

Algorithms and Data Structures

C5. Shortest Paths: Foundations

Gabriele Röger and Patrick Schneider

University of Basel

May 13, 2026

1 / 26

Algorithms and Data Structures

May 13, 2026 — C5. Shortest Paths: Foundations

C5.1 Introduction

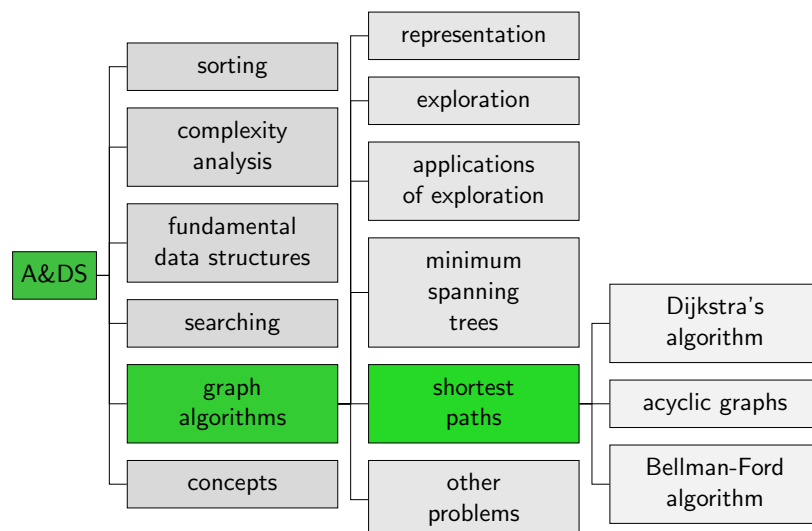
C5.2 Foundations

C5.3 Optimality Criterion and Generic Algorithm

C5.4 Summary

2 / 26

Content of the Course



3 / 26

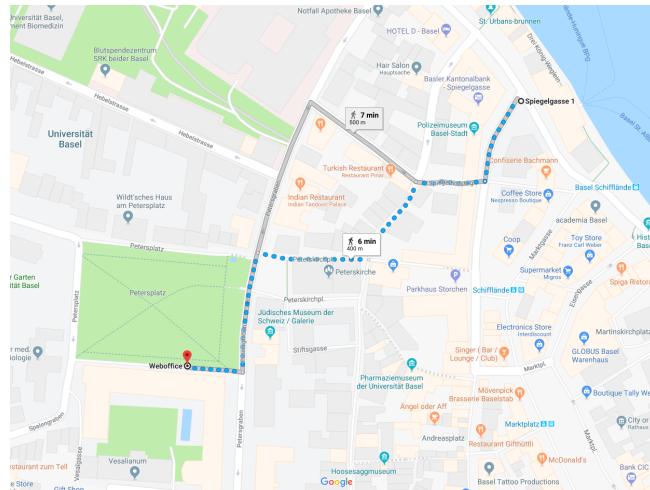
C5. Shortest Paths: Foundations

Introduction

C5.1 Introduction

4 / 26

Google Maps



5 / 26

Seam Carving



6 / 26

Applications

- ▶ Route planning
- ▶ Path planning in games
- ▶ robot navigation
- ▶ seam carving
- ▶ automated planning
- ▶ typesetting in TeX
- ▶ routing protocols in networks (OSPF, BGP, RIP)
- ▶ routing of telecommunication messages
- ▶ traffic routing

Source (partially): Network Flows: Theory, Algorithms, and Applications,
R. K. Ahuja, T. L. Magnanti, and J. B. Orlin, Prentice Hall, 1993

7 / 26

Variants

What are we interested in?

- ▶ **Single source:** from one vertex s to all other vertices
- ▶ **Single sink:** from all vertices to one vertex t
- ▶ **Source-sink:** from vertex s to vertex t
- ▶ **All pairs:** from every vertex to every vertex

Graph properties

- ▶ arbitrary / non-negative / Euclidean weights
- ▶ arbitrary / non-negative / no cycles

8 / 26

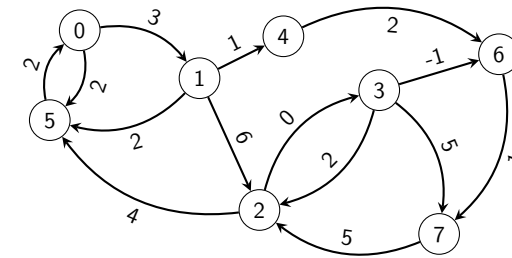
C5.2 Foundations

Weighted Directed Graphs

Same (high-level) definition of weighted graphs as before, but now we consider directed graphs.

Directed Graph

An (edge)-weighted graph associates every edge e with a weight (or cost) $weight(e) \in \mathbb{R}$.



Reminder: A directed graphs is also called a **digraph**.

API for Weighted Directed Edge

```

1 class DirectedEdge:
2     # Edge from n1 to n2 with weight w
3     def __init__(n1: int, n2: int, w: float) -> None
4
5     # Weight of the edge
6     def weight() -> float
7
8     # Initial vertex of the edge
9     def from_node() -> int
10
11    # Terminal vertex of the edge
12    def to_node() -> int

```

API for Weighted Digraphs

```

1 class EdgeWeightedDigraph:
2     # Graph with no_nodes vertices and no edges
3     def __init__(no_nodes: int) -> None
4
5     # Add weighted edge
6     def add_edge(e: DirectedEdge) -> None
7
8     # Number of vertices
9     def no_nodes() -> int
10
11    # Number of edges
12    def no_edges() -> int
13
14    # All outgoing edges of n
15    def outgoing_edges(n: int) -> Generator[DirectedEdge]
16
17    # All edges
18    def all_edges() -> Generator[DirectedEdge]

```

Shortest Path Problem

Single-source shortest path problem, SSSP

- ▶ Given: Graph and start vertex s
- ▶ Query for vertex v
 - ▶ Is there a path from s to v ?
 - ▶ If yes, what is the shortest path?
- ▶ In **weighted graphs**:
Shortest path is the one with **lowest weight**
 (= minimal sum of edge costs)

API for Shortest-path Implementation

The algorithms for shortest paths should implement the following interface:

```

1 class ShortestPaths:
2     # Initialization for start vertex s
3     def __init__(graph: EdgeWeightedDigraph, s: int) -> None
4
5     # Distance from s to v; infinity, if there is no path
6     def dist_to(v: int) -> float
7
8     # Is there a path from s to v?
9     def has_path_to(v: int) -> bool
10
11    # Path from s to v; None, if there is none
12    def path_to(v: int) -> Generator[DirectedEdge]

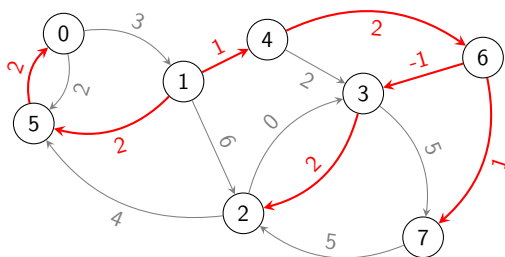
```

Shortest-path Tree

Shortest-path Tree

For a weighted digraph G and vertex s , a **shortest-path tree** is a subgraph that

- ▶ forms a directed tree with root s ,
- ▶ contains all vertices that are reachable from s , and
- ▶ for which every path in the tree is a shortest path in G .

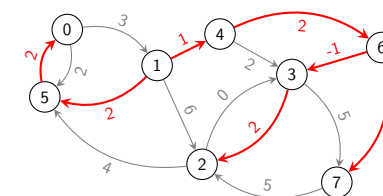


Shortest-path Tree: Representation

Representation: arrays indexed by vertex

- ▶ **parent** with reference to parent vertex
 None for unreachable vertices and start vertex
- ▶ **distance** with distance from the start vertex
 ∞ for unreachable vertices

	0	1	2	3	4	5	6	7
parent	5	None	3	6	1	1	4	6
distance	4	0	4	2	1	2	3	4



What about parallel edges?

Extracting Shortest Paths

```

1  def path_to(self, node):
2      if self.distance[node] == float('inf'):
3          yield None
4      elif self.parent[node] is None:
5          yield node
6      else:
7          # output path from start to parent node
8          self.path_to(self.parent[node])
9          # finish with node
10         yield node

```

This implementation generates a sequence of vertices. What do we have to change to generate a corresponding sequence of edges?

Relaxation

Relaxing edge (u, v)

- ▶ distance[u]: cost of the shortest **known** path to u
- ▶ distance[v]: cost of the shortest **known** path to v
- ▶ parent[v]: predecessor of v in the shortest known paths to v
- ▶ **Does edge (u, v) establish a shorter path to v (through u)?**
- ▶ If yes, update distance[v] and parent[v].

Illustration: Whiteboard

Relaxation

```

1  def relax(self, edge):
2      u = edge.from_node()
3      v = edge.to_node()
4      if self.distance[v] > self.distance[u] + edge.weight():
5          self.parent[v] = u
6          self.distance[v] = self.distance[u] + edge.weight()

```

C5.3 Optimality Criterion and Generic Algorithm

Optimality Criterion

Theorem

Let G be a weighted digraph without negative cycles. Array $distance[]$ contains the cost of the shortest paths from s if and only if

- 1 $distance[s] = 0$
- 2 $distance[w] \leq distance[v] + weight(e)$ for all edges $e = (v, w)$, and
- 3 for all vertices v , $distance[v]$ is the cost of *some* path from s to v , or ∞ if there is no such path.

Optimality Criterion (Continued)

Proof

“ \Rightarrow ”

- 1 Since the graph has no cycles of negative cost, no path from s to s can have negative cost. Thus, the empty path is optimal and $distance[s]$ is 0.
- 2 Consider an arbitrary edge e from u to v .
The shortest path from s to u has cost $distance[u]$. If we extend this path by edge e , we have a path from s to v of cost $distance[u] + weight(e)$. Thus, the cost of a shortest path from s to v cannot be larger and it holds that $distance[v] \leq distance[u] + weight(e)$.
- 3 Trivially true.

...

Optimality Criterion (Continued)

Proof (continued).

“ \Leftarrow ”

For unreachable vertices, the value is infinity by definition.

Consider an arbitrary vertex v and a shortest path $p = (v_0, \dots, v_n)$ from s to v , i.e. $v_0 = s$, $v_n = v$.

For $i \in \{1, \dots, n\}$, let e_i be a cheapest edge from v_{i-1} to v_i .

Since all inequalities are satisfied, we have

$$\begin{aligned} distance[v_n] &\leq distance[v_{n-1}] + weight(e_n) \\ &\leq distance[v_{n-2}] + weight(e_{n-1}) + weight(e_n) \\ &\leq \dots \leq weight(e_1) + \dots + weight(e_n) \\ &= \text{cost of an optimal path.} \end{aligned}$$

Due to 3, $distance[v_n]$ cannot be lower than the optimal cost. \square

Generic Algorithm

Generic Algorithm for Start Vertex s

- ▶ Initialize $distance[s] = 0$ and $distance[v] = \infty$ for all other vertices
- ▶ As long as the optimality criterion is not satisfied:
Relax an arbitrary edge

Correct:

- ▶ Finite $distance[v]$ always corresponds to the cost of a path from s to v .
- ▶ Every successful relaxation reduces $distance[v]$ for some v .
- ▶ For every vertex, the distance can only be reduced finitely often.

C5.4 Summary

Summary

- ▶ **Single-source shortest paths:** Compute in a weighted digraph the shortest paths from a given vertex to all reachable vertices.
- ▶ **Relaxation:** If for edge (u,v) the best known distance to v is larger than the one to u plus the edge cost, then update the distance to v (with predecessor u).
- ▶ **Generic algorithm**
 - ▶ Based on relaxation and optimality criterion.
 - ▶ Every instantiation is correct for all weighted digraphs **without negative-cost cycles**.
 - ▶ Specific instantiations: next chapter.