



# Recovering Visibility and Dodging Obstacles in Pursuit-Evasion Games

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## Introduction

- ▶ Visibility-based pursuit-evasion:
  - ▷ Motion: holonomic, max speed per player
  - ▷ Visibility: omnidirectional, optional range
  - ▷ Traditionally, the game ends when the pursuer loses sight of the evader.
- ▶ Applications:
  - ▷ Surveillance, monitoring, FPS and racing games, etc.

## Discretization and Strategy Matrix

- ▶ Represent the map as a grid of pixels (white: clear, black: obstacle).
- ▶ Assume the players take turns, with the evader moving first.
- ▶ For each pair of positions  $(p, e)$ ,  $S[p, e] = 1$  iff the evader can win.
- ▶ Easily accommodates different motion and sensing models.
- ▶ Solve for the optimal strategy by backward induction.
- ▶  $\mathcal{N}(x)$  denotes the neighboring locations player  $x$  can reach in **1** turn.

## The Classical (Primal) Game

- ▶ Is it possible to keep the evader in sight? How?
- ▶ Initialize with visibility queries then solve the recurrence:

$$S[p, e, i] = \begin{cases} \neg v_t(p, e) & \text{if } i = 0, \\ \bigvee_{e' \in \mathcal{N}(e)} \bigwedge_{p' \in \mathcal{N}(p)} S[p', e', i - 1] & \text{otherwise.} \end{cases}$$

## The Visibility Induction Loop

**Input** : An initialized strategy matrix  $S$ .

```

1 begin
2   S' ← 0;
3   iter ← 0;
4   while S' ≠ S do
5     S' ← S;
6     foreach p ∈ w × h do
7       foreach e ∈ w × h do
8         foreach e' ∈ N(e) do
9           isExit ← True;
10          foreach p' ∈ N(p) do
11            if S'[p', e'] = 0 then
12              isExit ← False;
13          if isExit = True then
14            S[p, e] ← 1;
15            break;
16        iter ← iter + 1;
17    return S;
```

## The Dual Game

- ▶ Is it possible to stay out of the pursuer's sight? How?
- ▶ If a pursuer fails to keep an evader in sight, it may be able to recover.
- ▶ We get a new recurrence as the logical negation of the previous one:

$$S[p, e, i] = \begin{cases} \neg v_t(p, e) & \text{if } i = 0, \\ \bigwedge_{e' \in \mathcal{N}(e)} \bigvee_{p' \in \mathcal{N}(p)} \neg S[p', e', i - 1] & \text{otherwise.} \end{cases}$$

## The Dual Induction Loop

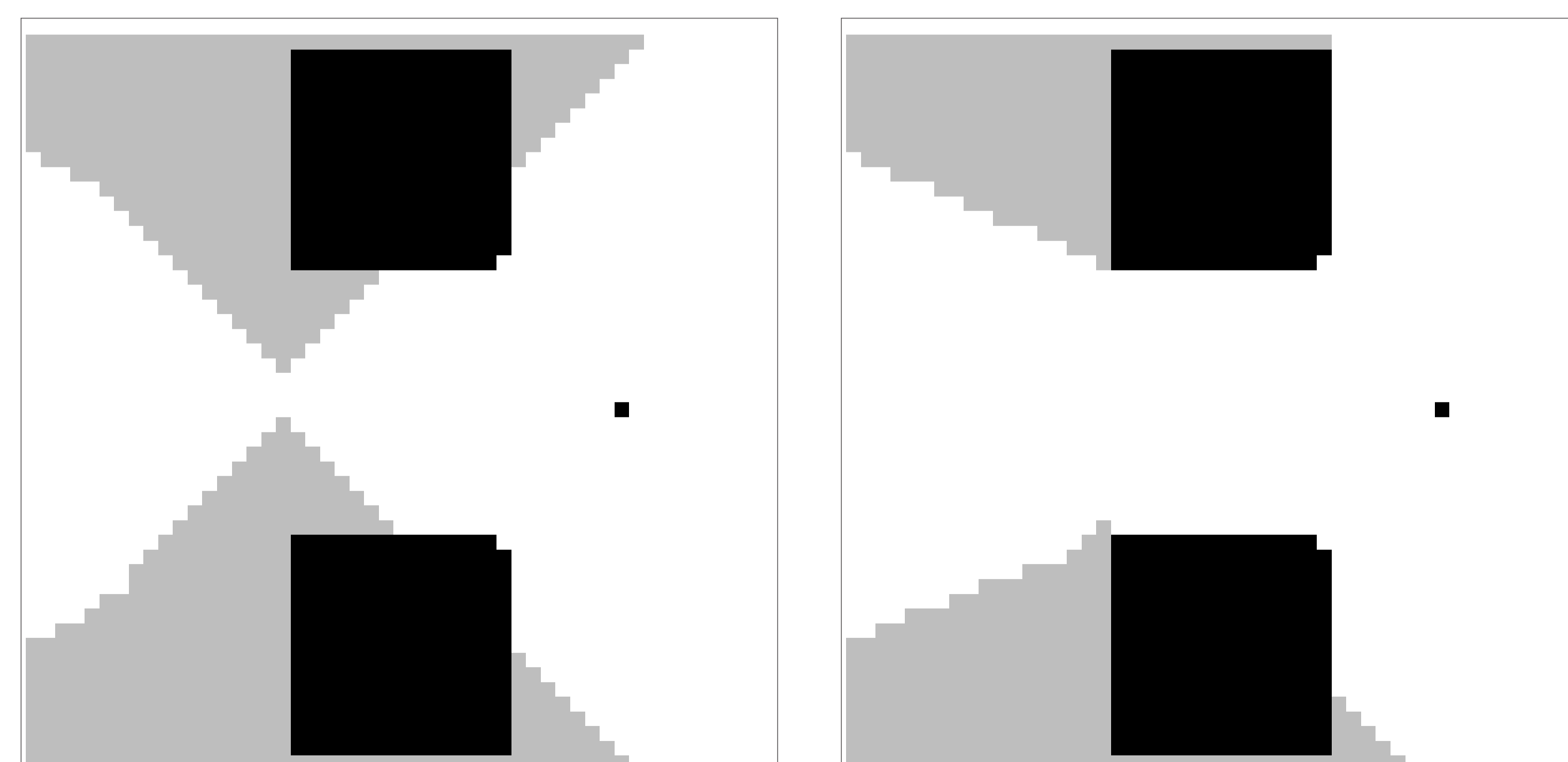
**Input** : A strategy matrix  $S$ , grid map  $M$ , tolerance  $d$ .

```

1 begin
2   iter ← 0;
3   while iter < d and S' ≠ S do
4     S' ← S;
5     foreach p ∈ w × h do
6       foreach e ∈ w × h do
7         hasExit ← False;
8         foreach e' ∈ N(e) do
9           isExit ← True;
10          foreach p' ∈ N(p) do
11            if M.vis(p', e') or S'[p', e'] = 0 then
12              isExit ← False;
13          if isExit = True then
14            hasExit ← True;
15          if hasExit = False then
16            S[p, e] ← 0;
17        iter ← iter + 1;
18    return S;
```

## Recovering Visibility

- ▶ Pursuer view in the case with  $d = 0$  (left) vs.  $d = 5$  (right).



## Dodging Obstacles

- ▶ Assume obstacle trajectories are known for a duration  $T$ .
- ▶ Time-varying visibility has to be incorporated into the recurrence:

$$S[p, e, t] = \begin{cases} \neg v_t(p, e) & \text{if } t = T, \\ \neg v_t(p, e) \vee \bigvee_{e' \in \mathcal{N}(e)} \bigwedge_{p' \in \mathcal{N}(p)} S[p', e', t + 1] & \text{otherwise.} \end{cases}$$

## The Dynamic Induction Loop

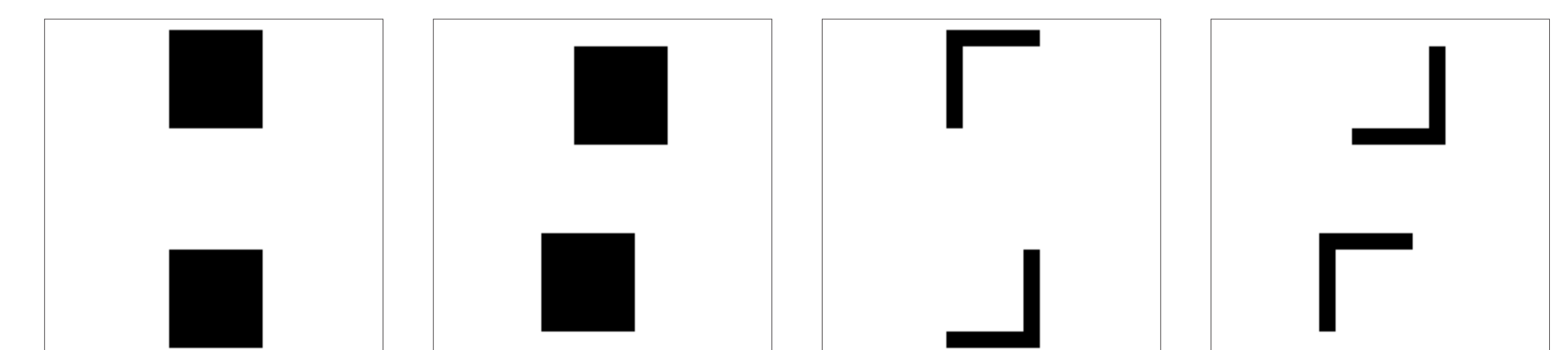
**Input** : A sequence of maps  $\{M_t\}$ ,  $t = 1 \dots T$ .

```

1 begin
2   S ← 0;
3   t ← T;
4   while t > 0 do
5     S' ← S;
6     foreach p ∈ w × h do
7       foreach e ∈ w × h do
8         foreach e' ∈ N(e) do
9           isExit ← True;
10          foreach p' ∈ N(p) do
11            if M_t.vis(p, e) and S'[p', e'] = 0 then
12              isExit ← False;
13          if isExit = True then
14            S[p, e] ← 1;
15            break;
16        t ← t - 1;
17    return S;
```

## Maintain optimal strategies after a single update in the map

- ▶ Move obstacles by add/remove. Use primal/dual induction to update  $S$ .



## Conclusions & Future Work

- ▶ Optimal strategies allowing visibility to be recovered within  $d$  turns
- ▶ Optimal strategies in dynamic environments
- ▶ Future directions
  - ▷ State space reduction for improved running times
  - ▷ Generate optimal strategies for more than two players