Algorithms and Data Structures Shortest Path, Dijkstra Algorithm

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Bioinformatics Group / Department of Computer Science Algorithms and Data Structures, March 2018

Structure



Graphs

Structure



Graphs

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For a graph G = (V, E):
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- A path of G is a sequence of edges $u_1, u_2, ..., u_i \in V$ with
 - Undirected graph: $\{u_1, u_2\}, \{u_2, u_3\}, \dots, \{u_{i-1}, u_i\} \in E$
 - Directed graph: $(u_1, u_2), (u_2, u_3), \dots, (u_{i-1}, u_i) \in E$
- The length of a path is
 - Without weights: number of edges taken
 - With weights: sum of weights of edges taken

For a graph G = (V, E):

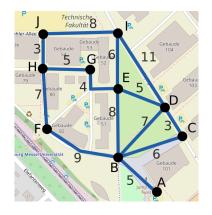
- The shortest path between two vertices u, v is the path P = (u, ..., v) with the shortest length d(u, v) or lowest costs
- The diameter of a graph is the longest shortest path

Structure



Graphs

- Wanted: Shortest path from M to all other points
- Place pearls on crossings and clamp strings between them



Shortest Path without Computer





Figure: Based on OpenStreetMaps; CC BY-SA 2.0





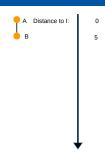
- Each node (pearl) now has a specific height
- The distance to M is exactly the shortest path

Shortest Path without Computer





Figure: Based on OpenStreetMaps; CC BY-SA 2.0





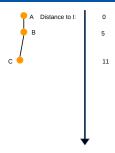
- Each node (pearl) now has a specific height
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Shortest Path without Computer





Figure: Based on OpenStreetMaps; CC BY-SA 2.0





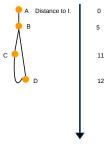
- Each node (pearl) now has a specific height
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Shortest Path without Computer



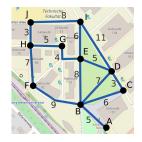


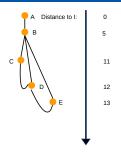
Figure: Based on OpenStreetMaps; CC BY-SA 2.0





- Each node (pearl) now has a specific height
- The distance to M is exactly the shortest path

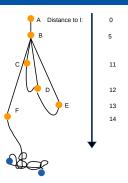






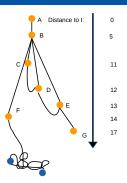
- Each node (pearl) now has a specific height
- The distance to M is exactly the shortest path



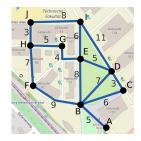


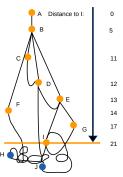
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- The distance to M is exactly the shortest path

Shortest Path



Figure: Shortest path from s to t

- Let r be the shortest path from s to t
- For each node u on path r the path from u to t is the shortest path

Proof:

- If there was a shorter path from s to u then we could choose this path to get faster to t
- Then *r* would not be the shortest path



Figure: Shortest path from *s* to *t*

- This is also correct for all sub paths on r
- If the shortest path from s to t passes u_1 and u_2 then the sub path (u_1, u_2) is the shortest path from u_1 to u_2

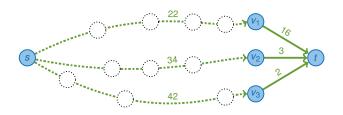


Figure: Shortest paths from s to t

If we know the shortest path form s to the preceding nodes of $t(v_1, v_2, v_3)$ we can determine the shortest path to t

Idea:

- Attach the cost of the shortest path to each node
- Let the information travel over the edges (message passing)
- In which order should we process the nodes?

Inventor:

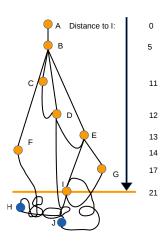
- Edsger Dijkstra (1930 2002)
- Computer scientist from Netherlands
- Won Turing-Award as one of few Europeans for his studies of structured programming
- Invented the Dijkstra-Algorithm in 1959



Figure: Portrait © Hamilton Richards - manuscripts of Edsger W. Dijkstra, University Texas at Austin

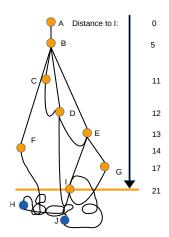
Example:

- Lift pearl A a little bit
- Connection to pearl B is hanging in the air
- Lift further until pearl B starts to lift at 5 m
- The shortest path to B is now known
- Lift further: The wires from C, D, E and F are now in the air
- One of the pearls C, D, E or F is the next one Which one?



Example:

- At 11 m pearl C gets lifted
- The wire to D is now in the air
- One of the pearls D, E and F is the next one
 Which one?
- At 12m pearl D gets lifted ...
- How to translate this into an computer algorithm?





High level description: Three types of nodes

Settled: For node u we know dist(s, u) (Pearl example: This pearl is hanging in the air)



Active: For node u we know a tentative distance $td(u) \ge dist(s, u)$ (Can be optimal but doesn't have to) (Pearl example: This pearl is laying on the table but one connected wire is already in the air)



Unreached: We have not reached the node yet (Pearl example: This preal is hanging in the air)



High level description:

- Each iteration take the active node u with the smallest td(u) (The pearl getting lifted next)
- We update the state of the node *u* to settled (The pearl gets lifted)
- We check for each neighbor v of node u if we can reach v faster than currently possible (Check all outgoing wires from this pearl: Activate all connected pearls, update tentative distance if smaller)
- Iterate until no active nodes exist anymore

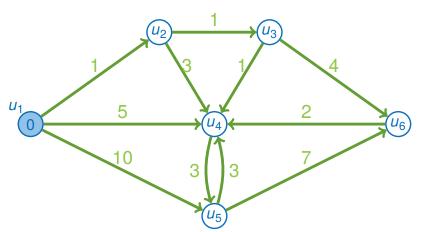


Figure: Start at u_1



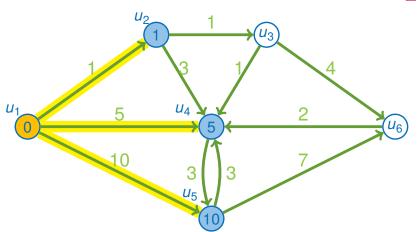


Figure: Iteration 1



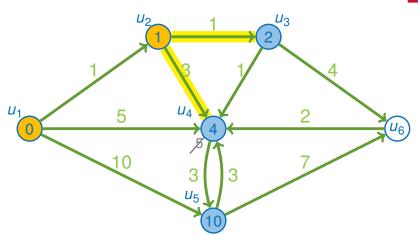


Figure: Iteration 2



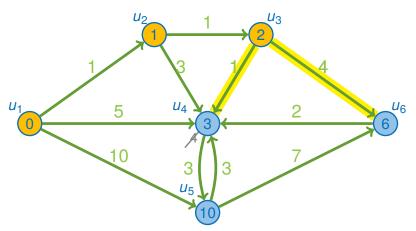


Figure: Iteration 3



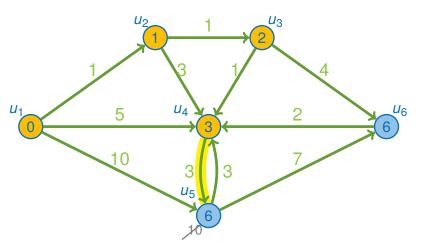


Figure: Iteration 4



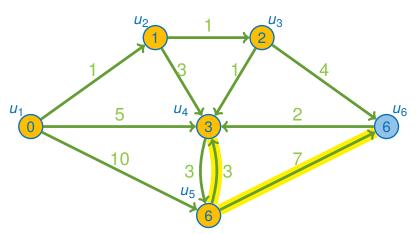


Figure: Iteration 5

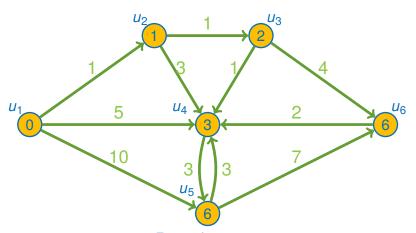


Figure: Iteration 6



- Assumption 1: All edges have a positive length
- **Assumption 2:** Each node has a unique distance dist(s, u) to the start s

(This was not the case on the previous slides)

This results in an easy and intuitive proof.

It is possible to show this without assumption 2. See references if interested

■ With assumption 2 there exists a sorting $u_1, u_2, ...$ with that:

$$\operatorname{dist}(s, u_1) < \operatorname{dist}(s, u_2) < \operatorname{dist}(s, u_3) < \dots$$

Proof:

With **assumption 2** there exists a sorting $u_1, u_2, ...$ with that:

$$\operatorname{dist}(s, u_1) < \operatorname{dist}(s, u_2) < \operatorname{dist}(s, u_3) < \dots$$

- We want to show that the *Dijkstra* algorithm finds the shortest path for each node u_i so that $td(u_i) = dist(s, u_i)$ holds
- Additionally we show that each node gets solved in order of the distance: Node u_i gets solved in iteration i

$$u_1, u_2, u_3, \dots$$

- **To show:** Node u_i gets solved in round i
 - Node u_i contains the correct distance $(td(u_i) = dist(s, u_i))$ and is active
 - Node u_i has the smallest value for $td(u_i)$ and gets selected by the algorithm

Induction start:

- Only the start node $s = u_1$ is active and td(s) = 0
 - Node u_1 gets solved and $td(u_1) = dist(s, u_1) = 0$
- 2 Only the start node u_1 is active

Induction step: i = i + 1

- **To show:** Node u_{i+1} contains the correct distance $(td(u_{i+1}) = dist(s, u_{i+1}))$ and is active
 - On the shortest path from s to u_{i+1} is a preceding node that:

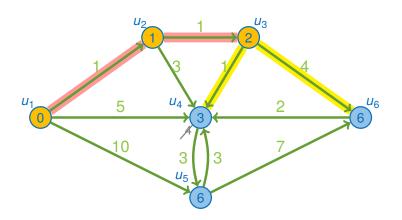
$$\operatorname{dist}(s, u_{i+1}) = \operatorname{dist}(s, v) + \operatorname{c}(v, u_{i+1})$$

 $(c(v, u_{i+1}))$ are the costs of the edge)



- Hence $\operatorname{dist}(s, v) < \operatorname{dist}(s, u_{i+1})$ because c > 0 (c=cost of edge)
- Because u_{i+1} is currently settled, the node v is one of the preceding nodes u_1, \ldots, u_i , hence $v = u_i$ with $0 \le i \le i$





- Preceding node of u_6 is $v = u_3$
- In round 3 $td(u_6) = 2 + 4 = 6$ was already solved



- **To show:** Node u_{i+1} contains the correct distance $td(u_{i+1}) = dist(s, u_{i+1})$ and is active
 - With **induction assumption**: v already contains the correct distance which was evaluated in round j (edge from v to u_{i+1}) and is stored in $td(u_{i+1})$
 - \blacksquare u_{i+1} is active because the preceding node was solved



- **To show:** Node u_{i+1} has the smallest value for $td(u_{i+1})$ and gets selected by the algorithm
 - All nodes with smaller dist are already solved
 - All other nodes u_k with k > i + 1 have a greater dist (s, u_k) and with that the $td(u_k)$ is greater or equal
 - $\Rightarrow u_{i+1}$ is the node with the smallest td and gets selected by the algorithm

Implementation:

Implementation

- We have to manage a set of active nodes
- We start with only the start node in our set
- At the start of each iteration we need the node u with the smallest td(u)

How to implement this?

Implementation:

- Using a priority queue with td(u) as keys
- The following problem occurs:
 - The tentative distance of an active node might change multiple times before it is settled
 - We have to change the key in our priority queue without removing the entry

Limitations:

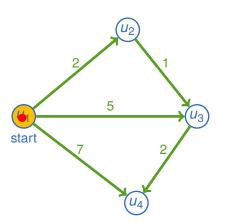
- Often only insert, getMin and deleteMin are implemented
- ⇒ We only have access to the first element and not any desired one

Alternative:

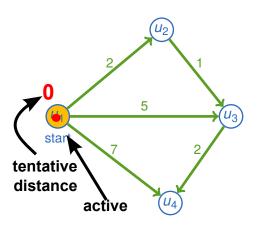
- If a node reoccurs with a smaller dist we insert the element one more time into the priority queue
 (We do nothing if the distance is greater or equal)
- We do not remove the old entry
- The node always gets solved with the smallest distance because of the smaller key
- If a settled node reoccurs with a higher dist we remove it and do simply nothing

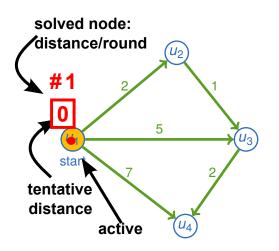
Implementation - Example

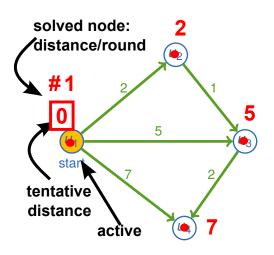




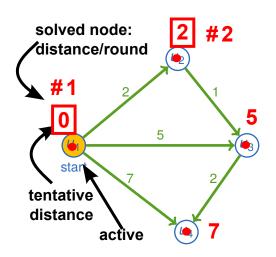






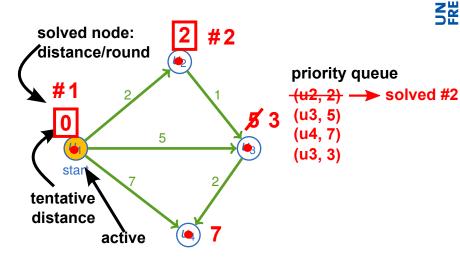


- (u2, 2)
- (u3, 5)
- (u4, 7)

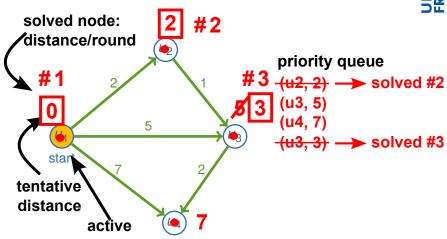


(u2, 2) → solved #2 (u3, 5)

(u4, 7)

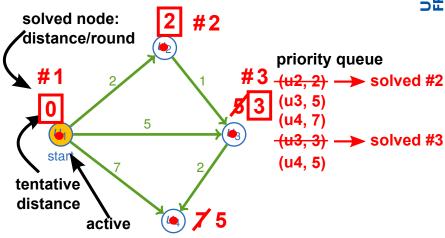


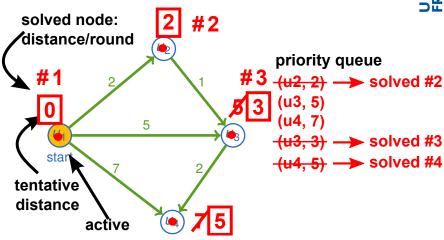


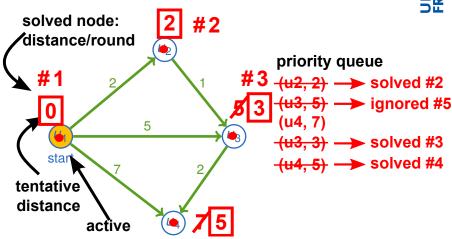




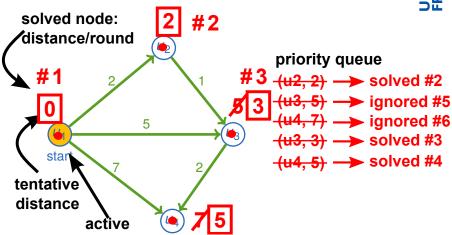
Implementation - Example











Runtime analysis

Graph with *n* nodes and *m* edges: $(m \ge n)$

- Each node gets solved exactly one time
- When solving a node it's outgoing edges are taken into account
- Each edge triggers at maximum one insert operation
- The number of operations on the priority queue is at maximum O(m)
- This results in a runtime of $O(m \cdot \log m)$ (log m because of at max. m elements in the priority queue)

Runtime of $O(m \cdot \log m)$:

- Because of $m \le n^2$ we have a maximum runtime of $O(m \cdot \log n)$, because $\log n^2 = 2 \log n$
- With a complex priority queue the runtime can be reduced to $O(m+n \log n)$
 - For example with a Fibonacci heap
 - This results in a better runtime for complex graphs $m \sim n^2$
 - Complex heaps create a management overhead
 - ⇒ In practice $m \in O(n)$ with a **binary heap** being faster (See lecture 6)

Termination criteria:

Terminate as soon as the target node t is settled ... never before because tentative distance might change:

$$td(t) \ge dist(s, t)$$

Before the node t is solved all nodes u with $dist(s, u) \leq dist(s, t)$ are settled

Termination criteria:

- Not only the single source single target shortest path problem is solved by the Dijkstra algorithm but also the single source all targets problem
- This sounds wasteful but there is not a (much) better method for general graphs
 Intuitive: We only know that there is no shorter path if all
 - nodes in the distance of dist(s, t) are evaluated

- With the current implementation of the Dijkstra algorithm we only get the length of the path How to get the path itself too?
- If we save the preceding node of the current shortest path on settling of each node we can reconstruct the path



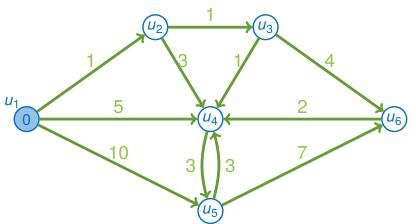


Figure: Start at u_1



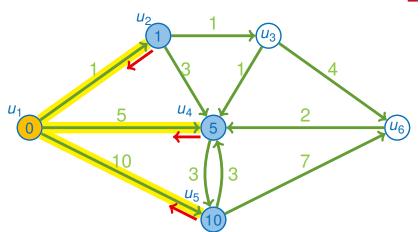


Figure: Iteration 1



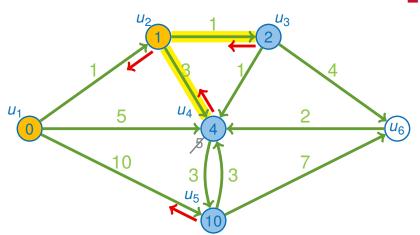


Figure: Iteration 2



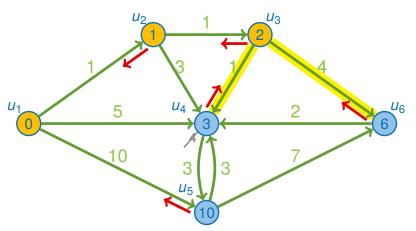


Figure: Iteration 3



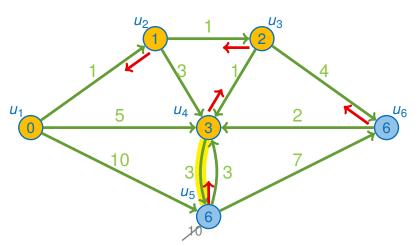


Figure: Iteration 4



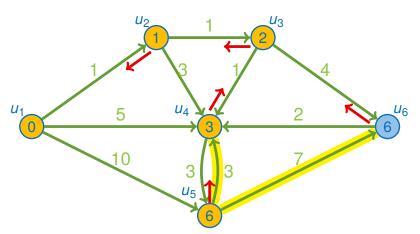


Figure: Iteration 5



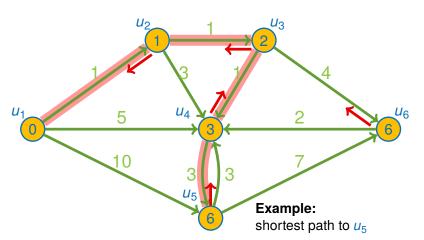


Figure: Iteration 6

Enhancement:

- In our proof we used the assumption that all costs are not negative (even > 0)
- With negative costs there might be negative cycles:

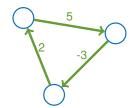


Figure: Here no problem ...

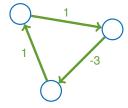
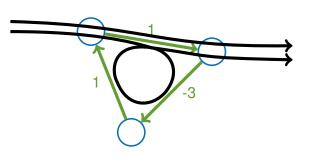


Figure: ... but here

Negative cycles:



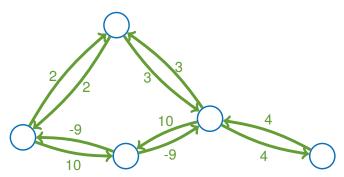
- No cycle: cost of 1
- 1 cycle:
 - cost of 0
- 2 cycles: cost of -1
- 3 cycles:
 - cost of -2
 -

Enhancement:

- We need a different algorithm to deal with negative edges
 - For example the **Bellman-Ford** algorithm
 - If the graph is acyclic we can simply use a topological sorting (with DFS) and settling the nodes in order of this sorting
- Another (not only) in artificial intelligence used variant of the Dijkstra algorithm is the A* algorithm Additional information given:

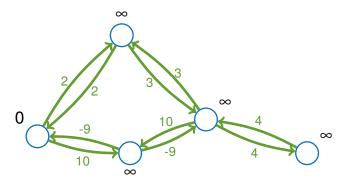
h(u) = estimated value for dist(u,t)

Message passing only from solved nodes



Bellman-Ford algorithm:

Message passing from all nodes until the path lengths are stable



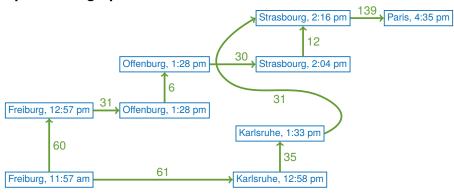
Application example:

- Route planner for car trips (exercise sheet)
- Route planner for bus / train connections What could the graph look like?

Application



Space-time graph:



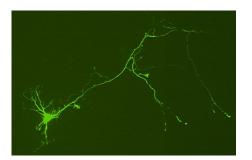


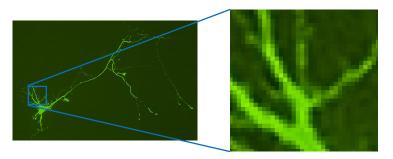


Figure: Neurons under fluorescence microscope

- Task: Measure length of axons (connections of neurons)
- Demo with ImageJ plugin NeuronJ http://www.imagescience.org/meijering/software/ neuronj/

Application: Trace axons





- Image as graph: Each pixel is a node
- Implicit edges: Each pixel has an edge to it's 8 neighbours (no need to save the edges)
- Costs for nodes (not edges): bright pixels are cheap, dark pixels are costly

■ General

- [CRL01] Thomas H. Cormen, Ronald L. Rivest, and Charles E. Leiserson. Introduction to Algorithms. MIT Press, Cambridge, Mass, 2001.
- [MS08] Kurt Mehlhorn and Peter Sanders.
 Algorithms and data structures, 2008.
 https://people.mpi-inf.mpg.de/~mehlhorn/
 ftp/Mehlhorn-Sanders-Toolbox.pdf.

■ Dijkstra's algorithm

[Wik] Dijkstra's algorithm

https:

//en.wikipedia.org/wiki/Dijkstra's_algorithm

- Shortest path problem
 - [Wik] Shortest path problem

https://en.wikipedia.org/wiki/Shortest_path_problem