



Algorithms and Data Structures

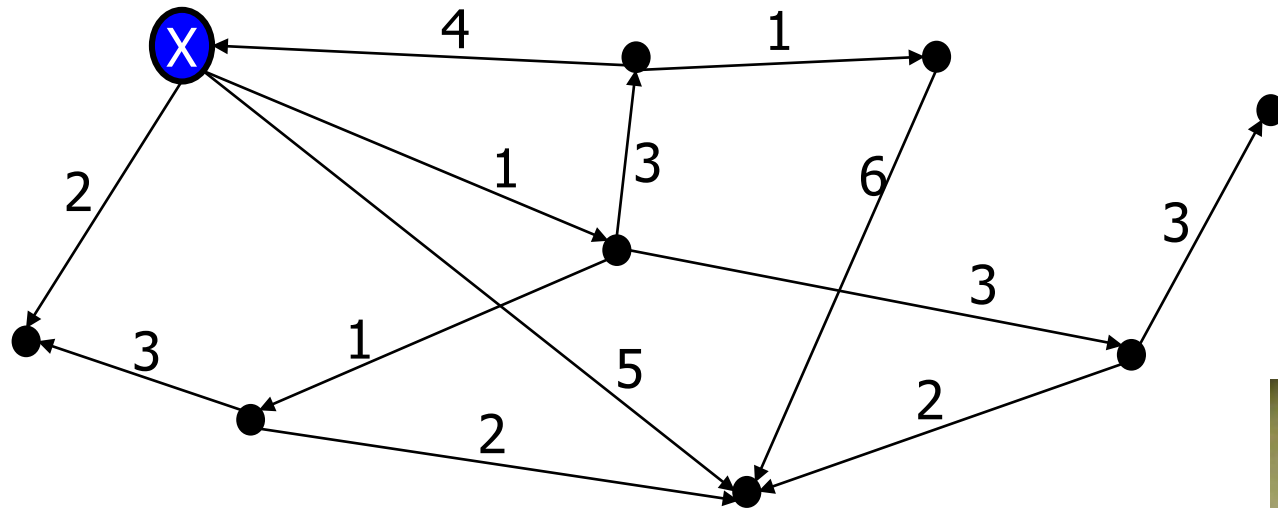
Graphs: Single-Source Shortest Paths

Ulf Leser

Content of this Lecture

- Shortest Paths
 - Single-Source-Shortest-Paths: Dijkstra's Algorithm
 - Shortest Path between two given nodes
 - Other

Shortest Paths in a Graph



- Task: Find the **distance between X** and **all other nodes**
 - Classical problem: Single-Source-Shortest-Paths
 - Famous solution: **Dijkstra's algorithm**
 - E. Dijkstra: A Note on Two Problems in Connexion with Graphs. Numerische Mathematik 1 (1959), S. 269–271



Computer Science is no more about
computers than astronomy is about
telescopes.

Attributed to Edsger Dijkstra, 1970.

Distance in Graphs

- Definition

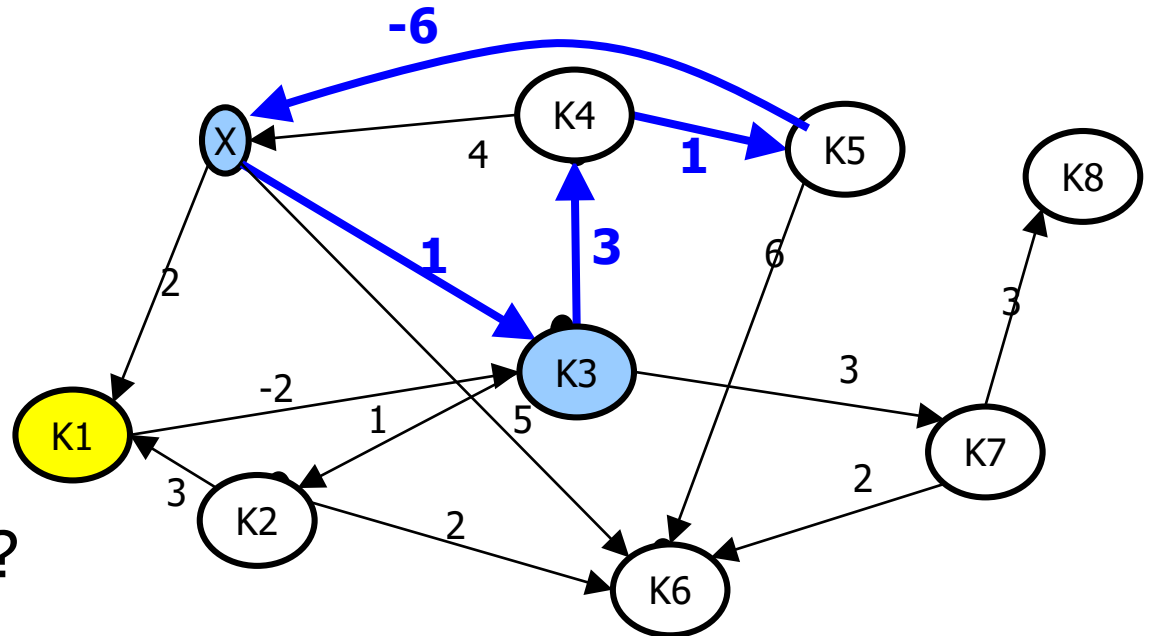
*Let $G=(V, E)$ be a graph. The **distance $d(u,v)$** between any two nodes $u,v \in V$ for $u \neq v$ is defined as*

- *G unweighted: The length of the **shortest path** from u to v , or ∞ if no path from u to v exists*
- *G weighted: The **minimal aggregated edge weight of all non-cyclic paths** from u to v , or ∞ if no path from u to v exists*
- *If $u=v$, $d(u,v)=0$*

- Remark

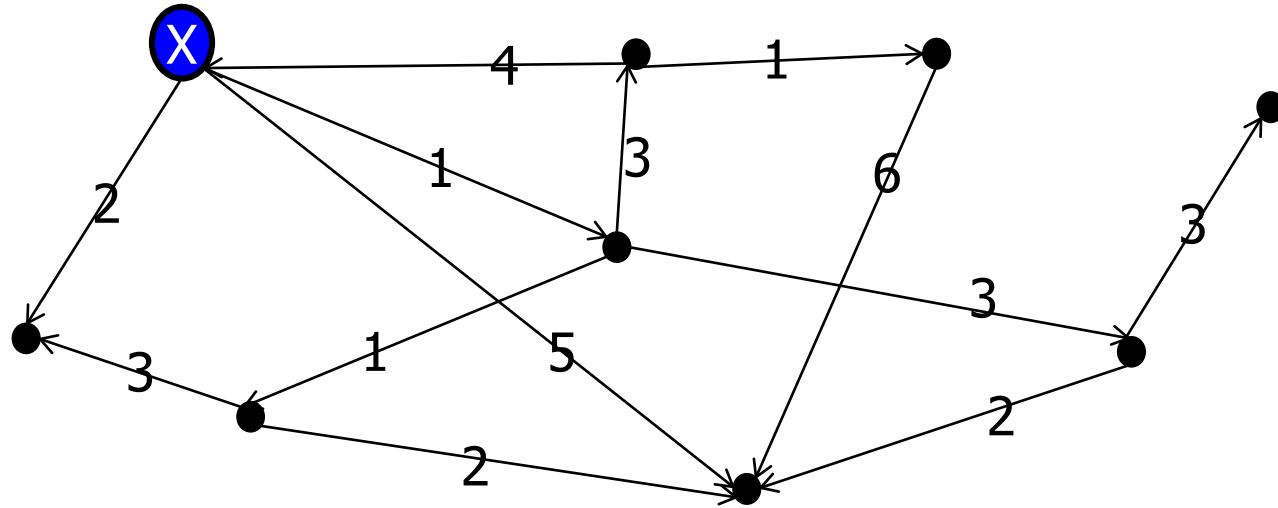
- Distance in unweighted graphs is the same as distance in weighted graphs with unit cost
- Beware of **negative cycles** in directed graphs

Negative Cycles



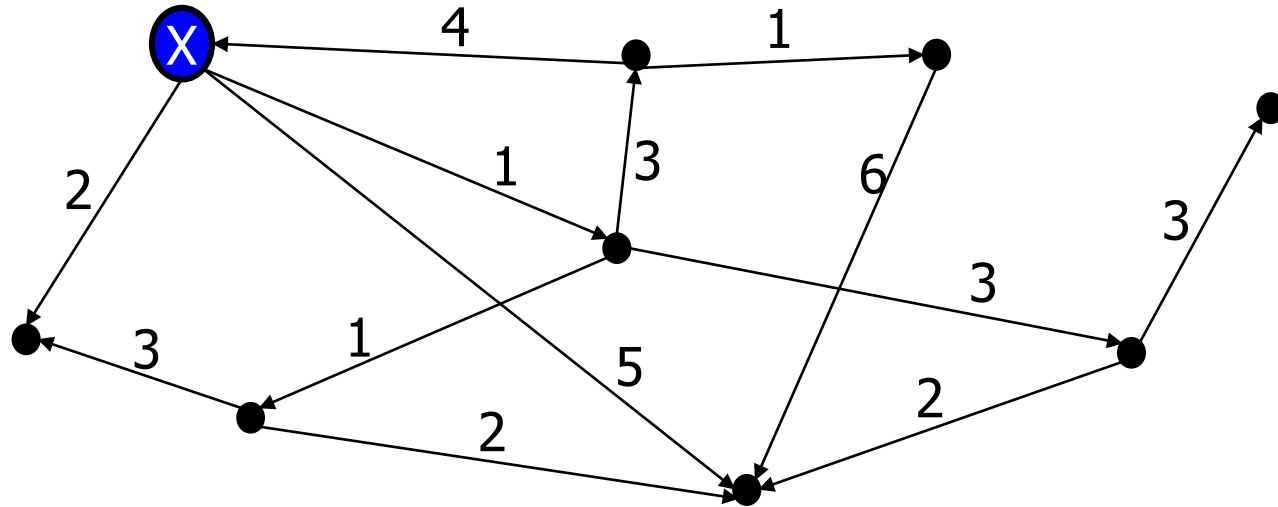
- Shortest path between X and K5?
 - X-K3-K4-K5: 5
 - X-K3-K4-K5-X-K3-K4-K5: 4
 - X-K3-K4-K5-X-K3-K4-K5-X-K3-K4-K5: 3
 - ...
- SP-Problem undefined if G contains a **negative cycle**

Single-Source Shortest Paths in a Graph



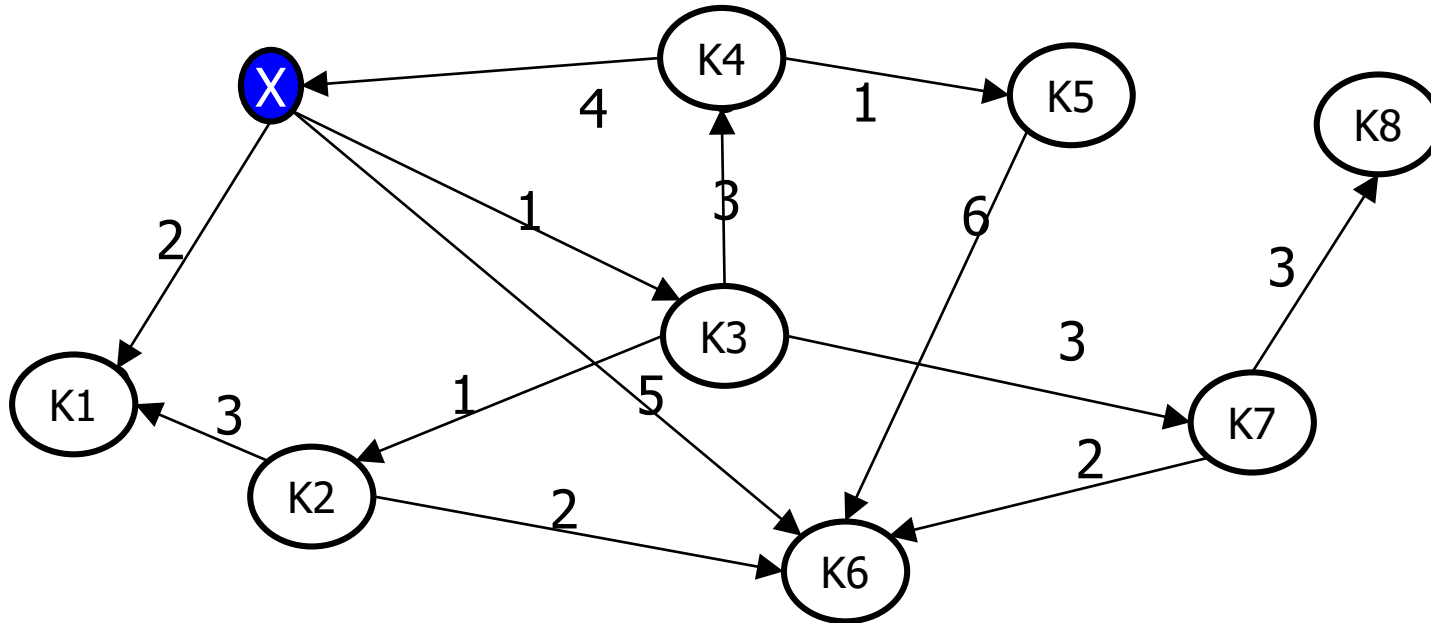
- Task: Find the **distance between X** and **all other nodes**
- Only **positive edge weights** allowed
 - Bellman-Ford algorithm solves the general case
- **Floyd-Warshall** finds distances between any pair of nodes

Assumptions



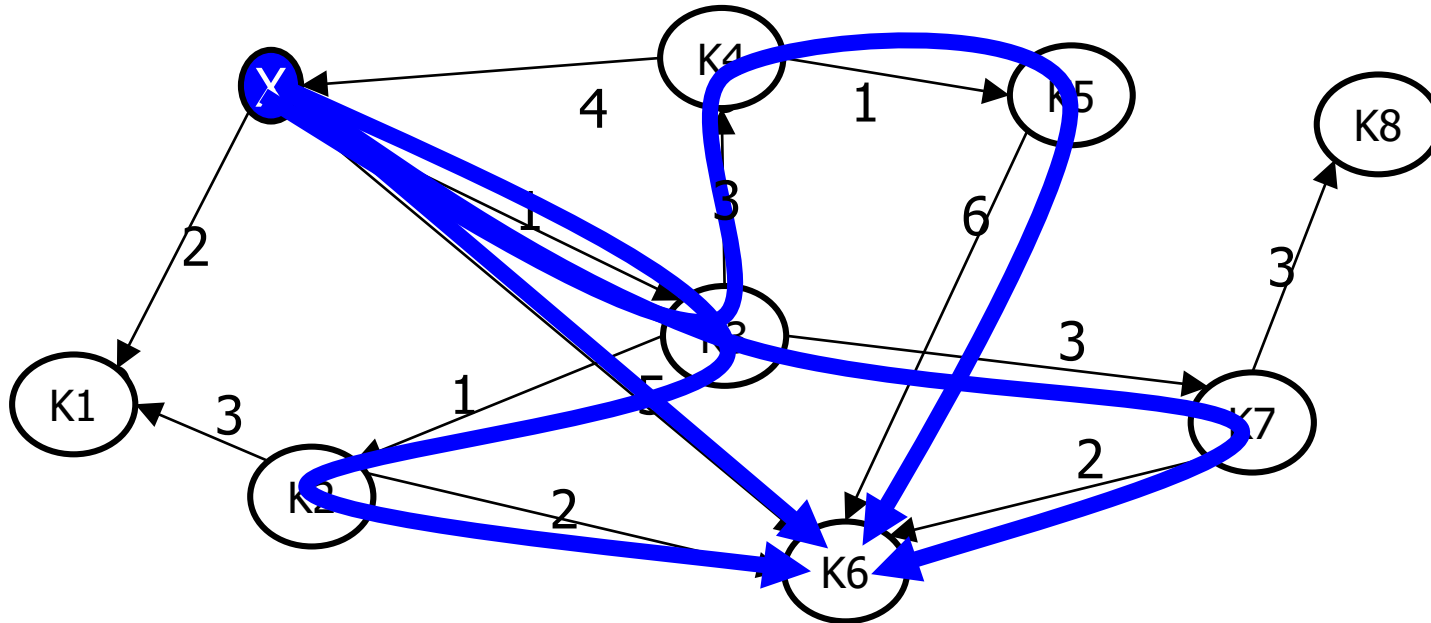
- We assume that every node is reachable from X
- There might be many shortest paths to node Y, but distance is unique
 - We only want the distances and need no “witness paths”
- Only positive edge weights
 - Whenever we extend a path with an edge, its length increases
 - Thus, no shortest path may contain a cycle

Exhaustive Solution



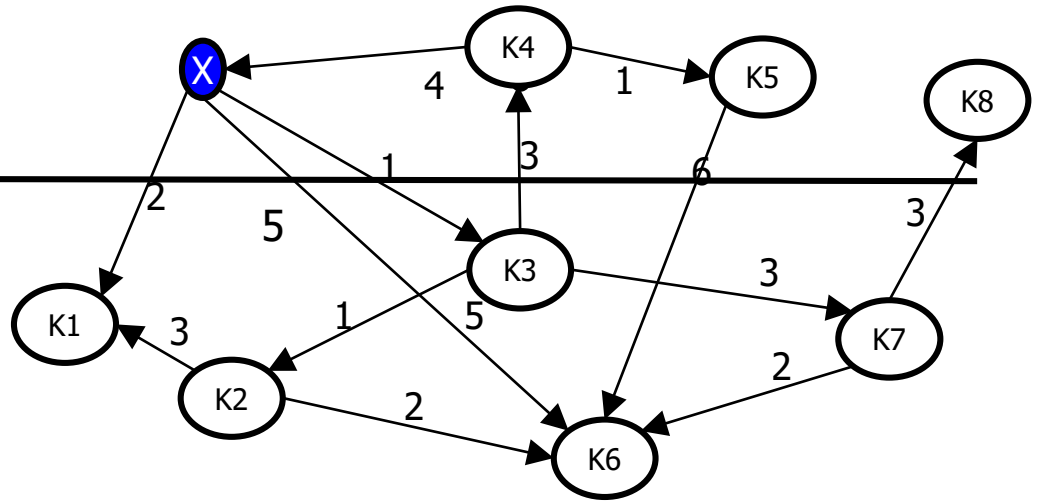
- First approach: **Enumerate all paths** ("BT": Backtrack)
 - Still need to **break cycles** (e.g. $X - K3 - K4 - X - K3 - \dots$)
 - Using DFS: $X - K3 - K4 - X$ [BT-K4] $- K5 - K6$ [BT-K5] [BT-K4] [BT-K3] $- K7 - K8$ [BT-K7] $- K6$ [BT-K7] [BT-K3] $- K2 - K6$ [BT-K2] $- K1$ [BT-K2] [BT-K3] [BT-X] $K6 - \dots$

Redundant work



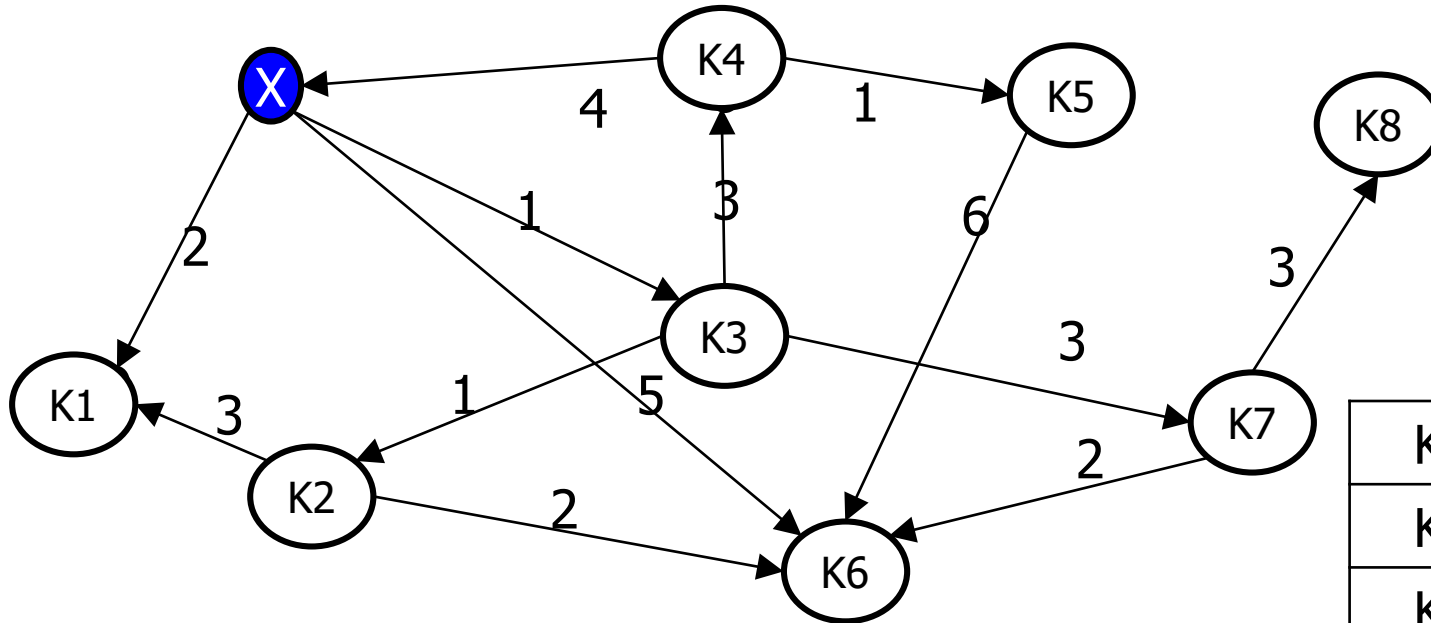
- First approach: Enumerate all paths
 - Need to break cycles (e.g. $X - K3 - K4 - X - K3 - \dots$)
 - Using DFS: $X - K3 - K4 - X$ [BT-K4] $- K5 - K6$ [BT-K5] [BT-K4]
 [BT-K3] $- K7 - K8$ [BT-K7] $- K6$ [BT-K7] [BT-K3] $- K2 - K6$ [BT-K2]
 $- K1$ [BT-K2] [BT-K3] [BT-X] $K6 - \dots$

Dijkstra's Idea



- Enumerate **paths from X by their length**
 - Neither DFS nor BFS
- Assume we reach a node **Y** by a **path p** of length **l** and we have already explored all paths from X with length $l' < l$ and that Y was not reached yet
- Then **p must be a shortest path** between X and Y
 - Because any p' between X and Y would have a **prefix of length at least l** and (a) a continuation with length > 0 (only positive weights) or (b) would not need a continuation (then p is as short as p')

Example for Idea



- 1: X – K3
- 2: X – K3 – K2
2: X – K1
- 4: X – K3 – K2 – K6
4: X – K3 – K4
4: X – K3 – K7

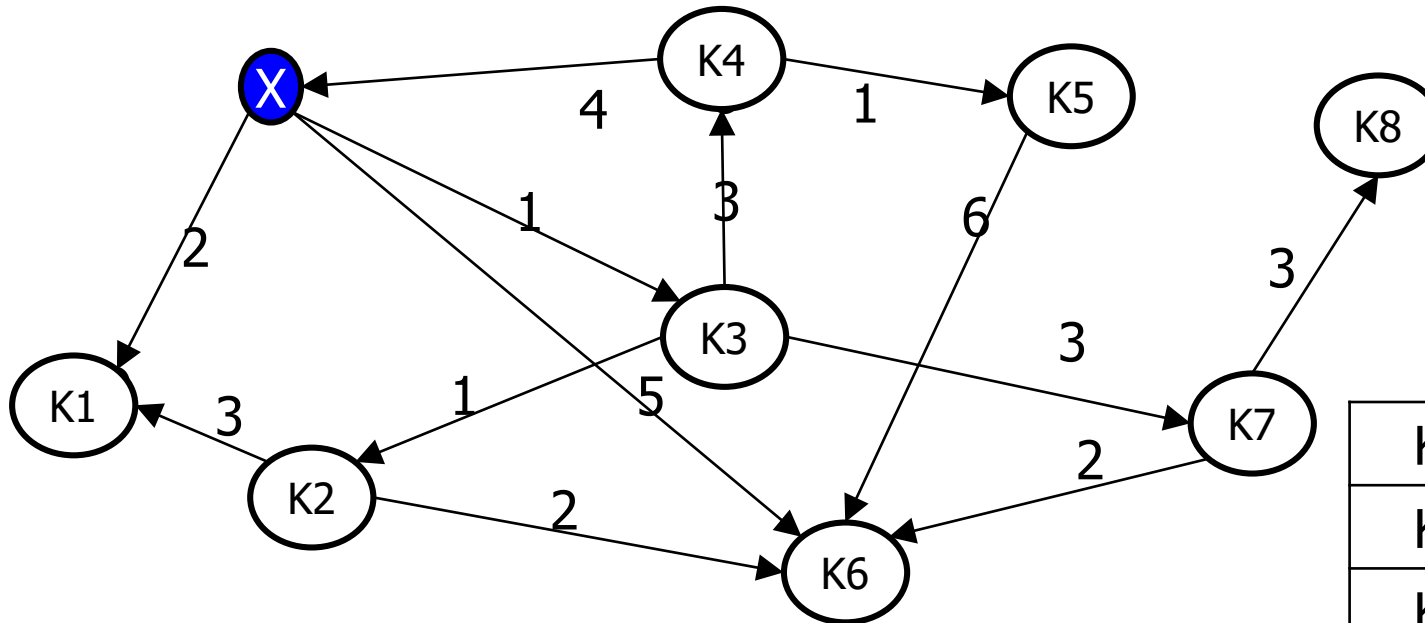
- 5: X – K3 – K4 – K5
- 7: X – K3 – K7 – K8
- Stop (all nodes found)

K3	1
K2	2
K1	2
K6	4
K4	4
K7	4
K5	5
K8	7

Algorithmic Idea

- Enumerate paths by iteratively extending already found shortest paths by all **possible extensions**
 - All edges outgoing from the end node of a short path
- These extensions
 - ... either lead to a node which we didn't reach before – then we found a path, but cannot yet be sure it is the shortest
 - ... or lead to a node which we already reached but we are not yet sure if we found the shortest path to it – **update current best distance**
 - ... or lead to a node which we already reached and for which we also surely found a shortest path already – these can be ignored
- Extensions are stored in a **priority queue** with prio=length
- We **enumerate nodes by their distance**

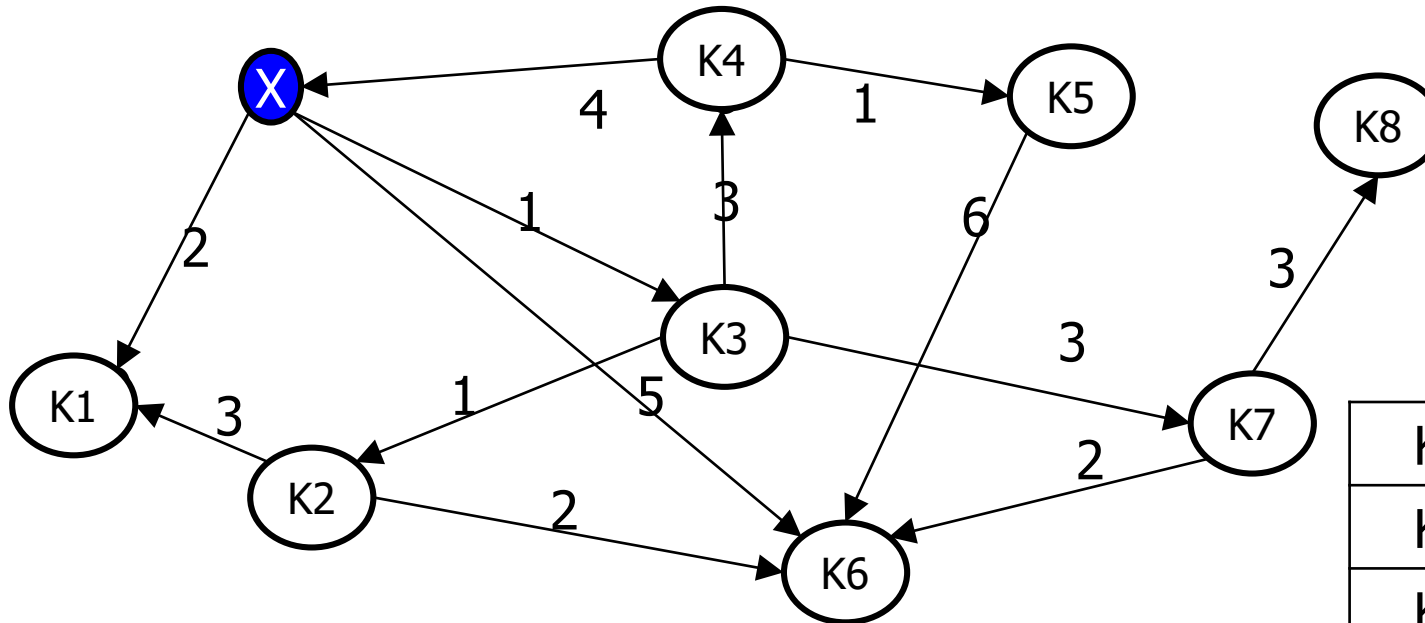
Example Step 2



- We first expand X
 - Path of length 1 to K3: New, not necessarily shortest
 - Path of length 2 to K1: New, not necessarily shortest
 - Path of length 5 to K6: New, not necessarily shortest

K3	1
K2	
K1	2
K6	5
K4	
K7	
K5	
K8	

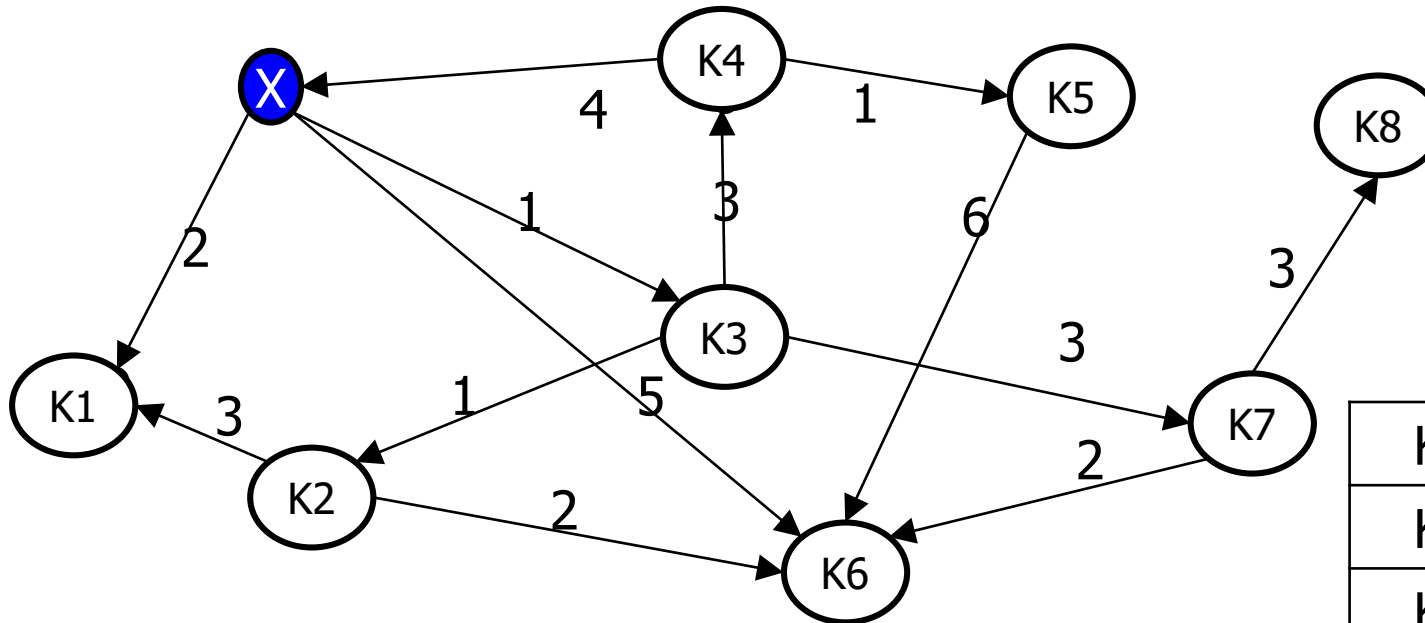
Example Step 1



- We expand K3 ("shortest" node)
 - Path of length $1+3$ to K4: New, not necessarily shortest
 - Path of length $1+3$ to K7: New, not necessarily shortest
 - Path of length $1+1$ to K2: New, not necessarily shortest

K3	1
K2	2
K1	2
K6	5
K4	4
K7	4
K5	
K8	

Example Step 2



- We expand K2 – currently shortest node
 - Path of length $2+3$ to K1: Discard, we have seen K1 already
 - Path of length $2+2$ to K6: Override, found shorter path
- We expand K1 – equally short node
 - ...

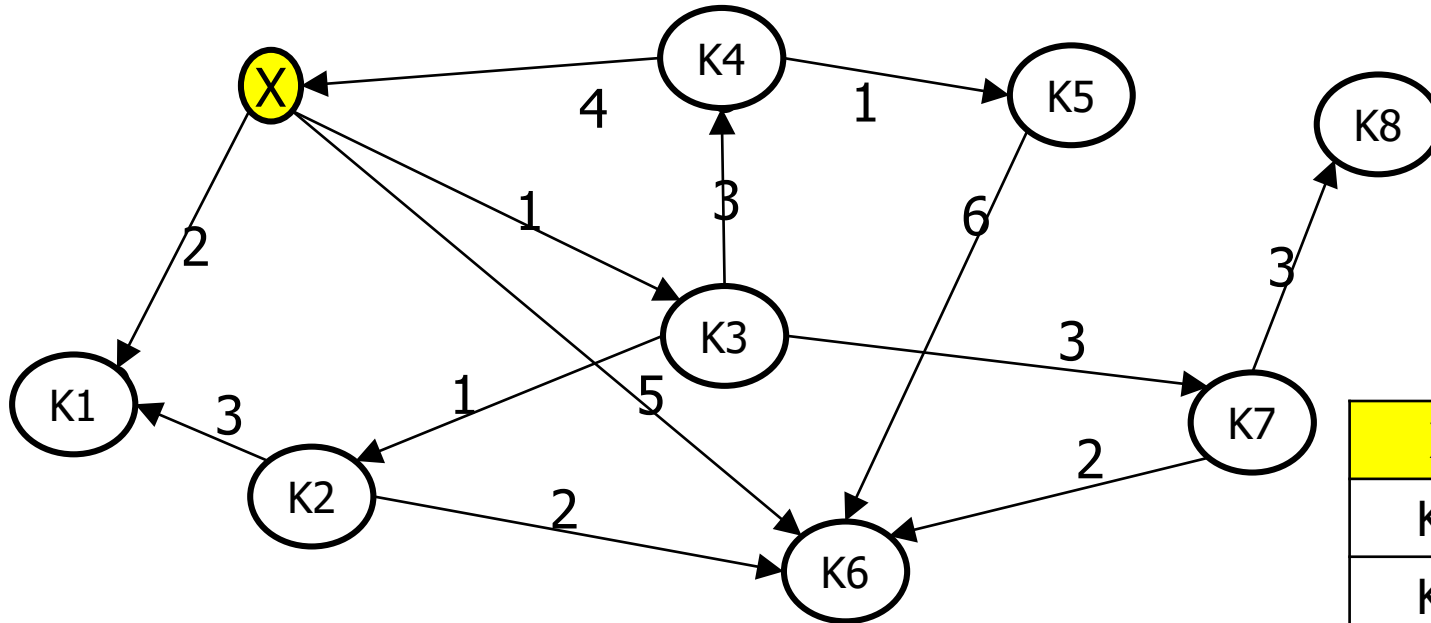
K3	1
K2	2
K1	2
K6	4
K4	4
K7	4
K5	
K8	

Algorithm

```
1. G = (V, E);
2. x : start_node;      # x ∈ V
3. A : array_of_distances;
4. ∀i: A[i] := ∞;
5. L := V;
6. A[x] := 0;
7. while L ≠ ∅
8.   k := L.get_closest_node(x);
9.   L := L \ k;
10.  forall (k, f, w) ∈ E do
11.    if f ∈ L then
12.      new_dist := A[k] + w;
13.      if new_dist < A[f] then
14.        A[f] := new_dist;
15.      end if;
16.    end if;
17.  end for;
18. end while;
```

- Assumptions
 - Nodes have IDs between 1 ... |V|
 - Edges are (from, to, weight)
- We enumerate nodes by length of their shortest paths
 - In the first iteration, we pick x and update distances A to all neighbors
 - When we pick a node k, we **already have computed its distance** to x in A
 - We adapt the current best distances to all neighbors of k we haven't picked yet
- Once we picked all nodes, we are done

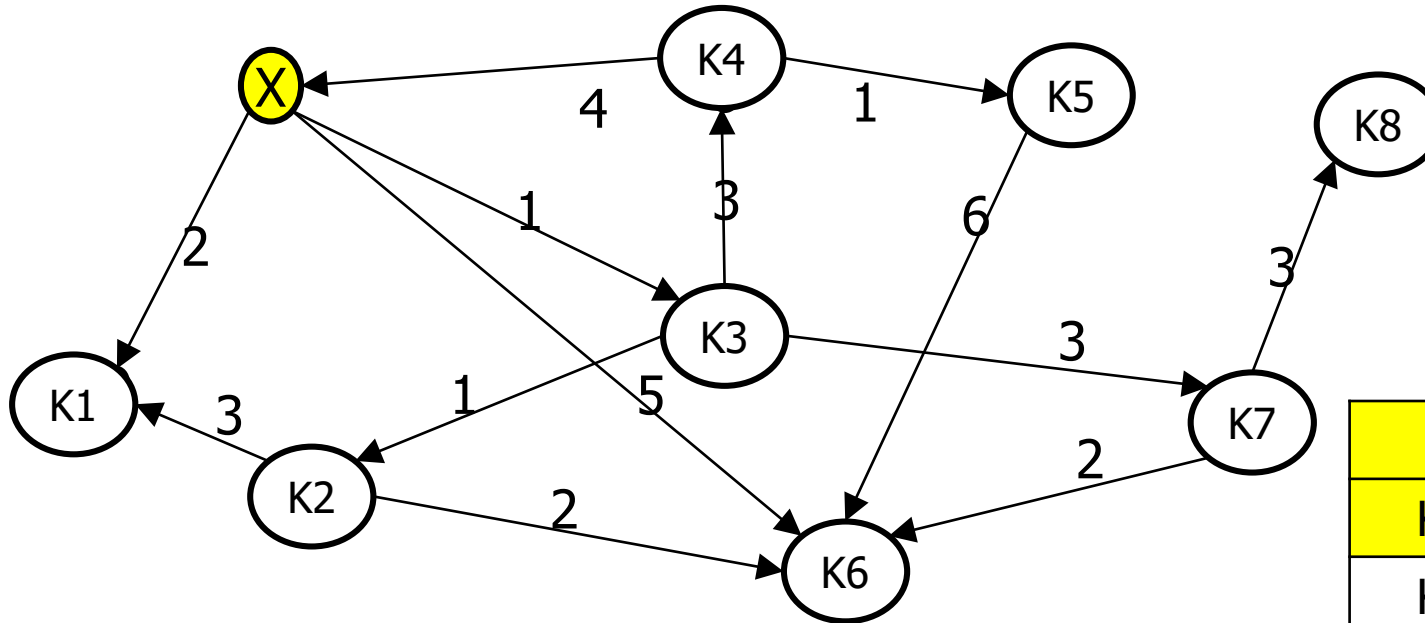
Example for Algorithm



- Pick x

X	0
K1	∞
K2	∞
K3	∞
K4	∞
K5	∞
K6	∞
K7	∞
K8	∞

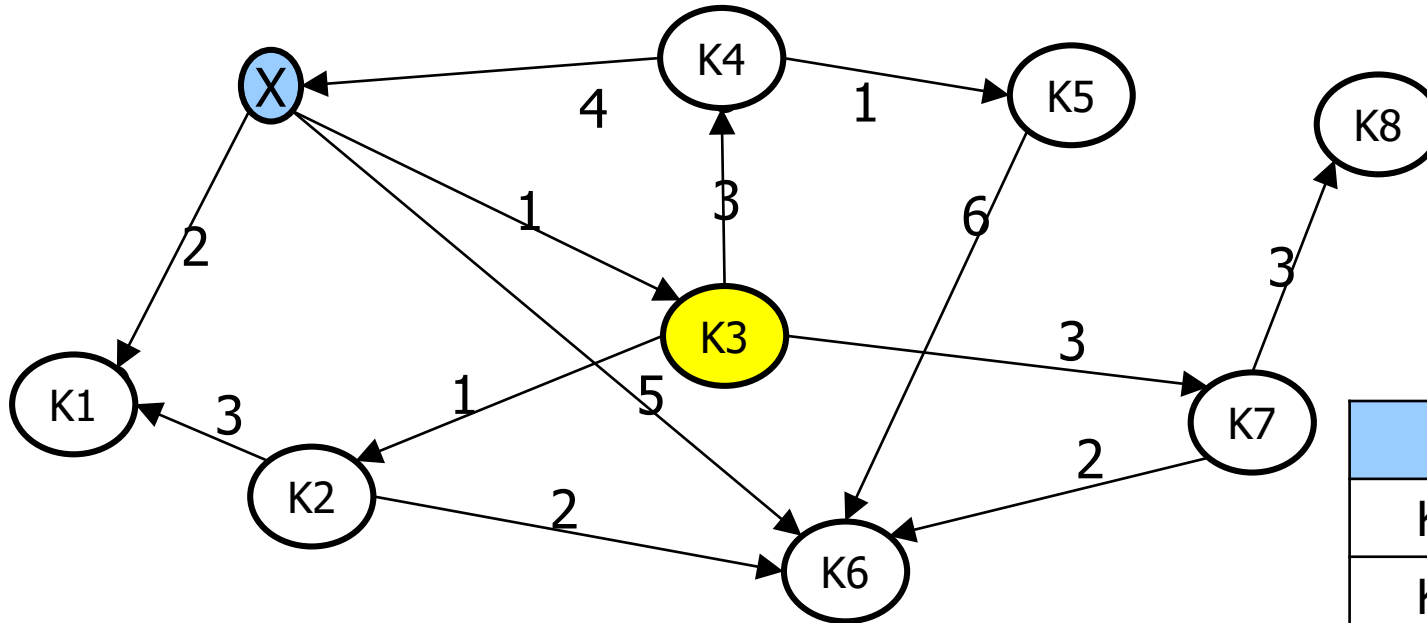
Example for Algorithm



- Pick x
- Adapt distances to all neighbors

X	0
K1	2
K2	∞
K3	1
K4	∞
K5	∞
K6	5
K7	∞
K8	∞

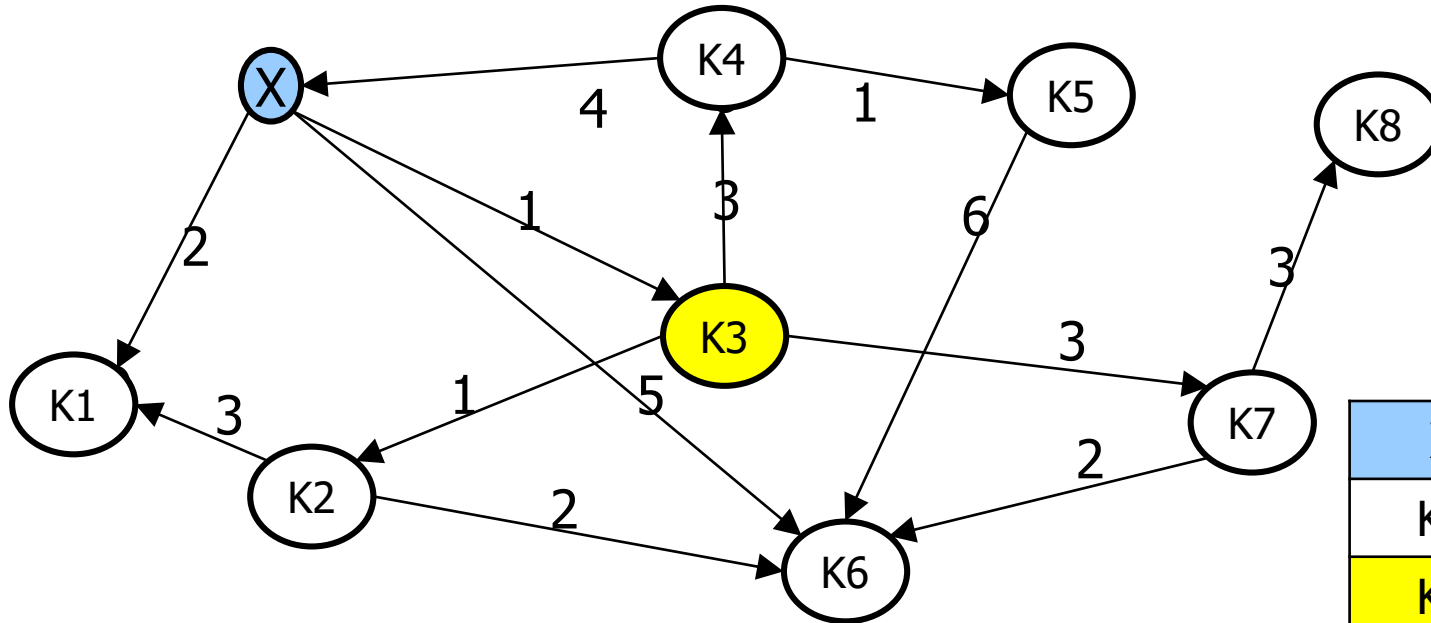
Example for Algorithm



X	0
K1	2
K2	∞
K3	1
K4	∞
K5	∞
K6	5
K7	∞
K8	∞

- X is **done** – remove from L
- Pick K3 (closest to x)

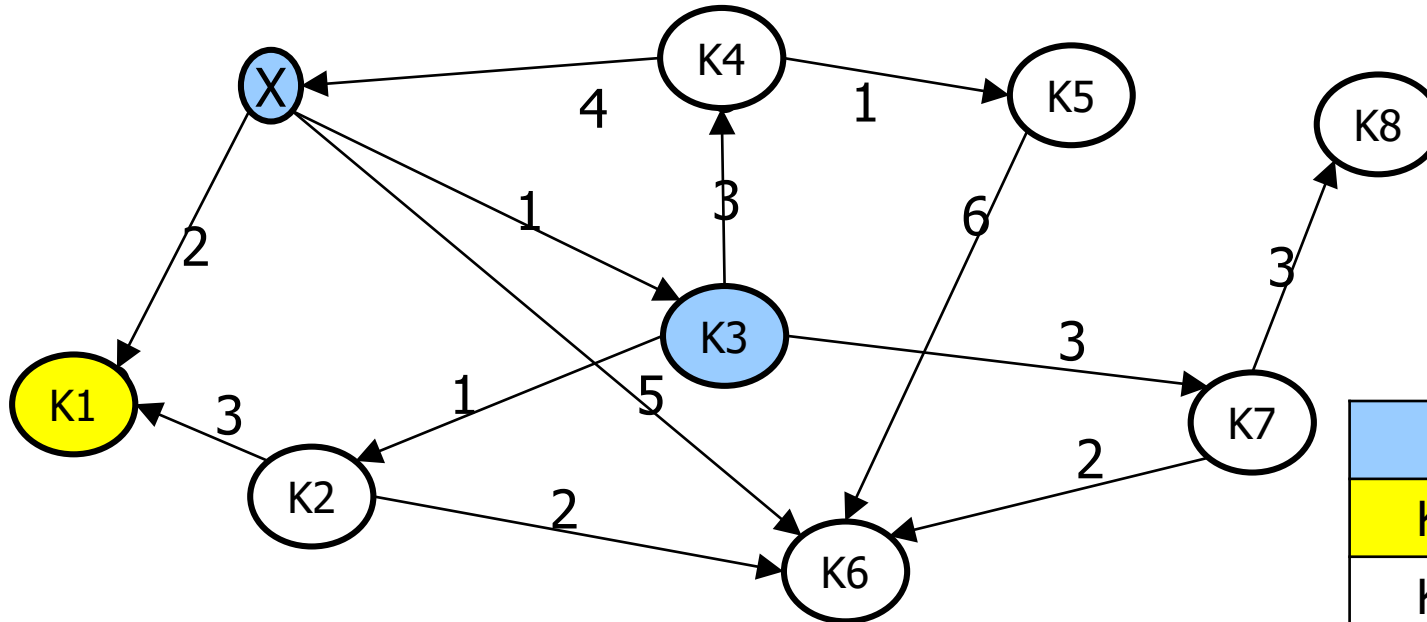
Example for Algorithm



- Pick K3
- Adapt distances (from x) to all neighbors (of K3)

X	0
K1	2
K2	2
K3	1
K4	4
K5	∞
K6	5
K7	4
K8	∞

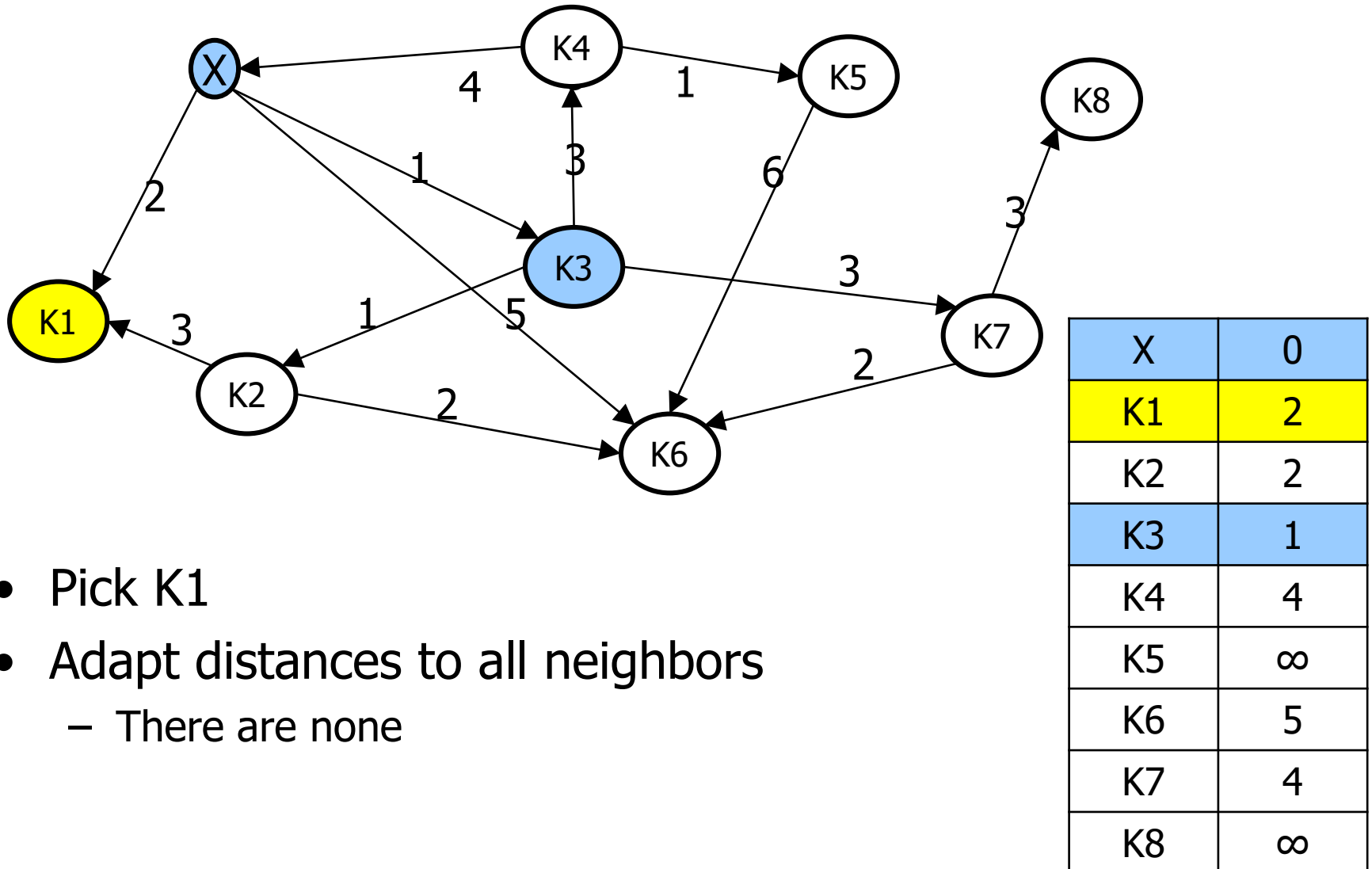
Example for Algorithm



X	0
K1	2
K2	2
K3	1
K4	4
K5	∞
K6	5
K7	4
K8	∞

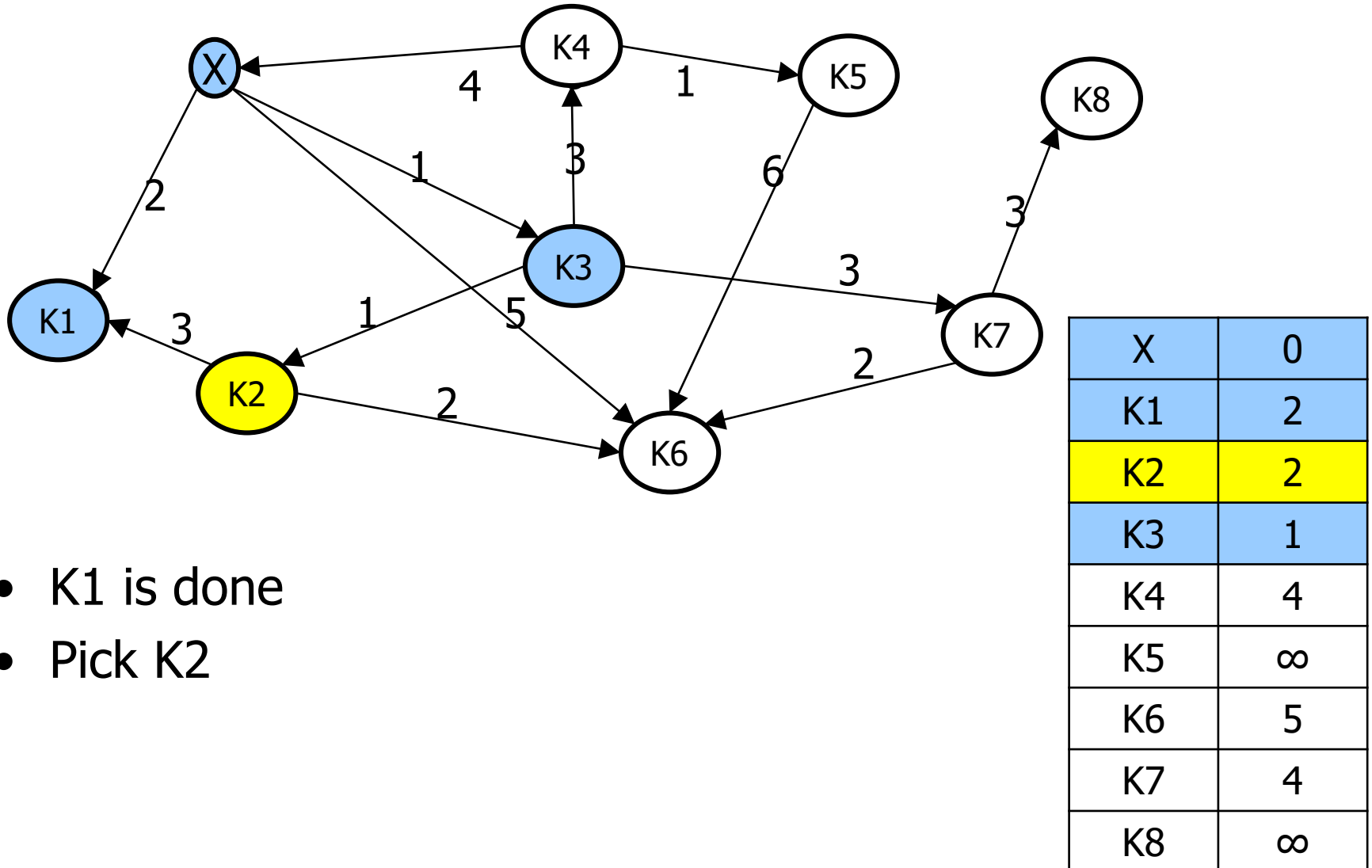
- K3 is done (we cannot find a shorter path)
- Pick K1 (or K2)

Example for Algorithm

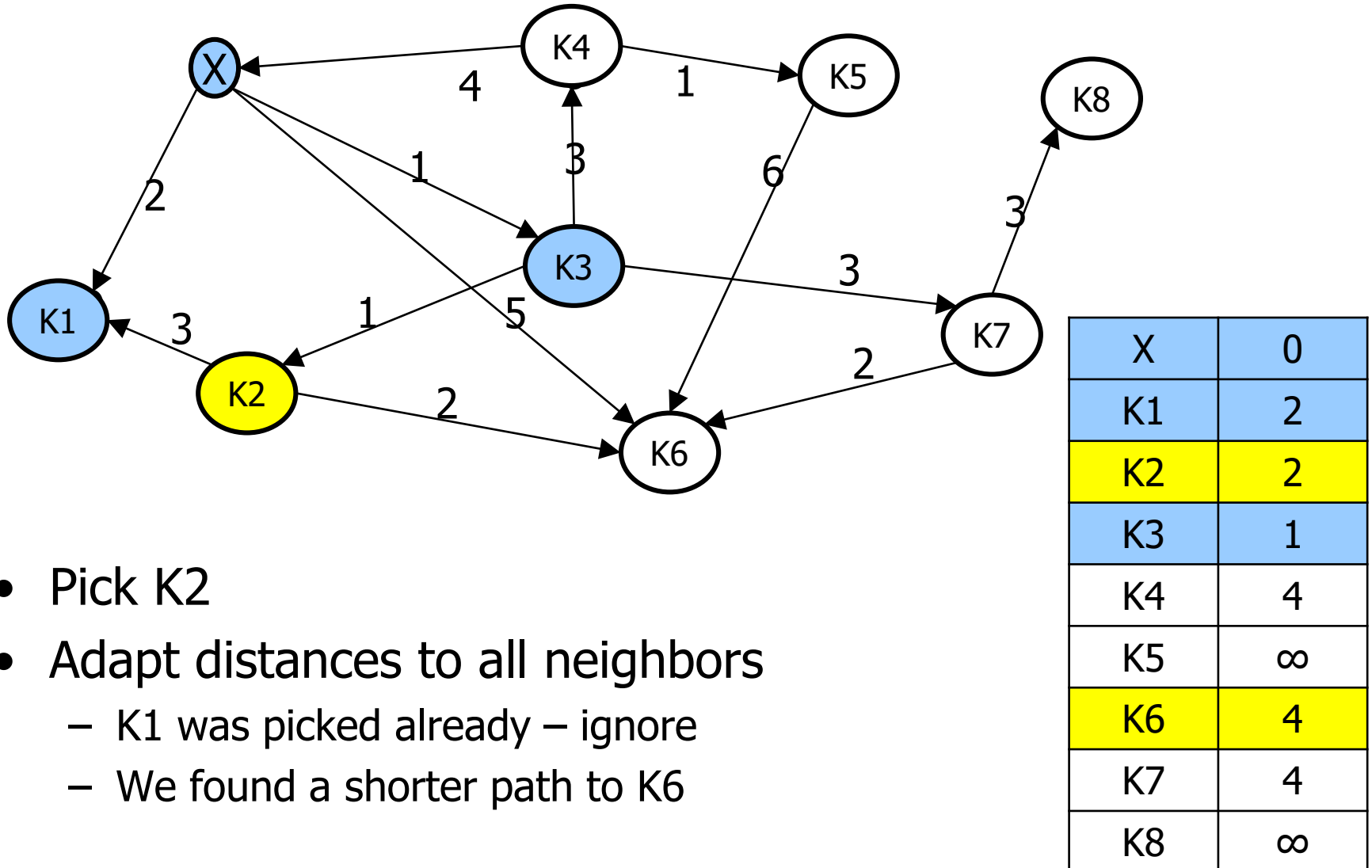


- Pick K1
- Adapt distances to all neighbors
 - There are none

Example for Algorithm

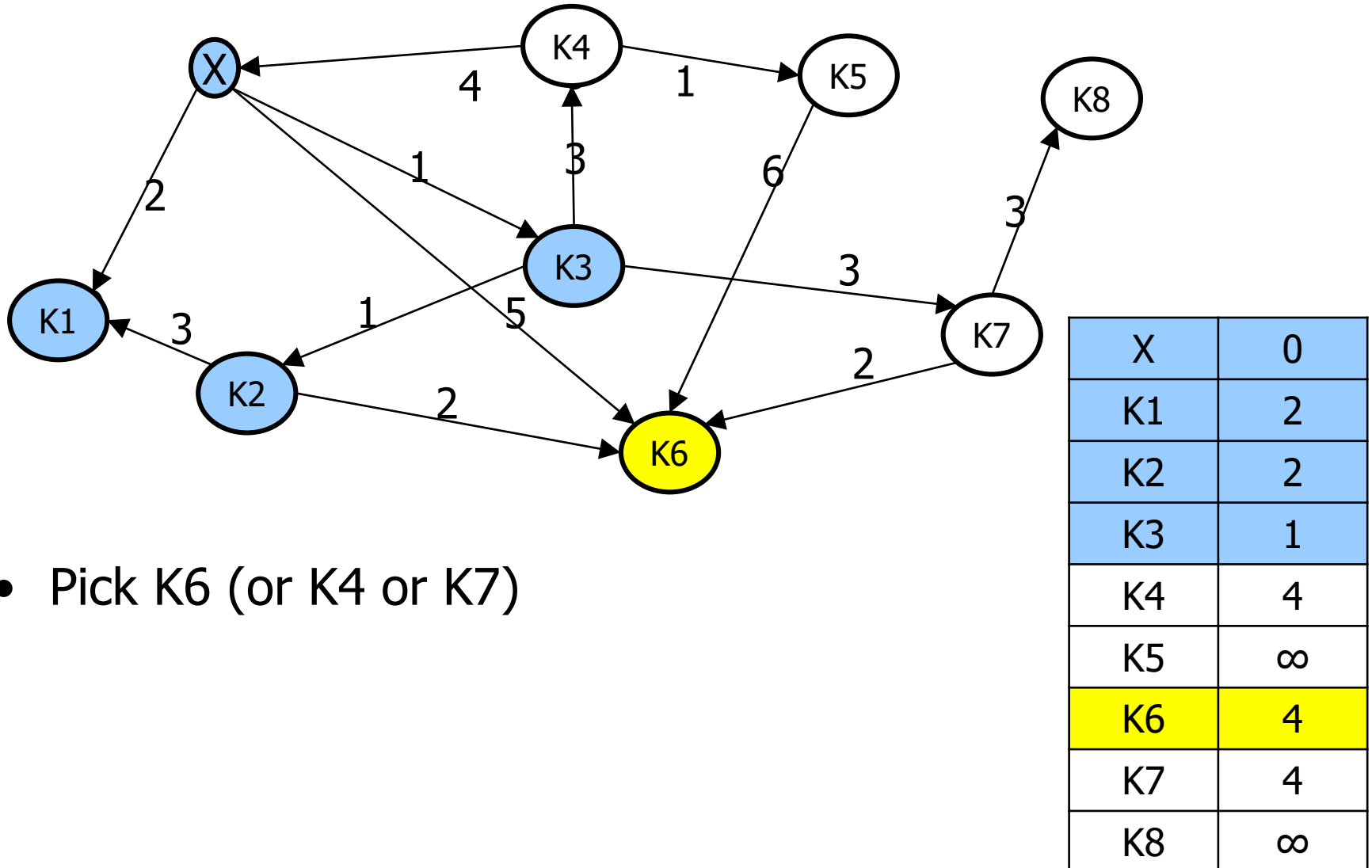


Example for Algorithm

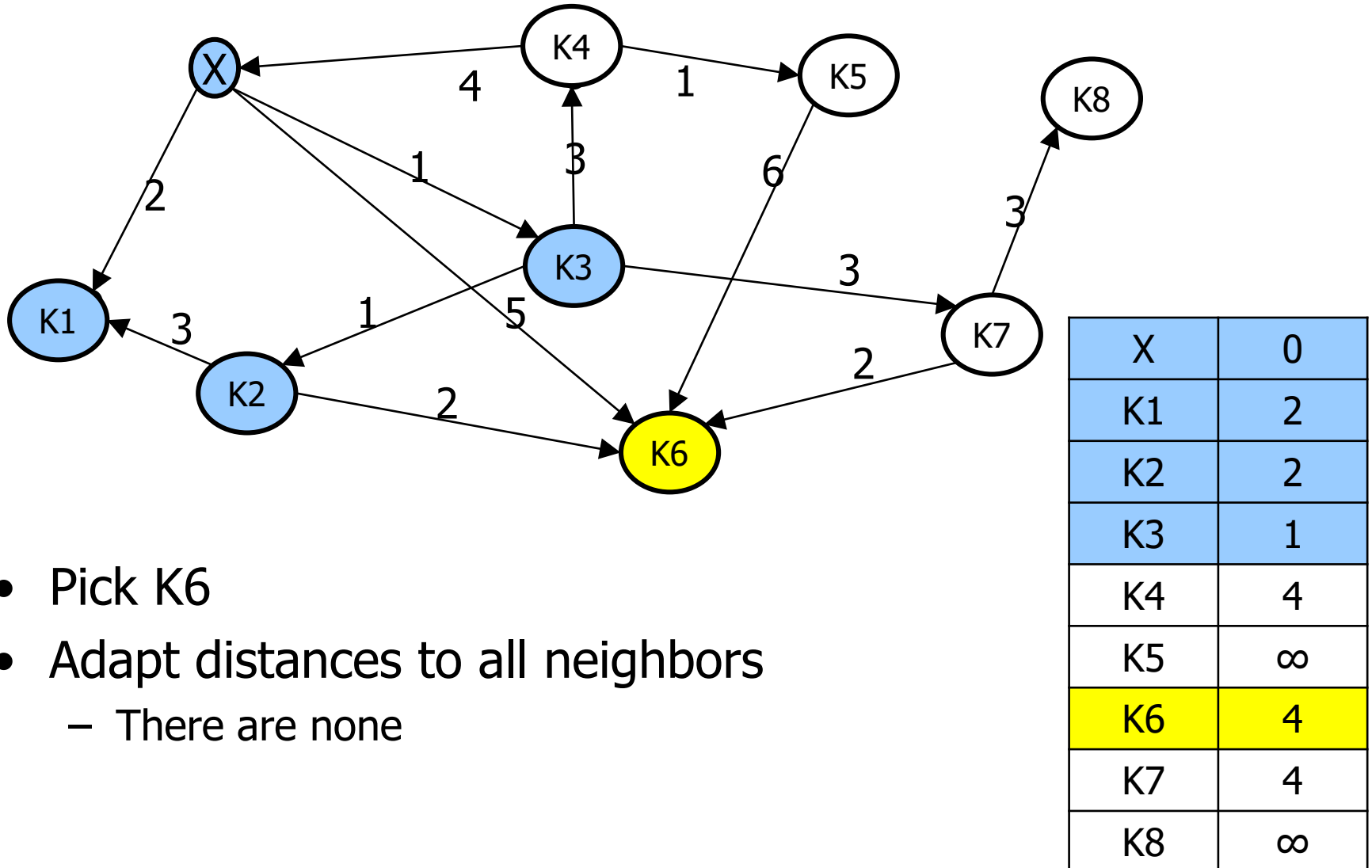


- Pick K2
- Adapt distances to all neighbors
 - K1 was picked already – ignore
 - We found a shorter path to K6

Example for Algorithm

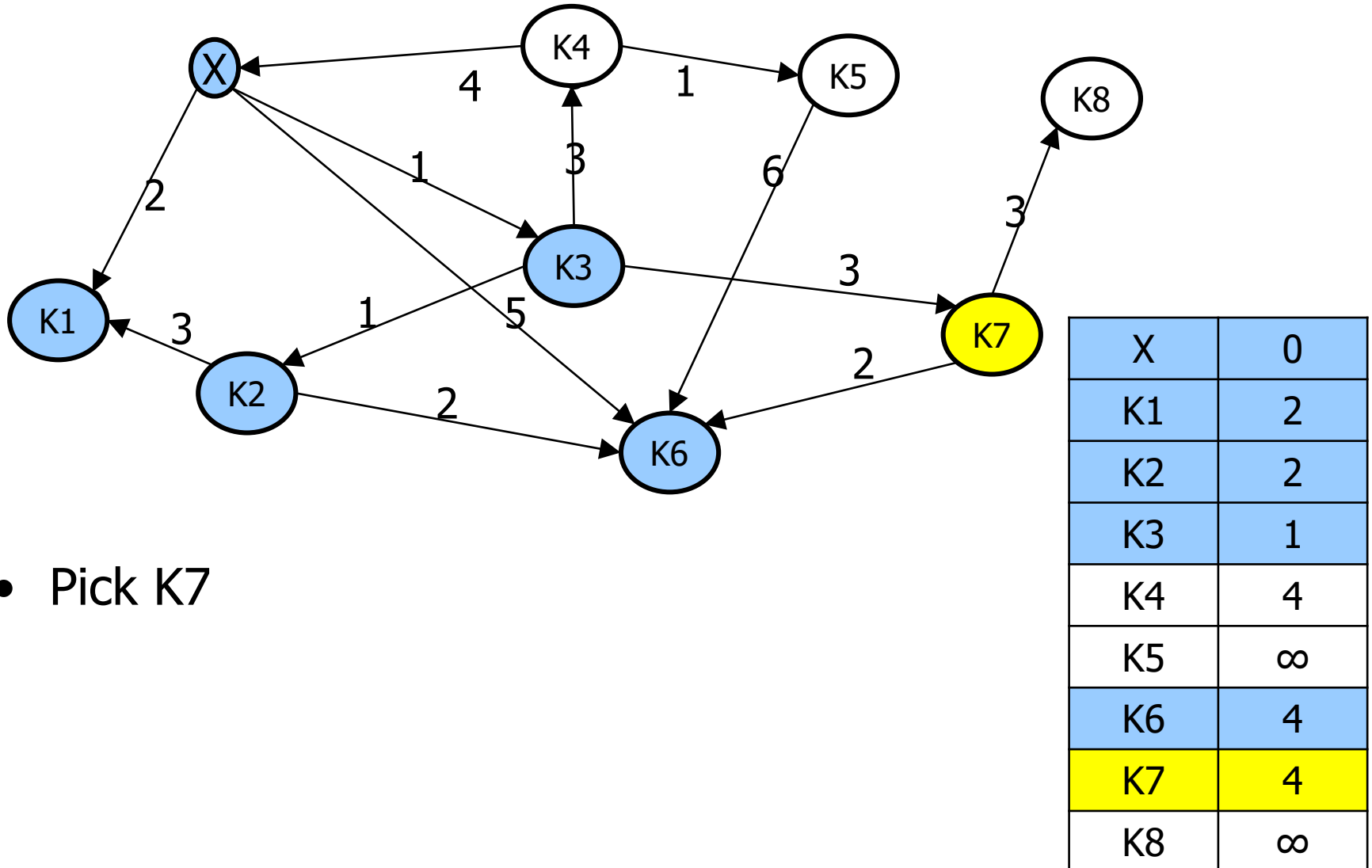


Example for Algorithm

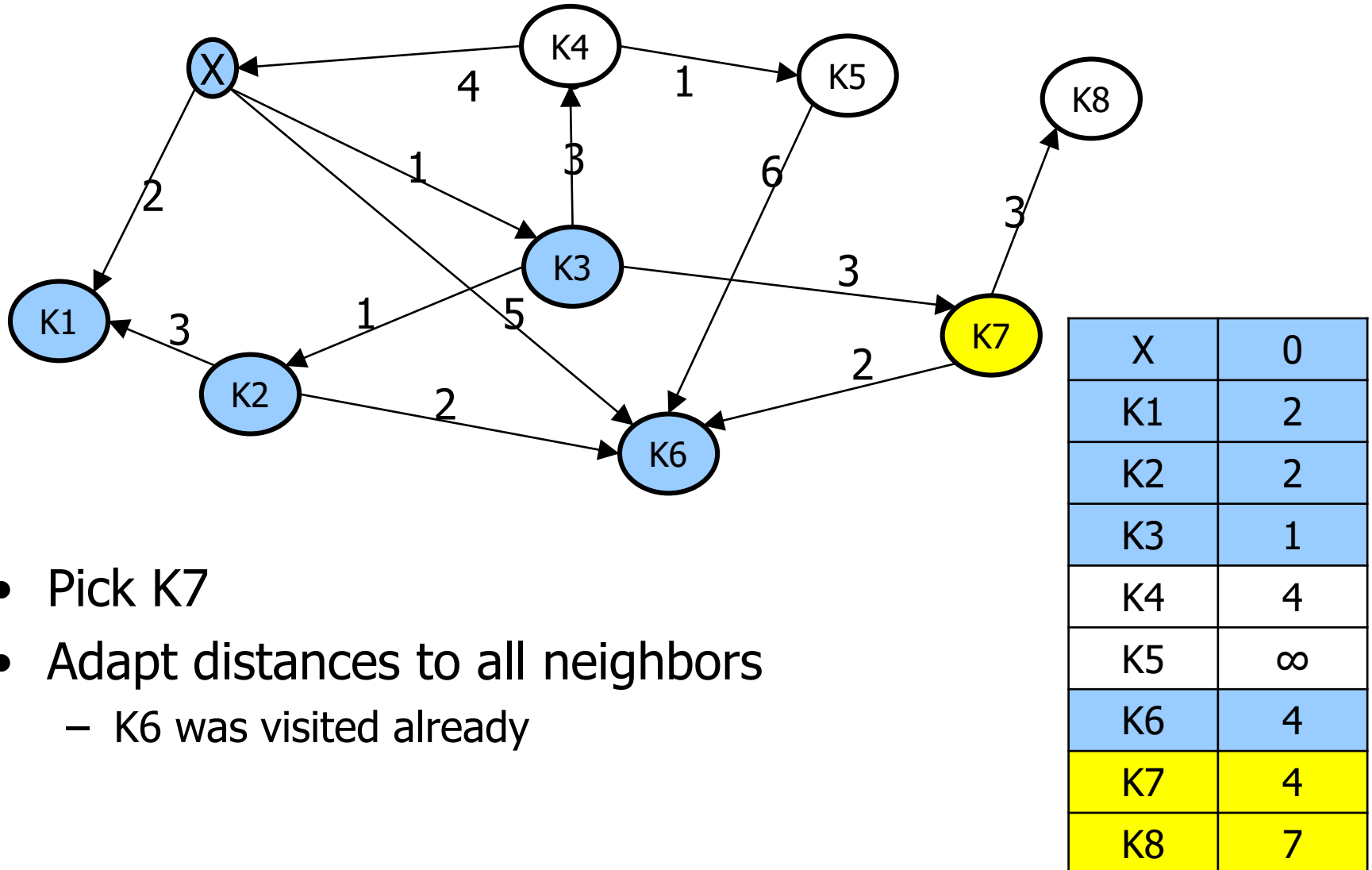


- Pick K6
- Adapt distances to all neighbors
 - There are none

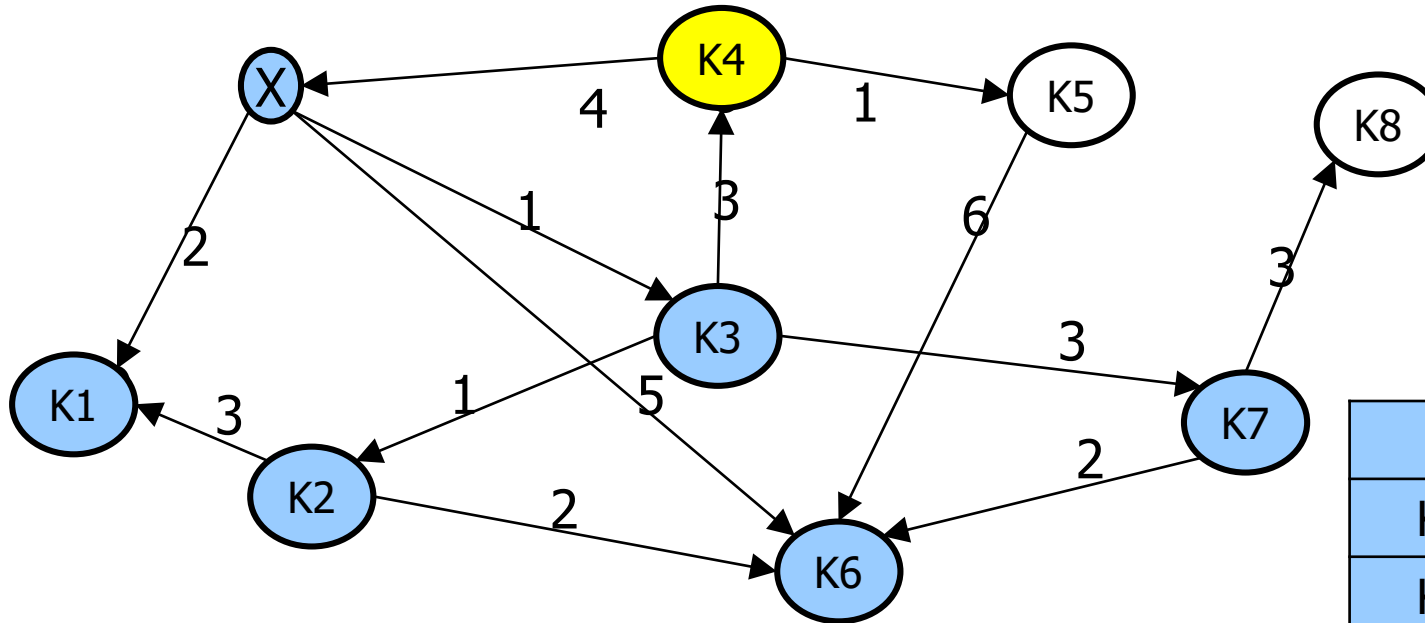
Example for Algorithm



Example for Algorithm



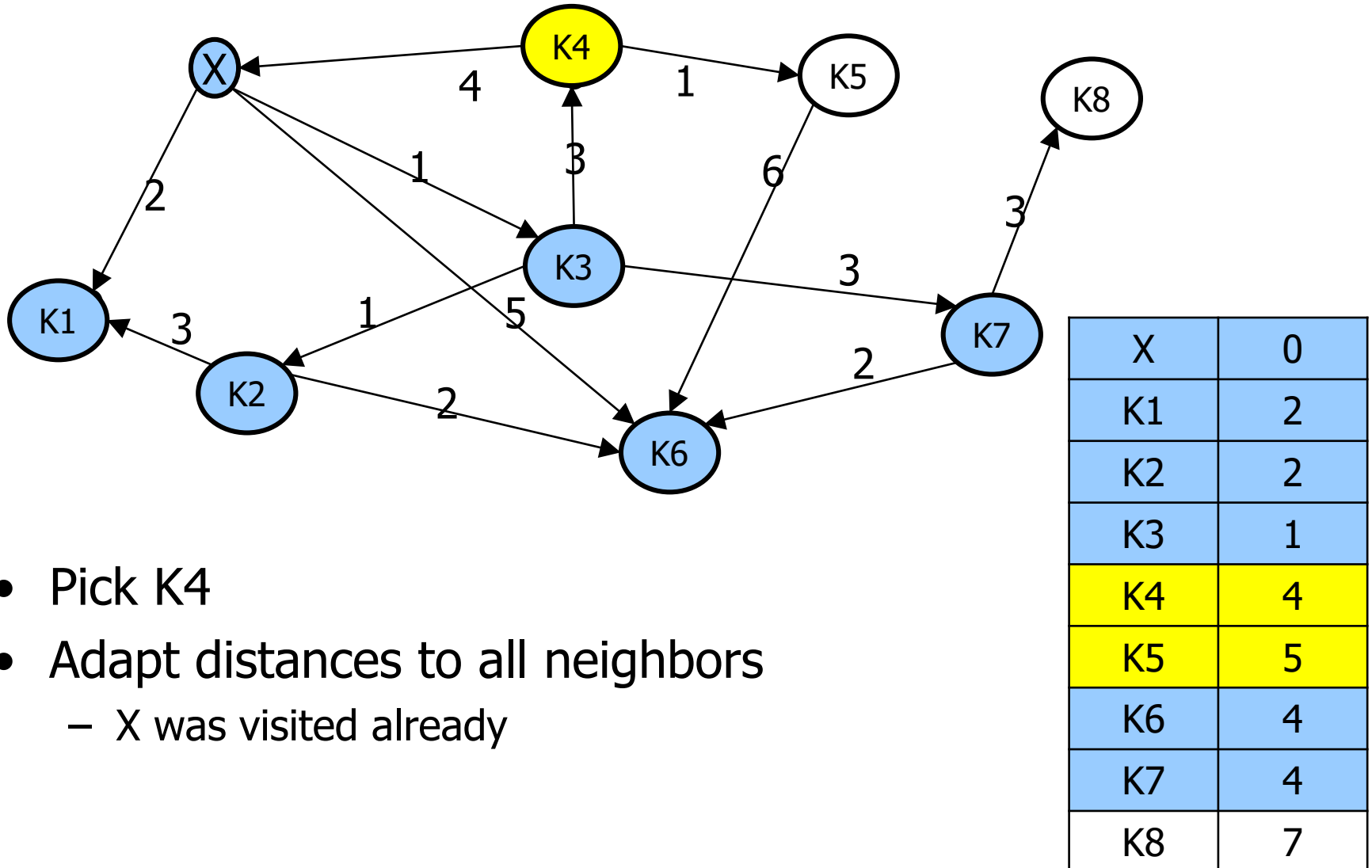
Example for Algorithm



- Pick K4

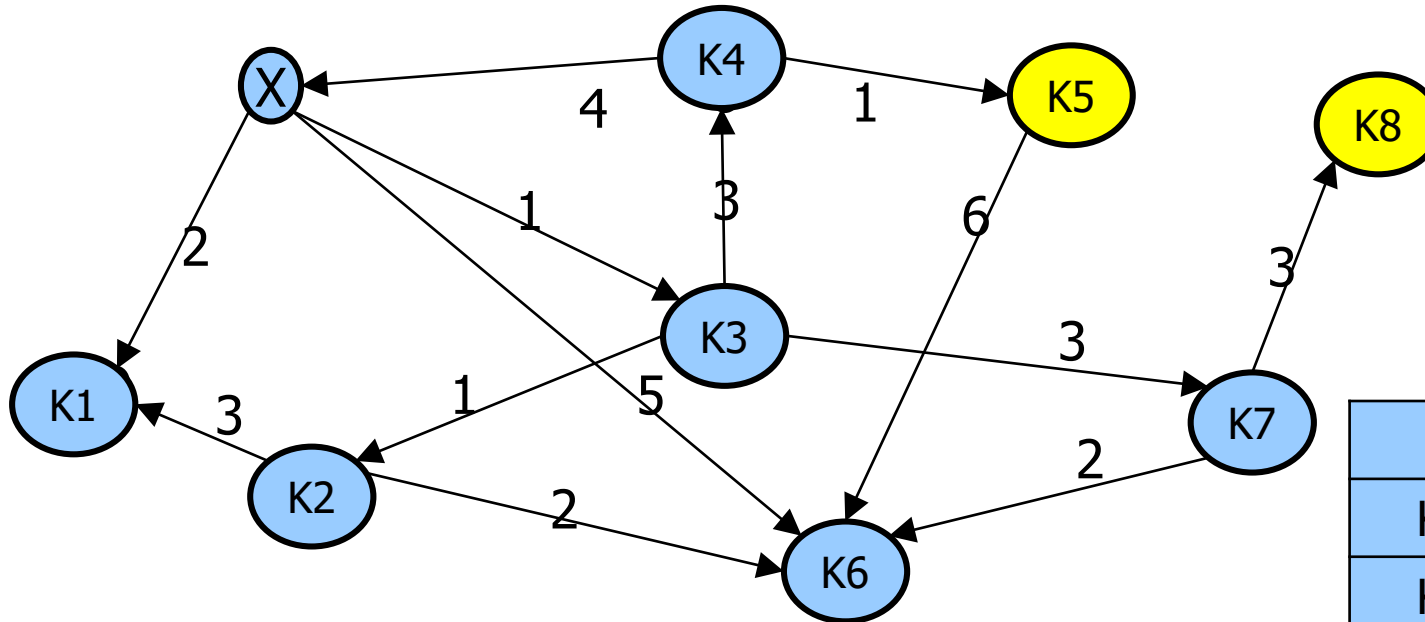
X	0
K1	2
K2	2
K3	1
K4	4
K5	∞
K6	4
K7	4
K8	7

Example for Algorithm



- Pick K4
- Adapt distances to all neighbors
 - X was visited already

Example for Algorithm



- Pick K5 ... Pick K8

X	0
K1	2
K2	2
K3	1
K4	4
K5	5
K6	4
K7	4
K8	7

A Closer Look

```
1. G = (V, E);
2. x : start_node;      # x ∈ V
3. A : array_of_distances_from_x;
4. ∀i: A[i] := ∞;
5. L := V;              # organized as PQ
6. A[x] := 0;
7. update( L );
8. while L ≠ ∅
9.   k := L.get_closest_node();
10.  L := L \ k;
11.  forall (k, f, w) ∈ E do
12.    if f ∈ L then
13.      new_dist := A[k] + w;
14.      if new_dist < A[f] then
15.        A[f] := new_dist;
16.        update( L );
17.      end if;
18.    end if;
19.  end for;
20. end while;
```

- Central: **get_closest_node(x)**
 - Needs to find the node k in L for which A[k] is the smallest
 - A[k] may change all the time
- Searching A? Resorting A?
- Trick: Organize L as “enhanced” **priority queue**
 - We need to be able to update the priority of nodes
 - Done in $O(\log(n))$ by removing then re-inserting the node in a min-heap

Dijkstra's Algorithm – Single Operations

```
1. G = (V, E);
2. x : start_node;      # x ∈ V
3. A : array_of_distances_from_x;
4. ∀i: A[i] := ∞;
5. L := V;              # organized as PQ
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```

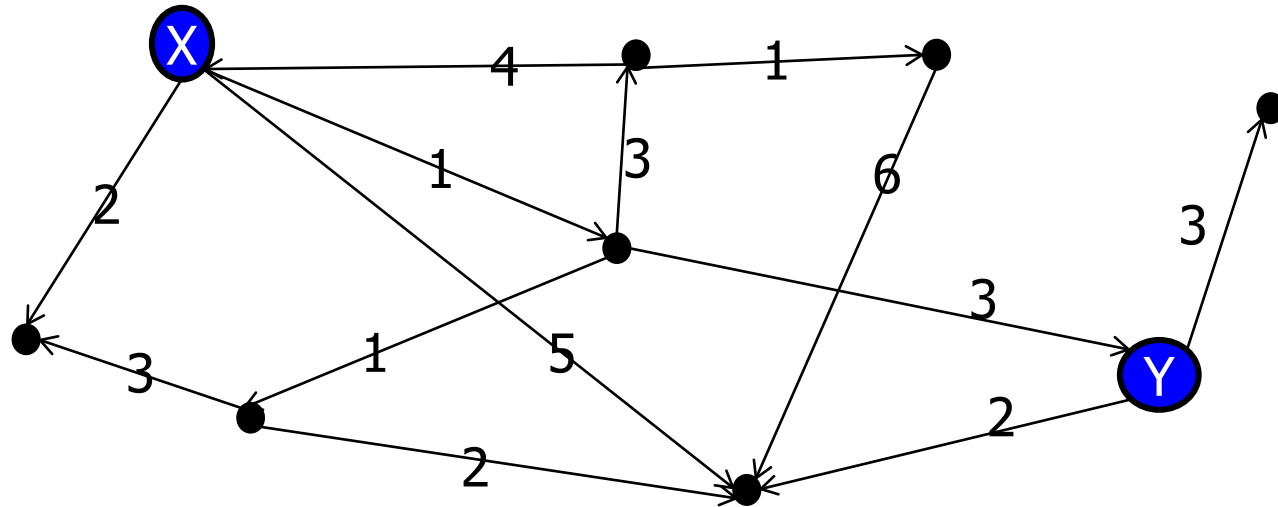
- Assume a heap-based Prio-Q L
 - L holds at most all nodes (n)
 - Line 4: $O(n)$
 - L5: $O(n)$ (build PQ)
 - L9: $O(1)$ (getMin)
 - L10: $O(\log(n))$ (deleteMin)
 - L11: $O(m)$ (with adjacency list)
 - L12: $O(1)$
 - Requires additional array LA of size $|V|$ storing membership of nodes in L
 - L16: $O(\log(n))$ (updatePQ)
 - Store in LA pointers to nodes in L; then remove/insert node

Dijkstra's Algorithm - Loops

```
1. G = (V, E);
2. x : start_node;      # x ∈ V
3. A : array_of_distances;
4. ∀i: A[i] := ∞;
5. L := V;              # organized as PQ
6. A[x] := 0;
7. update( L );
8. while L ≠ ∅
9.   k := L.get_closest_node();
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16.        update( L );
17.      end if;
18.    end if;
19.  end for;
20. end while;
```

- Central costs
 - L10: $O(\log(n))$ (deleteMin)
 - L16: $O(\log(n))$ (del+ins)
- Loops
 - Lines 8-19: $O(n)$
 - Line 11-18: **All edges** exactly once
 - Together: $O(m+n)$
- Altogether: **$O((n+m)*\log(n))$**
 - With Fibonacci heaps: Amortized costs are $O(n*\log(n)+m)$
 - Also possible in $O(n^2)$; this is better in dense graphs ($m \sim n^2$)

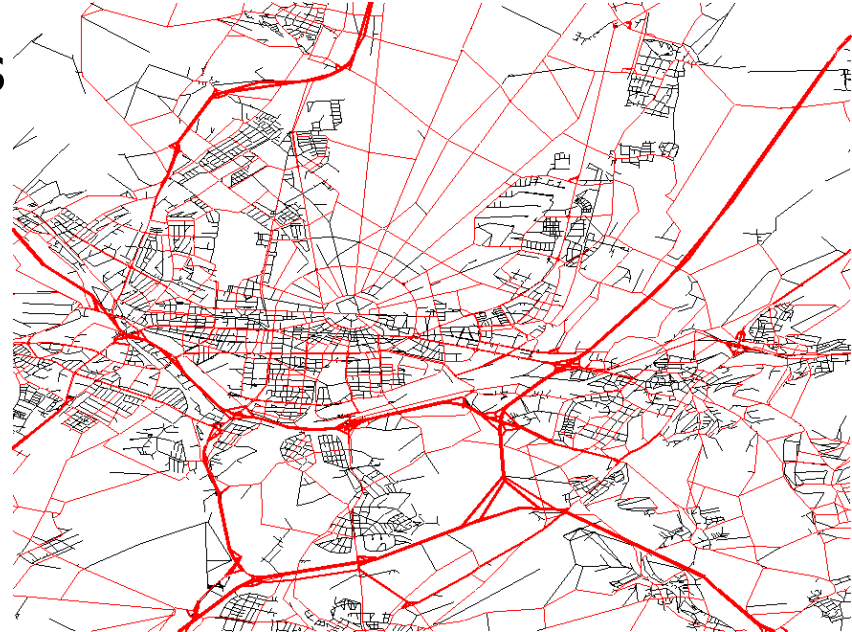
Single-Source, Single-Target



- Task: Find the **distance between X and only Y**
- Solution: Dijkstra as well
 - We can **stop as soon as Y** appears at the min position of the PQ
 - We can visit edges in order of increasing weight (might help)
 - Worst-case complexity unchanged
- Things are different in planar graphs (navigators!)

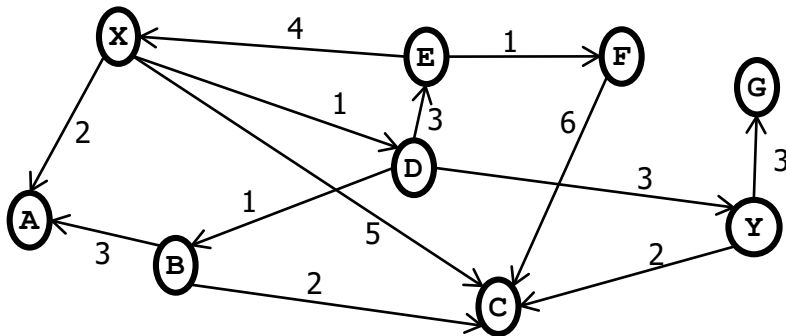
Outlook: Highway Hierarchies

- Shortest-Path Routing on maps
- Exploit Highway hierarchy
 - Autobahn, Bundesstrasse, Regionalstrasse, Strasse, Pfad ...
- Iterative refinement in layered maps
- “towards $O(1)$ ” [SS07]
- Extensions
 - Second best non-overlapping path
 - Fleet management: Traveling salesman
 - Logistics: Pick-up-and-delivery with intermediate stocks
 - Budget optimization (gasoline, empty trips, sleep-restrictions, road tolls, border / customs regulations, ...)



Faster SS-ST Algorithms

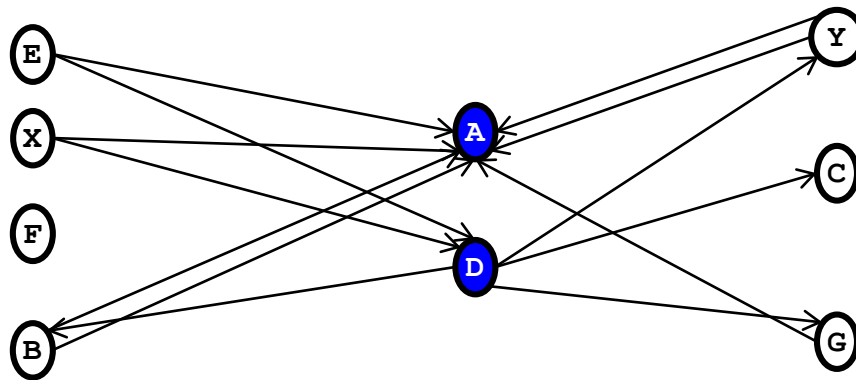
- Trick 1: Pre-compute all distances
 - Transitive closure with distances
 - Requires $O(|V|^2)$ space: Prohibitive for large graphs
 - How? See next lecture



→	A	B	C	D	E	F	G	X	Y
A	0	-	-	-	-	-	-	-	-
B	3	0	2	-	-	-	-	-	-
C	-	-	0	-	-	-	-	-	-
D	4	1	3	0	3	4	6	7	3
E	6	6	7	5	0	1	11	4	8
F	-	-	6	-	-	0	-	-	-
G	-	-	-	-	-	-	0	-	-
X	2	2	4	1	4	5	7	0	4
Y	-	-	2	-	-	-	3	-	0

Faster SS-ST Algorithms

- Trick 2: **Two-hop cover** with distances
 - Find a (hopefully small) set S of nodes such that
 - For every pair of nodes v_1, v_2 , at least **one shortest path from v_1 to v_2 goes through a node $s \in S$**
 - Thus, the distance between v_1, v_2 is $\min\{ d(v_1, s) + d(s, v_2) \mid s \in S \}$
 - S is called a 2-hop cover
 - Problem: Finding a **minimal S is NP-complete**
 - And S need not be small



More Distances

- Graphs with **negative edge** weights
 - Shortest paths (in terms of weights) may be very long (edges)
 - **Bellman-Ford** algorithm is in $O(n^2 \cdot m)$
- **All-pairs** shortest paths
 - Only positive edge weights: Use Dijkstra n times
 - With negative edge weights: **Floyd-Warshall** in $O(n^3)$
 - See next lecture
- **Reachability**
 - Simple in undirected graphs: Compute all connected components
 - In digraphs: Use graph traversal or a special **graph indexing method**