

# Online Motion Planning MA-INF 1314

## Searching

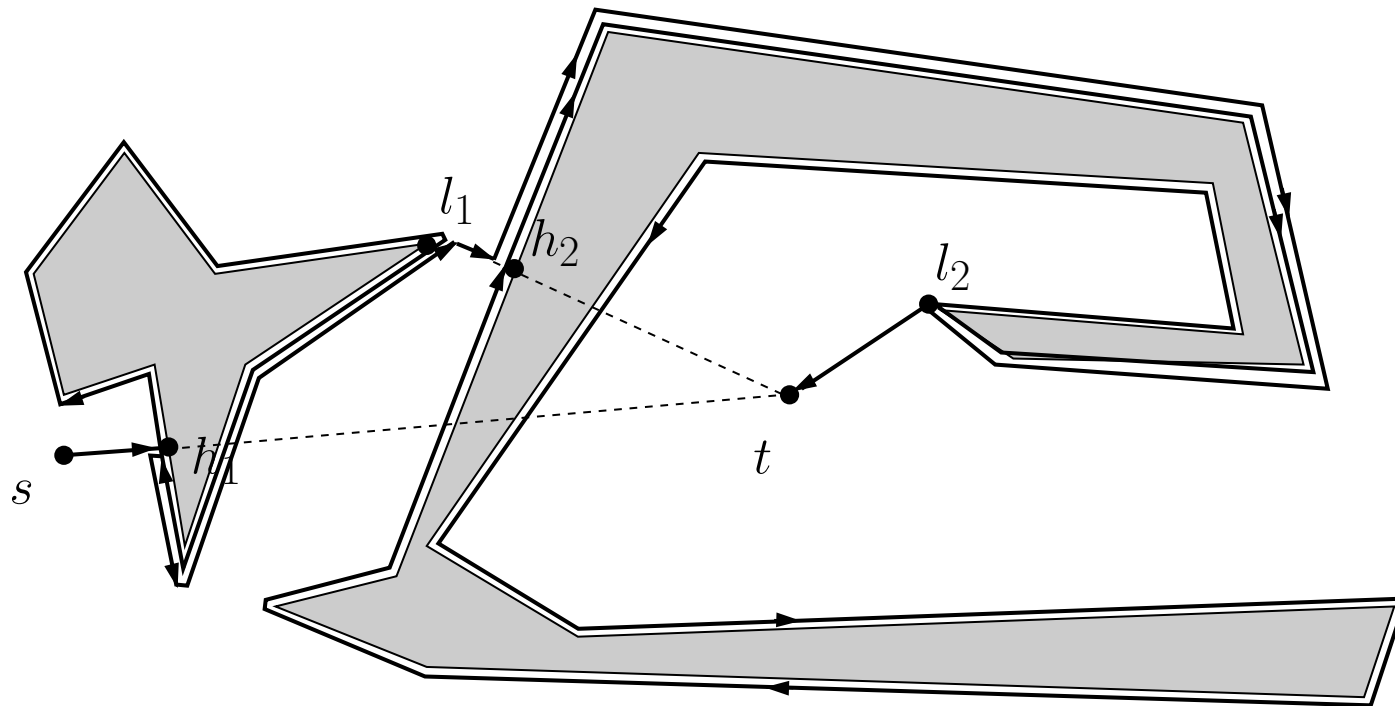
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# Rep.: Navigation

- Touch sensor, Target coordinates, Start  $s$ , Target  $t$ , Storage,
- Sojourner■
- Actions: ■
  - Move toward the target■
  - Move along the boundary ■
  - Sequence of Leave-Points  $l_i$ , Hit-Points  $h_i$ ■

# Rep.: BUG1 Strategy: Lumelsky/Stepanov

Toward target, surround obstacle, best leave point, toward target!

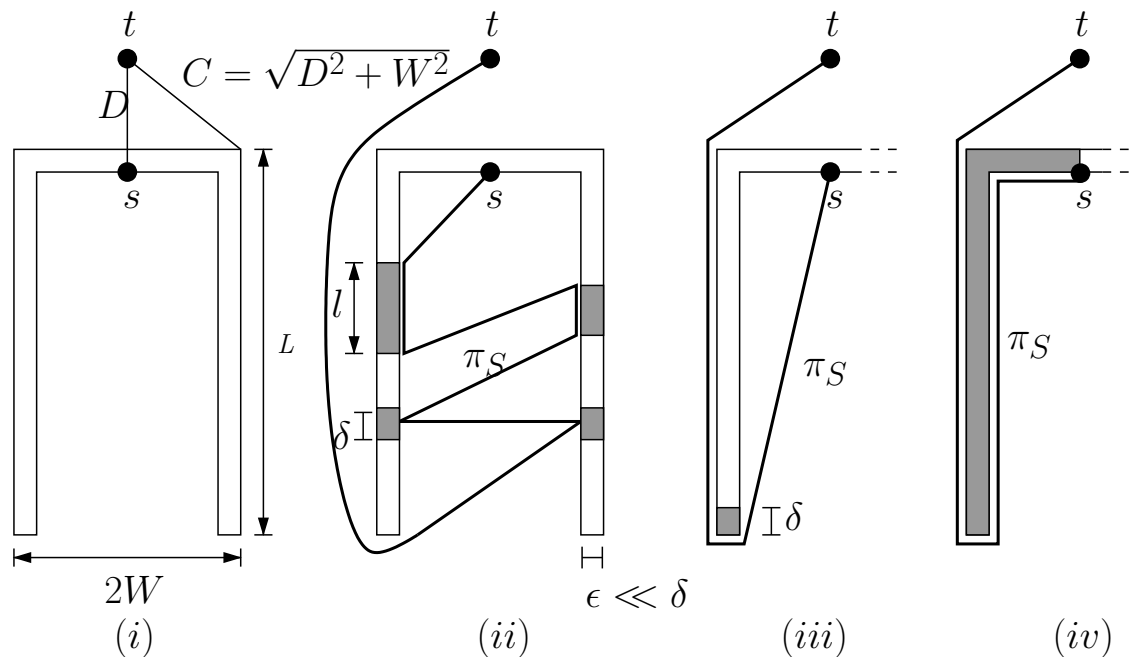


# Rep: Analysis BUG1 Strategy

- **Theorem** Strategy Bug1 is correct! ■
- **Theorem** Successful Bug1-path  $\Pi_{\text{Bug1}}$  from start  $s$  to target  $t$ :  
 $|\Pi_{\text{Bug1}}| \leq D + \frac{3}{2} \sum_i \text{UP}_i$ . ■
- **Theorem** For any strategy  $S$ , for arbitrary large  $K > 0$ , there exists examples for any  $D > 0$ , such that for any arbitrarily small  $\delta > 0$  we have:  $|\Pi_S| \geq K \geq D + \sum \text{UP}_i - \delta$ . ■
- **Korollar** Bug1 is  $\frac{3}{2}$ -competitive against any *online* strategy ■

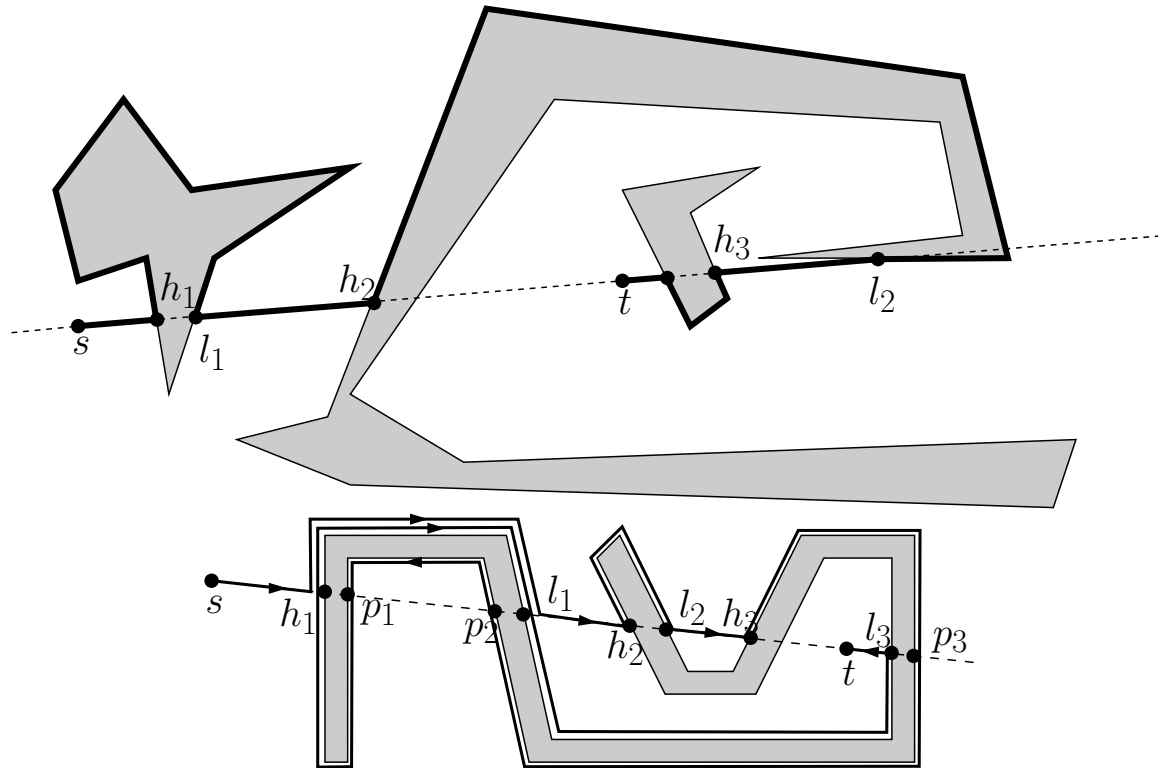
$$\text{Rep. LB: } |\Pi_S| \geq K \geq D + \sum \text{UP}_i - \delta$$

- Virtual horse-shoe , width  $2W$ , thickness  $\epsilon \ll \delta$ , length  $L$ , dist.  $D$
- Virtual gets concrete by touch
- Roughly surround any obstacle, by any strategy!



# Rep.: BUG2 Strategy

Line  $G$  passing  $st$ , toward target, surround obstacle, shorter distance on  $G$ , toward target!

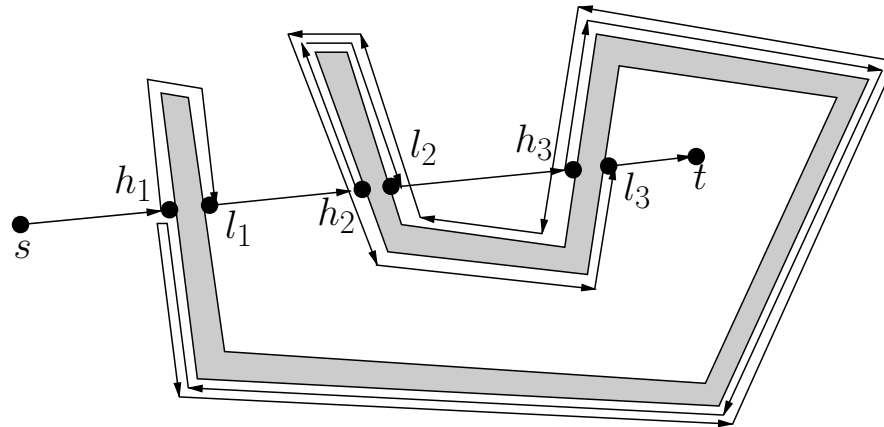


# Rep.: Analysis BUG2 Strategy

- **Lemma** Let  $n_i$  denote the number of intersection of  $G$  with relevant obstacle  $P_i$ . Bug2 meets any point on  $P_i$  at most  $\frac{n_i}{2}$  times.
- **Corollar** Bug2 is correct!
- **Theorem** Bug2-path  $\Pi_{\text{Bug2}}$  from  $s$  to  $t$ . We have:  
$$|\Pi_{\text{Bug2}}| \leq D + \sum_i \frac{n_i \text{UP}_i}{2}.$$

# Rep.: Change I

Change I, use former Leave/Hit Points once for !



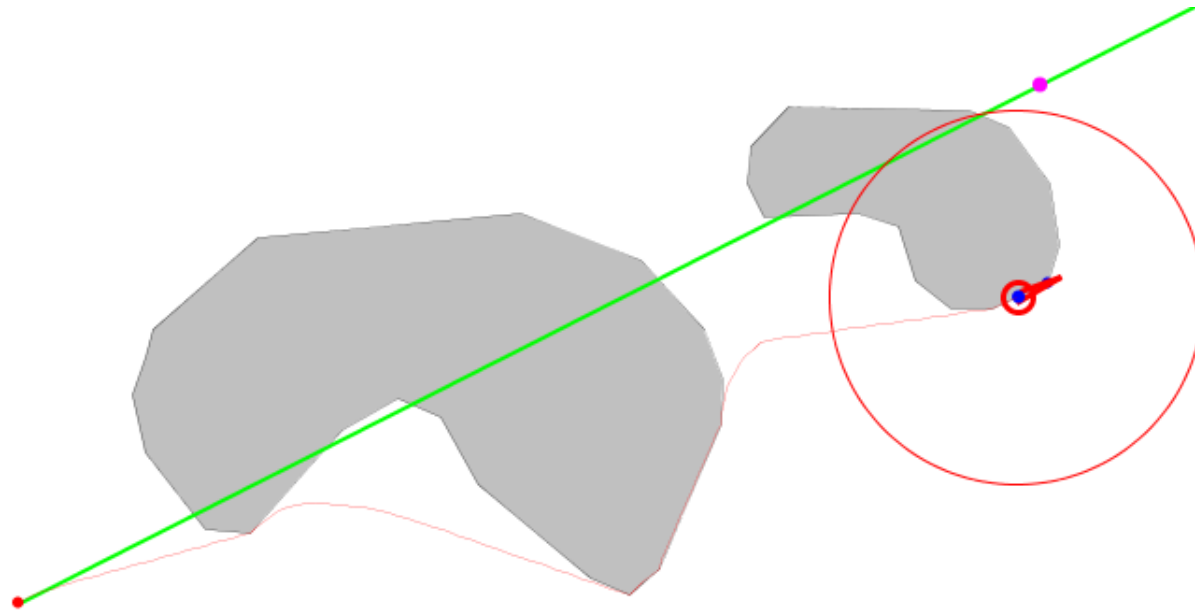
**Theorem:** Change I requires at most path length  $|\Pi_{\text{Change I}}| \leq D + 2 \sum_i \text{UP}_i$ . This is a tight bound! ■

Exercise! ■



# Different models

- Sensor with range: Circle around curr. point
- ● Short-cut for BUG2: VisBug
- Many others

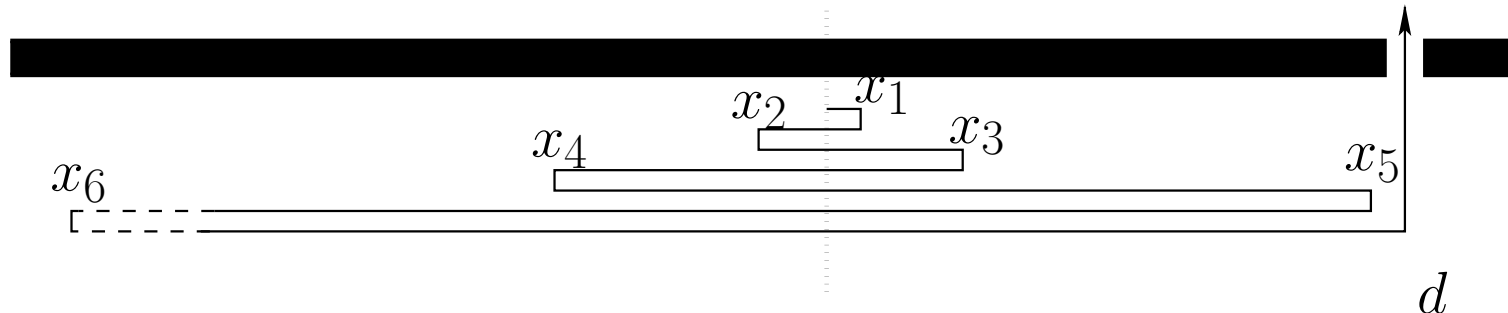


# Searching for a goal!

- Coordinates of the target unknown: Searching vs. Navigation■
- Polygonal environment■
- Full sight: Visibility polygon■
- **Def.** Let  $P$  be a simple polygon and  $r$  a point with  $s \in P$ . The visibility polygon of  $r$  w.r.t.  $P$ ,  $\text{Vis}_P(r)$ , is the set of all points  $q \in P$ , such that the segment  $\overline{r\bar{q}}$  is fully inside  $P$ .■
- Alg. Geom.: Compute in  $O(n)$  time! Offline!■

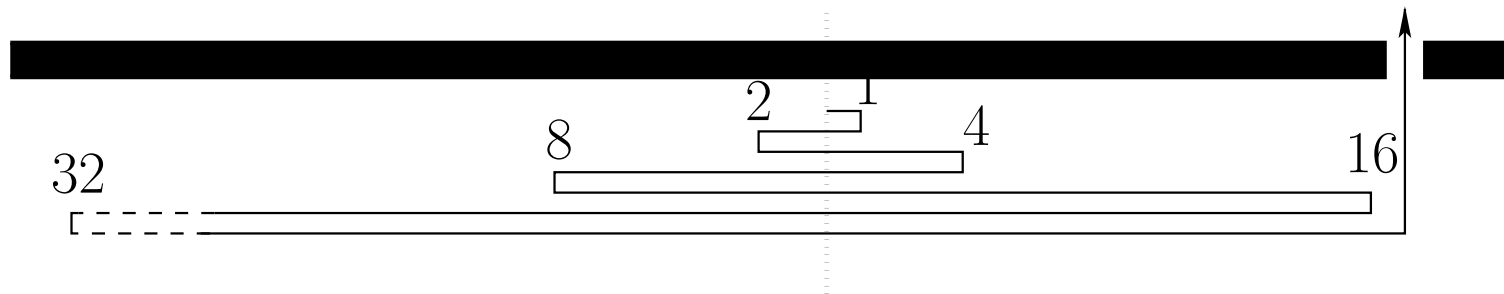
# Corridors (without sight)

- 2-ray search: Find door along a wall!■
- Compare to shortest path to the door, competitive?■
- Reasonable strategy: Depth  $x_1$  right, depth  $x_2$  left and so on ■
- Start-situation:  $2x_1 \geq C\epsilon$ , for any  $C > 0$  ex.  $\epsilon$ ■
- Additive constant or goal is at least step 1 away!■
- Local worst-case, not visited at  $d$ , once back!■
- Find strategy, such that:  $\sum_{i=1}^{k+1} 2x_i + x_k \leq Cx_k$  ■



# Corridors

- Worst-case, not visited at  $d$ , once back!
- Find strategy, such that:  $\sum_{i=1}^{k+1} 2x_i + x_k \leq Cx_k$  ■
- Minimize:  $\frac{\sum_{i=1}^{k+1} 2x_i + x_k}{x_k} = 1 + 2\frac{\sum_{i=1}^{k+1} x_i}{x_k}$  ■
- $x_i = 2^{i-1}$ , gives ratio  $C = 9$  ■
- Proof: Blackboard! ■



# Theorem Opt. of exponential solution: Gal 1980

- Strategy: Sequence  $X = f_1, f_2, \dots$  ■
- ● Minimize functional  $F_k(f_1, f_2, \dots) := \frac{\sum_{i=1}^{k+1} f_i}{f_k}$  for all  $k$  ■
- More precisely  $\inf_Y \sup_k F_k(Y) = C$  und  $\sup_k F_k(X) = C$  ■
- In general: Functional  $F_k$  continuous/unimodal: Unimodal:  
 $F_k(A \cdot X) = F_k(X)$  and  $F_k(X + Y) \leq \max\{F_k(X), F_k(Y)\}$  ■
- Some other helpful conditions! ■
- I.e.:  $F_{k+1}(f_1, \dots, f_{k+1}) \geq F_k(f_2, \dots, f_{k+1})$  ■
- **Theorem** Exponential function minimizes  $F_k$ :

$$\sup_k F_k(X) \geq \inf_a \sup_k F_k(A_a)$$

mit  $A_a = a^0, a^1, a^2, \dots$  und  $a > 0$ . ■

# Example: Exponential function

- $F_k(f_1, f_2, \dots) := \frac{\sum_{i=1}^{k+1} f_i}{f_k}$  for all  $k$ . ■
- Unimodal  $F_k(A \cdot X) = F_k(X)$  and  $F_k(X + Y) \leq \max\{F_k(X), F_k(Y)\}$ ? ■
- $\frac{\sum_{i=1}^{k+1} A \cdot f_i}{A \cdot f_k} = \frac{\sum_{i=1}^{k+1} f_i}{f_k}$
- $F_k(X + Y) \leq \max\{F_k(X), F_k(Y)\}$ ? ■
- Follows from  $\frac{a}{b} \geq \frac{c}{d} \Leftrightarrow \frac{a+c}{d+b} \leq \frac{a}{b}$  ■
- Simple equivalence ! ■
- Optimize:  $f_k(a) := \frac{\sum_{i=1}^{k+1} a^i}{a^k}$  ■
- Minimized by  $a = 2$  ■

# Theorem Gal 1980

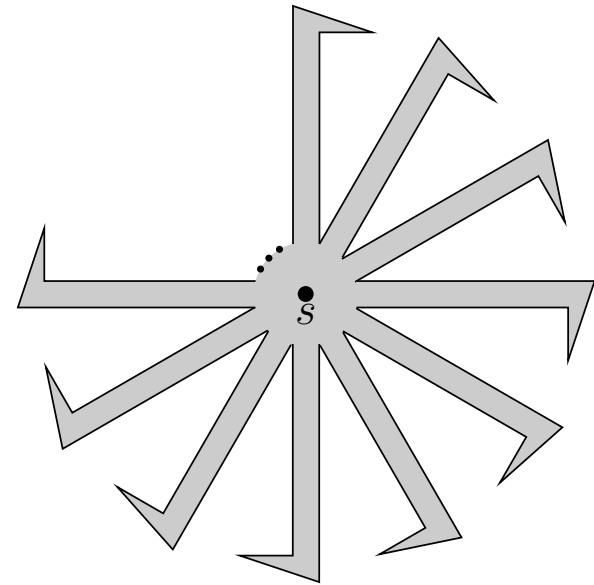
If functionally  $F_k$  has the following properties:

- i)  $F_k$  is continuous,
- ii)  $F_k$  is unimodal:  $F_k(A \cdot X) = F_k(X)$  and  $F_k(X + Y) \leq \max\{F_k(X), F_k(Y)\}$ ,
- iii)  $\liminf_{a \mapsto \infty} F_k \left( \frac{1}{a^{k+i}}, \frac{1}{a^{k+i-1}}, \dots, \frac{1}{a}, 1 \right) = \liminf_{\epsilon_{k+i}, \epsilon_{k+i-1}, \dots, \epsilon_1 \mapsto 0} F_k \left( \epsilon_{k+i}, \epsilon_{k+i-1}, \dots, \epsilon_1, 1 \right)$ ,
- iv)  $\liminf_{a \mapsto 0} F_k \left( 1, a, a^2, \dots, a^{k+i} \right) = \liminf_{\epsilon_{k+i}, \epsilon_{k+i-1}, \dots, \epsilon_1 \mapsto 0} F_k \left( 1, \epsilon_1, \epsilon_2, \dots, \epsilon_{k+i} \right)$ ,
- v)  $F_{k+1}(f_1, \dots, f_{k+i+1}) \geq F_k(f_2, \dots, f_{k+i+1})$ .

Then:  $\sup_k F_k(X) \geq \inf_a \sup_k F_k(A_a)$  with  $A_a = a^0, a^1, a^2, \dots$  and  $a > 0$ .

# Application m-ray search

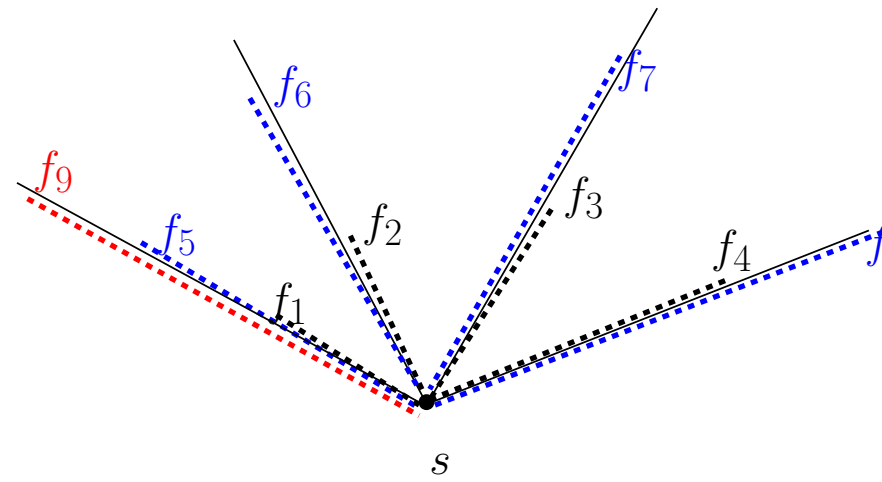
- Arbitrary  $m$ , not competitive, Fig.!
- $2m - 1$  vs.  $1!$
- Fixed  $m$ , infinite rays!
- Ass.: Rays in fixed order and increasing depth
- Tupel  $(f_j, J_j)$ : depth, next visit!





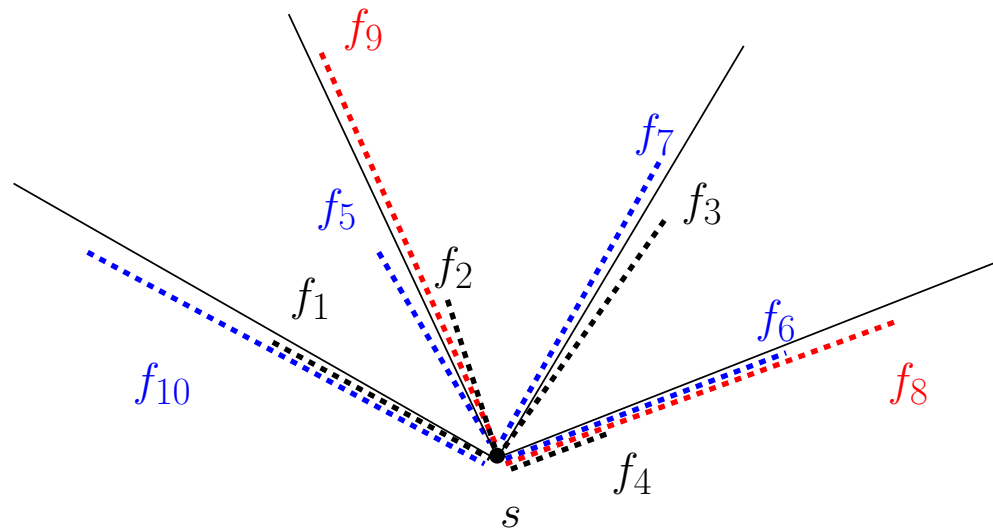
# Applicatiot m-ray search

- Ass.:  $(f_j, J_j)$ ,  $J_j = j + m$ ,  $f_j \geq f_{j-1}$  ■
- Visit rays in fixed order, increasing depth ■
- $F_k(f_1, f_2, \dots) := \frac{f_k + 2 \sum_{i=1}^{k+m-1} f_i}{f_k}$  for all  $k$  ■
- (Gal) Exp.-function minimizes  $F_k$ :  
 $\sup_k F_k(X) \geq \inf_a \sup_k F_k(A_a)$   
 with  $A_a = a^0, a^1, a^2, \dots$  and  $a > 1$ ,  
 optimal  $a = \frac{m}{m-1}$  ■
- Ratio:  $C = 1 + 2m \left( \frac{m}{m-1} \right)^{m-1}$  opt. ■



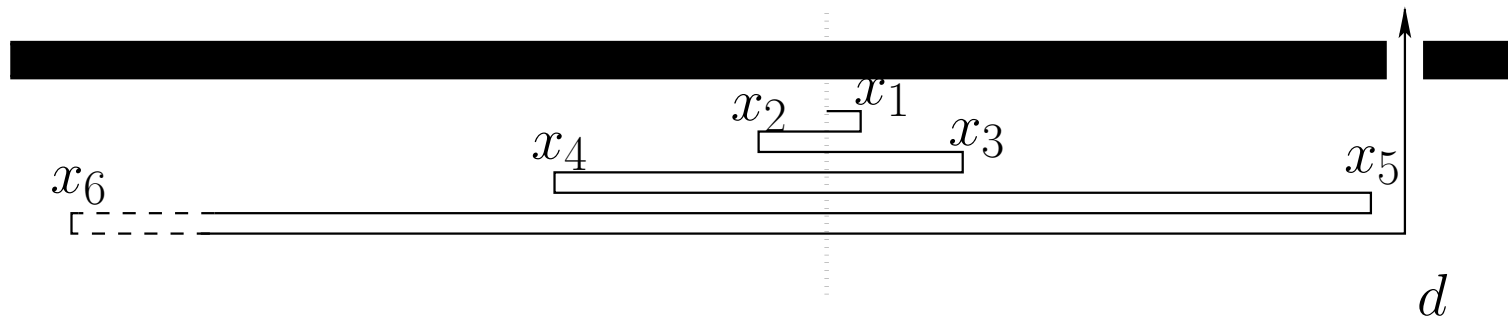
# m-ray search

- **Lemma** There is an optimal m-ray search strategy  $(f_1, f_2, \dots)$  that visits the rays in a fixed order and with increasing depth. ■
- periodic and monotone:  $(f_j, J_j)$ ,  $J_j = j + m$ ,  $f_j \geq f_{j-1}$  ■
- Second part: Proof blackboard! Change strategy! Conditions! ■



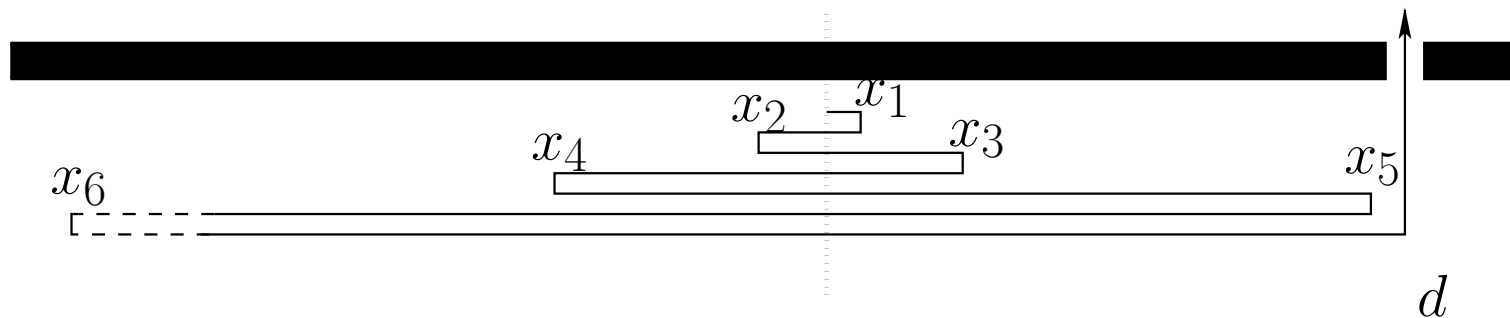
## Other approach: Optimality for equations!

- Reasonable strategy, ratio:  $\frac{\sum_{i=1}^{k+1} 2x_i + x_k}{x_k} = 1 + 2\frac{\sum_{i=1}^{k+1} x_i}{x_k}$  ■
- Ass.:  $C$  optimal,  $\frac{\sum_{i=1}^{k+1} x_i}{x_k} \leq \frac{(C-1)}{2}$  ■
- There is strategy  $(x'_1, x'_2, x'_3 \dots)$  s. th.  $\frac{\sum_{i=1}^{k+1} x'_i}{x'_k} = \frac{(C-1)}{2}$  for all  $k$  ■
- Monotonically increasing in  $x'_j$  ( $j \neq k$ ), decreasing in  $x'_k$  ■
- First  $k$  with:  $\frac{\sum_{i=1}^{k+1} x_i}{x_k} < \frac{(C-1)}{2}$ , decrease  $x_k$  ■
- $\frac{\sum_{i=1}^k x_i}{x_{k-1}} < \frac{(C-1)}{2}$ !,  $x_{k-1}$  decrease etc., monotonically decreasing sequence, bounde, converges! Non-constructive! ■



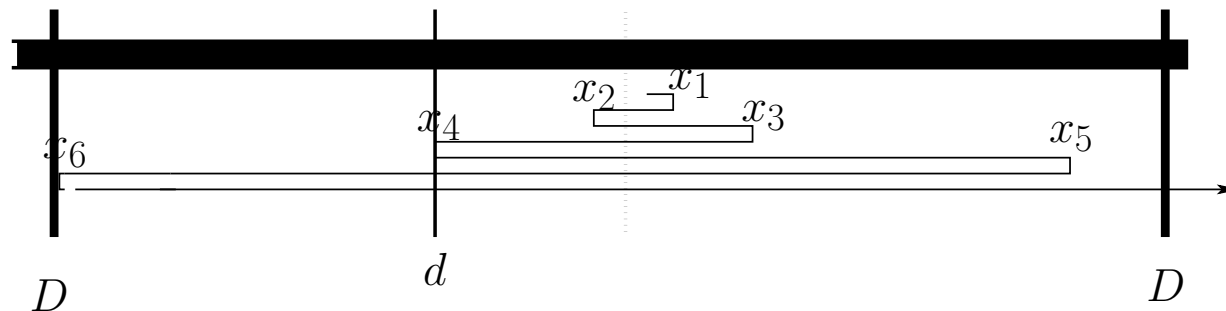
# Other approach: Optimality for equations!

- Set:  $\frac{\sum_{i=1}^{k+1} x'_i}{x'_k} = \frac{(C-1)}{2}$  for all  $k$ ■
- $\sum_{i=1}^{k+1} x'_i - \sum_{i=1}^k x'_i = \frac{(C-1)}{2} (x'_k - x'_{k-1})$ ■
- Thus:  $C' (x'_k - x'_{k-1}) = x'_{k+1}$ , Recurrence!■
- Solve a recurrence! Analytically! Blackboard!■
- Characteristical polynom: No solution  $C' < 4$ ■
- $x'_i = (i + 1)2^i$  with  $C' = 4$  is a solution! Blackboard! ■Optimal!■



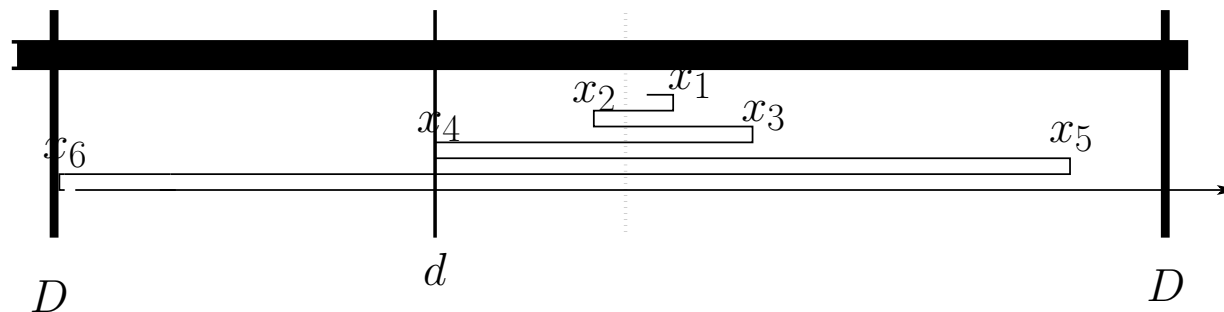
## 2-ray search, restricted distance

- Assume goal is no more than dist.  $\leq D$  away
- Exactly  $D$ ! Simple ratio 3!
- Find optimal strategy, minimize  $C$ !
- Vice-versa:  $C$  is given! Find the largest distance  $D$  (reach  $R$ ) that still allows  $C$  competitive search.
- One side with  $f_{\text{Ende}} = R$ , the other side arbitrarily large!



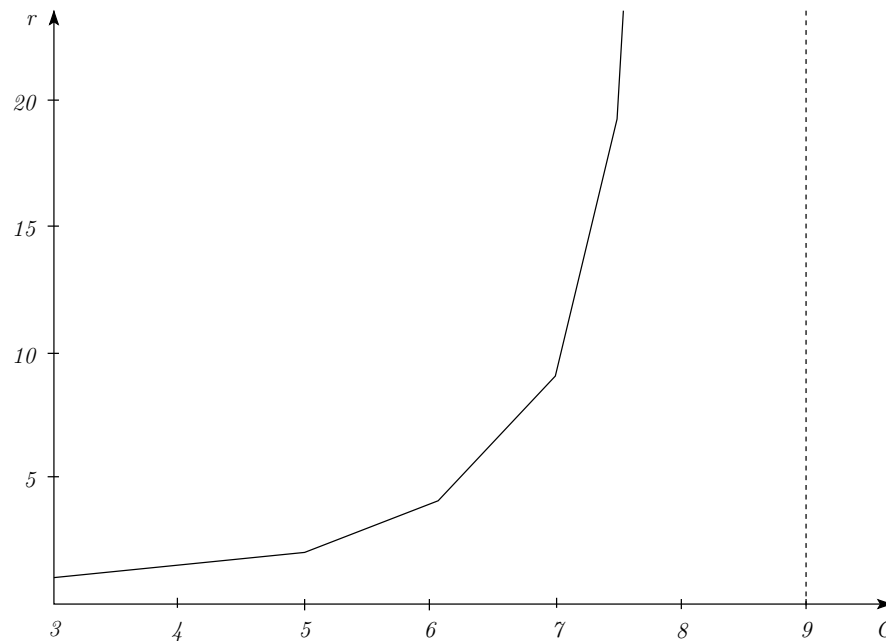
## 2-ray search, maximal reach $R$

- $C$  given, optimal reach  $R$ ! ■
- **Theorem** The strategy with equality in any step maximizes the reach  $R$  ! ■
- Strategy:  $\frac{\sum_{i=1}^{k+1} x_i}{x_k} = \frac{(C-1)}{2}$ , first step:  $x_1 = \frac{(C-1)}{2}$  ■
- Recurrence:  $x_0 = 1$ ,  $x_{-1} = 0$ ,  $x_{k+1} = \frac{(C-1)}{2}(x_k - x_{k-1})$  ■
- Strategy is optimal! By means of the Comp. Geom. lecture! ■



## 2-ray search, maximal reach $R$

- $f(C) :=$  maximal reach depending on  $C$
- Bends are more steps!



## 2-ray search, given distance $R$

- $f(C) :=$  maximal reach depending on  $C$
- Rotate,  $R$  given, binary search!

