

# Advanced Shortest Paths: A-star Algorithm ( $A^*$ )

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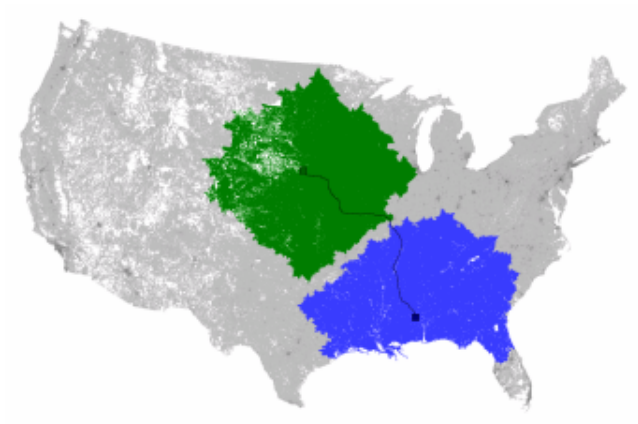
Higher School of Economics

Graph Algorithms  
Data Structures and Algorithms

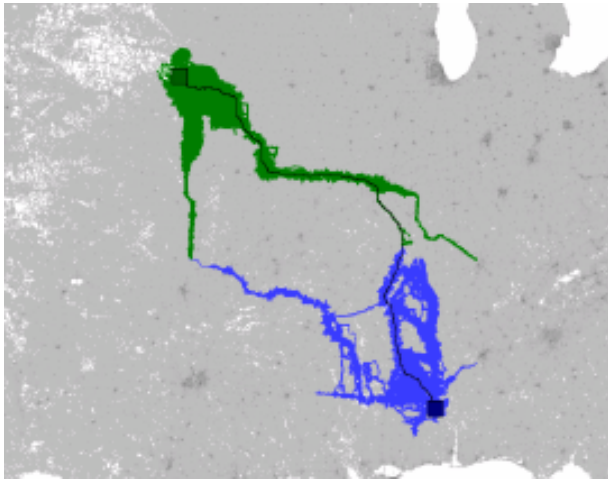
# Outline

- 1 Directed Search
- 2 Bidirectional  $A^*$
- 3 Lower Bounds
- 4 Landmarks

# Bidirectional Search



# Directed Search



# Potential Function

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- Take any potential function  $\pi(v)$  mapping vertices to real numbers.
- It defines new edge weights
$$\ell_{\pi}(u, v) = \ell(u, v) - \pi(u) + \pi(v)$$
- Replacing  $\ell$  by  $\ell_{\pi}$  does not change shortest paths

## Lemma

For any potential function  $\pi : V \rightarrow \mathbb{R}$ , for any two vertices  $s$  and  $t$  in the graph and any path  $P$  between them,

$$\ell_{\pi}(P) = \ell(P) - \pi(s) + \pi(t).$$



# Proof

$$P: s = v_1 \rightarrow v_2 \cdots \rightarrow v_k = t$$

$$\begin{aligned}\ell_\pi(P) &= \sum_{i=1}^{k-1} \ell_\pi(v_i, v_{i+1}) = \\&= \ell(v_1, v_2) - \pi(v_1) + \pi(v_2) + \\&+ \ell(v_2, v_3) - \pi(v_2) + \pi(v_3) + \\&+ \dots + \\&+ \ell(v_{k-2}, v_{k-1}) - \pi(v_{k-2}) + \pi(v_{k-1}) + \\&+ \ell(v_{k-1}, v_k) - \pi(v_{k-1}) + \pi(v_k) = \\&= \sum_{i=1}^{k-1} \ell(v_i, v_{i+1}) - \pi(v_1) + \pi(v_k) = \\&= \ell(P) - \pi(s) + \pi(t)\end{aligned}$$

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# Dijkstra with Potentials

- Take some potential function  $\pi$
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- The resulting shortest path is also a shortest path initially
- Does any  $\pi$  fit us?
- For any edge  $(u, v)$ , the new length  $\ell_\pi(u, v)$  must be non-negative — such  $\pi$  is called **feasible**

# Intuition

- $\pi(v)$  is an estimation of  $d(v, t)$  — “how far is it from here to  $t$ ?”
- If we have such estimation, we can often avoid going wrong direction — directed search
- Typically  $\pi(v)$  is a lower bound on  $d(v, t)$
- I.e., on a real map a path from  $v$  to  $t$  cannot be shorter than the straight line segment from  $v$  to  $t$

# $A^* \equiv \text{Dijkstra}$

- On each step, pick the vertex  $v$  minimizing  $\text{dist}[v] - \pi(s) + \pi(v)$
- $\pi(s)$  is the same for all  $v$ , so  $v$  minimizes  $\text{dist}[v] + \pi(v)$  — the most promising vertex
- $\pi(v)$  is an estimate of  $d(v, t)$
- Pick the vertex  $v$  with the minimum current estimate of  $d(s, v) + d(v, t)$
- Thus the search is directed



# Performance of $A^*$

If  $\pi(v)$  gives lower bound on  $d(v, t)$

- Worst case:  $\pi(v) = 0$  for all  $v$  — the same as Dijkstra
- Best case:  $\pi(v) = d(v, t)$  for all  $v$  — then  $\ell_\pi(u, v) = 0$  iff  $(u, v)$  is on a shortest path to  $t$ , so search visits only the edges of shortest  $s - t$  paths
- It can be shown that the tighter are the lower bounds — the fewer vertices will be scanned

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# Bidirectional $A^*$

- Same as Bidirectional Dijkstra, but with potentials
- Needs two potential functions:  $\pi_f(v)$  estimates  $d(v, t)$ ,  $\pi_r(v)$  estimates  $d(s, v)$
- Problem: different edge weights:  
$$\ell_{\pi_f}(u, v) = \ell(u, v) - \pi_f(u) + \pi_f(v),$$
$$\ell_{\pi_r}(u, v) = \ell(u, v) - \pi_r(v) + \pi_r(u)$$

# Bidirectional $A^*$

- We need  $\ell_{\pi_f}(u, v) = \ell_{\pi_r}(u, v) \Rightarrow$   
 $\pi_f(u) + \pi_r(u) = \pi_f(v) + \pi_r(v)$  for any  
 $(u, v)$
- Need constant  $\pi_f(u) + \pi_r(u)$  for any  $u$
- Use  $p_f(u) = \frac{\pi_f(u) - \pi_r(u)}{2}$ ,  $p_r(u) = -p_f(u)$
- Then  $p_f(u) + p_r(u) = 0$  for any  $u$

## Lemma

If  $\pi_f$  is a feasible potential for forward search, and  $\pi_r$  is a feasible potential for reverse search, then  $p_f = \frac{\pi_f - \pi_r}{2}$  is a feasible potential for forward search.

## Proof

- $\ell(u, v) - \pi_f(u) + \pi_f(v) \geq 0$

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- $\ell(u, v) - \pi_r(v) + \pi_r(u) \geq 0$
- $2\ell(u, v) - (\pi_f(u) - \pi_r(u)) + (\pi_f(v) - \pi_r(v)) \geq 0$



## Proof

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- $\ell(u, v) - \frac{\pi_f(u) - \pi_r(u)}{2} + \frac{\pi_f(v) - \pi_r(v)}{2} \geq 0$

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- $\ell(u, v) - \frac{\pi_f(u) - \pi_r(u)}{2} + \frac{\pi_f(v) - \pi_r(v)}{2} \geq 0$
- $\ell(u, v) - p_f(u) + p_f(v) \geq 0$



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## Lemma

If  $\pi$  is feasible, and  $\pi(t) \leq 0$ , then  
 $\pi(v) \leq d(v, t)$  for any  $v$

## Proof

- $\ell_\pi(x, y) \geq 0$  for any  $x, y$ , so  $\ell_\pi(P) \geq 0$  for any path  $P$
- Take a  $v - t$  shortest path  
 $P = (v, w_1, w_2, \dots, w_k, t)$
- $0 \leq \ell_\pi(P) = \ell(P) - \pi(v) + \pi(t) \leq \ell(P) - \pi(v) \Rightarrow \pi(v) \leq \ell(P) = d(v, t)$



# Euclidean Potential

## Lemma

Consider a road network on a plane map with each vertex  $v$  having coordinates  $(v.x, v.y)$ .

The potential given by Euclidean distance (length of a line segment) between  $v$  and  $t$

$\pi(v) = d_E(v, t) = \sqrt{(v.x - t.x)^2 + (v.y - t.y)^2}$  is feasible, and  $\pi(t) = 0$ .

## Proof

- For any edge  $(u, v) \in E$ ,  
 $\ell(u, v) \geq d_E(u, v)$ , because line segment is the shortest path between two points on a plane

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- $\pi(u) = d_E(u, t) \leq_{(\text{triangle inequality})} d_E(u, v) + d_E(v, t) \leq \ell(u, v) + \pi(v) \Rightarrow$   
 $\ell(u, v) - \pi(u) + \pi(v) \geq 0$



# Proof

- For any edge  $(u, v) \in E$ ,  
 $\ell(u, v) \geq d_E(u, v)$ , because line segment is the shortest path between two points on a plane
- $\pi(u) = d_E(u, t) \leq_{(\text{triangle inequality})} d_E(u, v) + d_E(v, t) \leq \ell(u, v) + \pi(v) \Rightarrow \ell(u, v) - \pi(u) + \pi(v) \geq 0$
- $\pi(t) = d_E(t, t) = 0$



# $A^*$ on a Plane Map

- Need to find the shortest path from  $s$  to  $t$
- For each  $v$ , compute  $\pi(v) = d_E(v, t)$
- Launch Dijkstra with potentials  $\pi(v)$

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# Landmarks

## Lemma

Fix some vertex  $A \in V$ , we will call it a **landmark**. Then the potential  $\pi(v) = d(A, t) - d(A, v)$  is feasible, and  $\pi(t) = 0$ .

## Proof

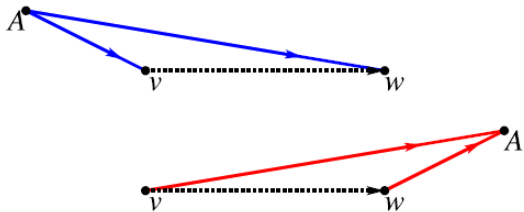
- $\ell(u, v) - \pi(u) + \pi(v) = \ell(u, v) - d(A, t) + d(A, u) + d(A, t) - d(A, v) = d(A, u) + \ell(u, v) - d(A, v) \geq_{(\text{triangle inequality})} 0$
- $\pi(t) = d(A, t) - d(A, t) = 0$  □

# Landmarks

- Select several landmarks and precompute their distances to all other vertices
- For any landmark  $A$ ,  
$$d(v, t) \geq d(A, t) - d(A, v),$$
$$d(v, t) \geq d(v, A) - d(t, A)$$
- Tightest lower bound  $d(v, t) \geq \max(d(A, t) - d(A, v), d(v, A) - d(t, A))$  over all  $A$

# Landmark Selection

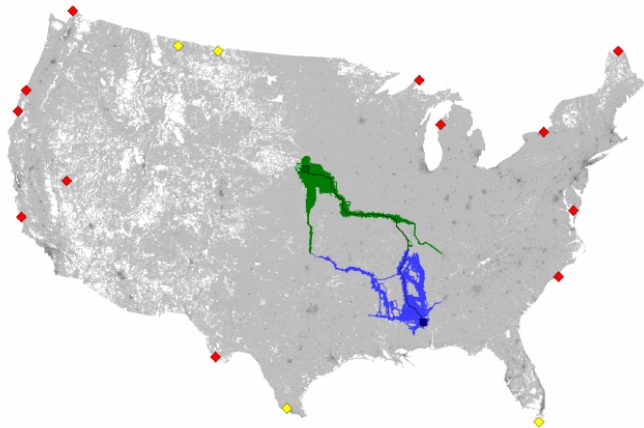
Good landmark appears “before”  $v$  or “after”  $w$ :



For any query  $(s, t)$ , we need some landmarks before  $s$  and after  $t$

# Landmark Selection

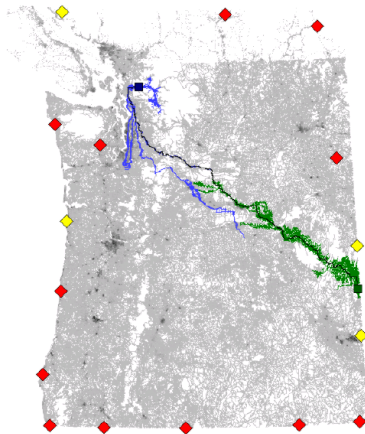
Choosing landmarks on the border seems reasonable:





# Landmark Selection

Choosing landmarks on the border seems reasonable:



# Conclusion

- Directed search can scan fewer vertices
- $A^*$  is a directed search algorithm based on Dijkstra and potential functions
- $A^*$  can also be bidirectional
- Euclidean distance is a potential for a plane (road networks)
- Landmarks can be used for good potential function, but we need preprocessing to use them