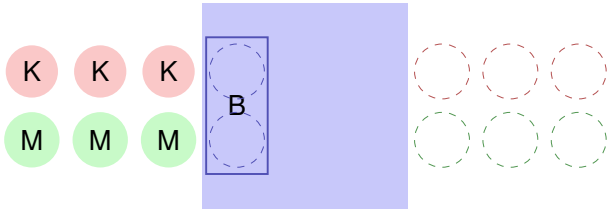
24. Shortest Paths

Motivation, Dijkstra's algorithm on distance graphs, Bellman-Ford Algorithm, Floyd-Warshall Algorithm

[Ottman/Widmayer, Kap. 9.5 Cormen et al, Kap. 24.1-24.3, 25.2-25.3]

River Crossing (Missionaries and Cannibals)

Problem: Three cannibals and three missionaries are standing at a river bank. The available boat can carry two people. At no time may at any place (banks or boat) be more cannibals than missionaries. How can the missionaries and cannibals cross the river as fast as possible? [36]



[36] There are slight variations of this problem. It is equivalent to the jealous husbands problem.

Problem as Graph

Enumerate permitted configurations as nodes and connect them with an edge, when a crossing is allowed. The problem then becomes a shortest path problem.

Example

	links	rechts
Missionare	3	0
Kannibalen	3	0
Boot	x	

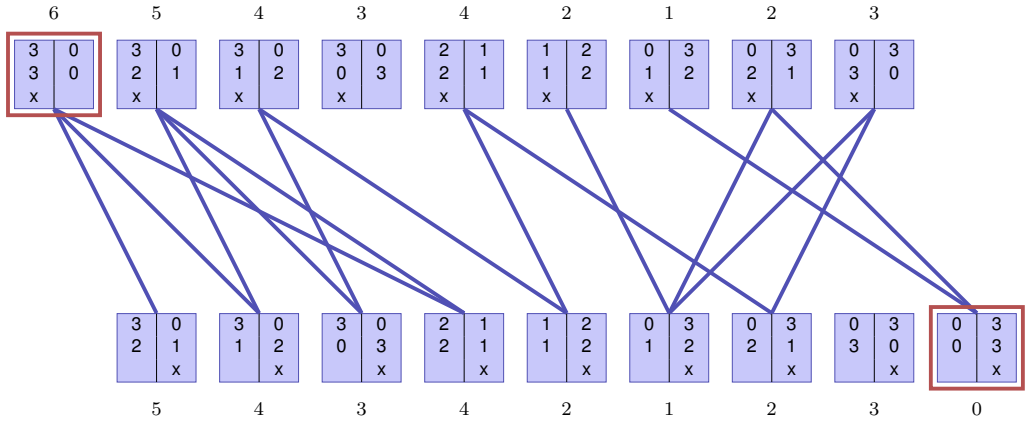
Überfahrt möglich

	links	rechts
Missionare	2	1
Kannibalen	2	1
Boot		x

6 Personen am linken Ufer

4 Personen am linken Ufer

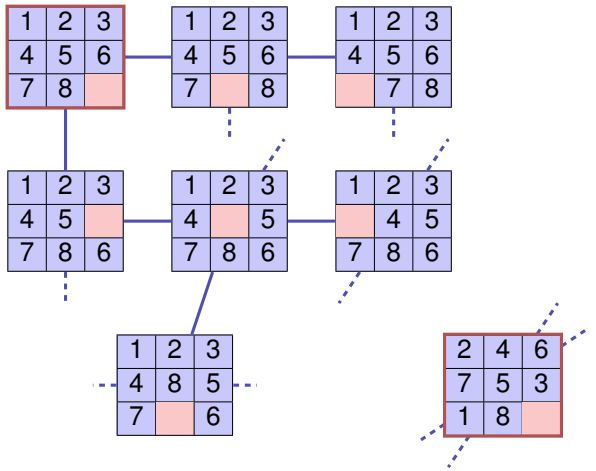
The whole problem as a graph



Example Mystic Square

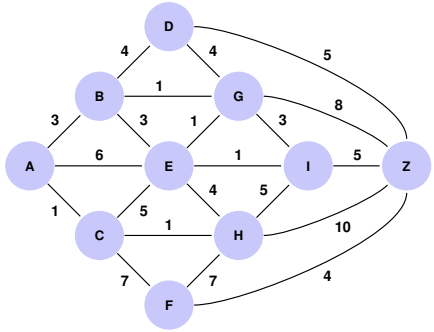
Want to find the fastest solution for

Problem as Graph



Route Finding

Provided cities A - Z and Distances between cities.

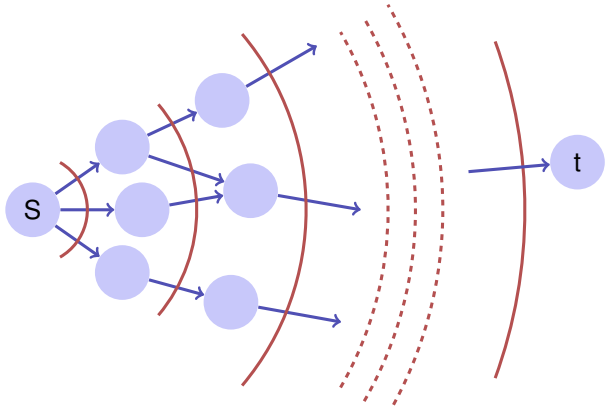


What is the shortest path from A to Z?

Simplest Case

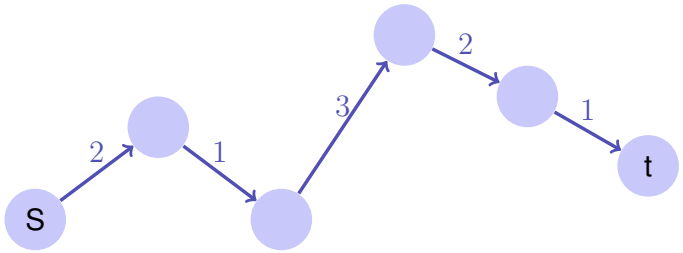
Constant edge weight 1 (wlog)

Solution: Breadth First Search



Graphs with positive weights

Given: $G = (V, E, c)$, $c : E \to \mathbb{R}^+$, $s, t \in V$.
Wanted: Length of a shortest path (weight) from s to t.
Path: $\langle s = v_0, v_1, \ldots, v_k = t \rangle$, $(v_i, v_{i+1}) \in E$ $(0 \leq i < k)$
Weight: $\sum_{i=0}^{k-1} c((v_i, v_{i+1}))$.



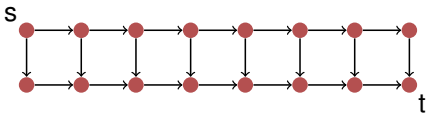
Path with weight 9

Existence of Shortest Path

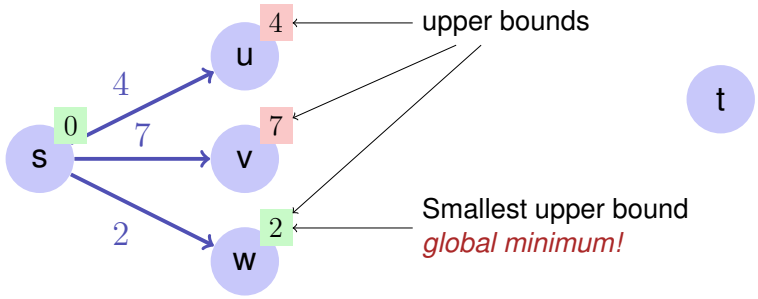
Assumption: There is a path from s to t in G.
Claim: There is a shortest path from s to t in G.

Proof: There can be infinitely many paths from s to t (cycles are possible). However, since c is positive, a shortest path must be acyclic. Thus the maximal length of a shortest path is bounded by some $n \in \mathbb{N}$ and there are only finitely many candidates for a shortest path.

Remark: There can be exponentially many paths. Example



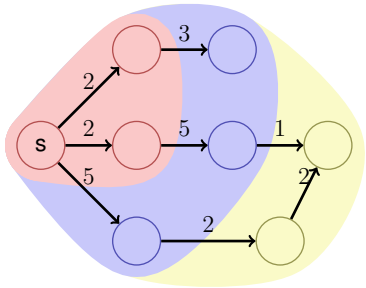
Observation



Basic Idea

Set V of nodes is partitioned into

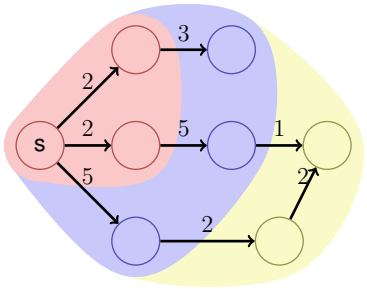
- the set M of nodes for which a shortest path from s is already known,
- the set $R = \bigcup_{v \in M} N^+(v) \setminus M$ of nodes where a shortest path is not yet known but that are accessible directly from M,
- the set $U = V \setminus (M \cup R)$ of nodes that have not yet been considered.



Induction

Induction over $|M|$: choose nodes from R with smallest upper bound. Add r to M and update R and U accordingly.

Correctness: if within the "wavefront" a node with minimal weight has been found then no path with greater weight over different nodes can provide any improvement.



Algorithm Dijkstra(G, s) [formal]

Input : Positively weighted Graph $G = (V, E, c)$, starting point $s \in V$,
Output : Minimal weights d of the shortest paths.

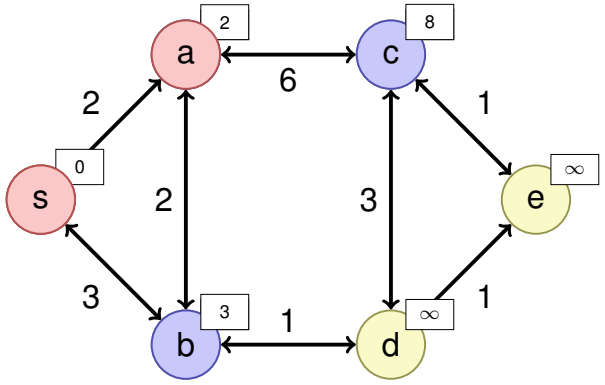
$M = \{s\}$; $R = N^+(s)$, $U = V \setminus R$
$d(s) \leftarrow 0$; $d(u) \leftarrow \infty \ \forall u \neq s$
while $R \neq \emptyset$ **do**
  $r \leftarrow \arg\min_{r \in R} \min_{m \in N^-(r) \cap M} d(m) + c(m, r)$
  $d(r) \leftarrow \min_{m \in N^-(r) \cap M} d(m) + c(m, r)$
  $M \leftarrow M \cup \{r\}$
  $R \leftarrow R - \{r\} \cup N^+(r) \setminus M$
return d

Algorithmus Dijkstra

Initial: $\mathrm{PL}(n) \leftarrow \infty$ für alle Knoten.

- Set $\mathrm{PL}(s) \leftarrow 0$
- Start with $M = \{s\}$. Set $k \leftarrow s$.
- While a new node k is added and this is not the target node
  1. For each neighbour node n of k:
     - compute path length x to n via k
     - If $\mathrm{PL}(n) = \infty$, than add n to R
     - If $x < \mathrm{PL}(n) < \infty$, then set $\mathrm{PL}(n) \leftarrow x$ and adapt R .
  2. Choose as new node k the node with smallest path length in R.

Example



$M = \{s, a\}$
$R = \{b, c\}$
$U = \{d, e\}$

Implementation: Naive Variant

- Find minimum: traverse all edges (u, v) for $u \in M, v \in R$.
- Overal costs: $\mathcal{O}(|V| \cdot |E|)$

Implementation: Better Variant

- Update of all outgoing edges when inserting new w in M:

  foreach $v \in N^+(w)$ **do**
    if $d(w) + c(w, v) < d(v)$ **then**
      $d(v) \leftarrow d(w) + c(w, v)$

- Costs of updates: $\mathcal{O}(|E|)$, Find minima: $\mathcal{O}(|V|^2)$, overal costs $\mathcal{O}(|V|^2)$

Implementation: Data Structure for R?

Required operations:

- Insert (add to R)
- ExtractMin (over R) and DecreaseKey (Update in R)

  foreach $v \in N^+(m)$ **do**
    if $d(m) + c(m, v) < d(v)$ **then**
      $d(v) \leftarrow d(m) + c(m, v)$
      if $v \in R$ **then**
        DecreaseKey(R, v) // Update of a $d(v)$ in the heap of R
      else
        $R \leftarrow R \cup \{v\}$ // Update of $d(v)$ in the heap of R

MinHeap!

DecreaseKey

- DecreaseKey: climbing in MinHeap in $\mathcal{O}(\log |V|)$
- Position in the heap?
  - alternative (a): Store position at the nodes
  - alternative (b): Hashtable of the nodes
  - alterantive (c): re-insert node after update-operation and mark it "deleted" once extracted (Lazy Deletion)

Runtime

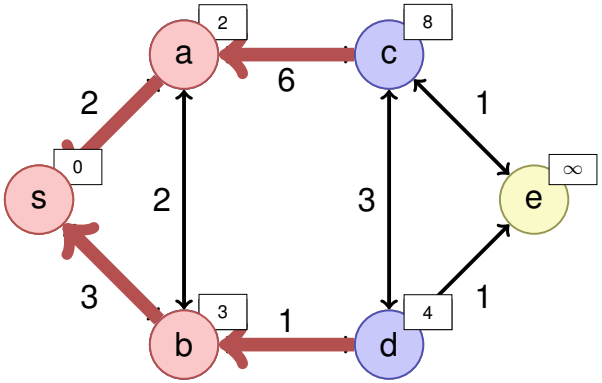
- $|V|\times$ ExtractMin: $\mathcal{O}(|V|\log|V|)$
- $|E|\times$ Insert or DecreaseKey: $\mathcal{O}(|E|\log|V|)$
- $1\times$ Init: $\mathcal{O}(|V|)$
- Overal: $\mathcal{O}(|E|\log|V|)$.

Can be improved when a data structure optimized for ExtractMin and DecreaseKey ist used (Fibonacci Heap), then runtime $\mathcal{O}(|E|+|V|\log|V|)$.

Reconstruct shortest Path

- Memorize best predecessor during the update step in the algorithm above. Store it with the node or in a separate data structure.
- Reconstruct best path by traversing backwards via best predecessor

Example

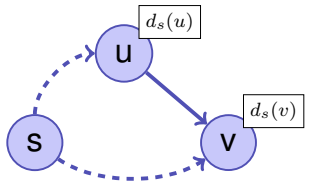


$$M = \{s, a, b\}$$
$$R = \{c, d\}$$
$$U = \{e\}$$

General Weighted Graphs

Relaxing Step as with Dijkstra:

Relax(u, v) $(u, v \in V, (u, v) \in E)$
if $d_s(v) > d_s(u) + c(u, v)$ **then**
  $d_s(v) \leftarrow d_s(u) + c(u, v)$
  return true
return false



Problem: cycles with negative weights can shorten the path, a shortest path is not guaranteed to exist.

Observations

- *Observation 1:* Sub-paths of shortest paths are shortest paths.
  Let $p = \langle v_0, \ldots, v_k \rangle$ be a shortest path from v_0 to v_k. Then each of the sub-paths $p_{ij} = \langle v_i, \ldots, v_j \rangle$ $(0 \leq i < j \leq k)$ is a shortest path from v_i to v_j.
  Proof: if not, then one of the sub-paths could be shortened which immediately leads to a contradiction.
- *Observation:* If there is a shortest path then it is simple, thus does not provide a node more than once.
  Immediate Consequence of observation 1.

Dynamic Programming Approach (Bellman)

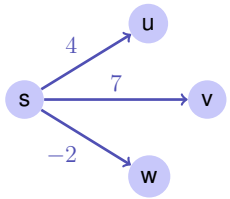
Induction over number of edges $d_s[i, v]$: Shortest path from s to v via maximally i edges.

$$d_s[i, v] = \min\{d_s[i-1, v], \min_{(u,v)\in E}(d_s[i-1, u] + c(u, v))$$

$$d_s[0, s] = 0, d_s[0, v] = \infty \; \forall v \neq s.$$

Dynamic Programming Approach (Bellman)

	s	$\cdots$	v	$\cdots$	w
0	0	∞	∞	∞	∞
1	0	∞	7	∞	-2
$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$
$n-1$	0	$\cdots$	$\cdots$	$\cdots$	$\cdots$



Algorithm: Iterate over last row until the relaxation steps do not provide any further changes, maximally $n-1$ iterations. If still changes, then there is no shortest path.

Algorithm Bellman-Ford(G, s)

Input : Graph $G = (V, E, c)$, starting point $s \in V$
Output : If return value true, minimal weights d for all shortest paths from s, otherwise no shortest path.

$d(v) \leftarrow \infty \; \forall v \in V$; $d(s) \leftarrow 0$
for $i \leftarrow 1$ **to** $|V|$ **do**
  $f \leftarrow$ false
  foreach $(u, v) \in E$ **do**
    $f \leftarrow f \vee \text{Relax}(u, v)$
  if $f =$ false **then return** true
return false;

Runtime $\mathcal{O}(|E| \cdot |V|)$.

All shortest Paths

Compute the weight of a shortest path for each pair of nodes.

- $|V|\times$ Application of Dijkstra's Shortest Path algorithm $\mathcal{O}(|V| \cdot |E| \cdot \log |V|)$ (with Fibonacci Heap: $\mathcal{O}(|V|^2 \log |V| + |V| \cdot |E|)$)
- $|V|\times$ Application of Bellman-Ford: $\mathcal{O}(|E| \cdot |V|^2)$
- There are better ways!

Induction via node number[37]

Consider weights of all shortest paths S^k with intermediate nodes in $V^k := \{v_1, \ldots, v_k\}$, provided that weights for all shortest paths S^{k-1} with intermediate nodes in V^{k-1} are given.

- v_k no intermediate node of a shortest path of $v_i \rightsquigarrow v_j$ in V^k: Weight of a shortest path $v_i \rightsquigarrow v_j$ in S^{k-1} is then also weight of shortest path in S^k.
- v_k intermediate node of a shortest path $v_i \rightsquigarrow v_j$ in V^k: Sub-paths $v_i \rightsquigarrow v_k$ and $v_k \rightsquigarrow v_j$ contain intermediate nodes only from S^{k-1}.

[37] like for the algorithm of the reflexive transitive closure of Warshall

DP Induction

$d^k(u, v)$ = Minimal weight of a path $u \rightsquigarrow v$ with intermediate nodes in V^k

Induktion

$$d^k(u, v) = \min\{d^{k-1}(u, v), d^{k-1}(u, k) + d^{k-1}(k, v)\} (k \geq 1)$$
$$d^0(u, v) = c(u, v)$$

DP Algorithm Floyd-Warshall(G)

Input : Acyclic Graph $G = (V, E, c)$
Output : Minimal weights of all paths d

$d^0 \leftarrow c$
for $k \leftarrow 1$ **to** $|V|$ **do**
  for $i \leftarrow 1$ **to** $|V|$ **do**
    for $j \leftarrow 1$ **to** $|V|$ **do**
      $d^k(v_i, v_j) = \min\{d^{k-1}(v_i, v_j), d^{k-1}(v_i, v_k) + d^{k-1}(v_k, v_j)\}$

Runtime: $\Theta(|V|^3)$

Remark: Algorithm can be executed with a single matrix d (in place).

Reweighting

Idea: Reweighting the graph in order to apply Dijkstra's algorithm.

The following does *not* work. The graphs are not equivalent in terms of shortest paths.

Reweighting

Other Idea: "Potential" (Height) on the nodes

- $G = (V, E, c)$ a weighted graph.
- Mapping $h : V \to \mathbb{R}$
- New weights

$$\tilde{c}(u, v) = c(u, v) + h(u) - h(v),\ (u, v \in V)$$

Reweighting

Observation: A path p is shortest path in in $G = (V, E, c)$ iff it is shortest path in in $\tilde{G} = (V, E, \tilde{c})$

$$\begin{aligned}\tilde{c}(p) &= \sum_{i=1}^{k} \tilde{c}(v_{i-1}, v_i) = \sum_{i=1}^{k} c(v_{i-1}, v_i) + h(v_{i-1}) - h(v_i) \\ &= h(v_0) - h(v_k) + \sum_{i=1}^{k} c(v_{i-1}, v_i) = c(p) + h(v_0) - h(v_k)\end{aligned}$$

Thus $\tilde{c}(p)$ minimal in all $v_0 \rightsquigarrow v_k \iff c(p)$ minimal in all $v_0 \rightsquigarrow v_k$.

Weights of cycles are invariant: $\tilde{c}(v_0, \ldots, v_k = v_0) = c(v_0, \ldots, v_k = v_0)$

Johnson's Algorithm

Add a new node $s \notin V$:

$$\begin{aligned}G' &= (V', E', c') \\ V' &= V \cup \{s\} \\ E' &= E \cup \{(s, v) : v \in V\} \\ c'(u, v) &= c(u, v),\ u \neq s \\ c'(s, v) &= 0 (v \in V)\end{aligned}$$

Johnson's Algorithm

If no negative cycles, choose as height function the weight of the shortest paths from s,

$$h(v) = d(s, v).$$

For a minimal weight d of a path the following triangular inequality holds:

$$d(s, v) \leq d(s, u) + c(u, v).$$

Substitution yields $h(v) \leq h(u) + c(u, v)$. Therefore

$$\tilde{c}(u, v) = c(u, v) + h(u) - h(v) \geq 0.$$

Algorithm Johnson(G)

Input : Weighted Graph $G = (V, E, c)$
Output : Minimal weights of all paths D.

New node s. Compute $G' = (V', E', c')$
if BellmanFord$(G', s) =$ false **then** return "graph has negative cycles"
foreach $v \in V'$ **do**
    $h(v) \leftarrow d(s, v)$ // d aus BellmanFord Algorithmus
foreach $(u, v) \in E'$ **do**
    $\tilde{c}(u, v) \leftarrow c(u, v) + h(u) - h(v)$
foreach $u \in V$ **do**
    $\tilde{d}(u, \cdot) \leftarrow$ Dijkstra$(\tilde{G}', u)$
    foreach $v \in V$ **do**
        $D(u, v) \leftarrow \tilde{d}(u, v) + h(v) - h(u)$

Analysis

Runtimes

- Computation of G': $\mathcal{O}(|V|)$
- Bellman Ford G': $\mathcal{O}(|V| \cdot |E|)$
- $|V|\times$ Dijkstra $\mathcal{O}(|V| \cdot |E| \cdot \log |V|)$
  (with Fibonacci Heap: $\mathcal{O}(|V|^2 \log |V| + |V| \cdot |E|)$)

Overal $\mathcal{O}(|V| \cdot |E| \cdot \log |V|)$
($\mathcal{O}(|V|^2 \log |V| + |V| \cdot |E|)$)