Lecture Overview

- Single-source single-target Dijkstra
- Bidirectional search
- Goal directed search - potentials and landmarks

Readings


**DIJKSTRA** single-source, single-target

```
Initialize()
Q ← V[G]
while Q ≠ φ
do u ← EXTRACT_MIN(Q) (stop if u = t!)
for each vertex v ∈ Adj[u]
do RELAX(u, v, w)
```

Observation: If only shortest path from s to t is required, stop when t is removed from Q, i.e., when u = t

**DIJKSTRA** Demo

![Dijkstra Demonstration](image)

Figure 1: Dijkstra Demonstration with Balls and String
Bi-Directional Search

Note: Speedup techniques covered here do not change worst-case behavior, but reduce the number of visited vertices in practice.

![Bi-Directional Search](image)

Figure 2: Bi-directional Search

**Bi-D Search**

Alternate forward search from $s$
backward search from $t$
(follow edges backward)
$d_f(u)$ distances for forward search
$d_b(u)$ distances for backward search

Algorithm terminates when some vertex $w$ has been processed, i.e., deleted from the queue of both searches, $Q_f$ and $Q_b$

![Bi-D Search](image)

Figure 3: Bi-D Search
**Subtlety:** After search terminates, find node $x$ with minimum value of $d_f(x) + d_b(x)$. $x$ may not be the vertex $w$ that caused termination as in example to the left!

Find shortest path from $s$ to $x$ using $\Pi_f$ and shortest path backwards from $t$ to $x$ using $\Pi_b$.

*Note:* $x$ will have been deleted from either $Q_f$ or $Q_b$ or both.

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**Figure 4:** Forward and Backward Search

Minimum value for $d_f(x) + d_b(x)$ over all vertices that have been processed in at least one search

$$d_f(u) + d_b(u) = 3 + 6 = 9$$
\[ d_f(u') + d_b(u') = 6 + 3 = 9 \]
\[ d_f(w) + d_b(w) = 5 + 5 = 10 \]

**Goal-Directed Search or \( A^* \)**

Modify edge weights with potential function over vertices.

\[ \overline{w}(u, v) = w(u, v) - \lambda(u) + \lambda(v) \]

Search toward target:

![Diagram](image)

**Figure 5: Targeted Search**

**Correctness**

\[ \overline{w}(p) = w(p) - \lambda_l(s) + \lambda_l(t) \]

So shortest paths are maintained in modified graph with \( \overline{w} \) weights.

![Diagram](image)

**Figure 6: Modifying Edge Weights**

To apply Dijkstra, we need \( \overline{w}(u, v) \geq 0 \) for all \((u, v)\).

Choose potential function appropriately, to be feasible.

**Landmarks**

Small set of landmarks \( LCV \). For all \( u \in V, l \in L \), pre-compute \( \delta(u, l) \). Potential \( \lambda^{(l)}_l(u) = \delta(u, l) = \delta(t, l) \) for each \( l \).

CLAIM: \( \lambda^{(l)}_l \) is feasible.
Feasibility

\[
\bar{w}(u, v) = w(u, v) - \lambda_t^{(l)}(u) + \lambda_t^{(l)}(v) \\
= w(u, v) - \delta(u, l) + \delta(t, l) + \delta(v, l) - \delta(t, l) \\
= w(u, v) - \delta(u, l) + \delta(v, l) \geq 0 \quad \text{by the } \Delta \text{-inequality} \\
\lambda_t(u) = \max_{l \in L} \lambda_t^{(l)}(u) \quad \text{is also feasible}
\]