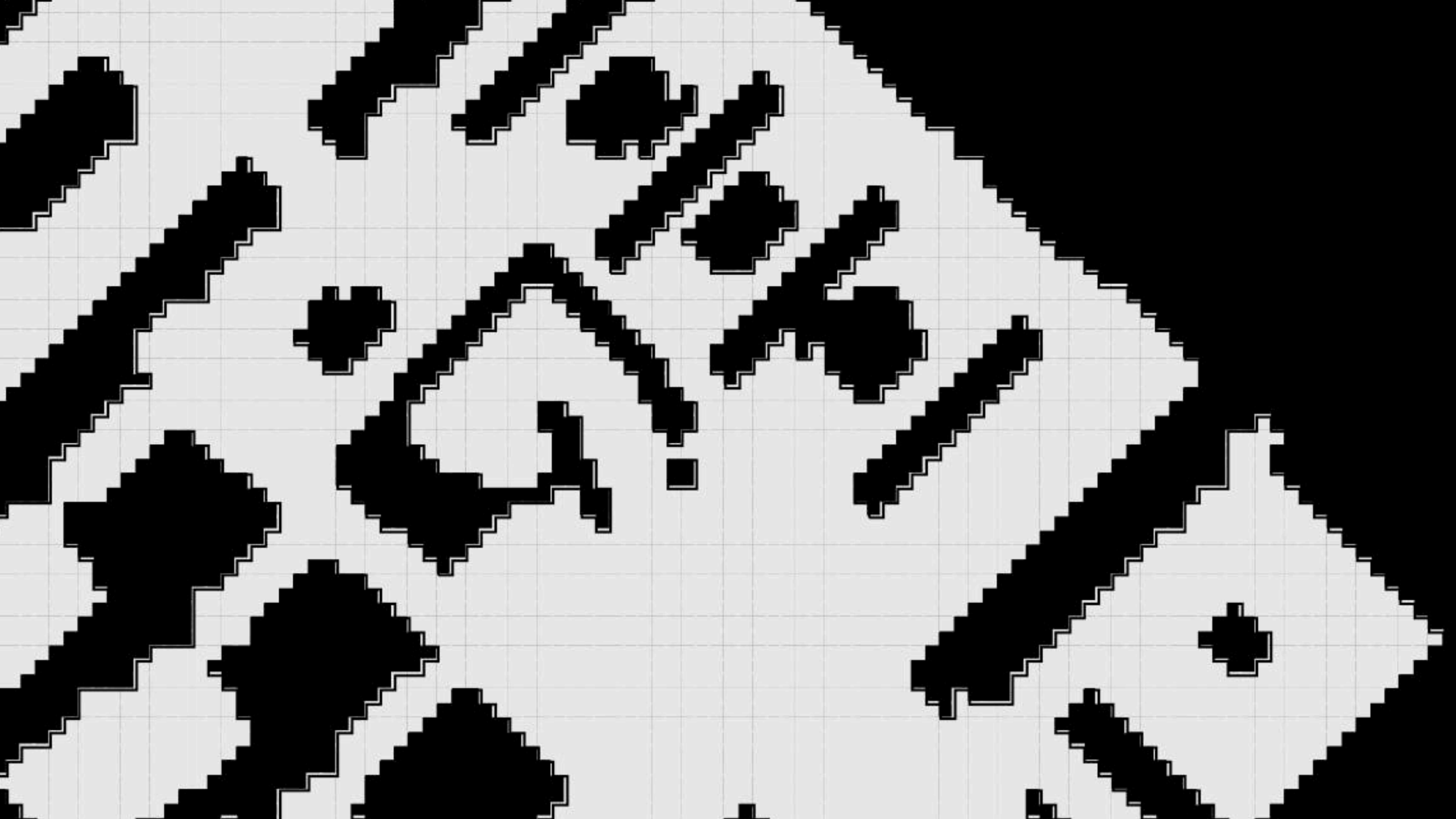
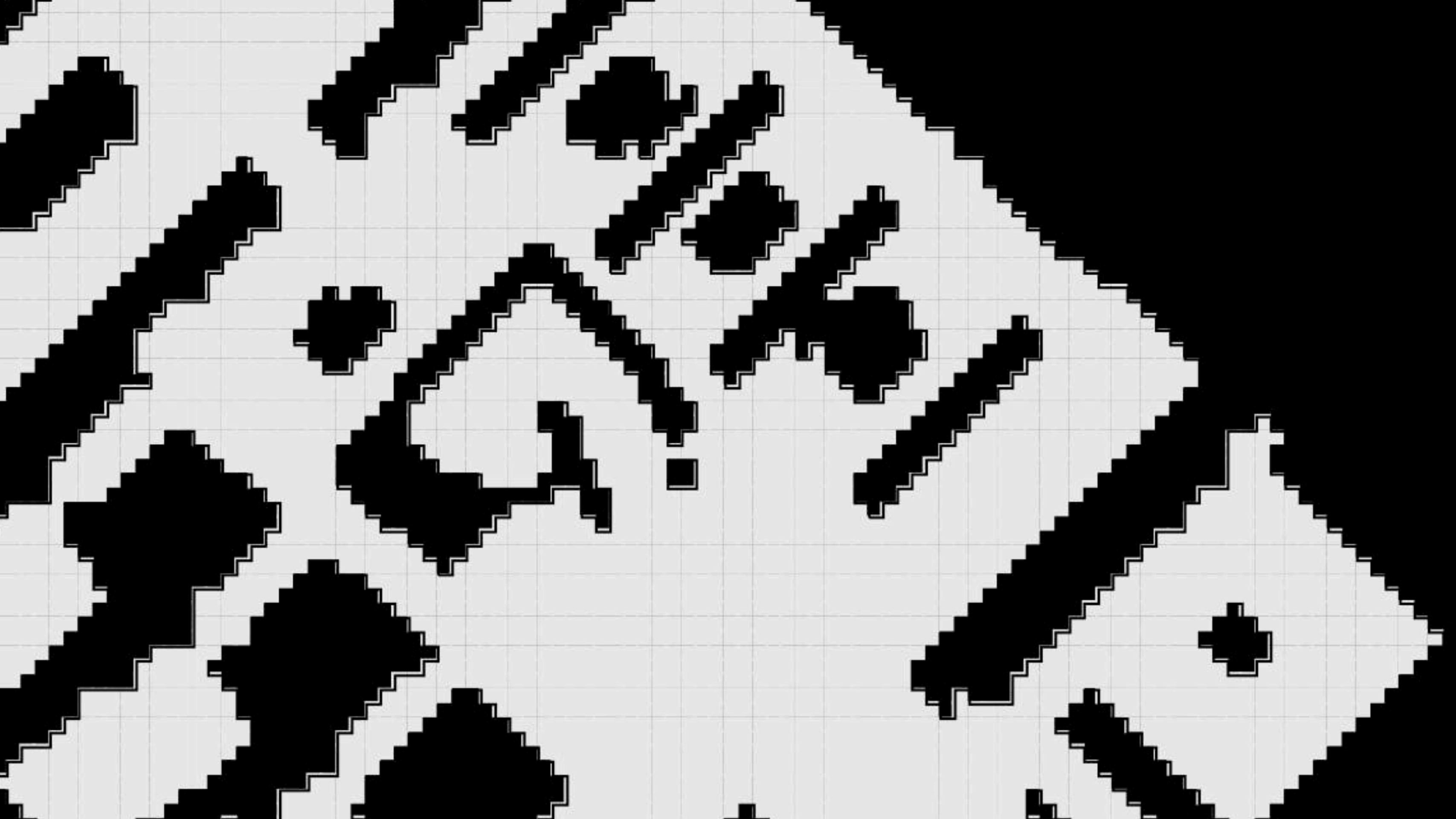
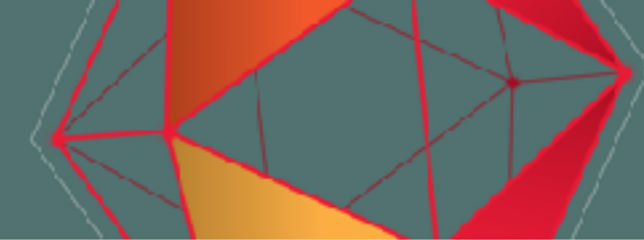


Bidirectional Search: Is It For Me?

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Sneha Sawlani



Ariel Felner



Sandra Zilles



Eshed Shaham



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Designed and
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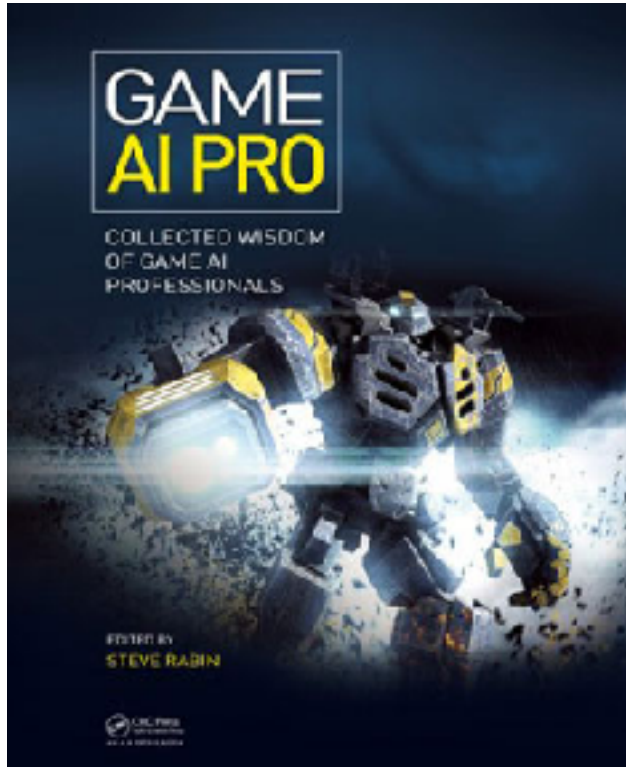




Lecture Takeaways

- When should I use bidirectional search?
- What algorithm should I use for bidirectional search?



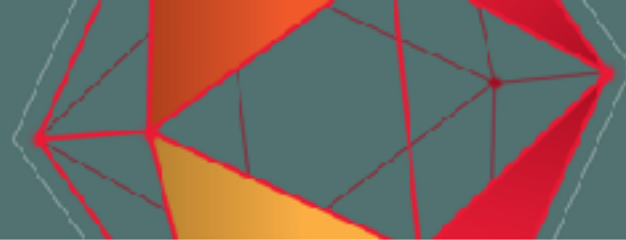


Pathfinding Architecture Optimizations

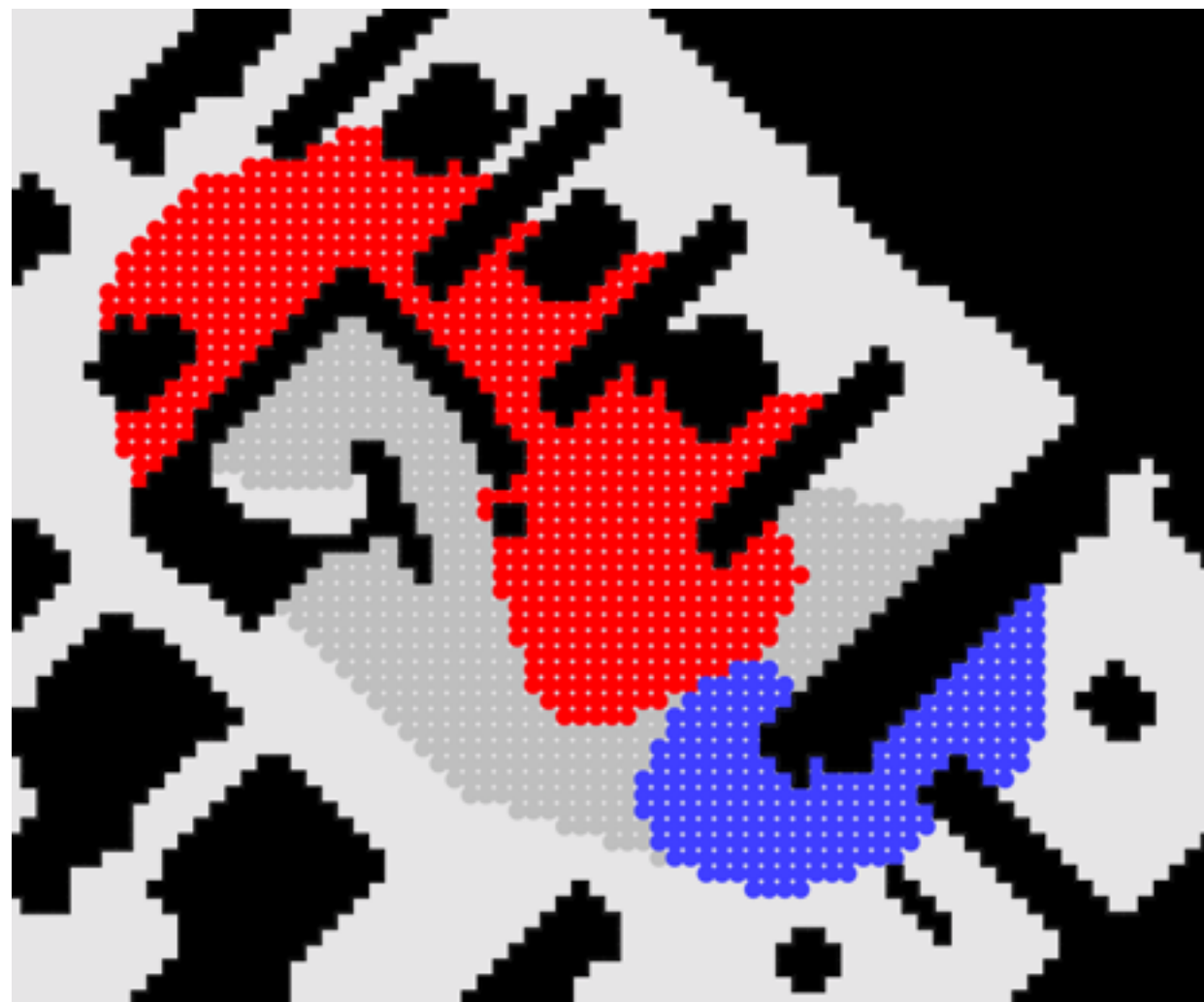
by Steve Rabin & Nathan Sturtevant

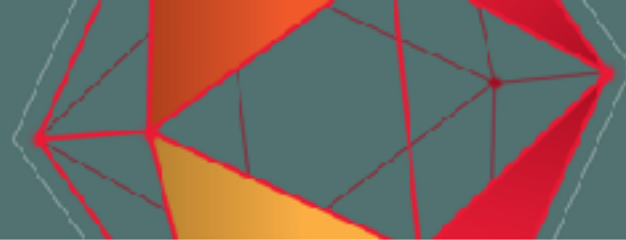
Bad Idea #2: Bidirectional Pathfinding



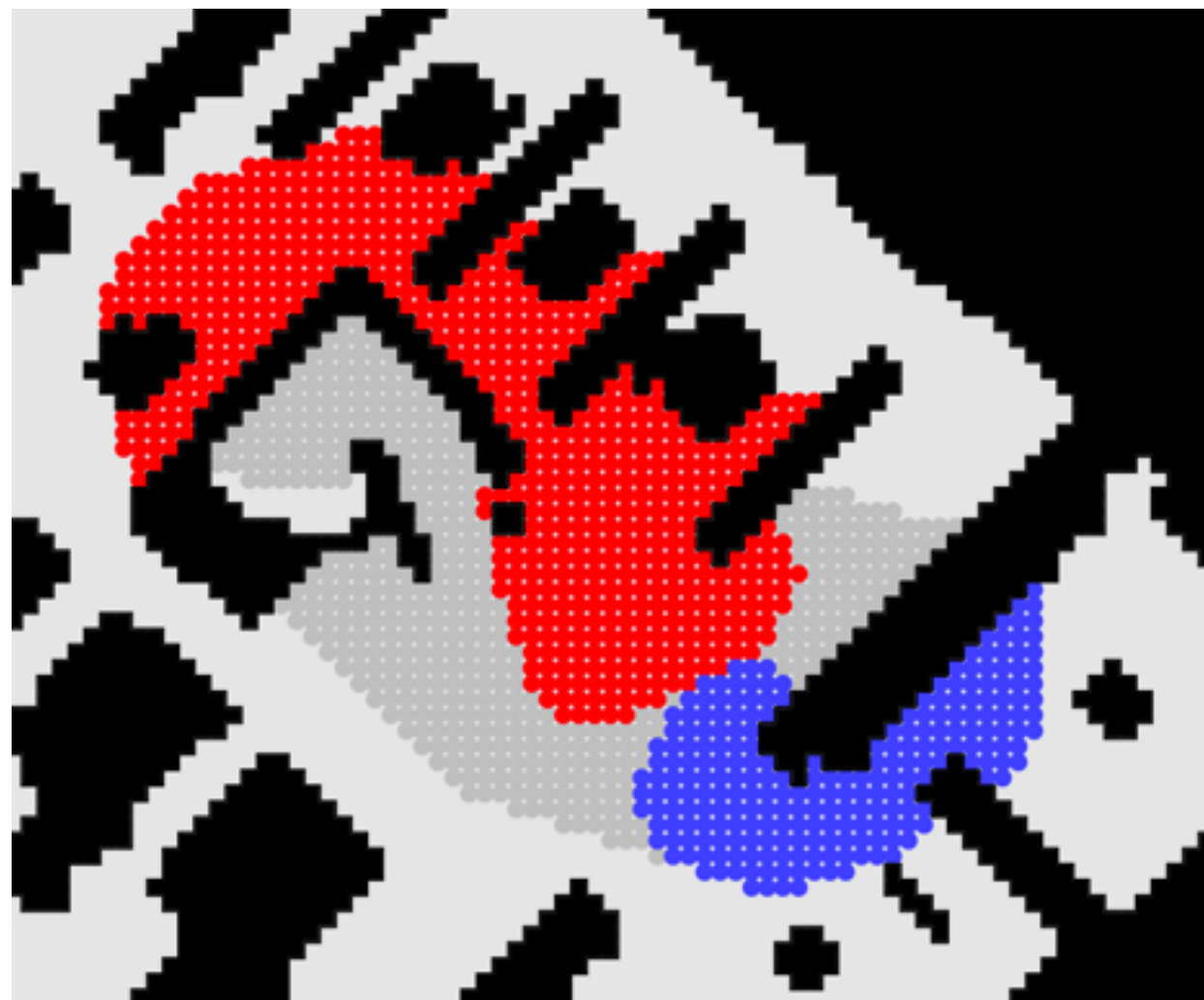


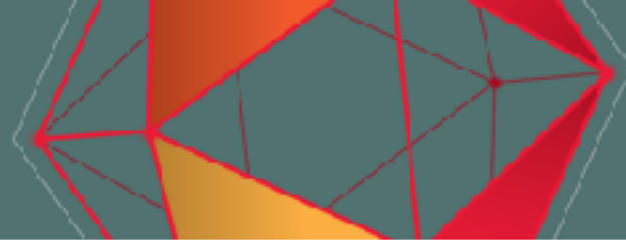
Optimal Bidirectional Search





Optimal Bidirectional Search

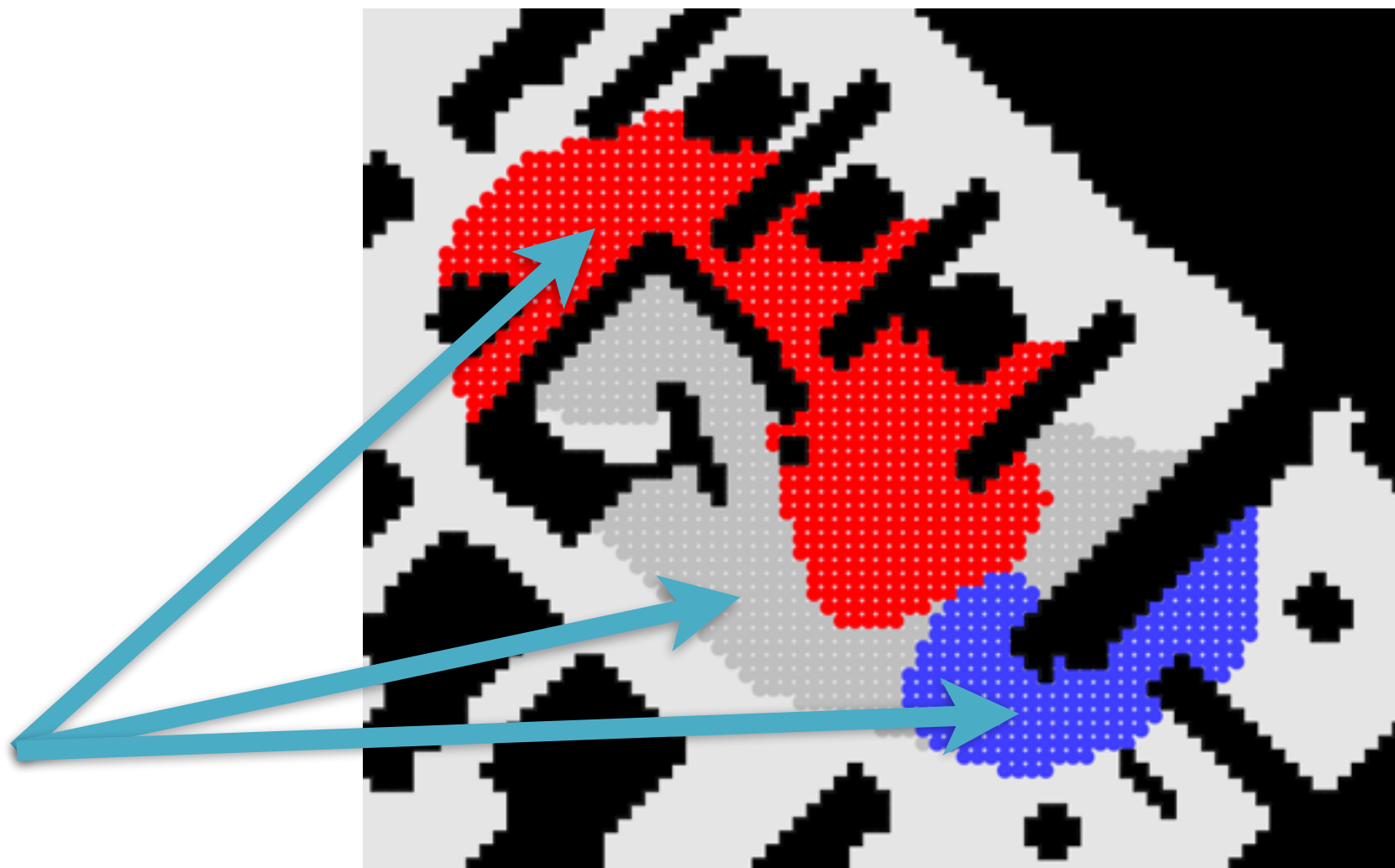


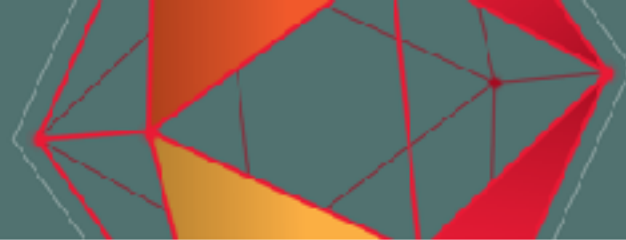


Optimal Bidirectional Search



All states that
could be expanded

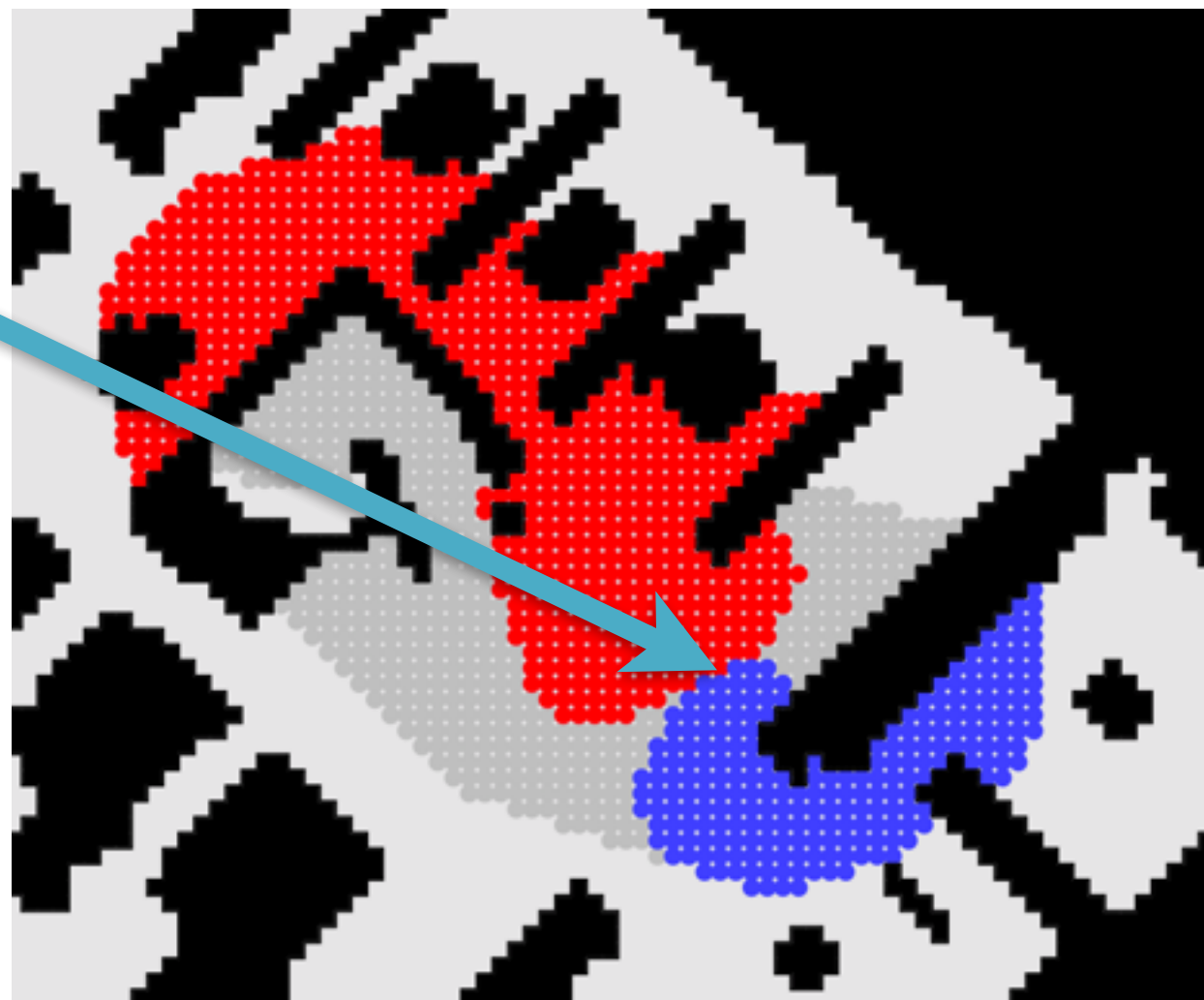


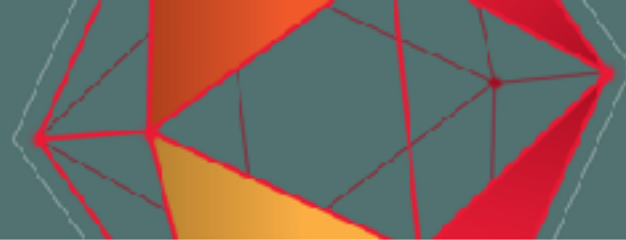


Optimal Bidirectional Search



Choose a meeting
point

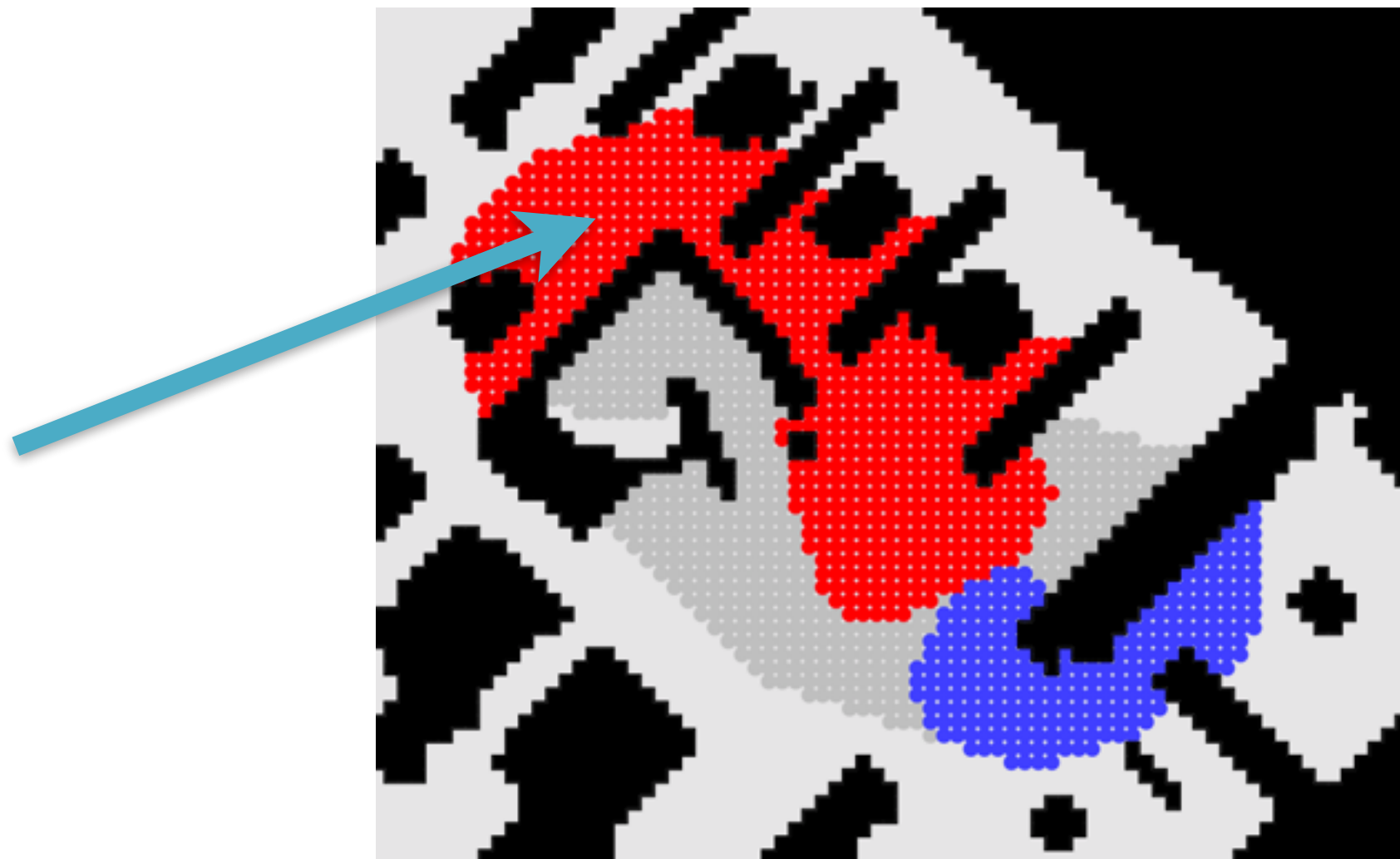


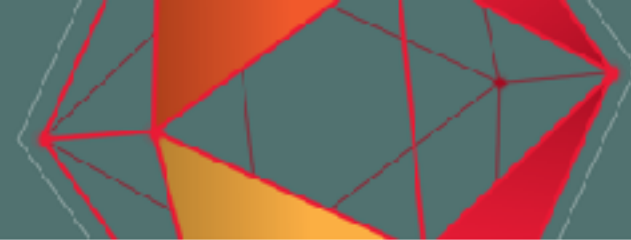


Optimal Bidirectional Search



Expand up to that
point forward



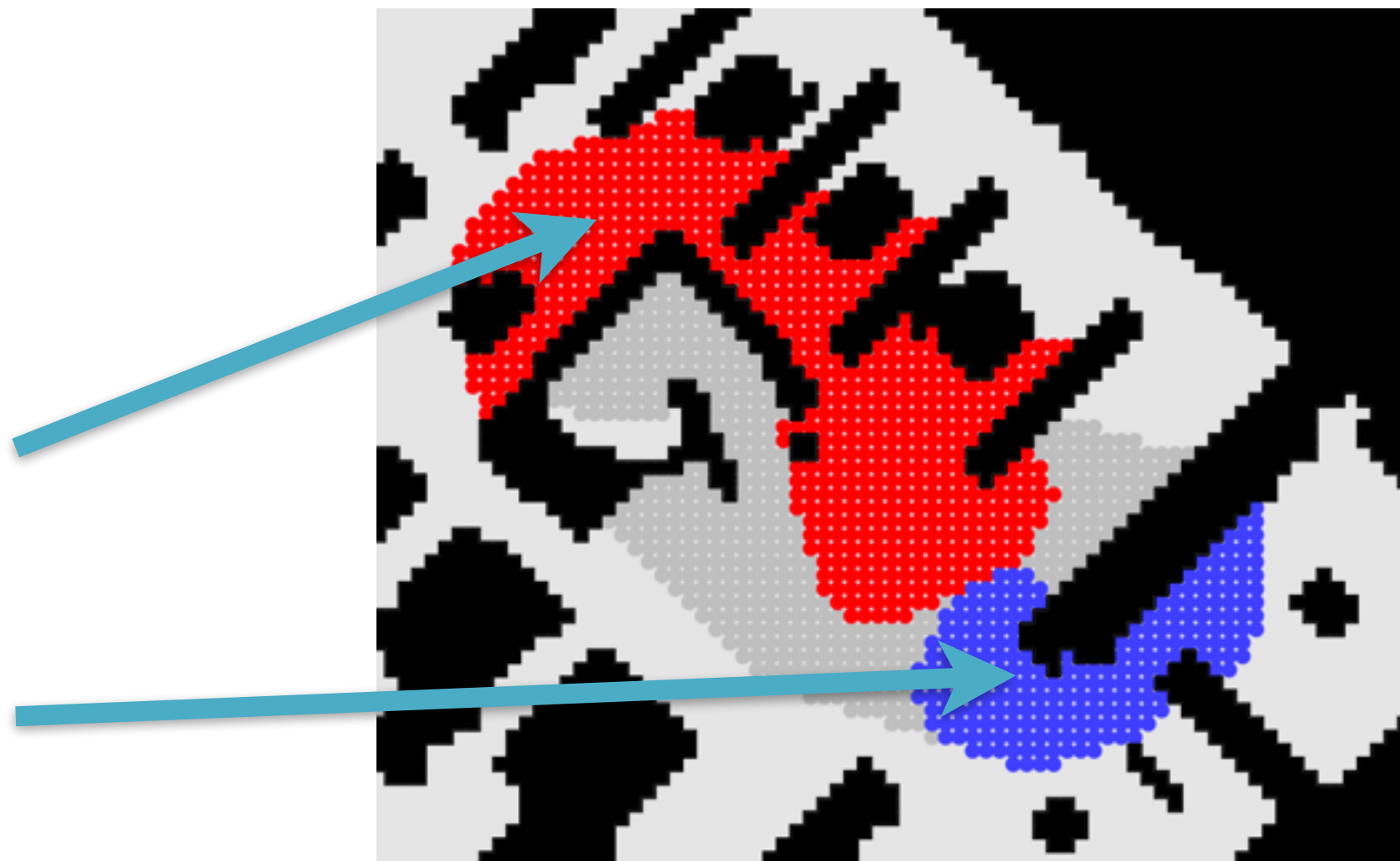


Optimal Bidirectional Search



Expand up to that
point forward

Expand up to that
point backward



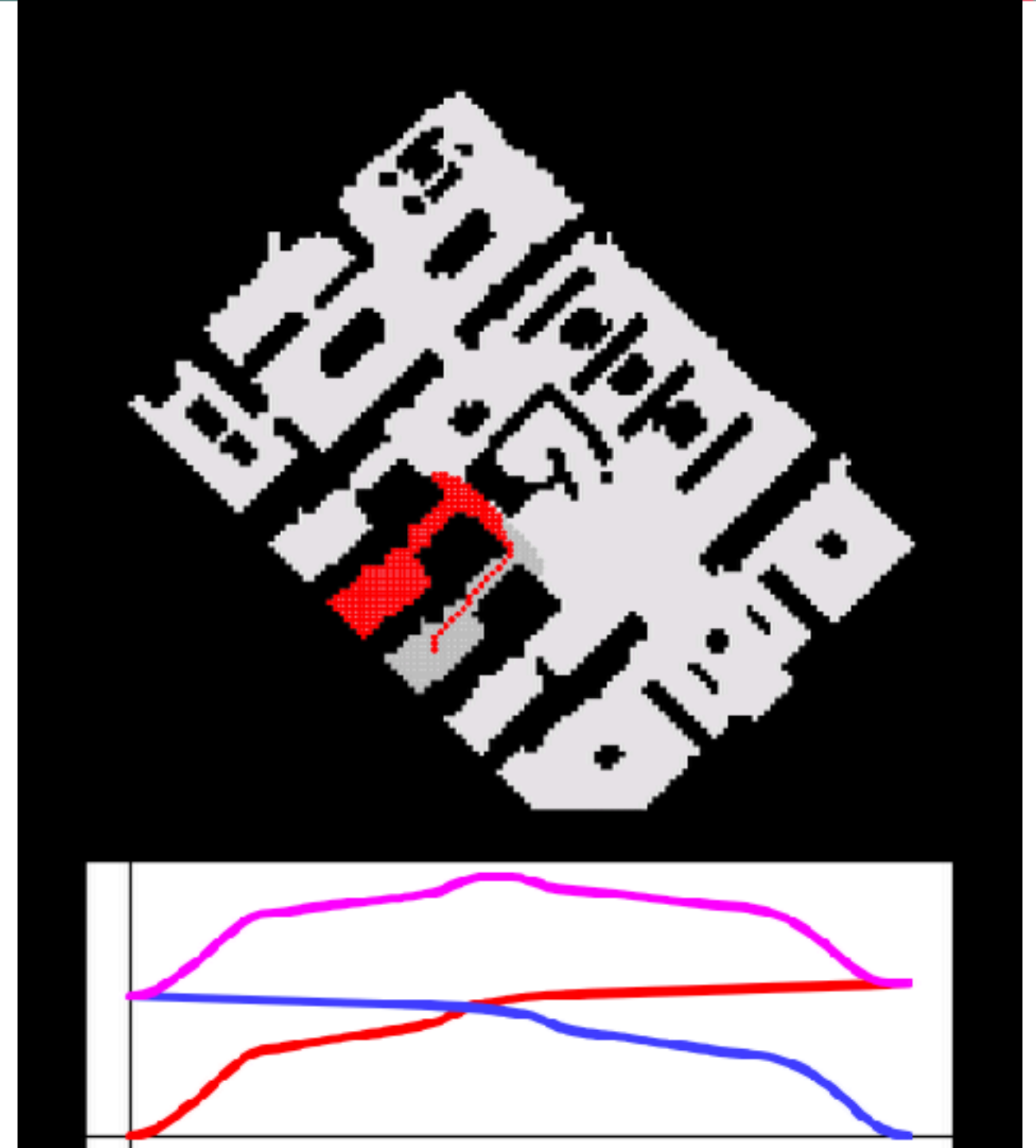


Demo



Explanation

- Perfect heuristic near goal
 - Open space
- Symmetric





New Algorithm: NBS



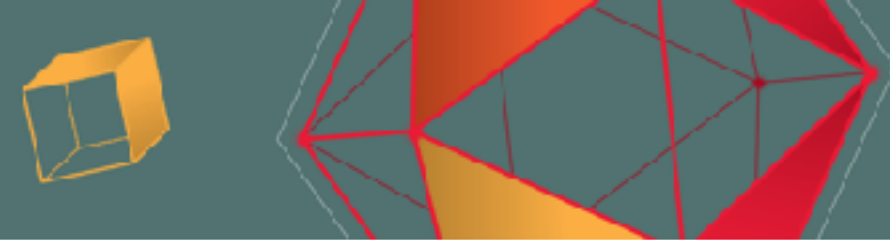
- NBS never expands more than 2x the states expanded by the **best possible** algorithm
 - *In our theoretical framework*
- NBS does equal work in each direction





When should we use NBS?





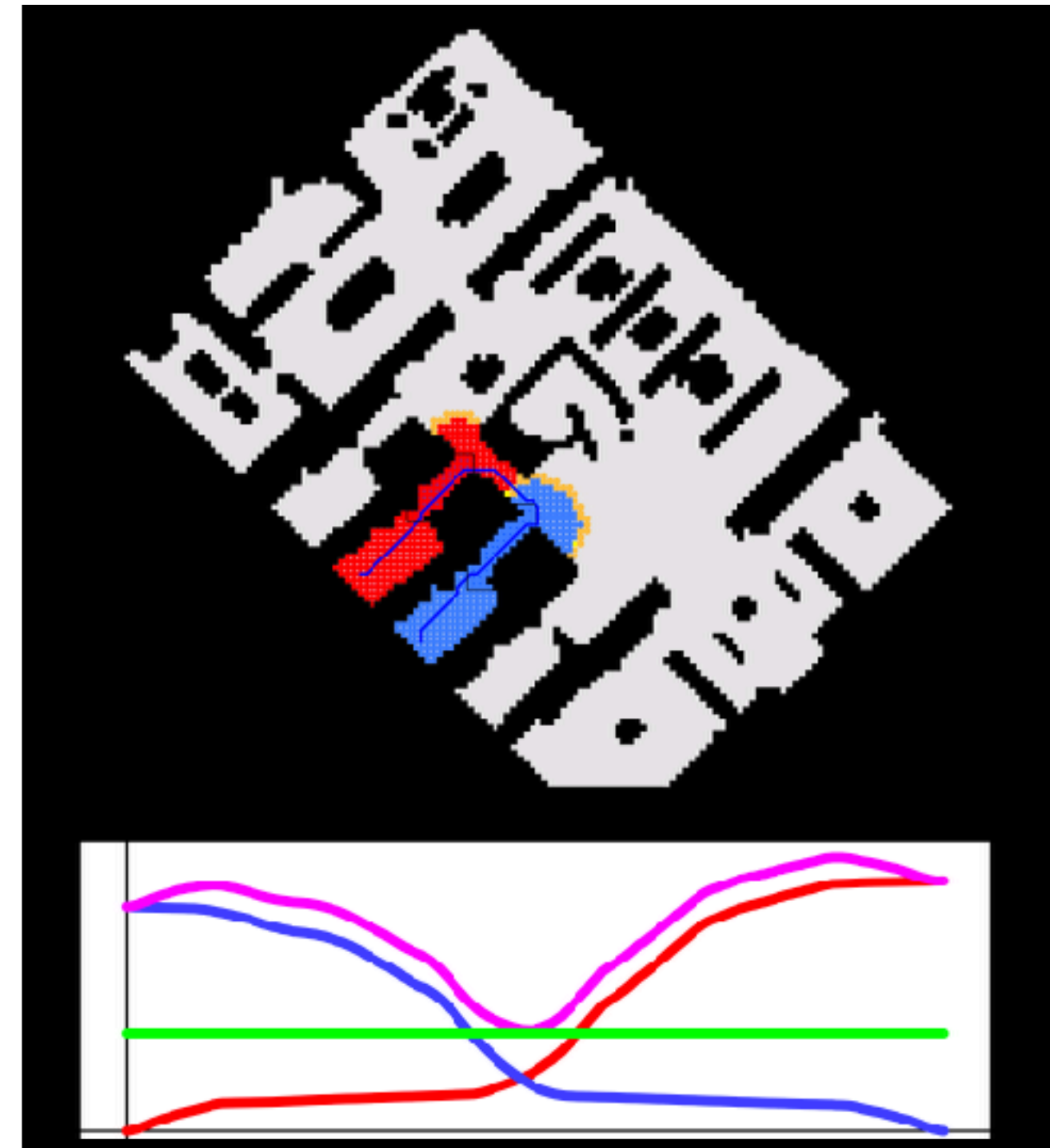
Scenario 1: Weighted terrain

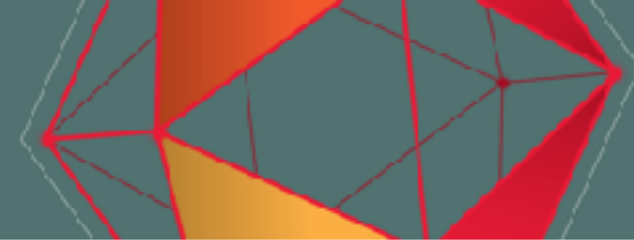




Weighted terrain

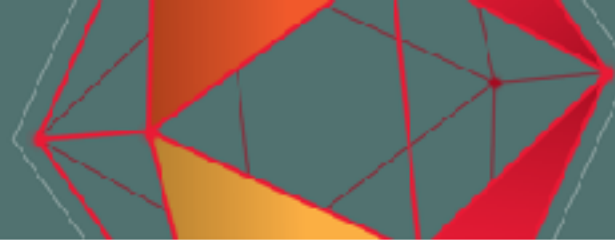
- Costly to look for alternate paths around weighted terrain





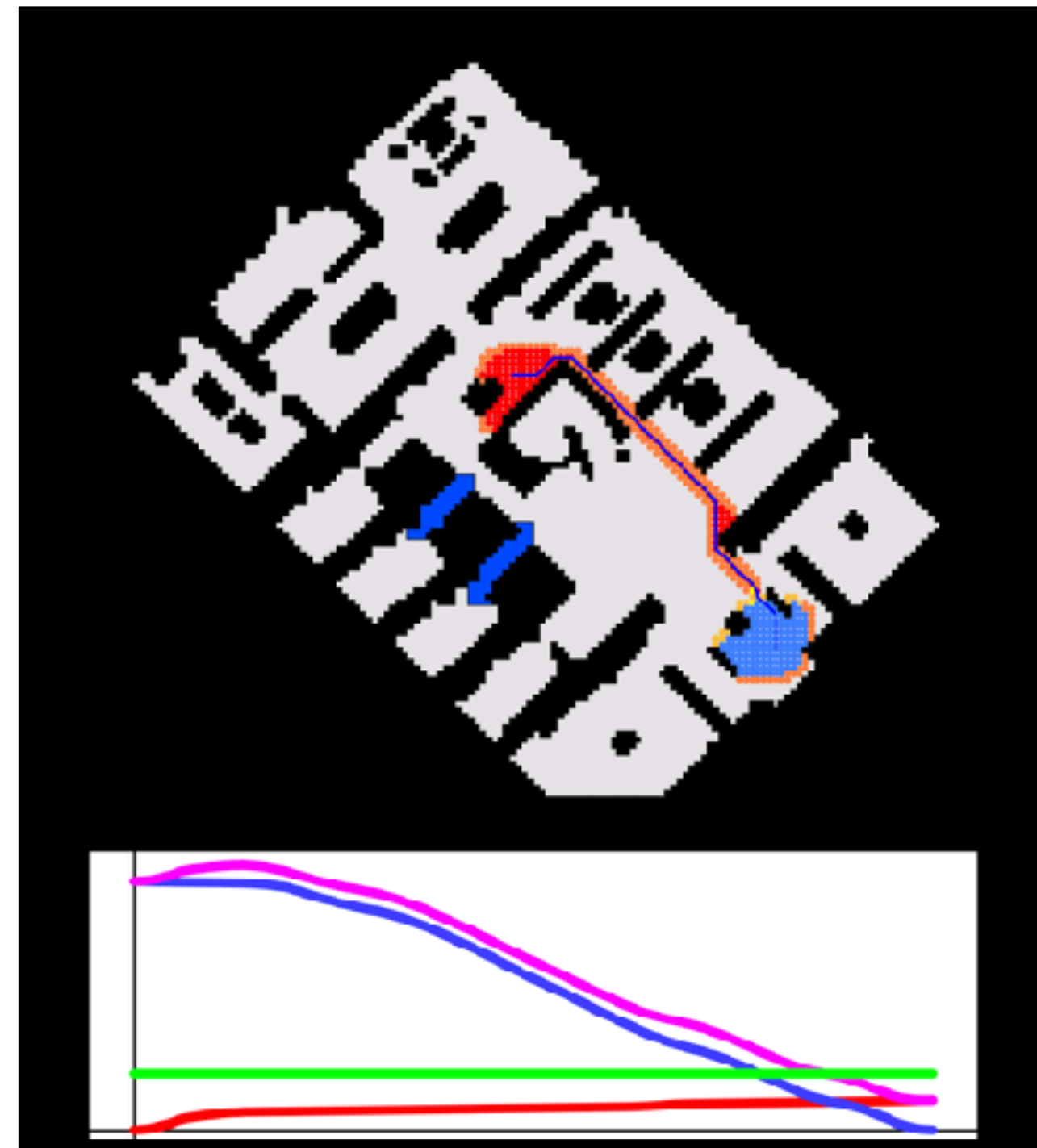
Scenario 2: Problem Asymmetry

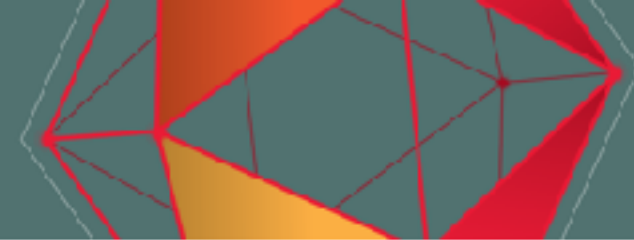




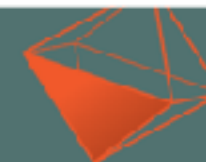
Problem Asymmetry

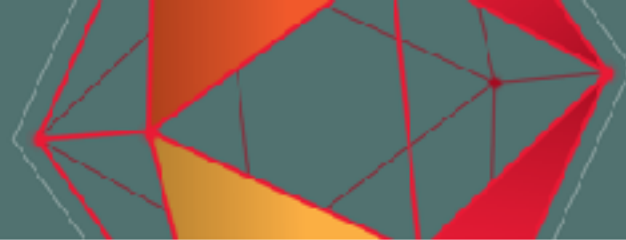
- When forward is much more expensive than backwards
 - 3x worse on average
- Also happens with weighted terrain





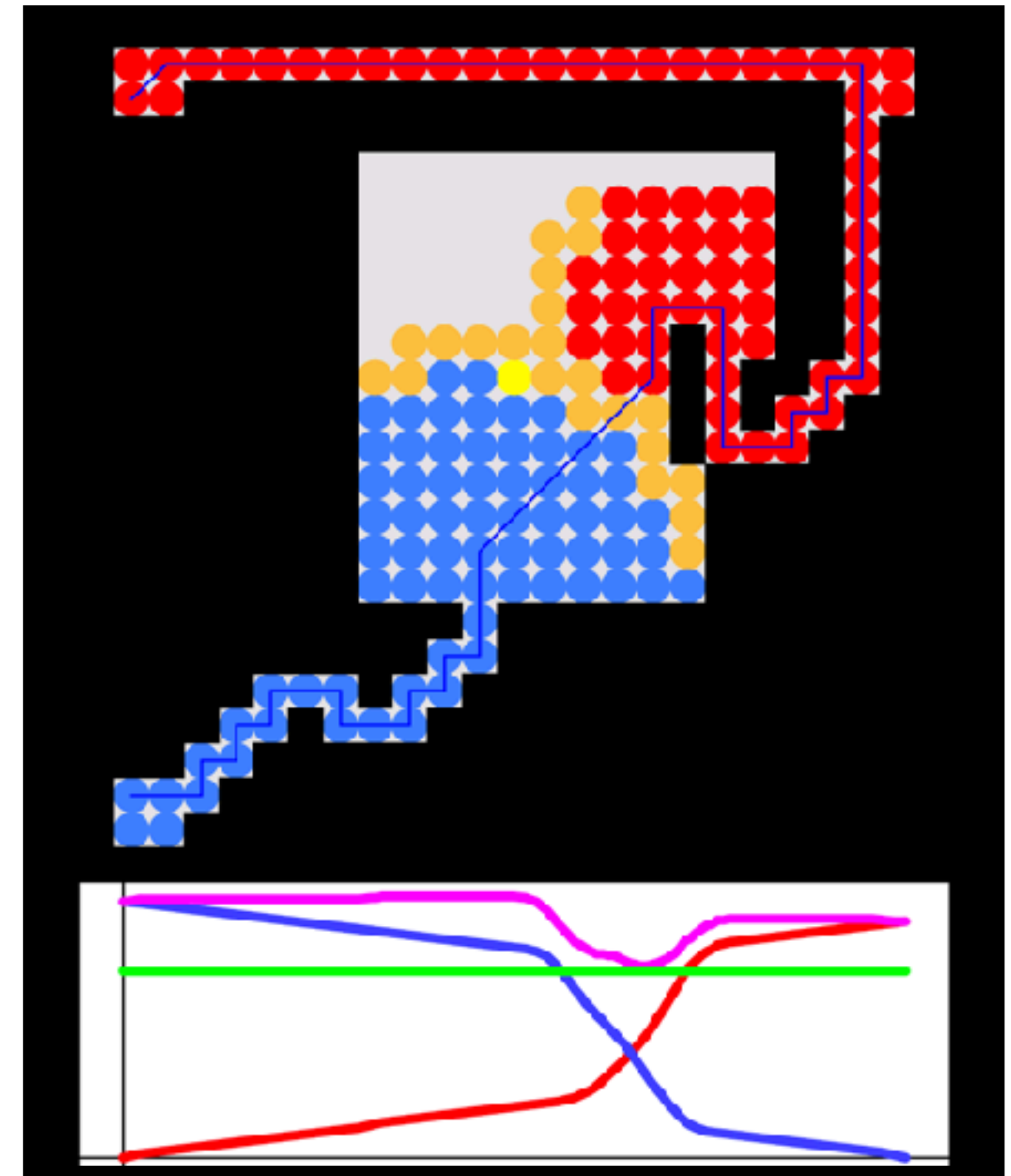
Scenario 3: Map Asymmetry

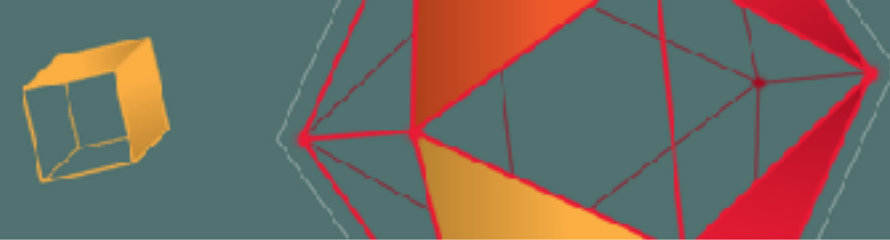




Map Asymmetry

- Common in city maps
 - Dense regions of pathfinding nodes
- Bidirectional search will avoid the densest region





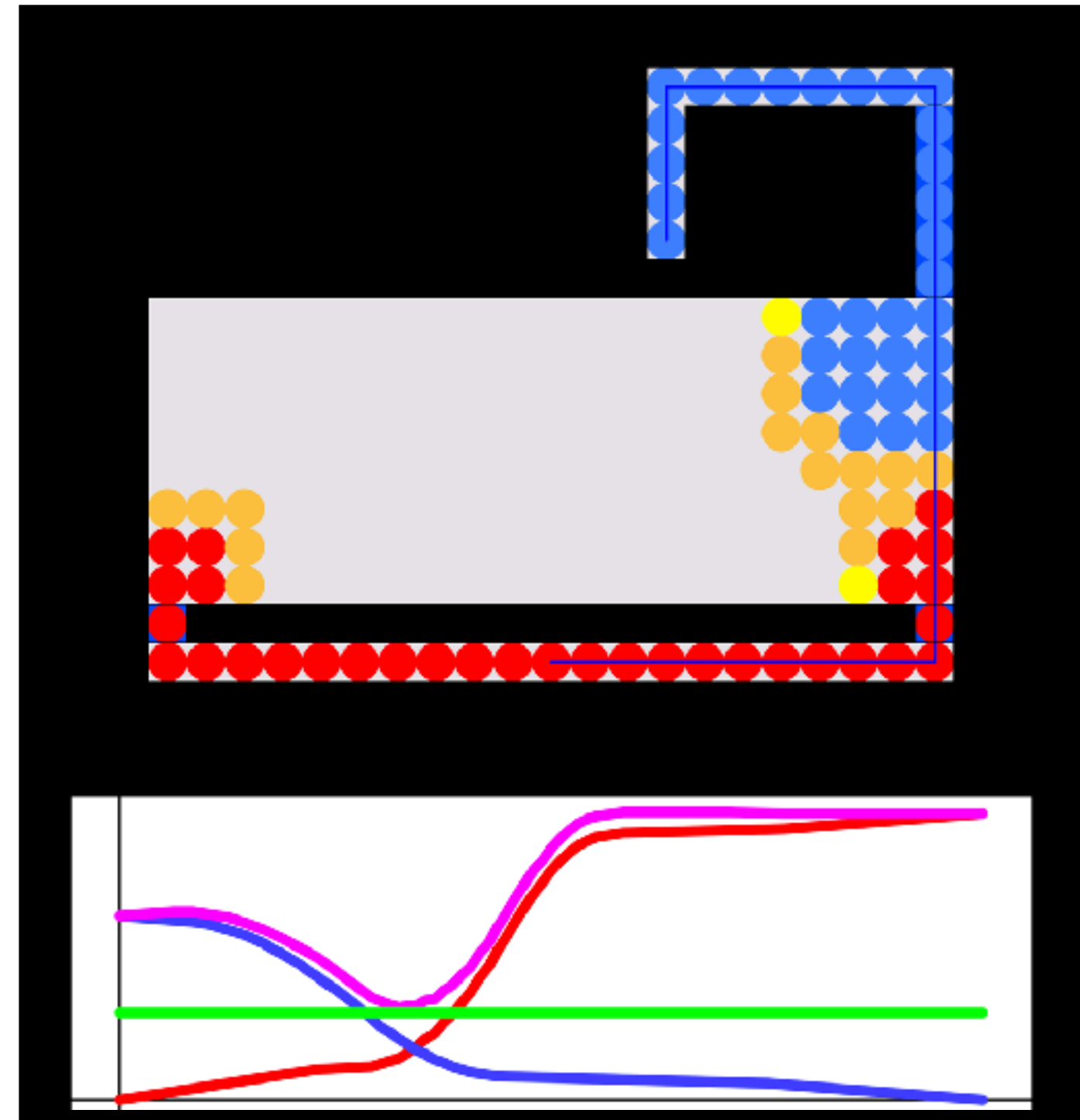
Scenario 4: Local Minima

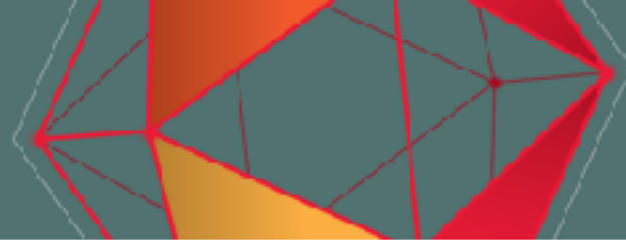




Local Minima

- Many states look close, but aren't
 - Could be fixed by a better heuristic





Testing in practice

- Web tool available for analysis
- <http://www.movingai.com/GDC18/test.html>

search in the text boxes below. The resulting plot will help you understand whether bidirectional search will work well in your

The purple line is the total work required to solve the input problem given different meeting points. The far ends of the plot represent forward A* and backward A* respectively. Any points in between are achieved by bidirectional search.

Guidelines:

- [Case 1] If the purple line is U shaped, then bidirectional search will work very well in your domain. (NBS is currently a bidirectional search algorithm.)
- [Case 2] If the purple line is an upside-down U or mostly flat, then bidirectional search will not work well.
- [Case 3] If the purple line is sloped significantly to the left or right, then your problem has significant forward/backward bias. Bidirectional search (NBS) will work well because it will find the right direction to search.

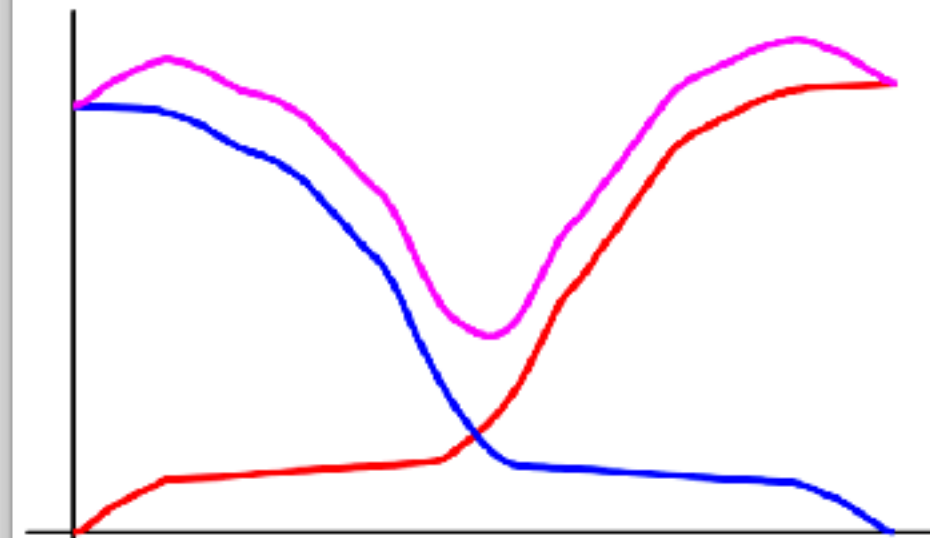
0.0 1.0 1.0 1.5 1.5 1.0 1.5 1.5	0.0 1.0 1.0 1.5 1.5 1.0 1.5 1.5
1.0 2.5 2.0 2.0 3.0 2.5 4.0	1.0 2.5 2.0 2.5 2.0 3.0 3.5
3.5 3.5 4.0 3.0 3.0 2.5 2.5	2.5 2.5 3.0 3.5 2.5 2.0 3.0
4.0 4.5 4.5 3.5 4.5 5.5 5.0	2.0 2.5 2.5 2.5 4.0 4.5 5.0

Plot Data

Load Sample Case 1 Data

Load Sample Case 2 Data

Load Sample Case 3 Data



Display a menu





NBS Details









A*





A*

- Put start onto priority queue





A*

- Put start onto priority queue
- While queue not empty / solution not found





A*

- Put start onto priority queue
- While queue not empty / solution not found
 - Among all states on queue:





A*

- Put start onto priority queue
- While queue not empty / solution not found
 - Among all states on queue:
 - Select the state with lowest **f-cost**






A*

- Put start onto priority queue
- While queue not empty / solution not found
 - Among all states on queue:
 - Select the state with lowest **f-cost**
 - Expand it





A^* : f-cost

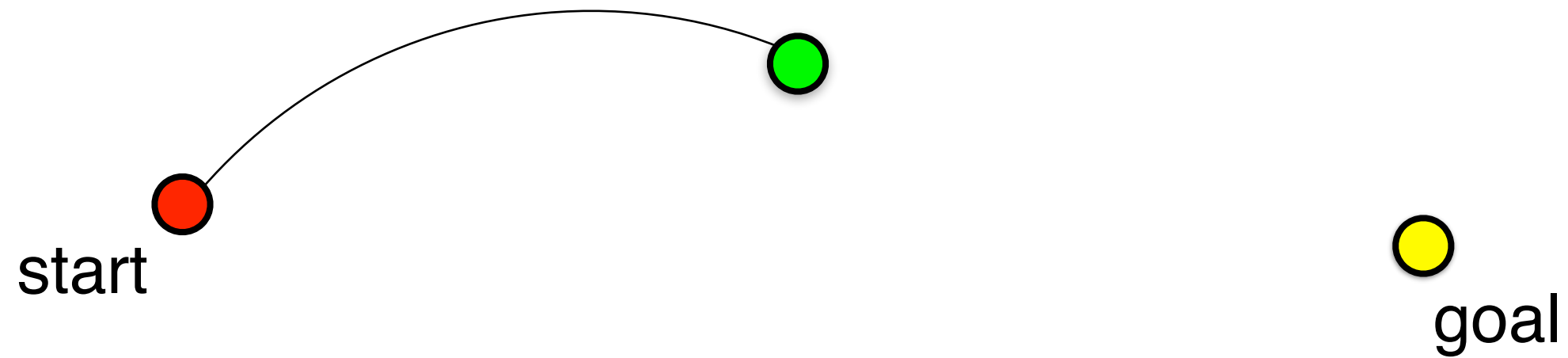
start 

 goal



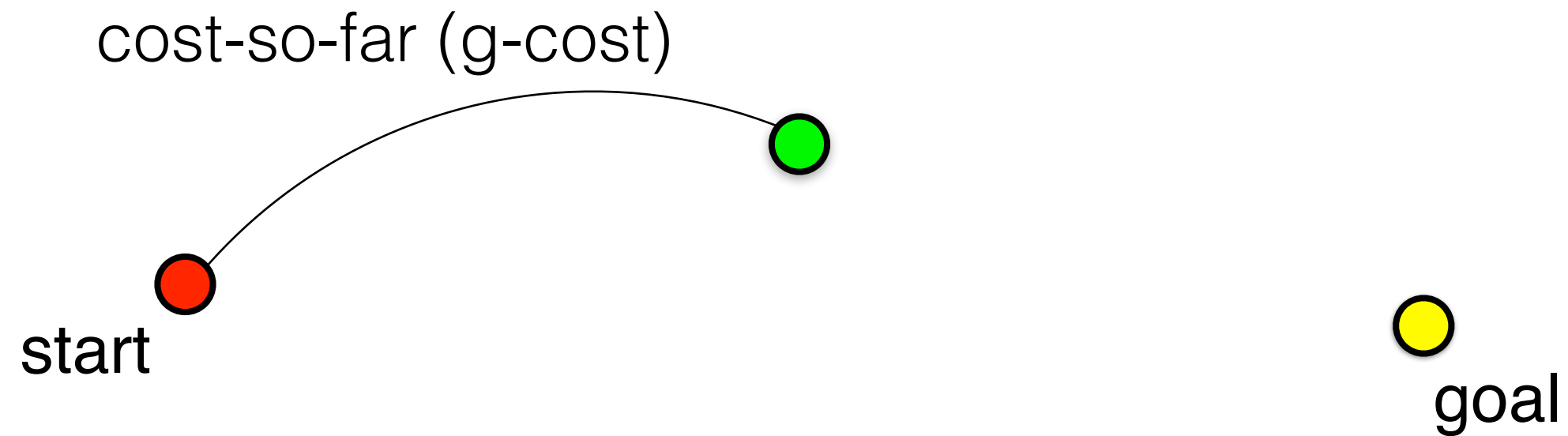


A^* : f-cost



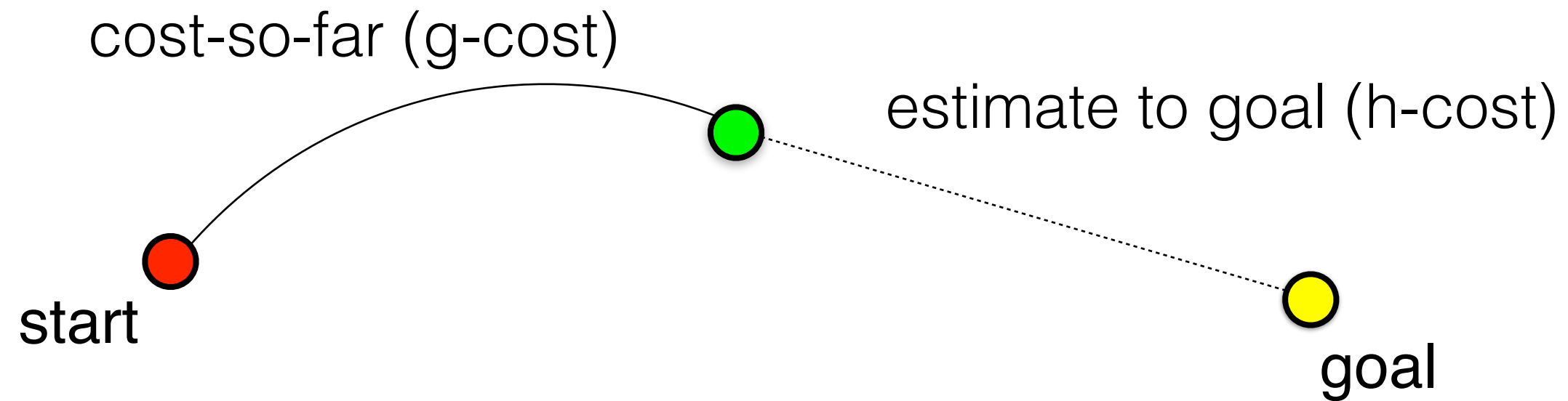


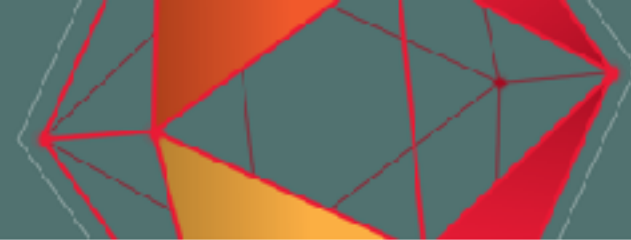
A*: f-cost



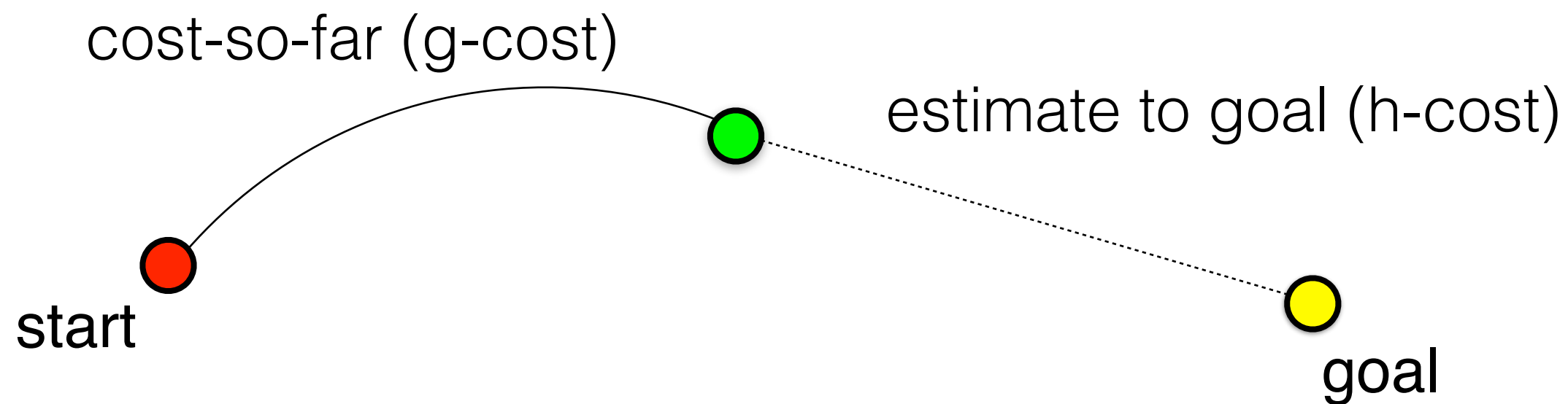


A^* : f-cost





A*: f-cost



$$\text{f-cost} = \text{g-cost} + \text{h-cost} = \text{estimated path length}$$

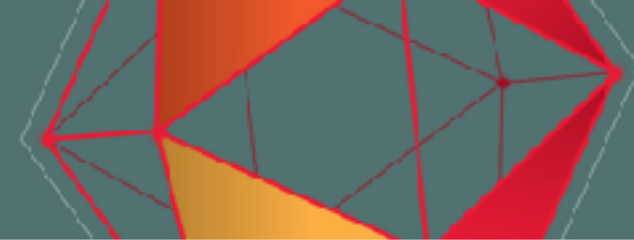




A*

- Put start onto priority queue
- While queue not empty / solution not found
 - Among all states on queue:
 - Select the state with lowest **f-cost**
 - Expand it





$A^* \rightarrow \text{NBS}$



- Put start onto priority queue
- While queue not empty / solution not found
 - Among all states on queue:
 - Select the state with lowest **f-cost**
 - Expand it

Front-to-End Bidirectional Heuristic Search with Near-Optimal Node Expansions,
Jingwei Chen, Robert C. Holte, Sandra Zilles and Nathan R. Sturtevant,
International Joint Conference on Artificial Intelligence (IJCAI), **2017**





$A^* \rightarrow \text{NBS}$



- Put start/**goal** onto forward/**backward** priority queues
- While queue not empty / solution not found
 - Among all states on queue:
 - Select the state with lowest **f-cost**
 - Expand it

Front-to-End Bidirectional Heuristic Search with Near-Optimal Node Expansions,
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$A^* \rightarrow \text{NBS}$



- Put start/**goal** onto forward/**backward** priority queues
- While **queues** not empty / solution not found
 - Among all states on queue:
 - Select the state with lowest **f-cost**
 - Expand it

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$A^* \rightarrow \text{NBS}$



- Put start/**goal** onto forward/**backward** priority queues
- While **queues** not empty / solution not found
 - Among all states on **queues**:
 - Select the state with lowest **f-cost**
 - Expand it

Front-to-End Bidirectional Heuristic Search with Near-Optimal Node Expansions,
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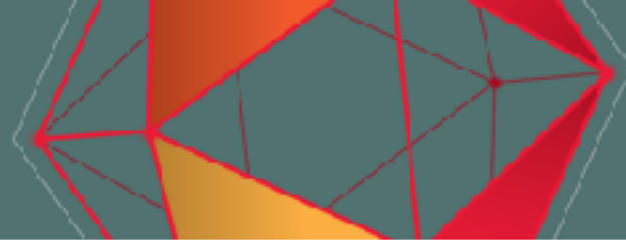
$A^* \rightarrow \text{NBS}$



- Put start/**goal** onto forward/**backward** priority queues
- While **queues** not empty / solution not found
 - Among all states on **queues**:
 - Select the **pair** with lowest ***lower bound***
 - Expand it

Front-to-End Bidirectional Heuristic Search with Near-Optimal Node Expansions,
Jingwei Chen, Robert C. Holte, Sandra Zilles and Nathan R. Sturtevant,
International Joint Conference on Artificial Intelligence (IJCAI), **2017**





$A^* \rightarrow \text{NBS}$



- Put start/**goal** onto forward/**backward** priority queues
- While queues not empty / solution not found
 - Among all states on queues:
 - Select the **pair** with lowest ***lower bound***
 - Expand **both** of them

Front-to-End Bidirectional Heuristic Search with Near-Optimal Node Expansions,
Jingwei Chen, Robert C. Holte, Sandra Zilles and Nathan R. Sturtevant,
International Joint Conference on Artificial Intelligence (IJCAI), **2017**





NBS: lower bound

start ●

● goal



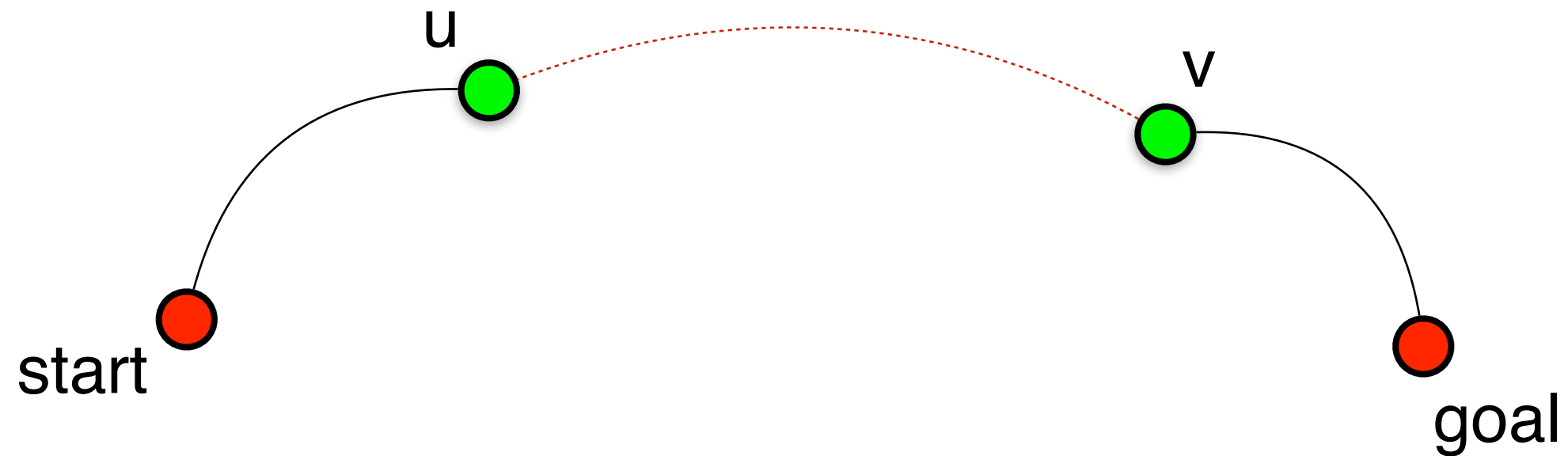


NBS: lower bound



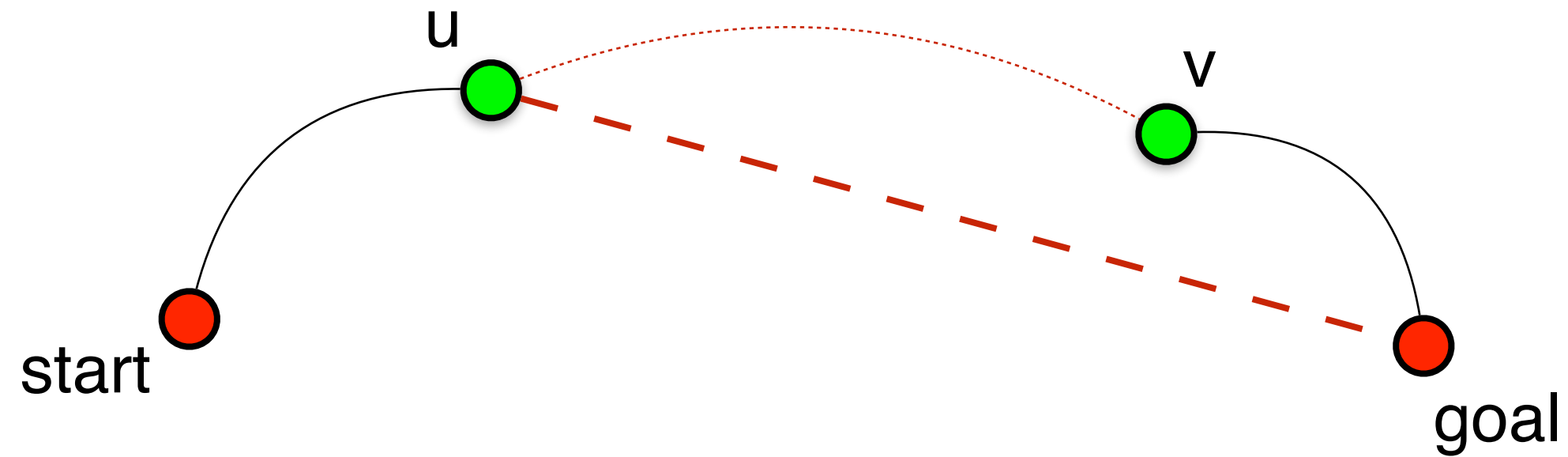


NBS: lower bound



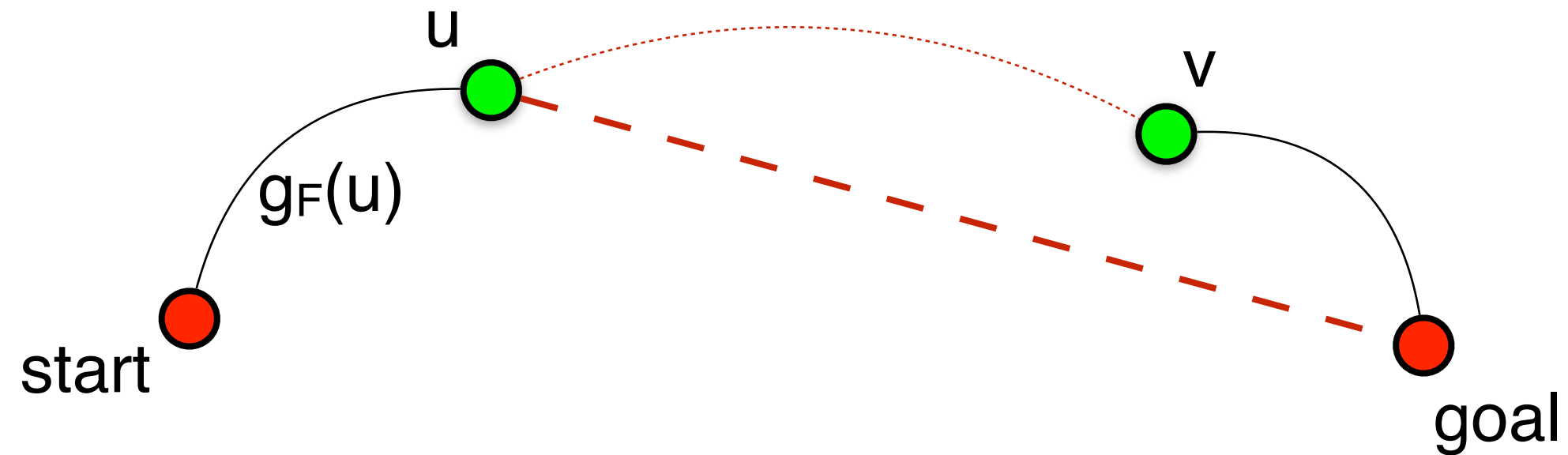


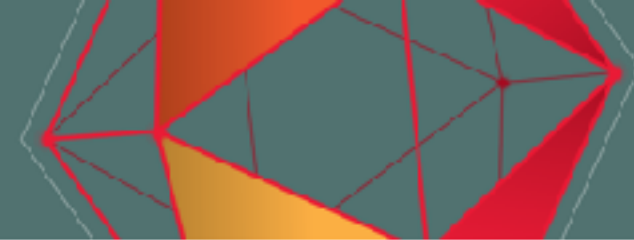
NBS: lower bound



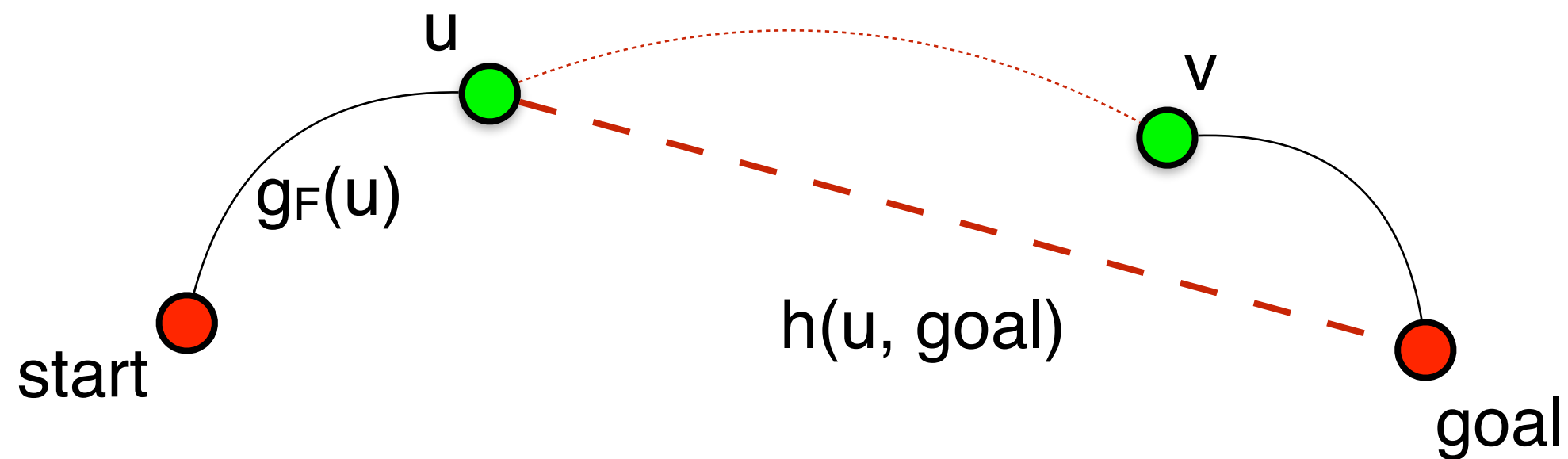


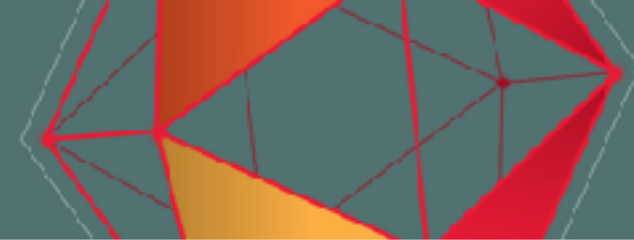
NBS: lower bound



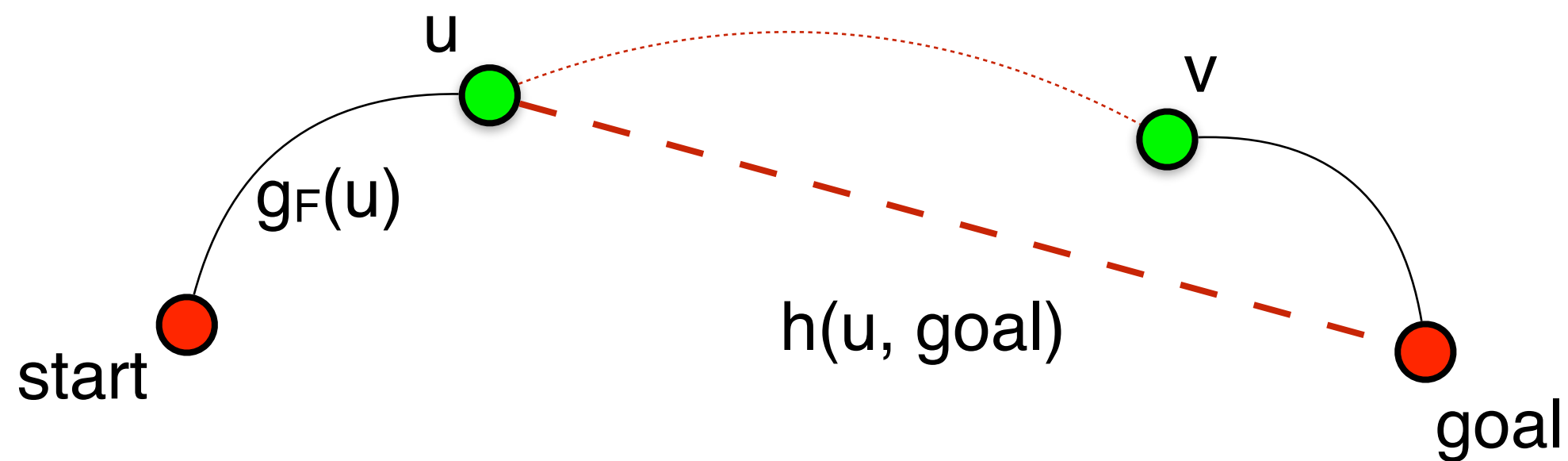


NBS: lower bound





NBS: lower bound

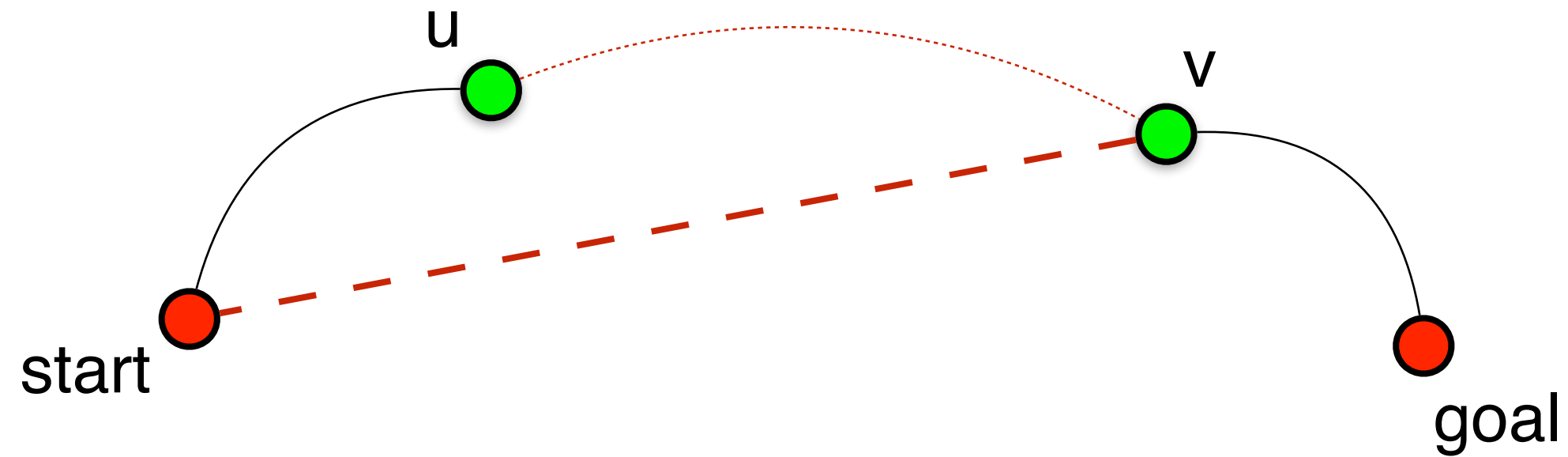


$$f_F(u) = g_F(u) + h(u, \text{goal})$$



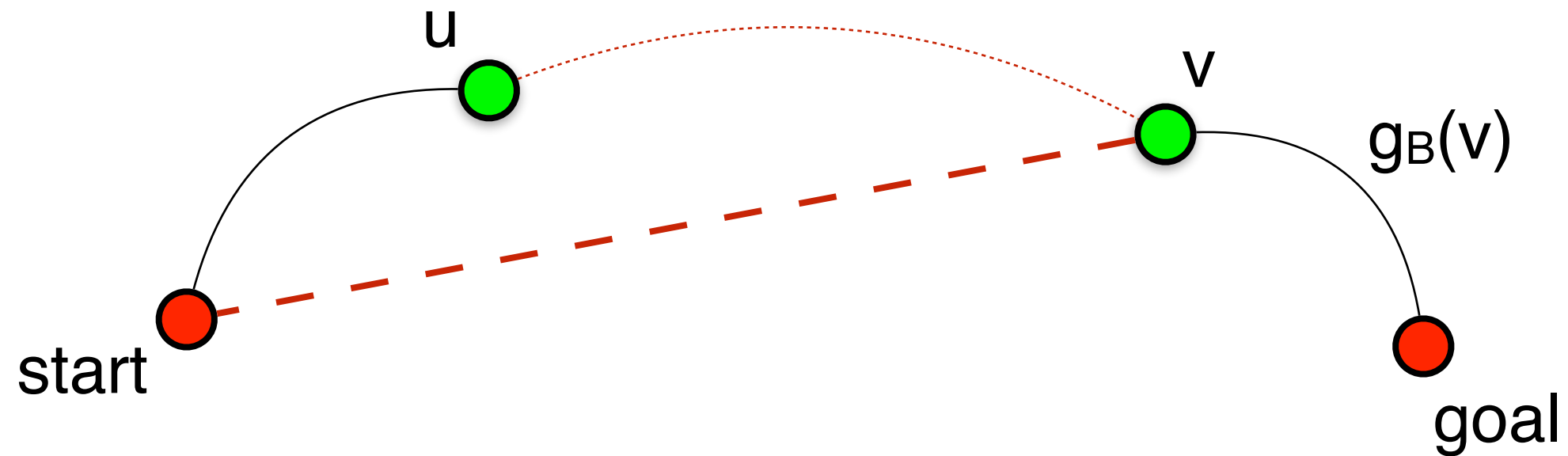


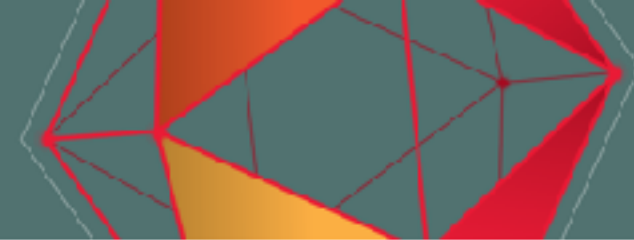
NBS: lower bound



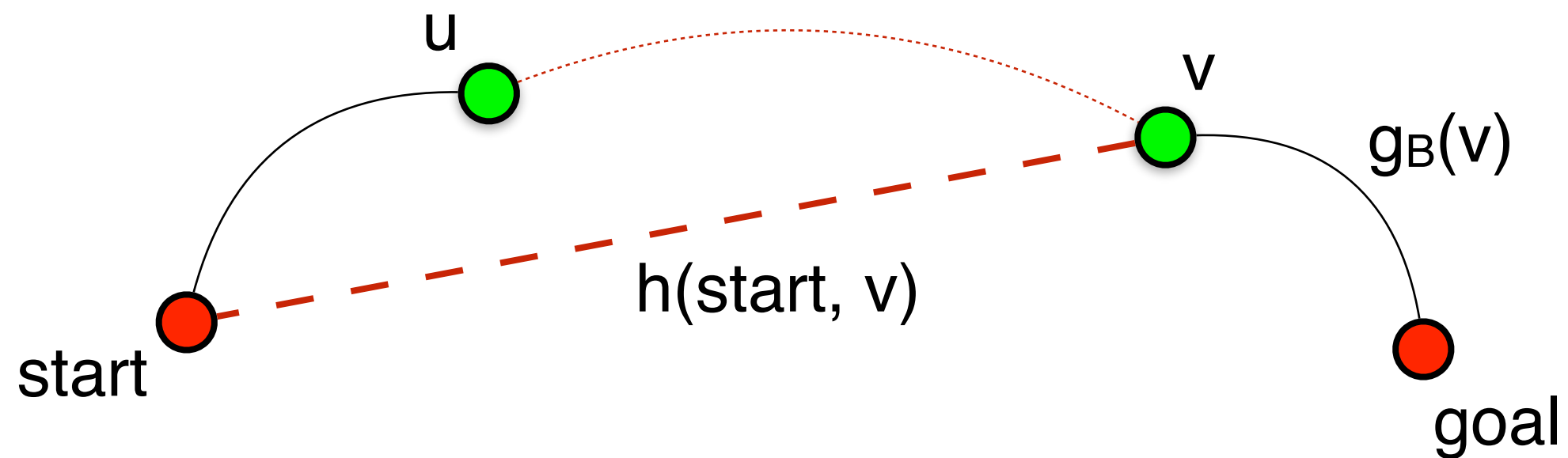


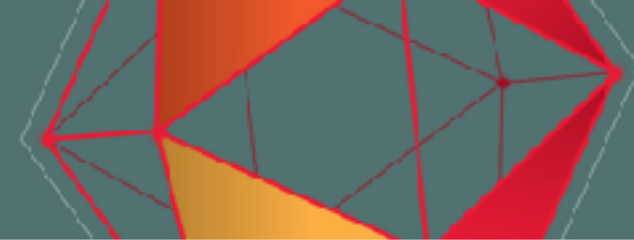
NBS: lower bound



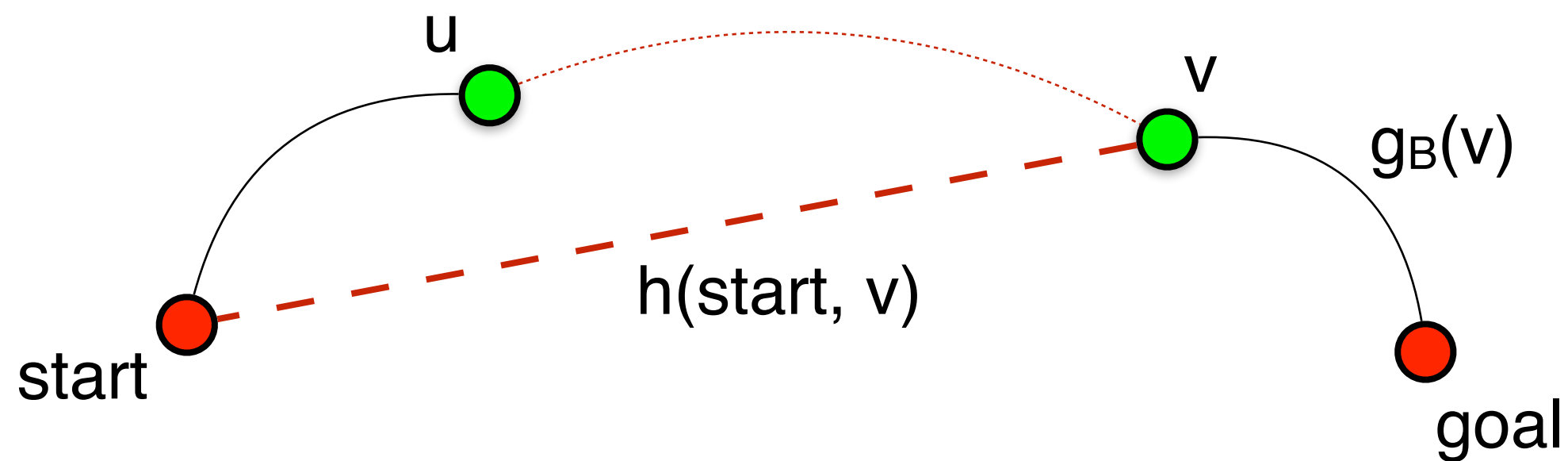


NBS: lower bound





NBS: lower bound

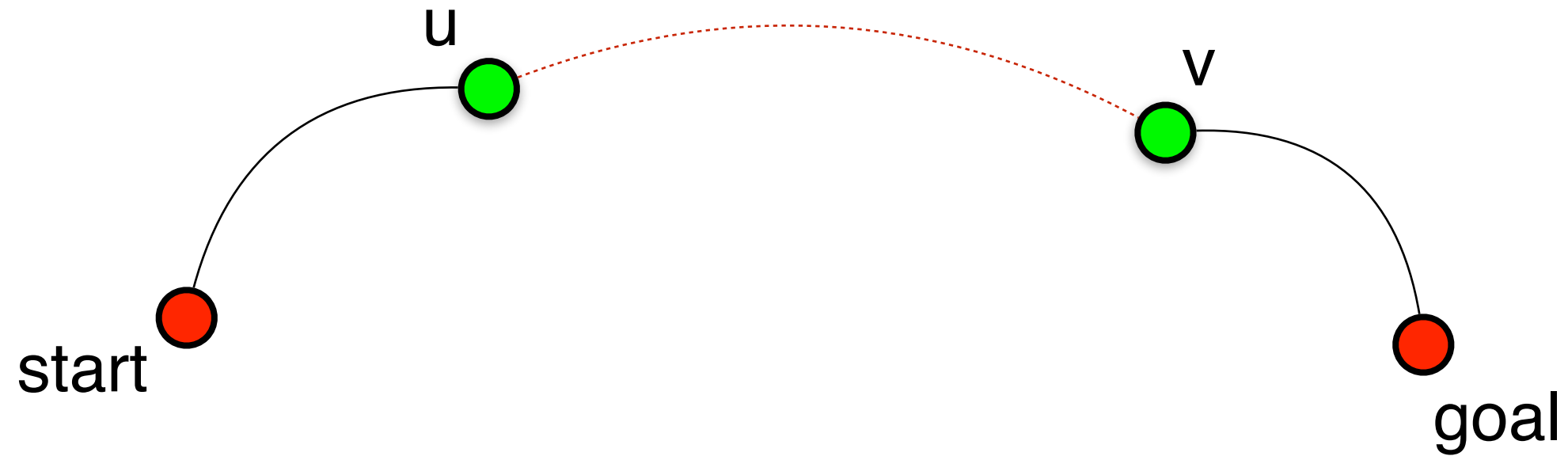


$$f_B(v) = g_B(v) + h(\text{start}, v)$$



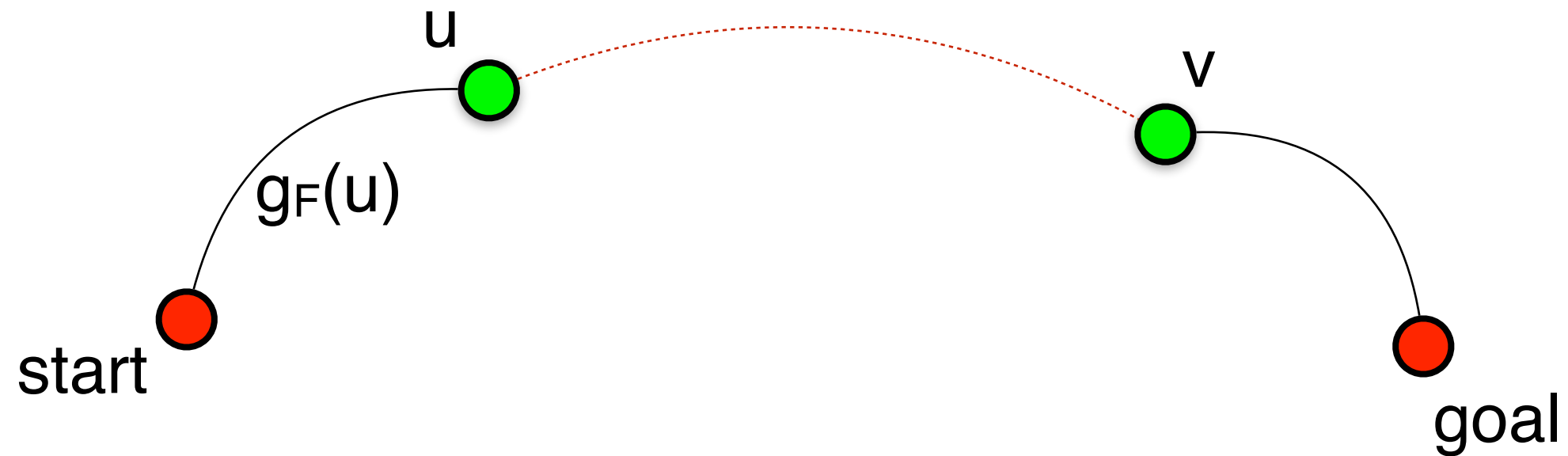


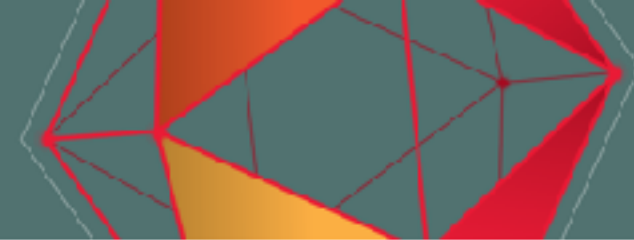
NBS: lower bound



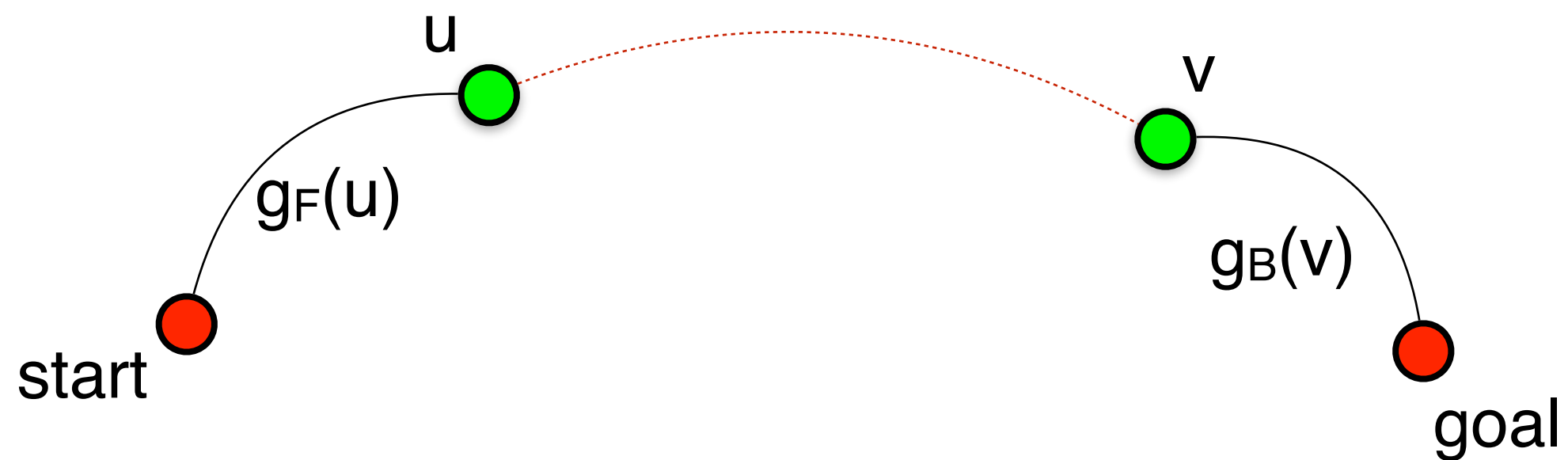


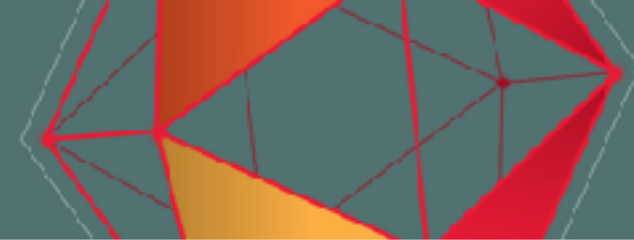
NBS: lower bound



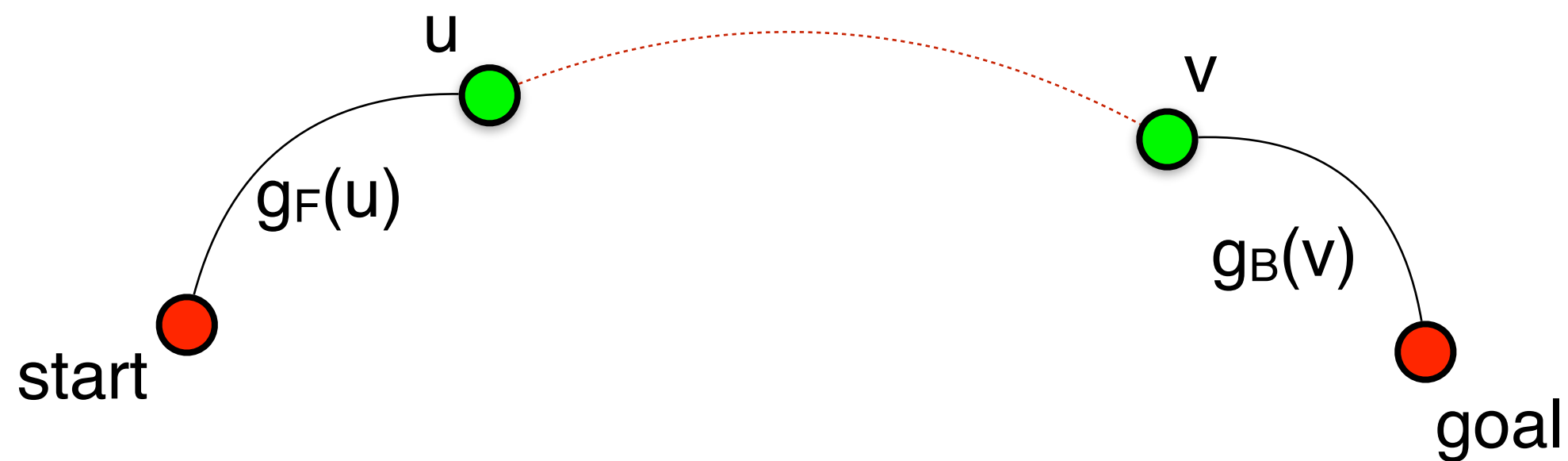


NBS: lower bound



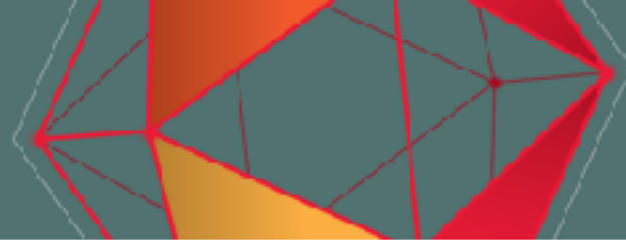


NBS: lower bound



$$g_F(u) + g_B(v)$$





NBS: lower bound



$$\text{lb}(u, v) = \max(f_F(u), \\ f_B(v), \\ g_F(u) + g_B(v))$$





NBS Data Structure



- Can efficiently find pair with minimum lower bound
 - Filter by f-cost then by g-cost





NBS Data Structure



- Can efficiently find pair with minimum lower bound
 - Filter by f-cost then by g-cost
- Cannot just select by f-cost (A^*) or g-cost (Dijkstra)



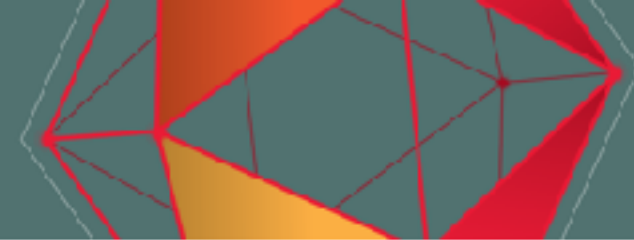


NBS Guarantee



- NBS never expands more than 2x the states expanded by the **best possible** algorithm
 - *In our theoretical framework*
- NBS does equal work in each direction





Suboptimal Solutions



- Use weighted A^* if path quality doesn't matter
- Terminate the search when the first solution is found in bidirectional search

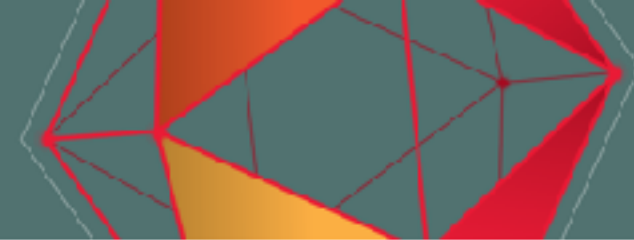




Summary / Conclusions

- Use NBS for bidirectional search
- May want bidirectional search for:
 - Weighted terrain
 - Problem Asymmetry
 - Map Asymmetry
 - Local Minima





Questions?

- <http://www.movingai.com/GDC18/>
 - Open-source implementation of NBS
 - Demo from this lecture*
 - Offline analyzer for analyzing pathfinding
 - Technical reference papers
- Find me on twitter:
 - @nathansttt

