

Algorithms and Datastructures

Shortest Path, Dijkstra Algorithm

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Graphs

Dijkstra Algorithm

For a graph $G = (V, E)$:

- A path of G is a sequence of edges $u_1, u_2, \dots, u_i \in V$ 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

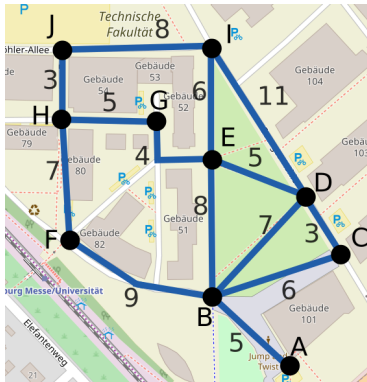
For a graph $G = (V, E)$:

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

Dijkstra Algorithm

Shortest Path without Computer

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



Dijkstra Algorithm

Shortest Path without Computer

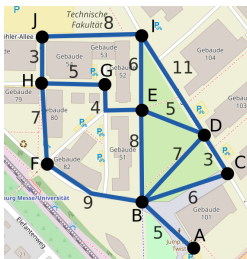
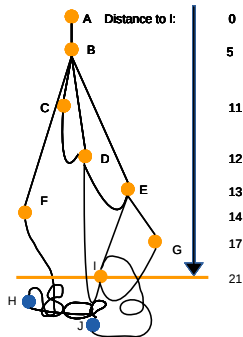


Figure: Based on
OpenStreetMaps; CC BY-SA 2.0

- Take the net and pull it slowly upwards until fully lifted



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

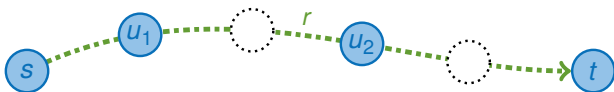


Figure: Shortest path from s to t

- 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

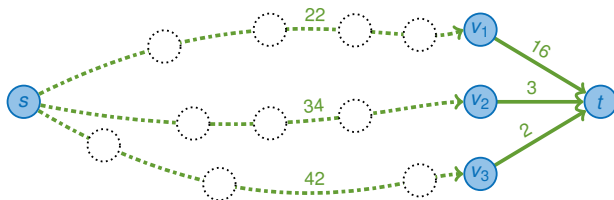


Figure: Shortest paths from s to t

- If we know the shortest path from 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



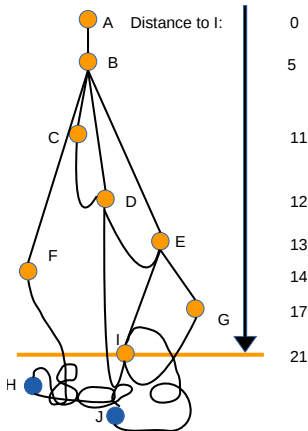
Figure: Portrait © Hamilton Richards - manuscripts of Edsger W. Dijkstra, University Texas at Austin

Dijkstra Algorithm



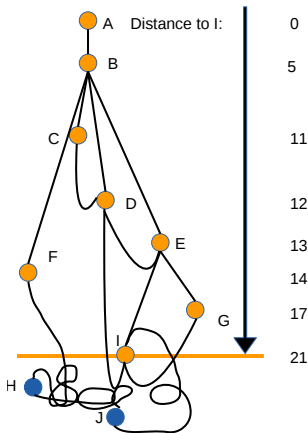
Example:

- Lift pearl **A** a little bit
 - Connection to pearl **B** is hanging in the air
 - Lift further until pearl **B** starts to lift at 5m
 - 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 $\text{dist}(s, u)$
(Pearl example: This pearl is hanging in the air) 
- **Active:** For node u we know a tentative distance $\text{td}(u) \geq \text{dist}(s, u)$ (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

Dijkstra Algorithm

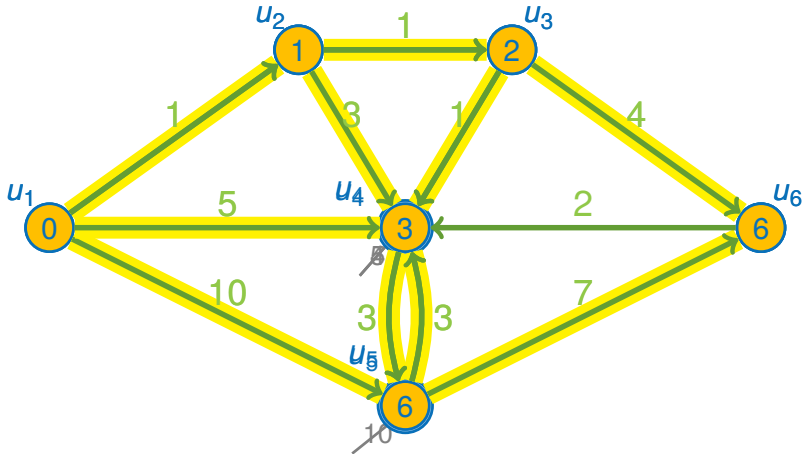


Figure: Start at u_1 Iteration 1 Iteration 2 Iteration 3 Iteration 4 Iteration 5 Iteration 6

Proof:

- **Assumption 1:** All edges have a positive length
- **Assumption 2:** Each node has a unique distance $\text{dist}(s, u)$ 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, \dots with that:

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

Proof:

- With **assumption 2** there exists a sorting u_1, u_2, \dots with that:

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

- We want to show that the *Dijkstra* algorithm finds the shortest path for each node u_i so that $\text{td}(u_i) = \text{dist}(s, u_i)$ 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

- 1 Node u_i contains the correct distance ($td(u_i) = dist(s, u_i)$) and is active
- 2 Node u_i has the smallest value for $td(u_i)$ and gets selected by the algorithm

Induction start:

- 1
 - 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: $i = i + 1$

- 1 **To show:** Node u_{i+1} contains the correct distance ($\text{td}(u_{i+1}) = \text{dist}(s, u_{i+1})$) and is active

- On the shortest path from s to u_{i+1} is a preceding node that:

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

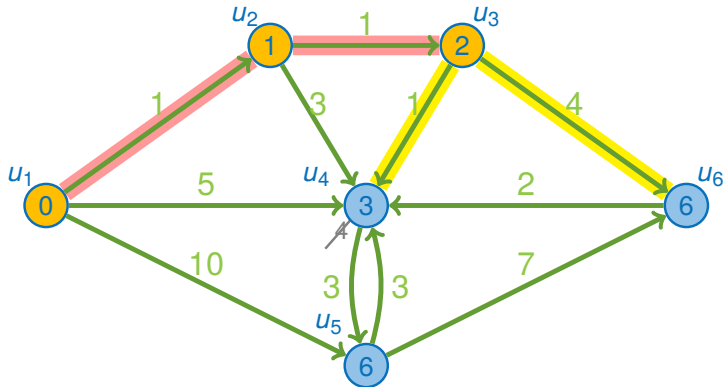
($c(v, u_{i+1})$ are the costs of the edge)



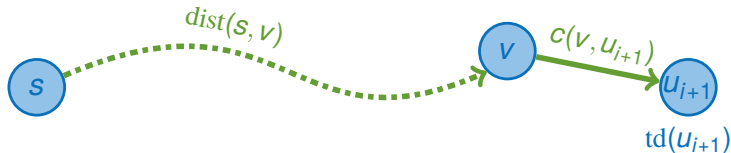
- Hence $\text{dist}(s, v) < \text{dist}(s, u_{i+1})$ because $c > 0$ (c =cost of edge)
- Because u_{i+1} is currently settled, the node v is one of the preceding nodes u_1, \dots, u_i , hence $v = u_j$ with $0 \leq j \leq i$

Dijkstra Algorithm

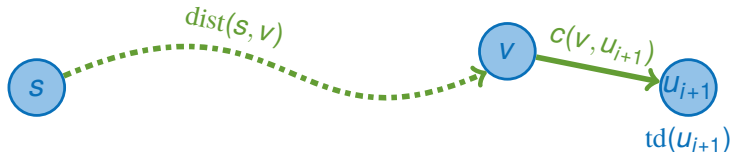
Proof - Example of Iteration 6



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



- 1 **To show:** Node u_i contains the correct distance $\text{td}(u_i) = \text{dist}(s, u_i)$ 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 $\text{td}(u_{i+1})$
 - u_{i+1} is active because the preceding node was solved



2 **To show:** Node u_{i+1} has the smallest value for $\text{td}(u_{i+1})$ 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 $\text{dist}(s, u_k)$ and with that the $\text{td}(u_k)$ is greater or equal
- $\Rightarrow u_{i+1}$ is the node with the smallest td and gets selected by the algorithm

Implementation:

- 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:

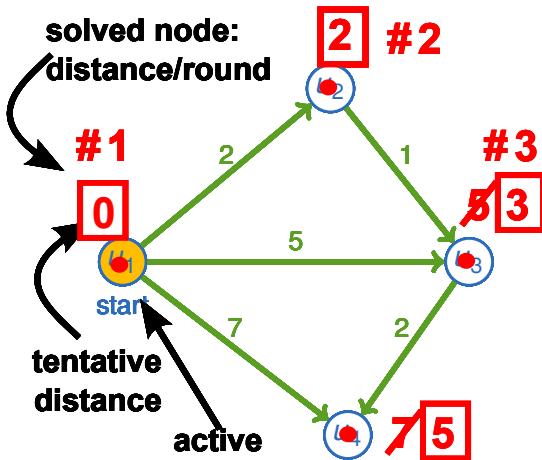
- Often only insert, getMin and deleteMin are implemented
- ⇒ We only have access to the first element and not any desired one

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**

Dijkstra Algorithm

Implementation - Example



priority queue

- (u2, 2) → solved #2
- (u3, 5) → ignored #5
- (u4, 7) → ignored #6
- (u3, 3) → solved #3
- (u4, 5) → solved #4

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

- Each node gets solved exactly **one time**
- When solving a node its 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 $m \leq n^2$ 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:

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

- 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 $\text{dist}(s, t)$ are evaluated

Calculate the shortest path:

- 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**

Dijkstra Algorithm

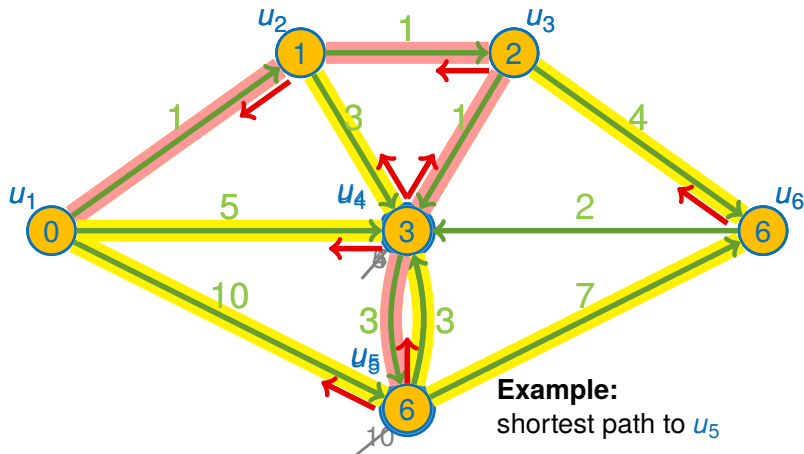


Figure: Start at u_1 Iteration 1 Iteration 2 Iteration 3 Iteration 4 Iteration 5 Iteration 6

Enhancement:

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

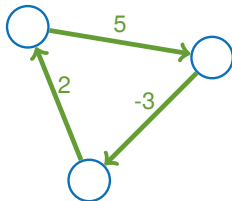


Figure: Here no problem ...

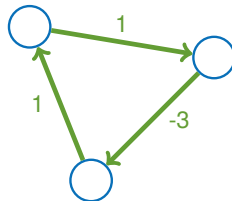
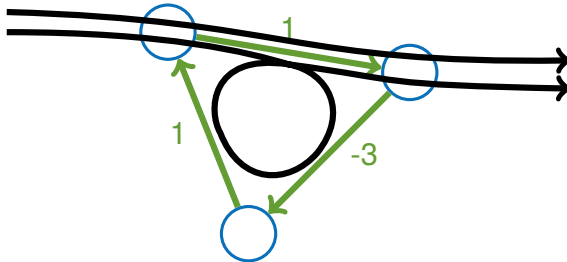


Figure: ... but here

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:

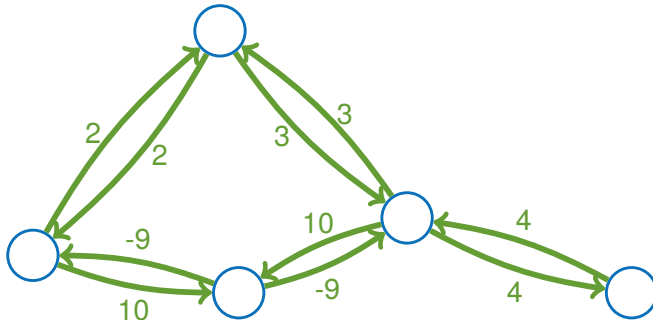
$h(u)$ = estimated value for $\text{dist}(u, t)$

Dijkstra Algorithm

Example - Negative costs (e-car consumption)

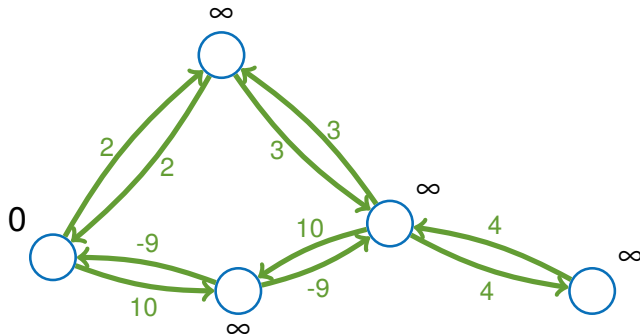
Dijkstra algorithm:

Message passing only from solved nodes



Bellman-Ford algorithm:

Message passing from all nodes until the path lengths are stable

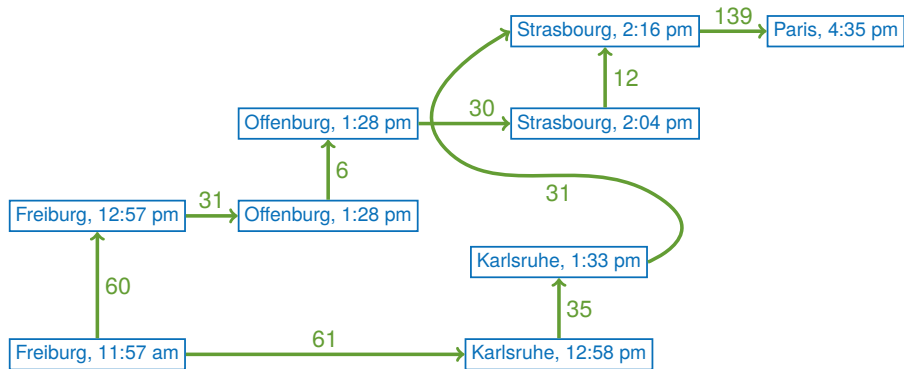


Application example:

- Route planner for car trips (exercise sheet)
- Route planner for bus / train connections

What could the graph look like?

Space-time graph:



Dijkstra Algorithm

Application in image processing



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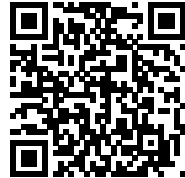
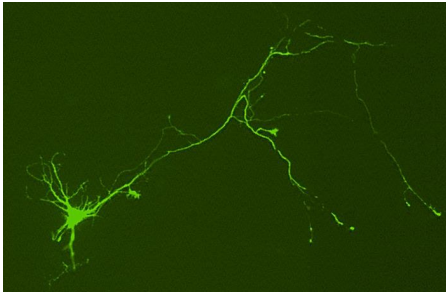
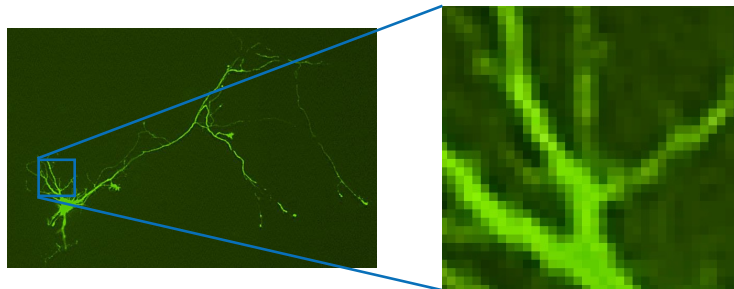


Figure: Neurons under fluorescence microscope

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

Dijkstra Algorithm

Application: Trace axons



- Image as graph: Each pixel is a node
- Implicit edges: Each pixel has an edge to its 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`