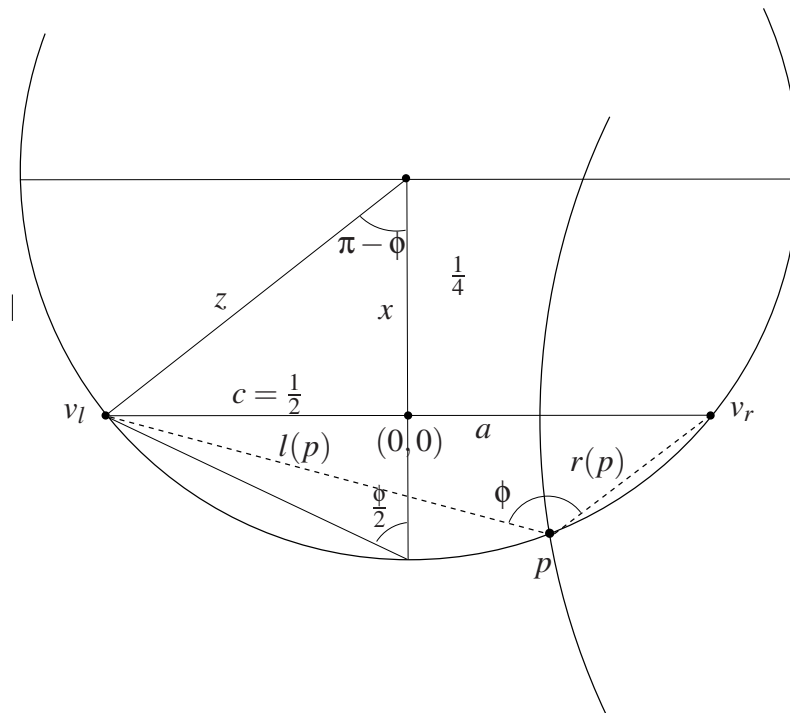


Figure 3.24: The left arc of the hyperbola is defined by  $v_l$ ,  $v_r$  and  $(l(p) - r(p)) = \frac{A}{K_\phi}$  and the circle running through  $v_l$  and  $v_r$  is defined by the opening angle  $\phi$ .

The hyperbola is defined by

$$\frac{X^2}{a^2} - \frac{Y^2}{b^2} = 1,$$



where  $2a = (l(p) - r(p)) = \frac{A}{K_\phi}$  and  $b^2 + a^2 = c^2 = \frac{1}{4}$  holds. This gives  $a^2 = \left(\frac{A}{2K_\phi}\right)^2$  and  $b^2 = \frac{1}{4} - \left(\frac{A}{2K_\phi}\right)^2$ . The circle is defined by

$$X^2 + (Y - x)^2 = z^2. \quad (3.10)$$

This means that we have to calculate  $x$  and  $z$ . From the law of sine we conclude

$$\begin{aligned} \frac{z}{\sin \frac{\pi}{2}} &= \frac{1}{2 \sin(\pi - \phi)} = \frac{1}{2 \sin \phi} \\ \frac{z - x}{\sin\left(\pi - \frac{\pi}{2} - \frac{\phi}{2}\right)} &= \frac{z - x}{\cos \frac{\phi}{2}} = \frac{1}{2 \sin \frac{\phi}{2}} \end{aligned}$$

and therefore  $z = \frac{1}{2 \sin \phi}$  and

$$x = z - \frac{1}{2} \cot \frac{\phi}{2} = \frac{1}{2 \sin \phi} - \frac{1}{2} \cot \frac{\phi}{2} = \frac{1 - 2 \cos^2 \frac{\phi}{2}}{4 \sin \frac{\phi}{2} \cos \frac{\phi}{2}} = -\frac{\cot \phi}{2}.$$

The intersection of the hyperbola and the circle is indeed given by the above functions  $X(\phi)$  and  $Y(\phi)$ . We have found the solutions by a computer algebra system. Here we simply verify that the solutions are correct. We insert the values into the hyperboly and the circle description.

$$\frac{X^2}{\left(\frac{A}{2K_\phi}\right)^2} - \frac{Y^2}{\left(\frac{1}{2}\right)^2 - \left(\frac{A}{2K_\phi}\right)^2} = 1 \quad (3.11)$$

$$X^2 + \left(Y + \frac{\cot \phi}{2}\right)^2 = \frac{1}{4 \sin^2 \phi}. \quad (3.12)$$

For (3.11) we have

$$\begin{aligned} &\frac{\left(\frac{A}{2} \cdot \frac{\cot \frac{\phi}{2}}{1 + \sin \phi} \sqrt{\left(1 + \tan \frac{\phi}{2}\right)^2 - A^2}\right)^2}{\left(\frac{A}{2K_\phi}\right)^2} - \frac{\left(\frac{1}{2} \cot \frac{\phi}{2} \left(\frac{A^2}{1 + \sin \phi} - 1\right)\right)^2}{\left(\frac{1}{2}\right)^2 - \left(\frac{A}{2K_\phi}\right)^2} = \\ &\left(\frac{\cot \frac{\phi}{2}}{K_\phi}\right)^2 \left(\left(1 + \tan \frac{\phi}{2}\right)^2 - A^2\right) - \frac{\cot^2 \frac{\phi}{2} \left(\left(\frac{A}{K_\phi}\right)^2 - 1\right)^2}{1 - \left(\frac{A}{K_\phi}\right)^2} = \\ &\left(\frac{\cot \frac{\phi}{2}}{K_\phi}\right)^2 \left(\left(1 + \tan \frac{\phi}{2}\right)^2 - A^2\right) + \cot^2 \frac{\phi}{2} \left(\left(\frac{A}{K_\phi}\right)^2 - 1\right) = \\ &\cot^2 \frac{\phi}{2} \left(\frac{\left(1 + \tan \frac{\phi}{2}\right)^2}{1 + \sin \phi} - 1\right) = 1. \end{aligned}$$

The conclusion is valid since the following identity holds.

$$1 + \sin \phi = 1 + \frac{2 \tan \frac{\phi}{2}}{1 + \tan^2 \frac{\phi}{2}} = \frac{\left(1 + \tan \frac{\phi}{2}\right)^2}{1 + \tan^2 \frac{\phi}{2}} \quad (3.13)$$

For showing (3.12) we proceed as follows:

$$\begin{aligned}
& \left( \frac{A}{2} \cdot \frac{\cot \frac{\phi}{2}}{1 + \sin \phi} \sqrt{\left(1 + \tan \frac{\phi}{2}\right)^2 - A^2} \right)^2 + \left( \frac{1}{2} \cot \frac{\phi}{2} \left( \frac{A^2}{1 + \sin \phi} - 1 \right) + \frac{\cot \phi}{2} \right)^2 = \\
& \quad \left( \frac{A}{2} \cdot \frac{\cot \frac{\phi}{2}}{1 + \sin \phi} \right)^2 \left( \left(1 + \tan \frac{\phi}{2}\right)^2 - A^2 \right) + \\
& \quad \left( \frac{1}{2} \cot \frac{\phi}{2} \left( \frac{A^2}{1 + \sin \phi} - 1 \right) \right)^2 + \cot \frac{\phi}{2} \left( \frac{A^2}{1 + \sin \phi} - 1 \right) \frac{\cot \phi}{2} + \left( \frac{\cot \phi}{2} \right)^2 = \\
& \quad \left( \frac{A}{2} \cdot \frac{\cot \frac{\phi}{2}}{1 + \sin \phi} \right)^2 \left(1 + \tan \frac{\phi}{2}\right)^2 + \\
& \quad \left( \frac{1}{2} \cot \frac{\phi}{2} \right)^2 \left( -2 \frac{A^2}{1 + \sin \phi} + 1 \right) + \cot \frac{\phi}{2} \left( \frac{A^2}{1 + \sin \phi} - 1 \right) \frac{\cot \phi}{2} + \left( \frac{\cot \phi}{2} \right)^2 = \\
& \quad \left( \frac{\cot \frac{\phi}{2}}{2} - \frac{\cot \phi}{2} \right)^2 + \frac{A^2 \cot^2 \frac{\phi}{2}}{4(1 + \sin \phi)} \left( \frac{\left(1 + \tan \frac{\phi}{2}\right)^2}{1 + \sin \phi} - 2 + 2 \frac{\cot \phi}{\cot \frac{\phi}{2}} \right) = \\
& \quad \frac{1}{4 \sin^2 \phi} + \frac{A^2 \cot^2 \frac{\phi}{2}}{4(1 + \sin \phi)} \left( \tan^2 \frac{\phi}{2} + 1 - 2 + \frac{1 - \tan^2 \frac{\phi}{2}}{\tan \frac{\phi}{2}} \tan \frac{\phi}{2} \right) = \\
& \quad \frac{1}{4 \sin^2 \phi} + \frac{A^2 \cot^2 \frac{\phi}{2}}{4(1 + \sin \phi)} \cdot 0 = \frac{1}{4 \sin^2 \phi}.
\end{aligned}$$

Here we make use of the identity (3.13) and the equations

$$\left( \frac{\cot \frac{\phi}{2}}{2} - \frac{\cot \phi}{2} \right)^2 = \frac{1}{4} \left( \frac{\sin \phi}{1 - \cos \phi} - \frac{\cos \phi}{\sin \phi} \right)^2 = \frac{1}{4} \frac{1}{\sin^2 \phi}$$

and

$$\cot \phi = \frac{1 - \tan^2 \frac{\phi}{2}}{2 \tan \frac{\phi}{2}}.$$

Finally, we have to prove that the above curve indeed fulfils the condition for any small piece  $w$ . Experimentally, we make use of the precise curve description and import it into Geogebra or Maple. Here we approximate the path between any two points by the corresponding segment. This procedure already indicates that assumption has to be true.

It can also be shown analytically. A lengthy, detailed proof is given in [IKL99] or [Lan00]. Figure 3.25 shows examples for the curve for different values of  $\phi$  and  $A$ . The figure stems from a Maple plot.

We obtain the following result:

**Corollary 3.17** *For a funnel polygon with opening angle  $\phi_0 > \frac{\pi}{2}$  we will find any unknown target within a competitive ratio  $K_{\phi_0}$ .*

Finally, for angles  $0 < \phi_0 < \frac{\pi}{2}$  we can apply the same approach. Of course we can also apply the condition

$$K_{\phi_2}(\ell_2 - r_2) = K_{\phi_1}(\ell_1 - r_1)$$

for  $\phi_1 < \phi_2 < \frac{\pi}{2}$ .

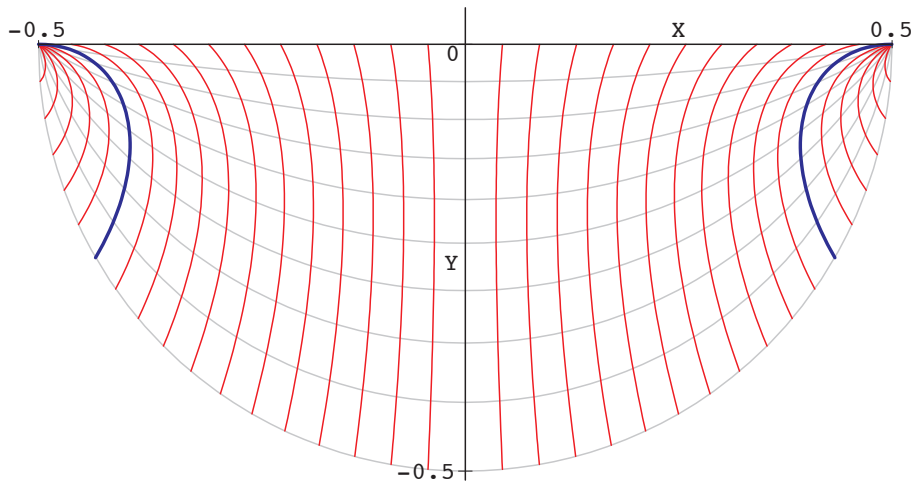


Figure 3.25: Curves  $(X(\phi), Y(\phi))$  depending from  $\phi$  and  $A$ .

Not that this will also result in a continuous extension of the curves of Figure 3.25. The problem is that these curve parts will not fulfil the condition Equation 3.9 because  $K_{\phi_1} < K_{\phi_2}$  holds. Therefore we just insert the fixed ratio  $\sqrt{2}$  which we would like to achieve at angle  $\frac{\pi}{2}$ . The factor  $\sqrt{2}$  dominates all  $K_\phi$ .

By the same arguments as before it is sufficient to guarantee

$$w \leq \min\{\sqrt{2}(\ell_1 - \ell_2), \sqrt{2}(r_1 - r_2)\}$$

for any small piece of our curve.

Again we would not prefer one side and set  $\ell_1 - \ell_2 = r_1 - r_2$ . This means that we are moving on the current angular bisector and call this strategy CAB (Current Angular Bisector); see also [IKL97, LOS96]. The analysis is also presented in [IKL99] oder [Lan00]. Note that if we apply the factor  $\sqrt{2}$  for the angles above  $\frac{\pi}{2}$  for the path  $w$  we will also define a curve but the above path length property for  $w$  does not hold.

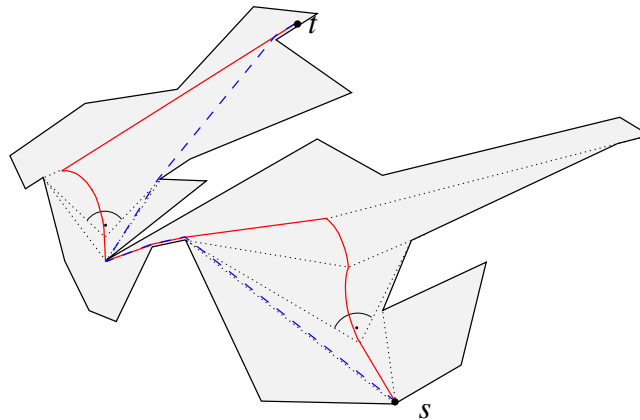


Figure 3.26: An example of the application of WCA.

Algorithm 3.1 summarizes the strategy, Figure 3.26 shows an example of its application. Altogether, the following result holds:

**Theorem 3.18** (Icking, Klein, Langetepe, Schuijver, Semrau, 1999)

Searching for the target  $t$  inside an unknown street polygon can be performed by an optimal  $\sqrt{2}$  competitive strategy. [IKL99, SS99, IKL<sup>+</sup>04]

We have implemented the optimal strategy under the name “WCA” (Worst-Case-Aware), an applet can be found here:

<http://www.geometrylab.de/>

---

**Algorithm 3.1** Searching for the target of a street.

---

While target  $t$  is not visible:

- Compute extreme reflex vertices  $v_\ell$  and  $v_r$ .
- If only one exists, move toward it.
- Otherwise repeat:
  - If a new reflex vertex  $v'_\ell$  or  $v'_r$  is detected: Replace  $v_\ell$  or  $v_r$  by  $v'_\ell$  or  $v'_r$ , respectively.
  - Let  $\phi$  be the current opening angle w.r.t.  $v_\ell$  and  $v_r$ .
  - If  $\phi \leq \frac{\pi}{2}$ : Follow the current angular bisector
  - If  $\phi > \frac{\pi}{2}$ : Follow the curve represented by  $X(\phi)$  and  $Y(\phi)$  with the current value  $A$ .
- Until either  $v_\ell$  or  $v_r$  is fully explored. Move to the vertex on the opposite side.

Move to the target  $t$ .

---



# Bibliography

- [Ad80] H. Abelson and A. A. diSessa. *Turtle Geometry*. MIT Press, Cambridge, 1980.
- [AFM00] E. M. Arkin, S. P. Fekete, and J. S. B. Mitchell. Approximation algorithms for lawn mowing and milling. *Comput. Geom. Theory Appl.*, 17:25–50, 2000.
- [AG03] Steve Alpern and Shmuel Gal. *The Theory of Search Games and Rendezvous*. Kluwer Academic Publications, 2003.
- [AKS02] Susanne Albers, Klaus Kursawe, and Sven Schuierer. Exploring unknown environments with obstacles. *Algorithmica*, 32:123–143, 2002.
- [BRS94] Margrit Betke, Ronald L. Rivest, and Mona Singh. Piecemeal learning of an unknown environment. Technical Report A.I. Memo No. 1474, Massachusetts Institute of Technology, March 1994.
- [BSMM00] Ilja N. Bronstein, Konstantin A. Semendjajew, Gerhard Musiol, and Heiner Mühlig. *Taschenbuch der Mathematik*. Verlag Harry Deutsch, Frankfurt am Main, 5th edition, 2000.
- [BYCR93] R. Baeza-Yates, J. Culberson, and G. Rawlins. Searching in the plane. *Inform. Comput.*, 106:234–252, 1993.
- [DHN97] G. Das, P. Heffernan, and G. Narasimhan. LR-visibility in polygons. *Comput. Geom. Theory Appl.*, 7:37–57, 1997.
- [DJMW91] G. Dudek, M. Jenkin, E. Milios, and D. Wilkes. Robotic exploration as graph construction. *Transactions on Robotics and Automation*, 7:859–865, 1991.
- [DKK01] Christian A. Duncan, Stephen G. Kobourov, and V. S. Anil Kumar. Optimal constrained graph exploration. In *Proc. 12th ACM-SIAM Symp. Discr. Algo.*, pages 307–314, 2001.
- [DKK06] Christian A. Duncan, Stephen G. Kobourov, and V. S. Anil Kumar. Optimal constrained graph exploration. *ACM Trans. Algor.*, 2:380–402, 2006.
- [EFK<sup>+</sup>06] Andrea Eubeler, Rudolf Fleischer, Tom Kamphans, Rolf Klein, Elmar Langetepe, and Gerhard Trippen. Competitive online searching for a ray in the plane. In Sándor Fekete, Rudolf Fleischer, Rolf Klein, and Alejandro López-Ortiz, editors, *Robot Navigation*, number 06421 in Dagstuhl Seminar Proceedings, 2006.
- [Gal80] Shmuel Gal. *Search Games*, volume 149 of *Mathematics in Science and Engineering*. Academic Press, New York, 1980.
- [GKP98] Ronald L. Graham, Donald E. Knuth, and Oren Patashnik. *Concrete Mathematics*. Addison-Wesley, 1998.
- [GR03] Yoav Gabriely and Elon Rimon. Competitive on-line coverage of grid environments by a mobile robot. *Comput. Geom. Theory Appl.*, 24:197–224, 2003.

- [HIKL99] Christoph Hipke, Christian Icking, Rolf Klein, and Elmar Langetepe. How to find a point on a line within a fixed distance. *Discrete Appl. Math.*, 93:67–73, 1999.
- [IKKL00a] Christian Icking, Thomas Kamphans, Rolf Klein, and Elmar Langetepe. Exploring an unknown cellular environment. In *Abstracts 16th European Workshop Comput. Geom.*, pages 140–143. Ben-Gurion University of the Negev, 2000.
- [IKKL00b] Christian Icking, Thomas Kamphans, Rolf Klein, and Elmar Langetepe. Exploring an unknown cellular environment. Unpublished Manuscript, FernUniversität Hagen, 2000.
- [IKKL05] Christian Icking, Tom Kamphans, Rolf Klein, and Elmar Langetepe. Exploring simple grid polygons. In *11th Internat. Comput. Combin. Conf.*, volume 3595 of *Lecture Notes Comput. Sci.*, pages 524–533. Springer, 2005.
- [IKL97] Christian Icking, Rolf Klein, and Elmar Langetepe. Searching for the kernel of a polygon: A competitive strategy using self-approaching curves. Technical Report 211, Department of Computer Science, FernUniversität Hagen, Germany, 1997.
- [IKL99] Christian Icking, Rolf Klein, and Elmar Langetepe. An optimal competitive strategy for walking in streets. In *Proc. 16th Sympos. Theoret. Aspects Comput. Sci.*, volume 1563 of *Lecture Notes Comput. Sci.*, pages 110–120. Springer-Verlag, 1999.
- [IKL<sup>+</sup>04] Christian Icking, Rolf Klein, Elmar Langetepe, Sven Schuierer, and Ines Semrau. An optimal competitive strategy for walking in streets. *SIAM J. Comput.*, 33:462–486, 2004.
- [IPS82] A. Itai, C. H. Papadimitriou, and J. L. Szwarcfiter. Hamilton paths in grid graphs. *SIAM J. Comput.*, 11:676–686, 1982.
- [KL03] Tom Kamphans and Elmar Langetepe. The Pledge algorithm reconsidered under errors in sensors and motion. In *Proc. of the 11th Workshop on Approximation and Online Algorithms*, volume 2909 of *Lecture Notes Comput. Sci.*, pages 165–178. Springer, 2003.
- [Kle91] Rolf Klein. Walking an unknown street with bounded detour. In *Proc. 32nd Annu. IEEE Sympos. Found. Comput. Sci.*, pages 304–313, 1991.
- [Kle97] Rolf Klein. *Algorithmische Geometrie*. Addison-Wesley, Bonn, 1997.
- [Lan00] Elmar Langetepe. *Design and Analysis of Strategies for Autonomous Systems in Motion Planning*. PhD thesis, Department of Computer Science, FernUniversität Hagen, 2000.
- [LB99] Sharon Laubach and Joel Burdick. RoverBug: Long range navigation for mars rovers. In Peter Corke and James Trevelyan, editors, *Proc. 6th Int. Symp. Experimental Robotics*, volume 250 of *Lecture Notes in Control and Information Sciences*, pages 339–348. Springer, 1999.
- [Lee61] C. Y. Lee. An algorithm for path connections and its application. *IRE Trans. on Electronic Computers*, EC-10:346–365, 1961.
- [LOS96] Alejandro López-Ortiz and Sven Schuierer. Walking streets faster. In *Proc. 5th Scand. Workshop Algorithm Theory*, volume 1097 of *Lecture Notes Comput. Sci.*, pages 345–356. Springer-Verlag, 1996.
- [LS87] V. J. Lumelsky and A. A. Stepanov. Path-planning strategies for a point mobile automaton moving amidst unknown obstacles of arbitrary shape. *Algorithmica*, 2:403–430, 1987.
- [Sch01] S. Schuierer. Lower bounds in on-line geometric searching. *Comput. Geom. Theory Appl.*, 18:37–53, 2001.

- [Sha52] Claude E. Shannon. Presentation of a maze solving machine. In H. von Foerster, M. Mead, and H. L. Teuber, editors, *Cybernetics: Circular, Causal and Feedback Mechanisms in Biological and Social Systems, Transactions Eighth Conference, 1951*, pages 169–181, New York, 1952. Josiah Macy Jr. Foundation. Reprint in [Sha93].
- [Sha93] Claude E. Shannon. Presentation of a maze solving machine. In Neil J. A. Sloane and Aaron D. Wyner, editors, *Claude Shannon: Collected Papers*, volume PC-03319. IEEE Press, 1993.
- [SM92] A. Sankaranarayanan and I. Masuda. A new algorithm for robot curvefollowing amidst unknown obstacles, and a generalization of maze-searching. In *Proc. 1992 IEEE Internat. Conf. on Robotics and Automation*, pages 2487–2494, 1992.
- [SS99] Sven Schuierer and Ines Semrau. An optimal strategy for searching in unknown streets. In *Proc. 16th Sympos. Theoret. Aspects Comput. Sci.*, volume 1563 of *Lecture Notes Comput. Sci.*, pages 121–131. Springer-Verlag, 1999.
- [Sut69] Ivan E. Sutherland. A method for solving arbitrary wall mazes by computer. *IEEE Trans. on Computers*, 18(12):1092–1097, 1969.
- [SV90a] A. Sankaranarayanan and M. Vidyasagar. A new path planning algorithm for a point object amidst unknown obstacles in a plane. In *Proc. 1990 IEEE Internat. Conf. on Robotics and Automation*, pages 1930–1936, 1990.
- [SV90b] A. Sankaranarayanan and M. Vidyasagar. Path planning for moving a point object amidst unknown obstacles in a plane: A new algorithm and a general theory for algorithm developments. In *Proceedings of 1990 IEEE Conf. on Decision and Control*, pages 1111–1119, 1990.
- [SV91] A. Sankaranarayanan and M. Vidyasagar. Path planning for moving a point object amidst unknown obstacles in a plane: The universal lower bound on the worst case path lengths and a classification of algorithms. In *Proc. 1991 IEEE Internat. Conf. on Robotics and Automation*, pages 1734–1741, 1991.
- [THL98] L. H. Tseng, P. Heffernan, and D. T. Lee. Two-guard walkability of simple polygons. *Internat. J. Comput. Geom. Appl.*, 8(1):85–116, 1998.
- [Wal86] Wolfgang Walter. *Gewöhnliche Differentialgleichungen*. Springer, 1986.
- [Web07] Maximilian Weber. Online suche auf beschränkten sternern. Diplomarbeit, Rheinische Friedrich-Wilhelms-Universität Bonn, 2007.





# Index

$\dot{\cup}$ .....	<i>see</i> disjoint union	<b>D</b>	
1-Layer .....	<b>14</b>	DFS .....	8, 11
1-Offset .....	<b>14</b>	diagonally adjacent .....	<b>8</b> , 27
2-Layer .....	<b>14</b>	<i>Dijkstra</i> .....	19
2-Offset .....	<b>14</b>	<i>diSessa</i> .....	43
		disjoint union .....	<b>15</b>
		doubling heuristic .....	58
		<i>Dudek</i> .....	40
		<i>Duncan</i> .....	35, 37
lower bound .....	5	<b>E</b>	
		error bound .....	43
<b>A</b>		<b>F</b>	
<i>Abelson</i> .....	43	<i>Fekete</i> .....	30
accumulator strategy .....	31	functionals .....	58
adjacent .....	<b>8</b>	funnel (polygon) .....	<b>78</b>
<i>Albers</i> .....	30	funnel polygons .....	78
<i>Alpern</i> .....	59	funnel situation .....	78
angular counter .....	41		
approximation .....	30	<b>G</b>	
<i>Arkin</i> .....	30	<i>Gabriely</i> .....	27, 29
		<i>Gal</i> .....	59
		grid-environment .....	8
		gridpolygon .....	<b>8</b> , 30
<b>B</b>		<b>H</b>	
Backtrace .....	19	Hit-Point .....	<b>50</b>
backward analysis .....	79	Hit-Points .....	44
<i>Betke</i> .....	30	<b>I</b>	
Bug-Algorithms .....	49	<i>Icking</i> .....	5, 18, 21, 84
		<i>Itai</i> .....	8
<b>C</b>		<b>J</b>	
CAB .....	84	Java-Applet .....	18
caves .....	76	Java-Applets .....	41
cell .....	<b>8</b>	<i>Jenkin</i> .....	40
$C_{\text{free}}$ -condition .....	44		
$C_{\text{half}}$ -condition .....	45		
columns .....	29		
competitive .....	35, 37		
configuration space .....	44		
constrained .....	31		
Constraint graph-exploration .....	31		
cow-path .....	58		
current angular bisector .....	84		

**K**

*Kamphans* ..... 5, 18, 21, 47  
*Klein* ..... 5, 18, 21, 76, 84  
*Kobourov* ..... 35, 37  
*Kumar* ..... 35, 37  
*Kursawe* ..... 30

**L**

*Langetepe* ..... 5, 18, 21, 47, 84  
 Layer ..... 15  
 layer ..... 27  
 Leave-Point ..... **50**  
 Leave-Points ..... 44  
*Lee* ..... 19  
 Left-Hand-Rule ..... 10–13, 42  
 lost-cow ..... 58  
 Lower Bound ..... 9  
 lower bound ..... 8, 51, 76, 78  
*Lumelsky* ..... 50, 51, 53, 55

**M**

*Milios* ..... 40  
*Mitchell* ..... 30  
*m-ray-search* ..... 59

**N**

narrow passages ..... 20  
 Navigation ..... 41, 49  
 navigation ..... 57  
 NP-hart ..... 8

**O**

Offline-Strategy ..... **5**  
 Online-Strategy ..... **5**  
 Online-Strategy ..... 8

**P**

*Papadimitriou* ..... 8  
 partially occupied cells ..... **23**  
 path ..... **8**  
 periodic order ..... 60  
 piecemeal-condition ..... 30  
*Pledge* ..... 42

**Q**

Queue ..... 19

**R**

recurrence ..... 62  
*Rimon* ..... 27, 29  
*Rivest* ..... 30  
 RoverBug ..... 50

**S**

*Sankaranarayanan* ..... 50, 54, 55  
*Schuieler* ..... 30, 84  
 Search Games ..... 58  
 Searching ..... 41, 49  
 searching ..... 57  
 searching depth ..... 58  
*Semrau* ..... 84  
*Shannon* ..... 3  
*Singh* ..... 30  
*Sleator* ..... 5  
 SmartDFS ..... 13, 14  
 spanning tree ..... 23  
 Spanning-Tree-Covering ..... 23  
 split-cell ..... **14**  
*Stepanov* ..... 50, 51, 53, 55  
 street ..... **75**  
 street polygon ..... 75  
 sub-cells ..... **23**  
*Sutherland* ..... 3  
*Szwarcfiter* ..... 8

**T**

*Tarjan* ..... 5  
 tether strategy ..... 31  
 tool ..... 23  
 touch sensor ..... 8

**U**

unimodal ..... 59

**V**

*Vidyasagar* ..... 54, 55  
 visibility polygon ..... 57, **57**  
 visible ..... 57

**W**

Wave propagation ..... 19  
 weakly visible ..... 75  
*Wilkes* ..... 40  
 work space ..... 44