# Weighted graphs

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### Weighted graphs

- Graphs with numbers, called **weights**, attached to each edge
  - Often restricted to **non-negative**
- Directed or undirected
- Examples

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- **Distance** between cities
- **Cost** of flight between airports
- **Time** to send a message between routers

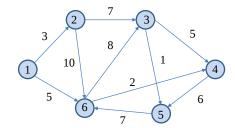
### Weighted graphs

• Adjacency matrix representation

$$T[i,j] = egin{cases} w_{i,j} & ext{if } i,j ext{ are connected} \ \infty & ext{if } i 
eq j ext{ are not connected} \ 0 & ext{if } i=j \end{cases}$$

• Similarly for adjacency lists

# Example weighted graph



### Example weighted graph

	1	2	3	4	5	6
1	0	3	$\infty$	∞	$\infty$	5
2	$\infty$	0	7	$\infty$	$\infty$	10
3	$\infty$	∞	0	5	1	∞
4	$\infty$	$\infty$	$\infty$	0	6	$\infty$
5	$\infty$	∞	$\infty$	∞	0	7
6	$\infty$	$\infty$	8	2	$\infty$	0

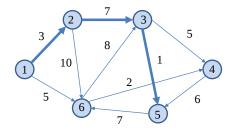
Adjacency matrix

### **Shortest paths**

- The **length** of a path is the **sum of the weights** of its edges
- Very common problem
  - find the **shortest path** from s to d
- Examples
  - Shortest route between cities
  - Cheapest connecting flight
  - Fastest network route
  - ...

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# Shortest path from vertex 1 to vertex 5



# Shortest path problem

Two main variants:

- Single source  $\boldsymbol{s}$ 
  - Find the shortest path from s to each node
  - **Dijkstra's** algorithm
    - Only for **non-negative** weights (important!)
- All-pairs
  - Find the shortest path between all pairs s,d
  - Floyd-Warshall algorithm
    - Any weights

### Dijkstra's algorithm

Main ideas:

- Keep a set  $\boldsymbol{W}$  of **visited** nodes
  - Start with  $W=\{s\}$  (or  $W=\{\}$ )
- Keep a matrix  $\Delta[u]$ 
  - Minimum distance from s to u passing only through W
  - Start with  $\Delta[u]=T[s,u]$  (or  $\Delta[s]=0,\Delta[u]=\infty$ )
- ullet At each step we **enlarge** W by adding a **new vertex**  $w
  ot\in W$ 
  - w is the one with  $oldsymbol{\mathsf{minumum}} \Delta[w]$

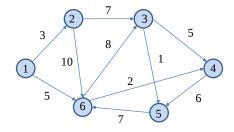
### Dijkstra's algorithm

Main ideas:

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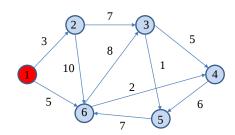
- Adding w to W might affect  $\Delta[u]$ 
  - For some **neighbour** u of w
- ullet We might now have a **shorter** path to u **passing through** w
  - Of the form  $s o \ldots o w o u$
  - If  $\Delta[u] > \Delta[w] + T[w,u]$
- In this case we update  $\Delta$ 
  - $\Delta[u] = \Delta[w] + T[w,u]$

### Example graph



# Expanding the vertex set w in stages

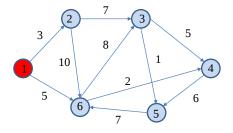
Stage	W	V-W	w	Δ(w)	Δ(1)	Δ(2)	Δ(3)	Δ(4)	Δ(5)	Δ(6)
Start	{1}	{2,3,4,5,6}	-	-	0	3	$\infty$	$\infty$	$\infty$	5



### Expanding the vertex set w in stages

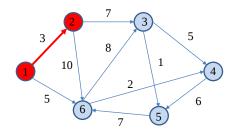
**W=2** is chosen for the second stage.

Stage	W	V-W	w	Δ(w)	Δ(1)	Δ(2)	Δ(3)	Δ(4)	Δ(5)	Δ(6)
Start	{1}	{2,3,4,5,6}	-	-	0	3	$\infty$	$\infty$	$\infty$	5



### Expanding the vertex set w in stages

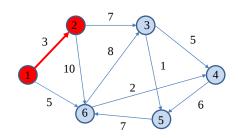
Stage	W	V-W	w	Δ(w)	Δ(1)	Δ(2)	Δ(3)	Δ(4)	Δ(5)	Δ(6)
Start	{1}	{2,3,4,5,6}	-	-	0	3	$\infty$	$\infty$	$\infty$	5
2	{1,2}	{3,4,5,6}	2	3	0	3	10	$\infty$	$\infty$	5



### Expanding the vertex set w in stages

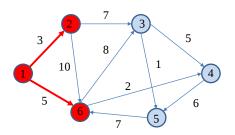
**W=6** is chosen for the third stage.

Stage	W	V-W	w	Δ(w)	Δ(1)	Δ(2)	Δ(3)	Δ(4)	Δ(5)	Δ(6)
Start	{1}	{2,3,4,5,6}	-	-	0	3	$\infty$	$\infty$	$\infty$	5
2	{1,2}	{3,4,5,6}	2	3	0	3	10	$\infty$	$\infty$	5



### Expanding the vertex set w in stages

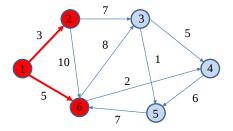
Stage	W	V-W	w	Δ(w)	Δ(1)	Δ(2)	Δ(3)	Δ(4)	Δ(5)	Δ(6)
Start	{1}	{2,3,4,5,6}	-	-	0	3	$\infty$	$\infty$	$\infty$	5
2	{1,2}	{3,4,5,6}	2	3	0	3	10	$\infty$	$\infty$	5
3	{1,2,6}	{3,4,5}	6	5	0	3	10	7	$\infty$	5



### Expanding the vertex set w in stages

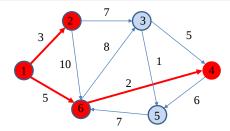
**W=4** is chosen for the fourth stage.

Stage	W	V-W	w	Δ(w)	Δ(1)	Δ(2)	Δ(3)	Δ(4)	Δ(5)	Δ(6)
Start	{1}	{2,3,4,5,6}	-	-	0	3	$\infty$	$\infty$	$\infty$	5
2	{1,2}	{3,4,5,6}	2	3	0	3	10	$\infty$	$\infty$	5
3	{1,2,6}	{3,4,5}	6	5	0	3	10	7	$\infty$	5



### Expanding the vertex set w in stages

Stage	W	V-W	w	Δ(w)	Δ(1)	Δ(2)	Δ(3)	Δ(4)	Δ(5)	Δ(6)
Start	{1}	{2,3,4,5,6}	-	-	0	3	$\infty$	$\infty$	$\infty$	5
2	{1,2}	{3,4,5,6}	2	3	0	3	10	$\infty$	$\infty$	5
3	{1,2,6}	{3,4,5}	6	5	0	3	10	7	$\infty$	5
4	{1,2,6,4}	{3,5}	4	7	0	3	10	7	13	5

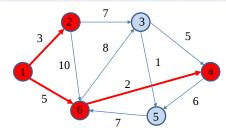


### Expanding the vertex set w in stages

**W=3** is chosen for the fifth stage.

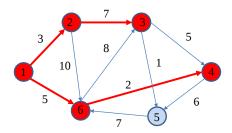
Stage	W	V-W	w	Δ(w)	Δ(1)	Δ(2)	Δ(3)	Δ(4)	Δ(5)	Δ(6)
Start	{1}	{2,3,4,5,6}	-	-	0	3	$\infty$	$\infty$	$\infty$	5
2	{1,2}	{3,4,5,6}	2	3	0	3	10	$\infty$	$\infty$	5
3	{1,2,6}	{3,4,5}	6	5	0	3	10	7	$\infty$	5
4	{1,2,6,4}	{3,5}	4	7	0	3	10	7	13	5

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# Expanding the vertex set w in stages

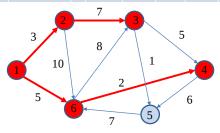
Stage	W	V-W	w	Δ(w)	Δ(1)	Δ(2)	Δ(3)	Δ(4)	Δ(5)	Δ(6)
Start	{1}	{2,3,4,5,6}	-	-	0	3	$\infty$	$\infty$	$\infty$	5
2	{1,2}	{3,4,5,6}	2	3	0	3	10	$\infty$	$\infty$	5
3	{1,2,6}	{3,4,5}	6	5	0	3	10	7	$\infty$	5
4	{1,2,6,4}	{3,5}	4	7	0	3	10	7	13	5
5	{1,2,6,4,3}	{5}	3	10	0	3	10	7	11	5



### Expanding the vertex set w in stages

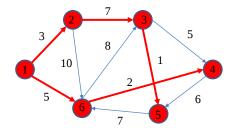
W=5 is chosen for the sixth stage.

Stage	W	V-W	w	Δ(w)	Δ(1)	Δ(2)	Δ(3)	Δ(4)	Δ(5)	Δ(6)
Start	{1}	{2,3,4,5,6}	-	-	0	3	$\infty$	$\infty$	$\infty$	5
2	{1,2}	{3,4,5,6}	2	3	0	3	10	$\infty$	$\infty$	5
3	{1,2,6}	{3,4,5}	6	5	0	3	10	7	$\infty$	5
4	{1,2,6,4}	{3,5}	4	7	0	3	10	7	13	5
5	{1,2,6,4,3}	{5}	3	10	0	3	10	7	11	5



### Expanding the vertex set w in stages

Stage	W	V-W	w	Δ(w)	Δ(1)	Δ(2)	Δ(3)	Δ(4)	Δ(5)	Δ(6)
Start	{1}	{2,3,4,5,6}	-	-	0	3	$\infty$	$\infty$	$\infty$	5
2	{1,2}	{3,4,5,6}	2	3	0	3	10	$\infty$	$\infty$	5
3	{1,2,6}	{3,4,5}	6	5	0	3	10	7	$\infty$	5
4	{1,2,6,4}	{3,5}	4	7	0	3	10	7	13	5
5	{1,2,6,4,3}	{5}	3	10	0	3	10	7	11	5
6	{1,2,6,4,3,5}	{}	5	11	0	3	10	7	11	5



### Dijkstra's algorithm in pseudocode

```
// Δεδομένα
src : αρχικός κόμβος

dest : τελικός κόμβος

// Πληροφορίες που κρατάμε για κάθε κόμβο ν

W[u] : 1 αν ο u είναι στο σύνολο W, 0 διαφορετικά

dist[u] : ο πίνακας Δ

prev[u] : ο προηγούμενος του ν στο βέλτιστο μονοπάτι

// Αρχικοποίηση: W={} (εναλλακτικά μπορούμε και W={src})

for each vertex u in Graph
   dist[u] = INT_MAX // infinity
   prev[u] = NULL
   W[u] = 0

dist[src] = 0
```

#### Dijkstra's algorithm in pseudocode

```
// Κυρίως αλγόριθμος
while true
    w = vertex with minimum dist[w], among those with W[w] = 0
    W[w] = 1
    if w == dest
        // optimal cost = dist[dest]
        // optimal path = dest <- prev[dest] <- ... <- src (inverse)</pre>
    for each neighbor u of w
        if W[u] == 1
            continue
        alt = dist[w] + weight(w, u)
                                         // κόστος του src -> ... -> w
        if alt < dist[u]</pre>
                                         // καλύτερο από πριν, update
            dist[u] = alt
            prev[u] = w
```

### Using a priority queue

- Finding the  $w \not\in W$  with **minumum**  $\Delta[w]$  is slow
  - O(n) time
- But we can use a **priority queue** for this!
  - We only keep vertices  $w \not\in W$  in the queue
  - They are compared based on their  $\Delta[w]$  (each w has "priority"  $\Delta[w]$ )
- Careful when  $\Delta[w]$  is modified!
  - Either use a priority queue that allows **updates**
  - Or insert multiple copies of each w with  ${f different\ priorities}$ 
    - $\circ$  the queue might contain **already visited** vertices: ignore them

#### Dijkstra's algorithm with priority queue

```
// Δεδομένα
src : αρχικός κόμβος

dest : τελικός κόμβος

// Πληροφορίες που κρατάμε για κάθε κόμβο u

W[u] : 1 αν ο ν είναι στο σύνολο W, Θ διαφορετικά

dist[u] : ο πίνακας Δ

prev[u] : ο προηγούμενος του ν στο βέλτιστο μονοπάτι

pq : Priority queue, εισάγουμε tuples {u, dist[u]}

συγκρίνονται με βάση το dist[u]

// Αρχικοποίηση: W={} (εναλλακτικά μπορούμε και W={src})

prev[src] = NULL

dist[src] = Θ

pqueue_insert(pq, {src,θ}) // dist[src] = Θ
```

#### Dijkstra's algorithm with priority queue

```
// Κυρίως αλγόριθμος
while pg is not empty
    w = pqueue_max(pq) // w with minimal dist[u]
    pqueue_remove_max(pq)
                        // το w μπορεί να υπάρχει πολλές φορές στην ο
    if exists(W[w])
                        // δεν κάνουμε replace), και να είναι ήδη vis
        continue
    W[w] = 1
    if w == dest
                        // optimal cost/path same as before
        stop
    for each neighbor u of w
        if exists(W[u])
            continue
       alt = dist[w] + weight(w,u) // cost \ of \ src -> ... -> w -> u
       if !exists(dist[u]) OR alt < dist[u]</pre>
            dist[u] = alt
            prev[u] = w
            pqueue_insert(pq, {u,alt}) // προαιρετικά: replace αν υπ
stop // pg άδειασε πριν βρούμε το dest => δεν υπάρχει μονοπάτι
```

### **Notation**

ullet s 
ightarrow d

- Direct step step from s to d
- $s \stackrel{W}{\longrightarrow} d$ 
  - Multiple steps  $s 
    ightarrow \ldots 
    ightarrow d$
  - All intermediate steps belong to the set  $W\subseteq V$
- $s \stackrel{W}{\Longrightarrow} d$ 
  - $^{ ext{-}}$  Shortest path among all  $s \overset{W}{\longrightarrow} d$
  - So  $s \stackrel{V}{\Longrightarrow} d$  is the overall shortest one

#### **Proof of correctness**

- We need to prove that  $\Delta[u]$  is the **minimum distance to** u
  - **after** the algorithm finishes
- But it's much easier to reason **step by step** 
  - we need a property that holds at every step
  - this is called an **invariant** (property that never changes)

#### **Proof of correctness**

#### Invariant of Dijkstra's algorithm

- $\Delta[u]$  is the cost of the shortest path **passing only through** W
- And the shortest **overall** when  $u \in W$

#### Formally:

- 1. For all  $u \in V$  the path  $s \stackrel{W}{\Longrightarrow} u$  has  $\cosh \Delta[u]$
- 2. For all  $u \in W$  the path  $s \stackrel{V}{\Longrightarrow} u$  has  $\mathrm{cost}\,\Delta[u]$

Proof: **induction** on the **size of** W, for both (1), (2) together

#### **Proof of correctness**

Base case  $W=\{s\}$ 

- ullet Trivial, the only path  $s \stackrel{W}{\longrightarrow} u$  is the direct one  $s \rightarrow u$
- For (1): its cost is exactly  $T[s,u]=\Delta[u]$ 
  - initial value of  $\Delta[u]$
- For (2): the only  $u \in W$  is s itself

#### **Proof of correctness**

Inductive case

- Assume |W|=k and (1),(2) hold
- The algorithm
  - Updates W , adding a new vertex w 
    otin W
  - Updates  $\Delta[u]$  for all neighbours u of w
- Let  $W', \Delta'$  be the values **after** the update
- Show that (1),(2) still hold for  $W',\Delta'$

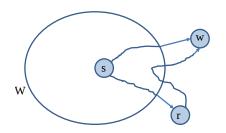
### **Proof of correctness**

We start showing that (2) still holds for  $W', \Delta'$ 

- ullet The interesting vertex is the w we just added
  - Vertices u 
    eq w are trivial from the induction hypothesis
- Show:  $s \stackrel{V}{\Longrightarrow} w$  has cost  $\Delta'[w]$ 
  - Note:  $\Delta'[w] = \Delta[w]$  (we do not update  $\Delta[w]$ )
  - $^{ ext{-}}$  We already know that  $s \stackrel{W}{\Longrightarrow} w$  has cost  $\Delta[w]$  (ind. hyp)
  - So we just need to prove that there is **no better** path **outside** W

#### **Proof of correctness**

- ullet Assuming such path exists, let r be its **first** vertex outside W
  - $^ ext{-}$  So the path  $s \stackrel{W}{\Longrightarrow} r \stackrel{V}{\Longrightarrow} w$  has cost  $c < \Delta[w]$
  - So the path  $s \stackrel{W}{\Longrightarrow} r$  has cost at most  $c < \Delta[w]$  (no negative weights!)
  - So  $\Delta[r] < \Delta[w]$
- Impossible! We chose w to be the one with min  $\Delta[w]$



### **Proof of correctness**

It remains to show (1) for  $W', \Delta'$ 

- ullet Take some arbitrary u
  - Let c be the cost of  $s \stackrel{W'}{\Longrightarrow} u$
  - Show:  $c=\Delta'[u]$
- $^{ullet}$  Three cases for the optimal path  $s \stackrel{W'}{\Longrightarrow} u$
- ullet Case 1: the path does not pass through w
  - So it is of the form  $s \stackrel{W}{\Longrightarrow} u$
  - This path has cost  $\Delta[u]$  (induction hypothesis)
  - No update:  $\Delta$  ' $[u] = \Delta[u] = c$

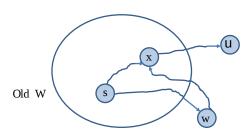
### **Proof of correctness**

• Case 2: w is right before u

- $^{ ext{-}}$  So the path is of the form  $s \overset{W}{\Longrightarrow} w o u$
- $^{ ext{-}}$  The cost of  $s \overset{W}{\Longrightarrow} w$  is  $\Delta[w]$  (induction hypothesis)
- So  $c=\Delta[w]+T[w,u]$
- So the algorithm will set  $\Delta$  '  $[u]=\Delta[w]+T[w,u]$  when updating the neighbours of w
- So  $c=\Delta '[u]$

#### **Proof of correctness**

- Case 3: some other x appears after w in the path
  - $^{ ext{-}}$  So the path is of the form  $s \stackrel{W}{\Longrightarrow} w o x \stackrel{W}{\Longrightarrow} u$
  - But the path  $s \stackrel{W}{\Longrightarrow} w \to x$  is no shorter than  $s \stackrel{W}{\Longrightarrow} x$ 
    - $\circ$  From the induction hypothesis for  $x \in W$
  - So  $s \stackrel{W}{\Longrightarrow} x \to u$  is also optimal, reducing to case 1!



## **Complexity**

Without a priority queue:

- Initialization stage: loop over vectices: O(n)
- ullet The while-loop adds one vertex every time: n iterations
- Finding the new vertex loops over vertices: O(n)
  - same for updating the neighbours
- So total  $O(n^2)$  time

### **Complexity**

With a priority queue:

- Initialization stage: loop over vectices, so O(n)
- Count the number of **updates** (steps in the **inner** loop)
  - Once for every neighbour of every node: e total
  - Each update is  $O(\log n)$  (at most n elements in the queue)
- Total  $O(e \log n)$ 
  - Assuming a connected graph  $(e \geq n)$
  - And an implementation using adjacency lists
- Only good for **sparse** graphs!
  - But  $O(n\log n)$  can be hugely better than  $O(n^2)$

### The all-pairs shortest path problem

- Find the shortest path between all pairs s,d
- Floyd-Warshall algorithm
- Any weights

- Even negative
- But no **negative loops** (why?)

### The all-pairs shortest path problem

Main idea

- At each step we compute the shortest path through a **subset of vertices** 
  - Similarly to W in Dijkstra
  - But now the set at step k is  $W_k = \{1, \dots, k\}$ 
    - Vectices are numbered in any order
- ullet Step k: the cost of  $i \stackrel{W_k}{\Longrightarrow} j$  is  $A_k[i,j]$ 
  - Similar to  $\Delta$  in Dijstra (but for all **pairs** i,j of nodes)

### Floyd-Warshall algorithm

- The algorithm at each step computes  $A_k$  from  $A_{k-1}$
- Initial step k=0

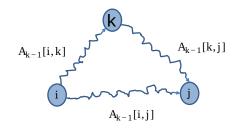
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- Start with  $A_0[i,j] = T[i,j]$
- Only direct paths are allowed

### Floyd-Warshall algorithm

k-th iteration: the optimal  $i \stackrel{W_k}{\Longrightarrow} j$  either **passes thorugh** k or not.

$$A_k[i,j] = \min egin{cases} A_{k-1}[i,j] \ A_{k-1}[i,k] + A_{k-1}[k,j] \end{cases}$$



### Floyd-Warshall algorithm in pseudocode

A is the **current**  $A_k$  at every step k.

# **Complexity**

- ullet Three simple loops of n steps
- So  $O(n^3)$
- Not better than n executions of Dijkstra in complexity
  - But much simpler
  - Often faster in practice
  - And works for **negative** weights

## Readings

- T. A. Standish. *Data Structures* , *Algorithms and Software Principles in C.* Chapter 10
- A. V. Aho, J. E. Hopcroft and J. D. Ullman. *Data Structures and Algorithms*. Chapters 6 and 7