



# Graduate AI

## Lecture 3: Search II

Teachers:

Zico Kolter

Ariel Procaccia (this time)

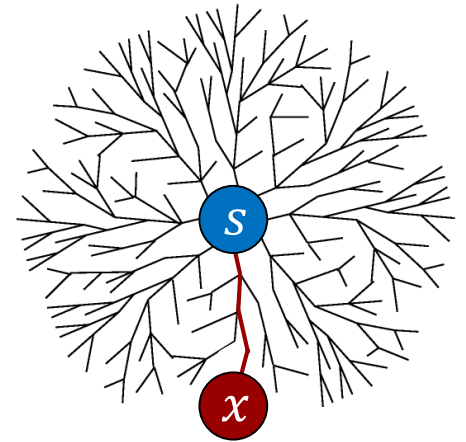
# A\* IS OPTIMALLY EFFICIENT

- We say that node  $x$  is **surely expanded** by A\* (tree search) if  $f(x) < f(t^*)$ , where  $t^*$  is the optimal goal
- **Theorem [Dechter and Pearl 1985]:** Any algorithm that is optimal given a consistent heuristic will expand, whenever the heuristic is consistent, all nodes surely expanded by A\*



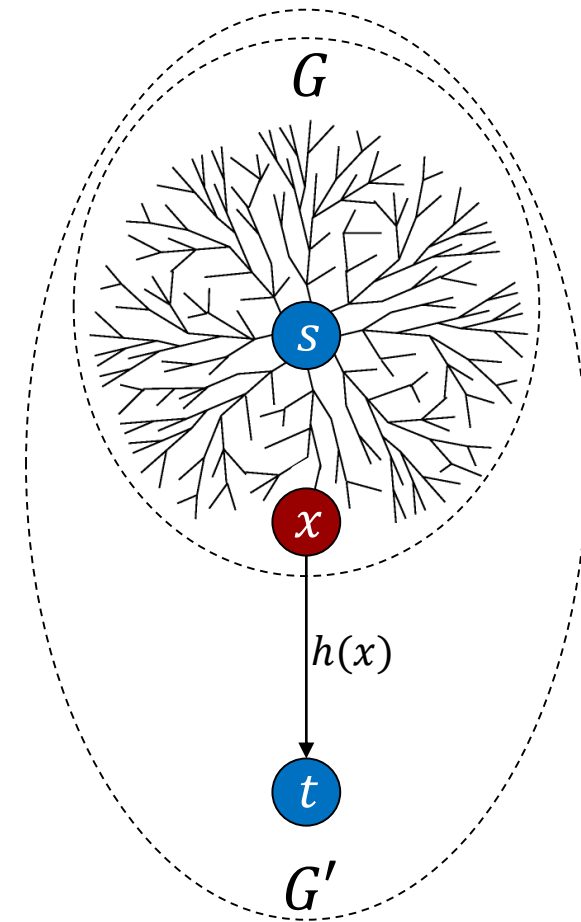
# PROOF OF THEOREM

- Let  $I$  be an instance with graph  $G$  and consistent heuristic  $h$
- Assume node  $x$  is surely expanded by  $A^*$
- Denote  $f(x) < f(t^*) = C^*$
- Let  $B$  be an optimal algorithm that does not expand  $x$



# PROOF OF THEOREM

- Create  $G'$  by adding a new goal  $t$ , and an edge  $(x, t)$  with cost  $h(x)$
- $h'$  is the same as  $h$ , and  $h'(t) = 0$
- **Lemma:**  $h'$  is consistent
  - Clearly true on pairs that do not include  $t$
  - For pairs  $(y, t)$ ,  
$$h'(y) = h(y) \leq c(y, x) + h(x)$$
$$= c(y, t) = c(y, t) + h'(t)$$

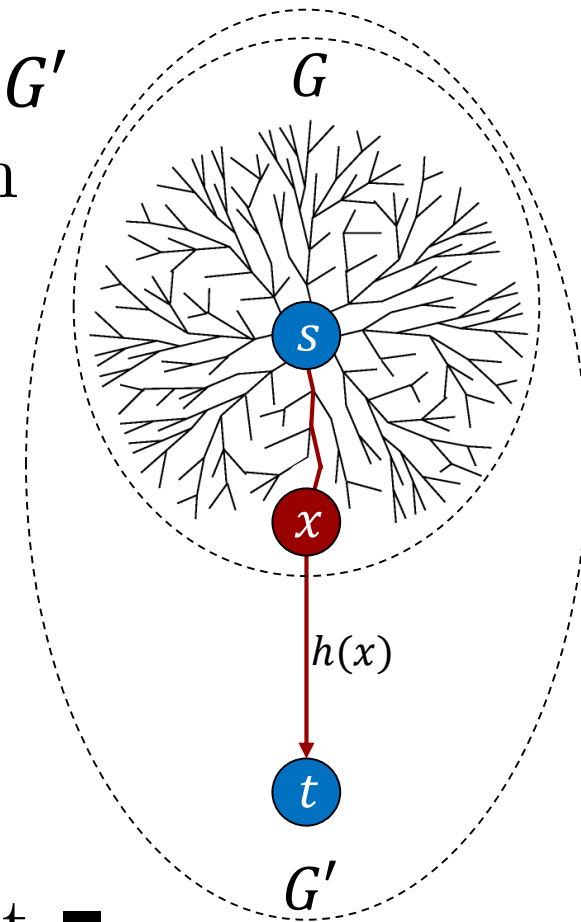


# PROOF OF THEOREM

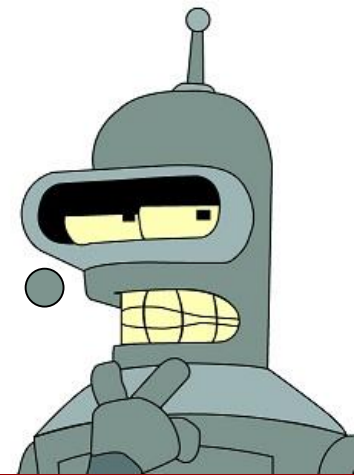
- On the new instance  $I'$  defined by  $G'$  and  $h'$ ,  $A^*$  will find the goal  $t$  with cost

$$g(t) + h'(t) = g(x) + h(x) < C^*$$

- Because  $B$  does not expand  $x$ ,  $I'$  looks identical to  $I$ , and  $B$  will find a solution of cost  $C^*$
- This is a contradiction to the assumption that  $B$  is optimal whenever the heuristic is consistent ■

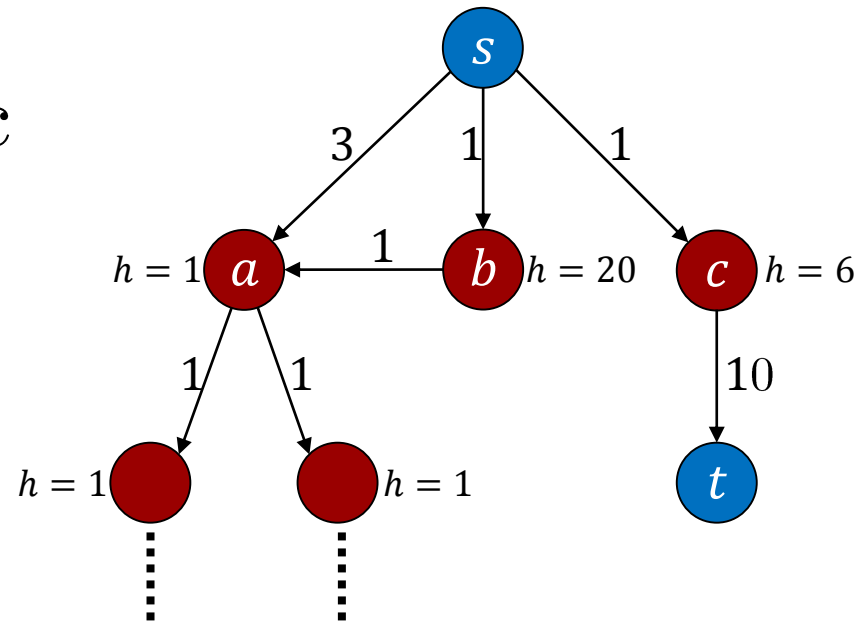


Why does the proof  
fail if we replace  
“consistent” with  
“admissible”?

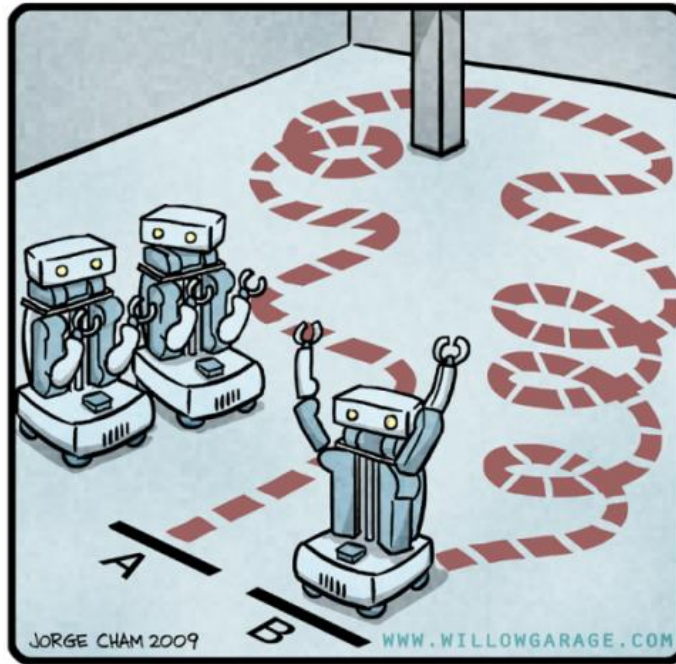


# A\* IS OPTIMALLY EFFICIENT

- In fact, the theorem is false when the heuristic is only admissible
- In the example on the right, algorithm B will find the optimal solution while expanding fewer nodes than A\*



Alg B: Conduct exhaustive search except for expanding  $a$ ; then expand  $a$  only if it has the potential to sprout cheaper solution



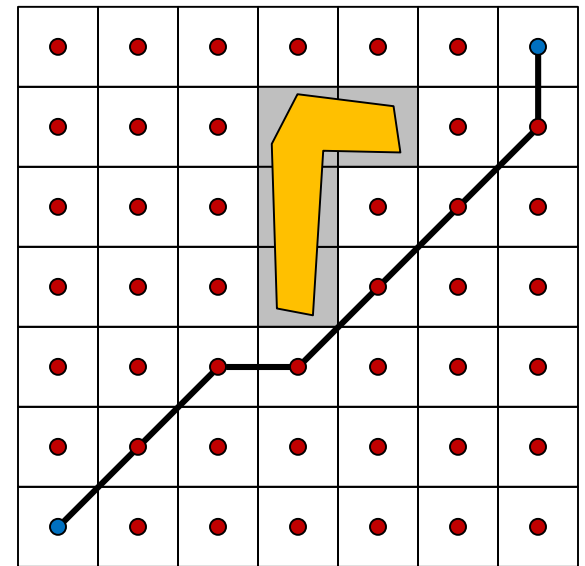
"HIS PATH-PLANNING MAY BE  
SUB-OPTIMAL, BUT IT'S GOT FLAIR."

# APPLICATION: MOTION PLANNING

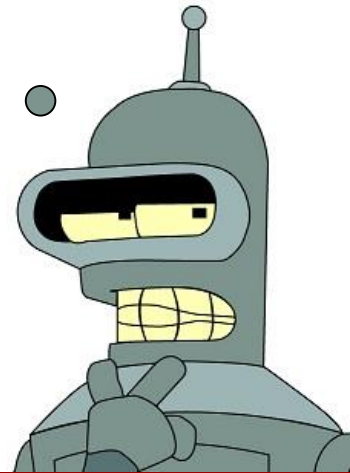


# MOTION PLANNING

- Navigating between two points while avoiding obstacles
- A first approach: define a discrete grid
- Mark cells that intersect obstacles as blocked
- Find path through centers of remaining cells

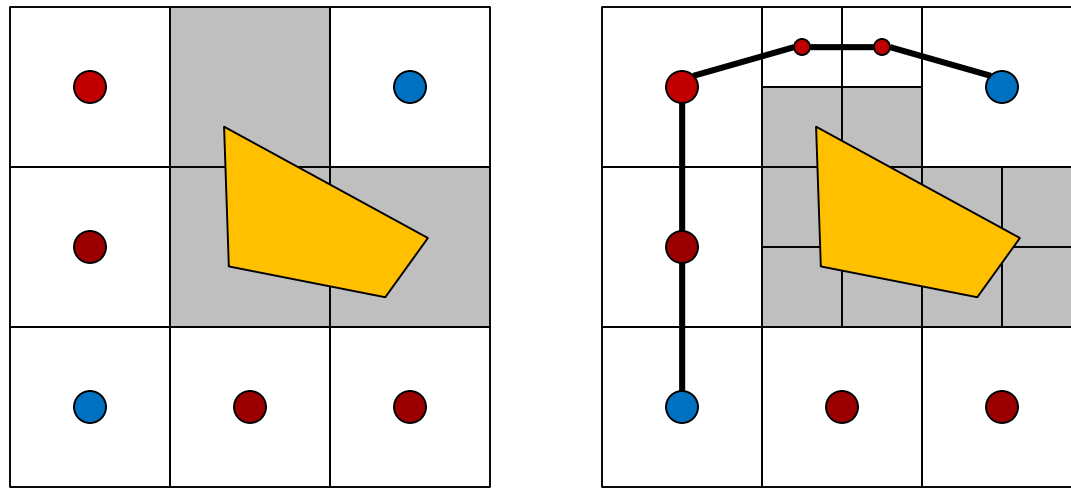


Is this approach  
optimal?  
Complete?



# CELL DECOMPOSITION

- Distinguish between
  - Cells that are contained in obstacles
  - Cells that intersect obstacles
- If no path found, subdivide the mixed cells

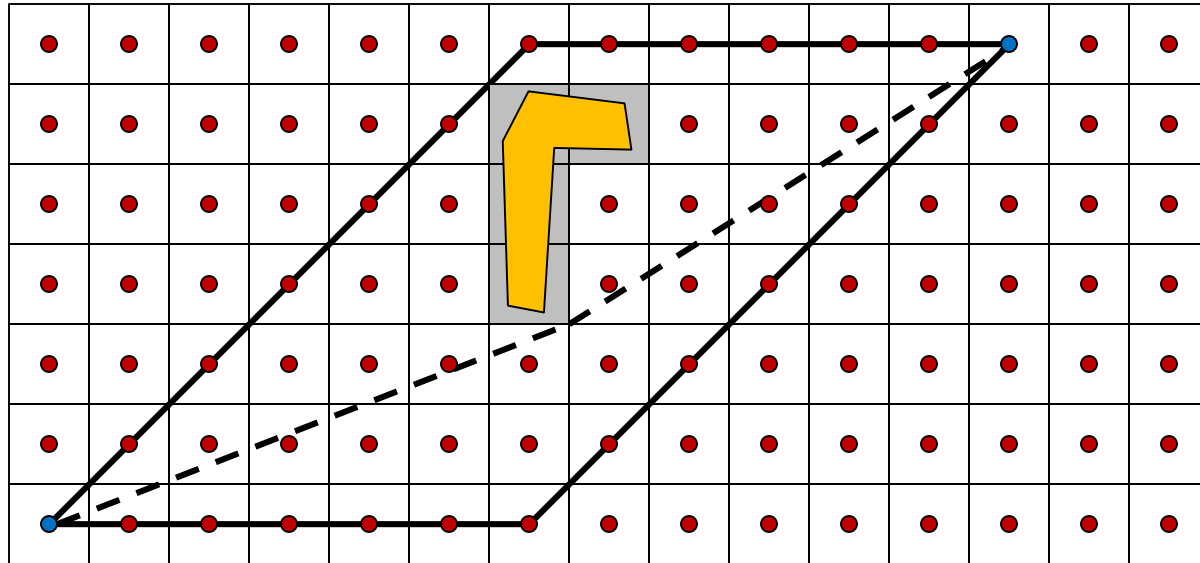


# IS IT COMPLETE NOW?

- An algorithm is **resolution complete** when:
  - a. If a path exists, it finds it in finite time
  - b. If a path does not exist, it returns in finite time
- Assume obstacles are closed sets, so free space is an open set
- **Poll 1:** Cell decomposition satisfies:
  1. a but not b
  2. b but not a
  3. Both a and b
  4. Neither a nor b



# CELL DECOMPOSITION

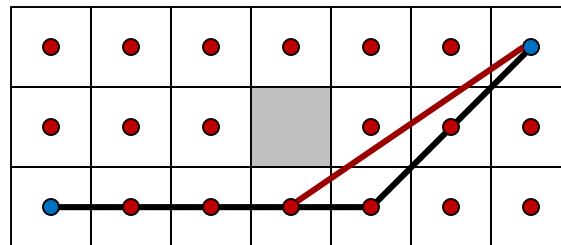


—— Shortest paths through cell centers

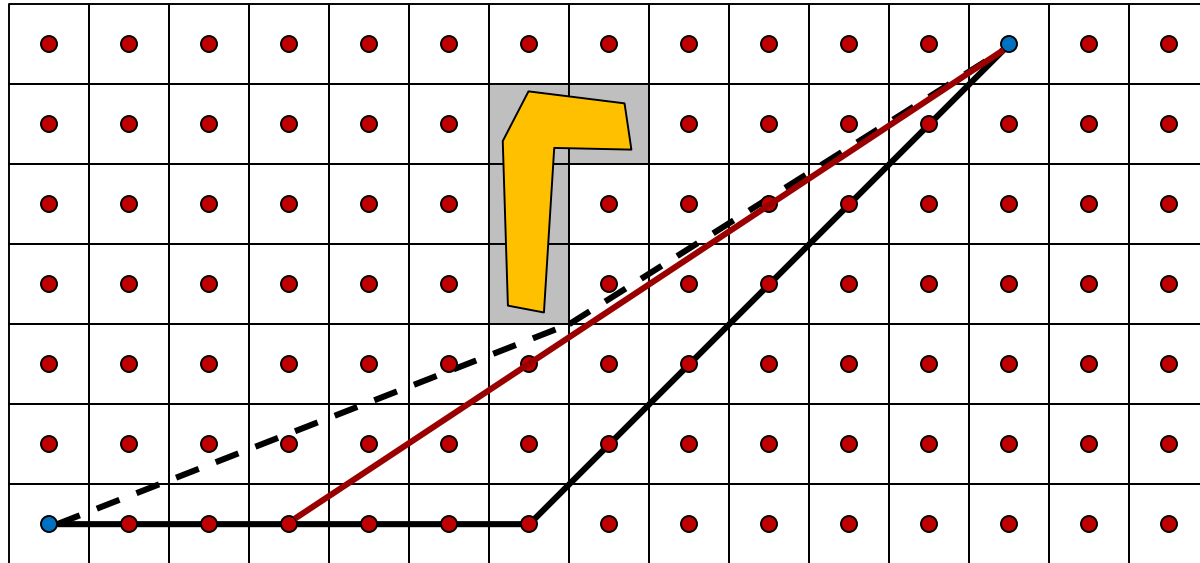
- - - - Shortest path

# SOLUTION 1: A\* SMOOTHING

- Allows connection to further states than neighbors on the grid
- Key observation:
  - If  $x_1, \dots, x_n$  is valid path
  - And  $x_k$  is visible from  $x_j$
  - Then  $x_1, \dots, x_j, x_k, \dots, x_n$  is a valid path



# SMOOTHING WORKS!

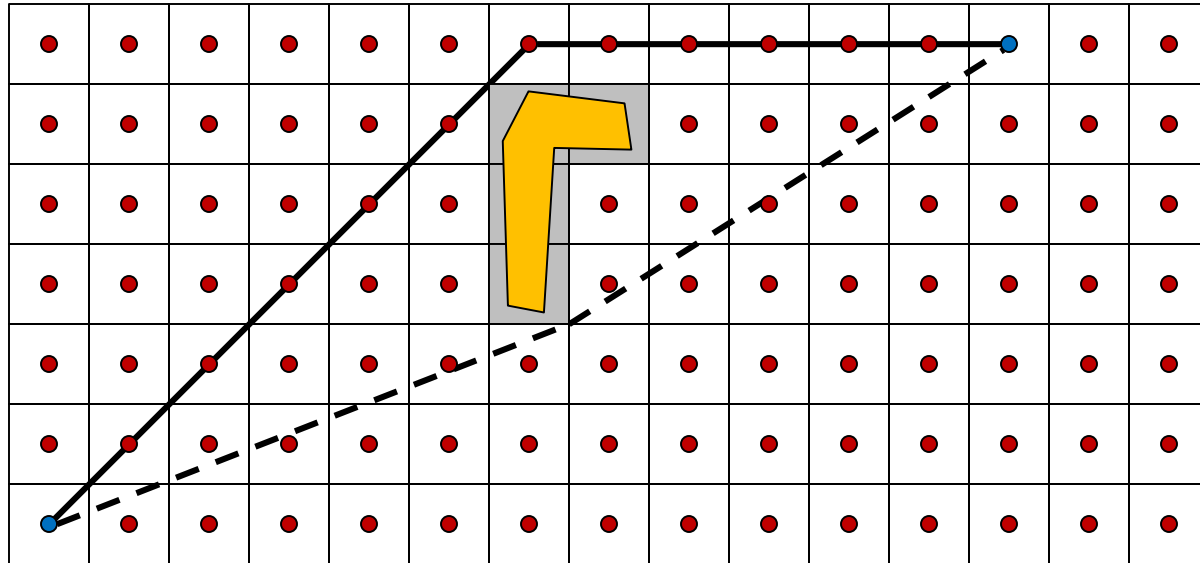


—— A shortest path through cell centers

- - - - Shortest path



# SMOOTHING DOESN'T WORK!



———— A shortest path through cell centers

----- Shortest path



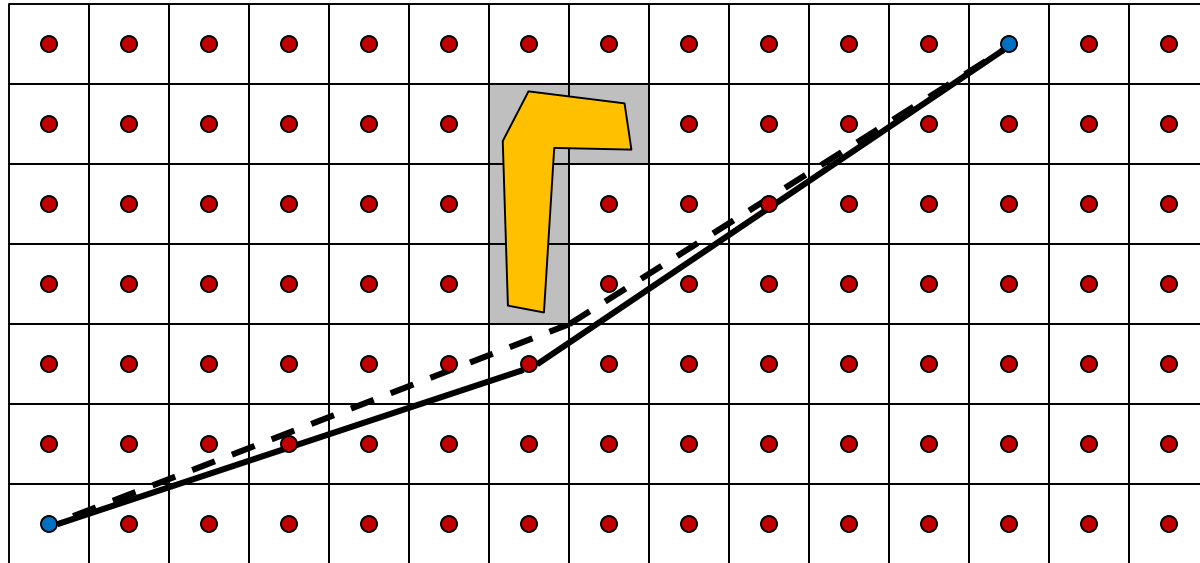


# SOLUTION 2: THETA\*

- Allow parents that are non-neighbors in the grid to be used during search
- Standard A\*
  - $g(y) = g(x) + c(x, y)$
  - Insert  $y$  with estimate
$$f(y) = g(x) + c(x, y) + h(y)$$
- Theta\*
  - If  $\text{parent}(x)$  is visible from  $y$ , insert  $y$  with estimate
$$f(y) = g(\text{parent}(x)) + c(\text{parent}(x), y) + h(y)$$



# THETA\* WORKS!

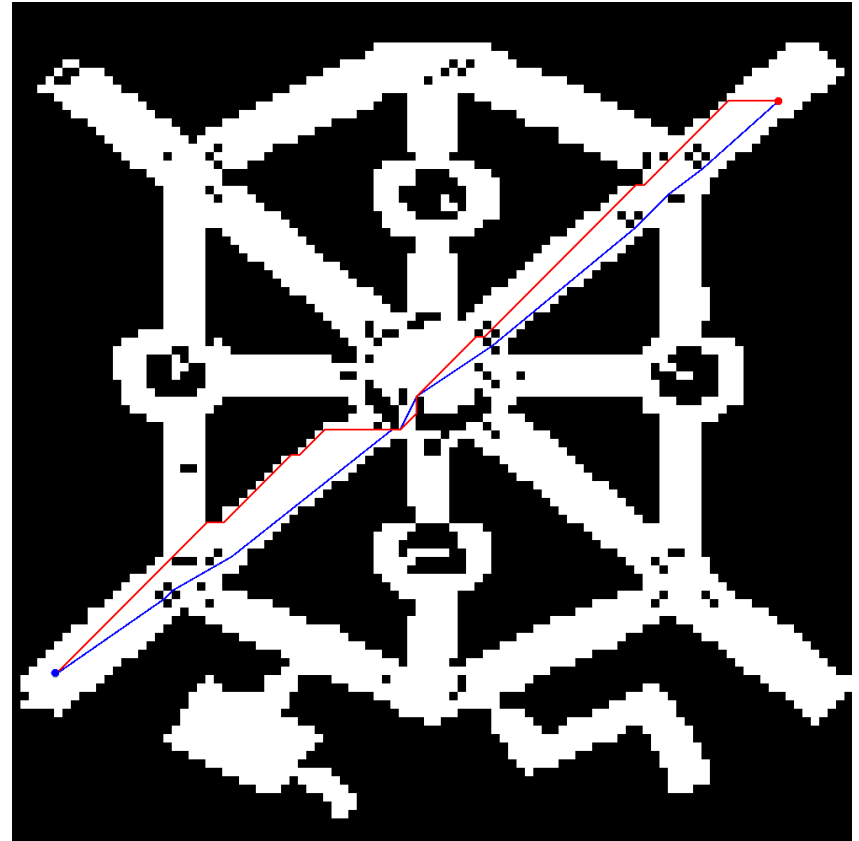
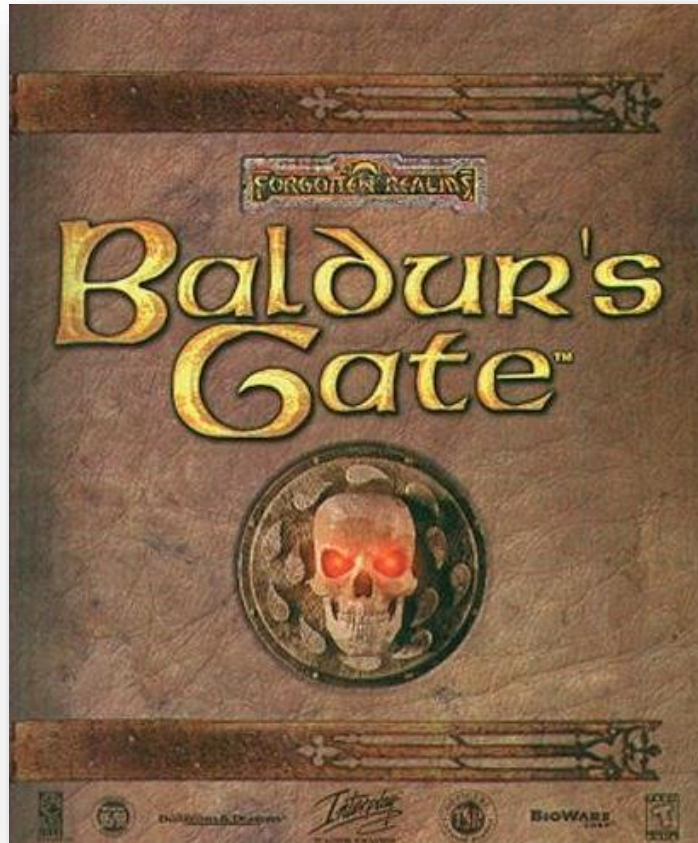


——— Theta\* path, I think 😊

- - - - Shortest path



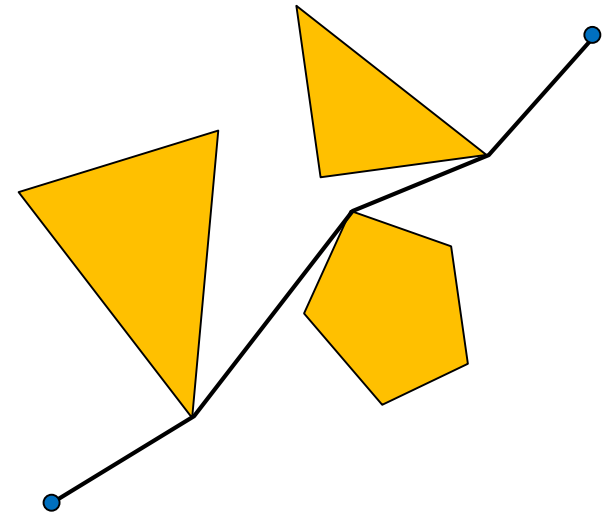
# THETA\* WORKS!



[Nash, AIGameDev 2010]

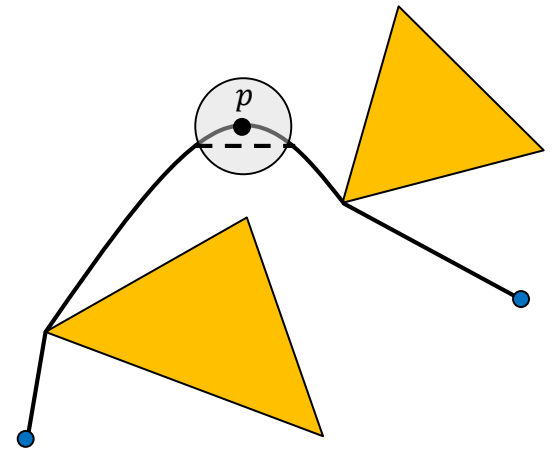
# THE OPTIMAL PATH

- **Polygonal path:** sequence of connected straight lines
- **Inner vertex of polygonal path:** vertex that is not beginning or end
- **Theorem:** assuming (closed) polygonal obstacles, shortest path is a polygonal path whose inner vertices are vertices of obstacles



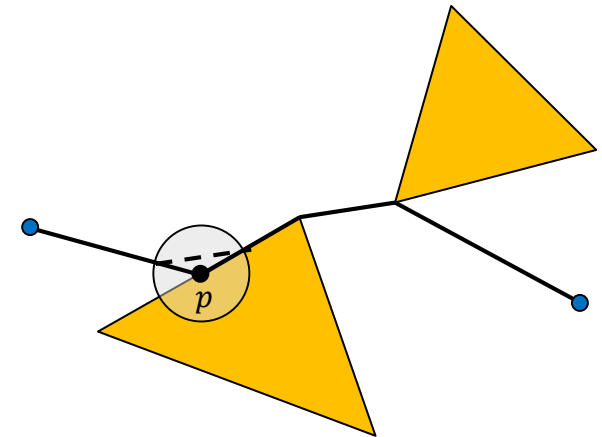
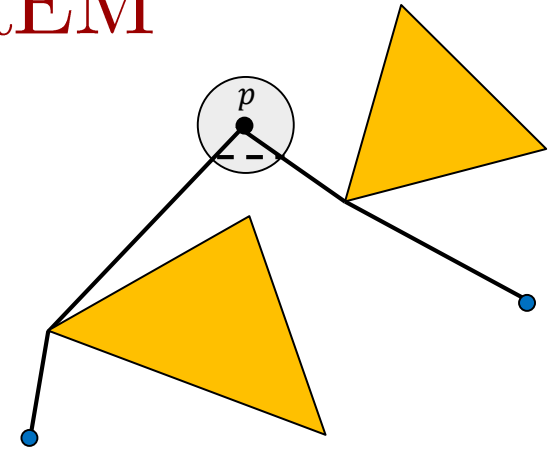
# PROOF OF THEOREM

- Suppose for contradiction that shortest path is not polygonal
- Obstacles are polygonal  $\Rightarrow$   
 $\exists$  point  $p$  in interior of free space such that “path through  $p$  is curved”
- $\exists$  disc of free space around  $p$
- Path through disc can be shortened by connecting points of entry and exit

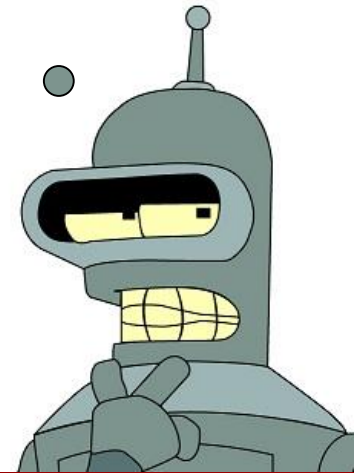


# PROOF OF THEOREM

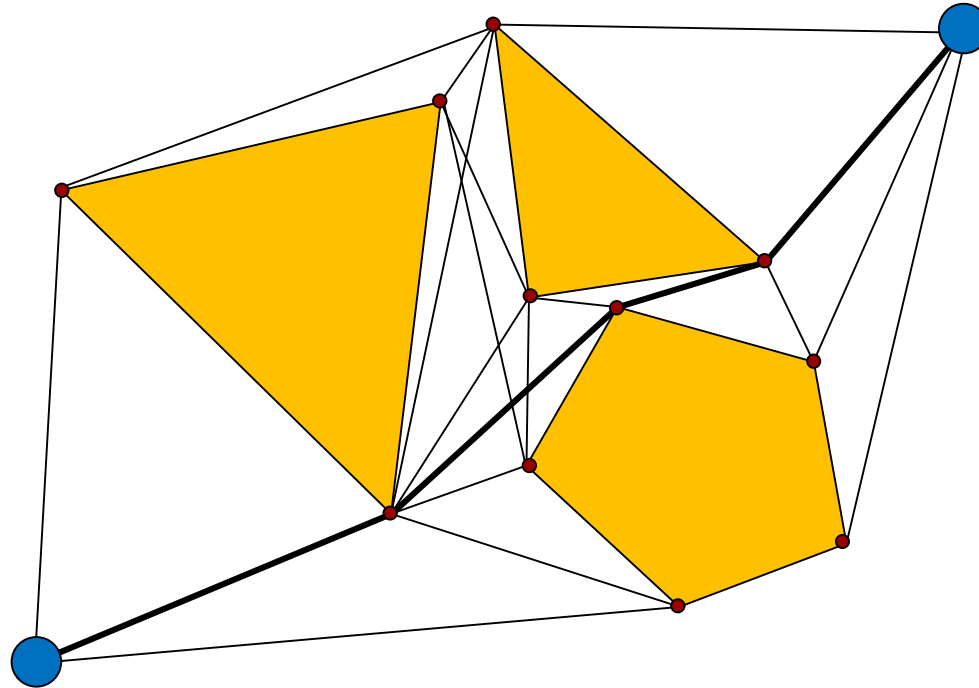
- Path is polygonal!
- Vertex cannot lie in interior of free space, otherwise we can do the same trick
- Vertex cannot lie on the interior of an edge, otherwise we can do the same trick ■



How would we define  
a graph on which  $A^*$   
would be optimal?



# VISIBILITY GRAPH



**Vertices** = vertices of polygons and  $s, t$   
**Edges** = all  $(x, y)$  such that  $y$  is visible from  $x$



# VISIBILITY GRAPH

- **Poll 2:** Let  $n$  be the total number of vertices of all polygons. How many edges will the optimal path in the visibility graph traverse in the worst case?
  1.  $\Theta(\sqrt{n})$
  2.  $\Theta(n)$
  3.  $\Theta(n^2)$
  4.  $\Theta(n^3)$



# SUMMARY

- Terminology and algorithms:
  - Cell decomposition
  - Resolution completeness
  - Theta\*
- Theorems:
  - A\* is optimally efficient
  - Geometry of shortest path with polygonal obstacles
- Big ideas:
  - A\* can be modified to work well in continuous spaces

