

Graph Algorithms

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Reading

M. T. Goodrich, R. Tamassia and M. H. Goldwasser,
Data Structures and Algorithms in Java, 6th Edition,
2014.

- Chapter 14. Graphs
- Sections 14.5-14.7
- pp. 609-638

Learning Objectives

- To be able to *understand* the topological sort algorithm, the minimal spanning tree algorithm and Dijkstra's shortest path algorithm;
- To be able to *analyze* the time complexity of Dijkstra's shortest path algorithm;
- To be able to *implement* these three graph algorithms;
- To be able to *apply* these graph algorithms to solve problems.

Topological Sort

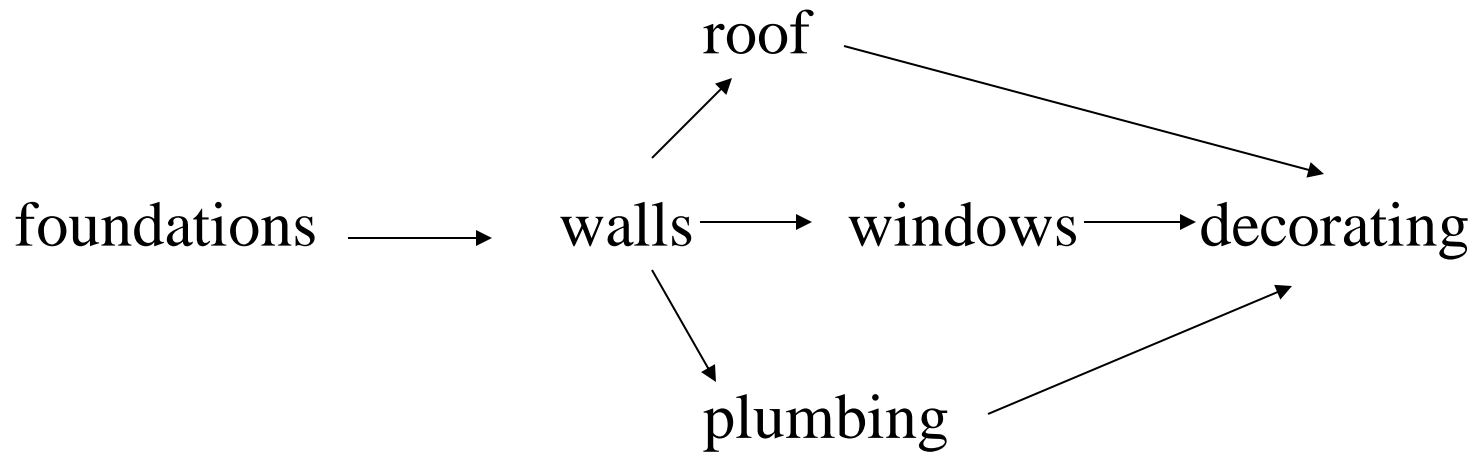
Given a directed acyclic graph, produce a linear sequence of vertices such that for any two vertices u and v , if there is an edge from u to v , then u is before v in the sequence.

拓扑排序 (Topological Sort) 是应用在**有向无环图 (DAG, Directed Acyclic Graph)**上的一种排序方法。
它的目标是：生成一个线性序列，使得每一条有向边 $u \rightarrow v$ 中，顶点 u 一定排在 v 之前。
条件：图必须是 有向 且 无环 (即 DAG)。

Topological Sort

- *Input* to the algorithm: directed acyclic graph 一个有向无环图 (DAG)
- *Output*: a linear sequence of vertices such that for any two vertices u and v , if there is an edge from u to v , then u is before v in the sequence.
一组顶点的线性序列，使得：对于任意边 $u \rightarrow v$ ，在排序中 u 必须排在 v 之前。
- Useful to think of this as: edges correspond to *dependencies* (pre-requisites), and a vertex could not precede its pre-requisites in the sequence.

Example: building a house



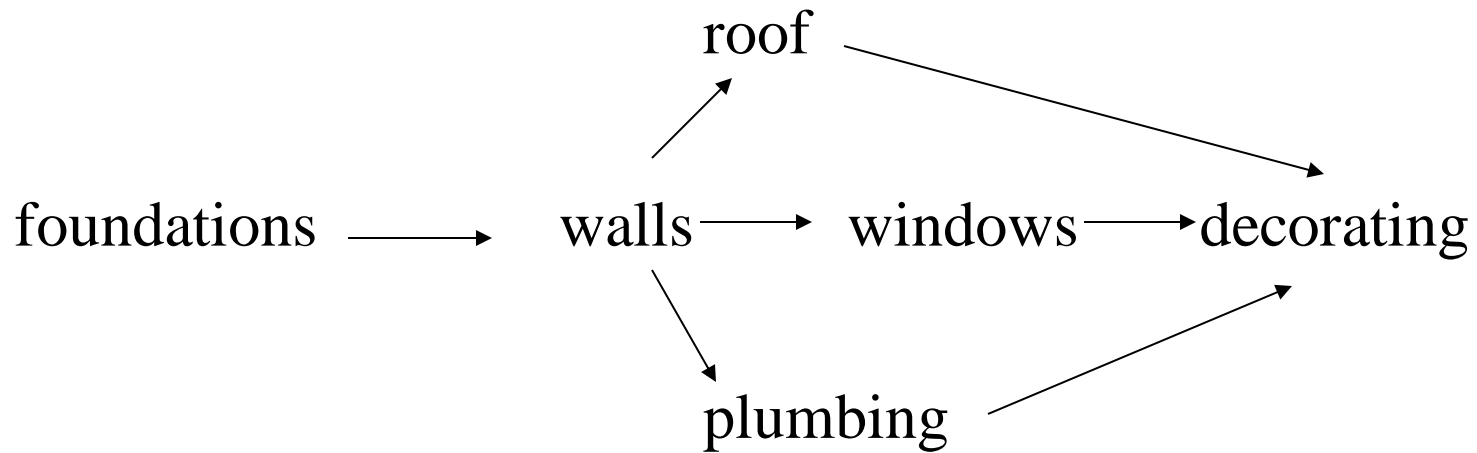
Possible sequence:

Foundations-Walls-Roof-Windows-Plumbing-Decorating

Applications

- Planning and scheduling 计划与调度
- The algorithm can also be modified to detect cycles. 检测环

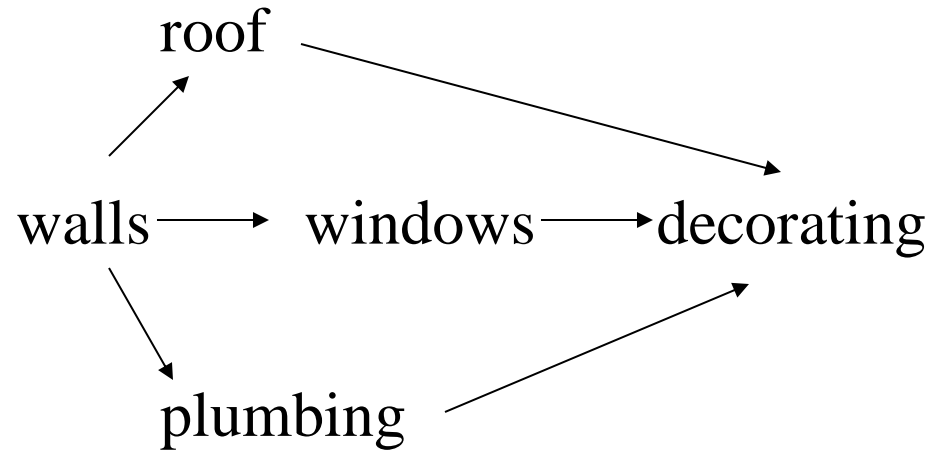
Example:



Array for the linear sequence: size 6

(Initially empty)

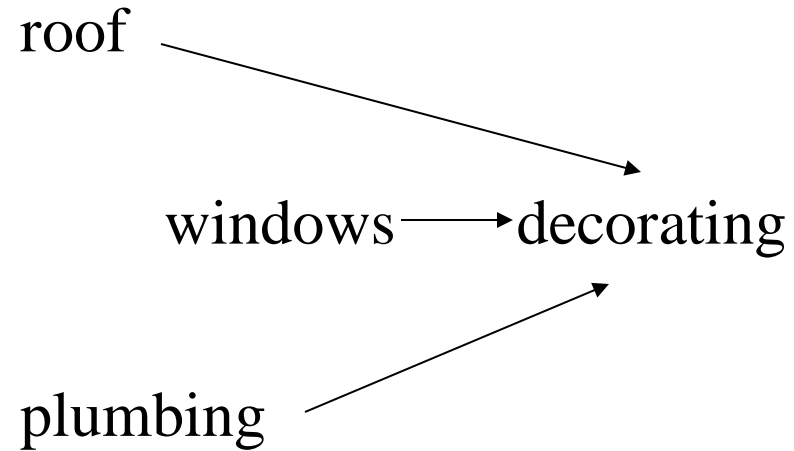
Example:



Array for the linear sequence: size 6

Foundations

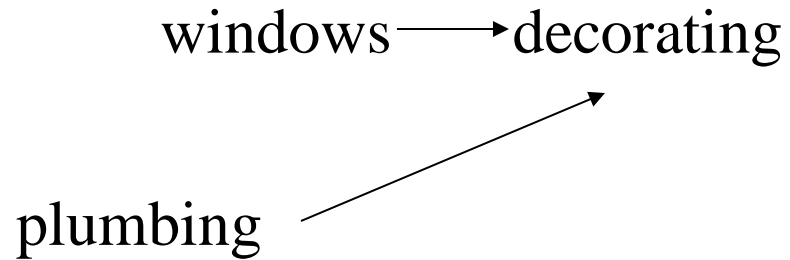
Example:



Array for the linear sequence: size 6

Foundations-Walls

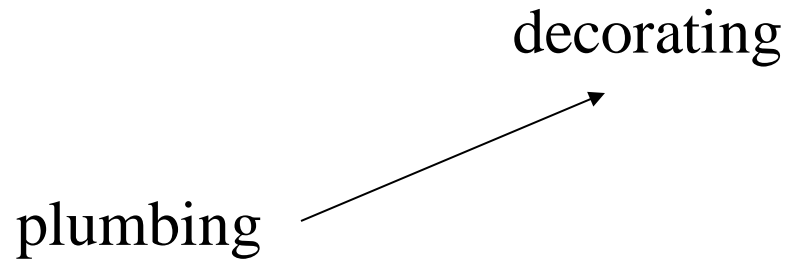
Example:



Array for the linear sequence: size 6

Foundations-Walls-Roof

Example:



Array for the linear sequence: size 6

Foundations-Walls-Roof-Windows

Example:

decorating

Array for the linear sequence: size 6

Foundations-Walls-Roof-Windows-Plumbing

Example:

Array for the linear sequence: size 6

Foundations-Walls-Roof-Windows-Plumbing-Decorating

创建一个空数组，长度为图中节点数（用于保存排序结果）。
当图中还有节点时，重复以下操作：找出 没有入边（即入度为 0、无先决条件）的节点 v 。
将这个节点 v 加入到结果数组 中。
从图中 删除该节点及其所有出边（即从其它节点的入边列表中删除）。
注意事项：这个过程会破坏原图结构（因为删除了节点），所以推荐先复制图的副本再进行操作，以保留原图。

Topological Sort algorithm

- Create an array of length equal to the number of vertices.
- While the number of vertices is greater than 0, repeat:
 - Find a vertex with no incoming edges (“no prerequisites”).
 - Put this vertex in the array.
 - Delete the vertex from the graph.
- Note that this destructively updates a graph; often this is a bad idea, so *make a copy* of the graph first and do topological sort on the copy.

Cycle detection with topological sort

- What happens if we run topological sort on a cyclic graph?

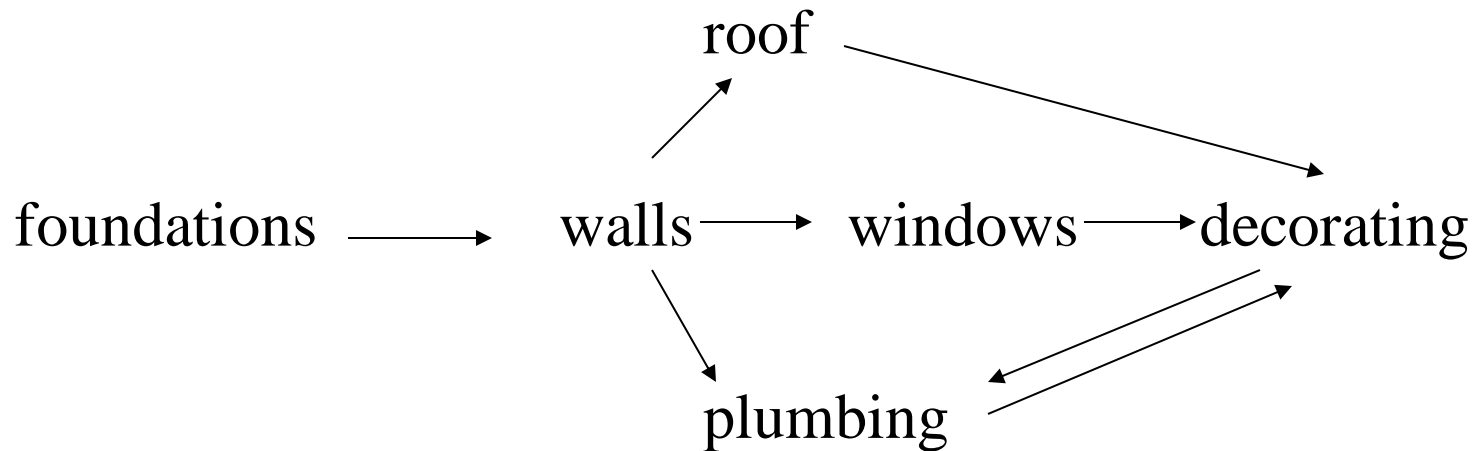
Cycle detection with topological sort

- What happens if we run topological sort on a cyclic graph?
- There will be either no vertex with 0 prerequisites to begin with, or at some point in the iteration.
- If we run a topological sort on a graph and there are vertices left undeleted, the graph contains a cycle.

不会找到入度为 0 的顶点，因为所有顶点都在某个环上，互相依赖，没有起点。
或者一开始能找到入度为 0 的节点，但在过程中会“卡住”找不到新的入度为 0 的节点。

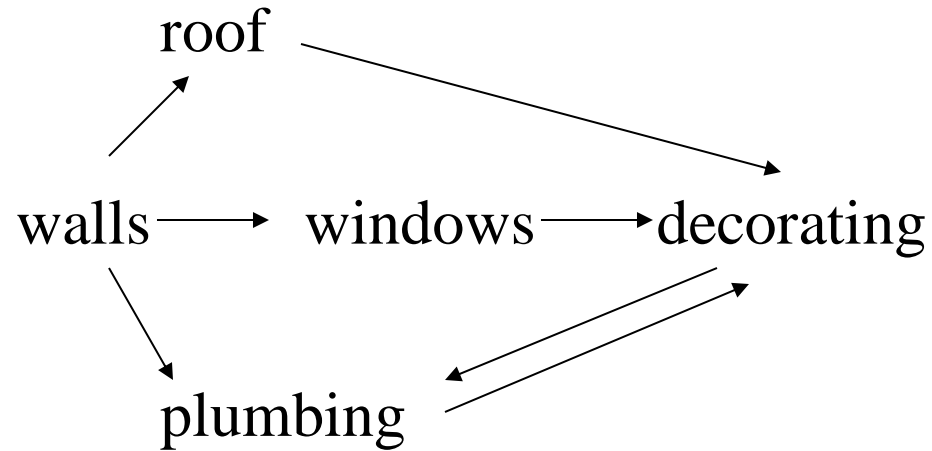
如果你运行拓扑排序，最终还剩下节点未被删除（或未排序），那么图中一定存在环。

Example: building a house with a vicious circle

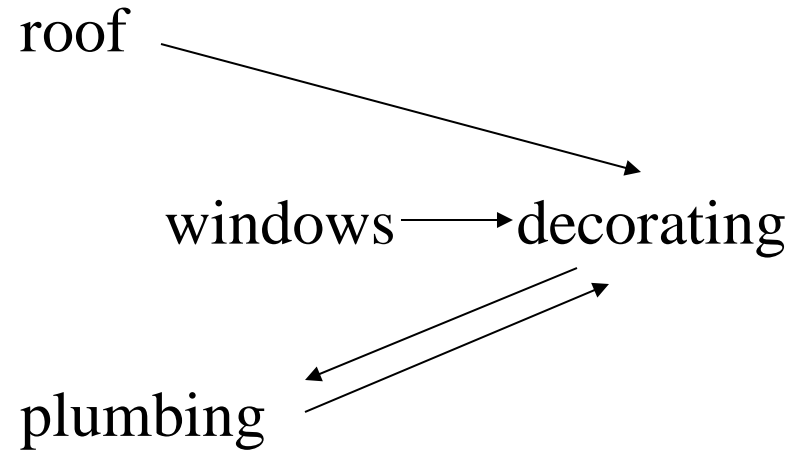


Plumbing depends on decorating and decorating on plumbing

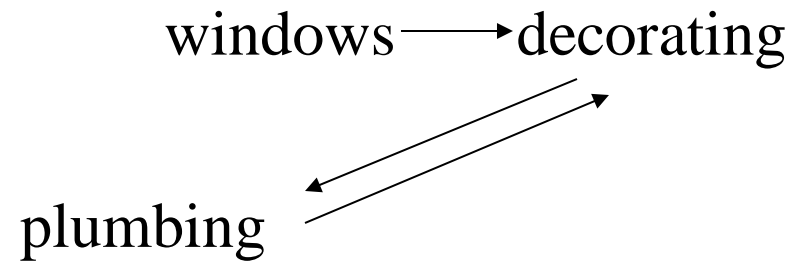
Example: building a house with a vicious circle



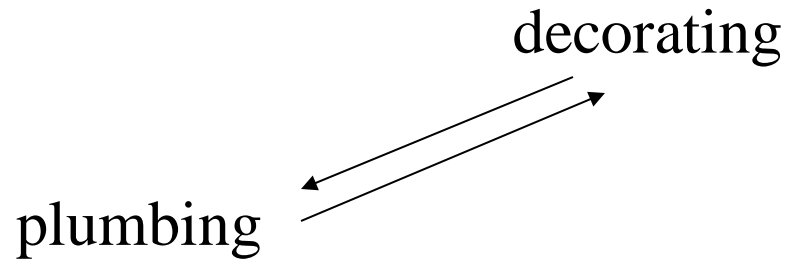
Example: building a house with a vicious circle



Example: building a house with a vicious circle



Example: building a house with a vicious circle



没有任何节点入度为0
拓扑排序无法继续
说明图中存在一个 环 (Cycle)

Stuck!

Why does it work?

- Topological sort: a vertex cannot be removed before all its prerequisites have been removed. So it cannot be inserted in the array before its prerequisite.
- Cycle detection: in a cycle, a vertex is its own prerequisite. So it can never be removed.

| 现象 | 含义 |
|------------------|---------------|
| 节点都能依次放入序列 | 图是有向无环图 (DAG) |
| 某一时刻找不到入度为 0 的节点 | 图中有环 (Cycle) |

Spanning Tree 生成树

一个连通的无向图 (connected, undirected graph)

- *Input*: connected, undirected graph
- *Output*: a tree which connects all vertices in the graph using only the edges present in the graph

一棵树 (无环图) :

连接所有顶点

只使用原图中的边

边数 = 顶点数 - 1 ($n-1$)

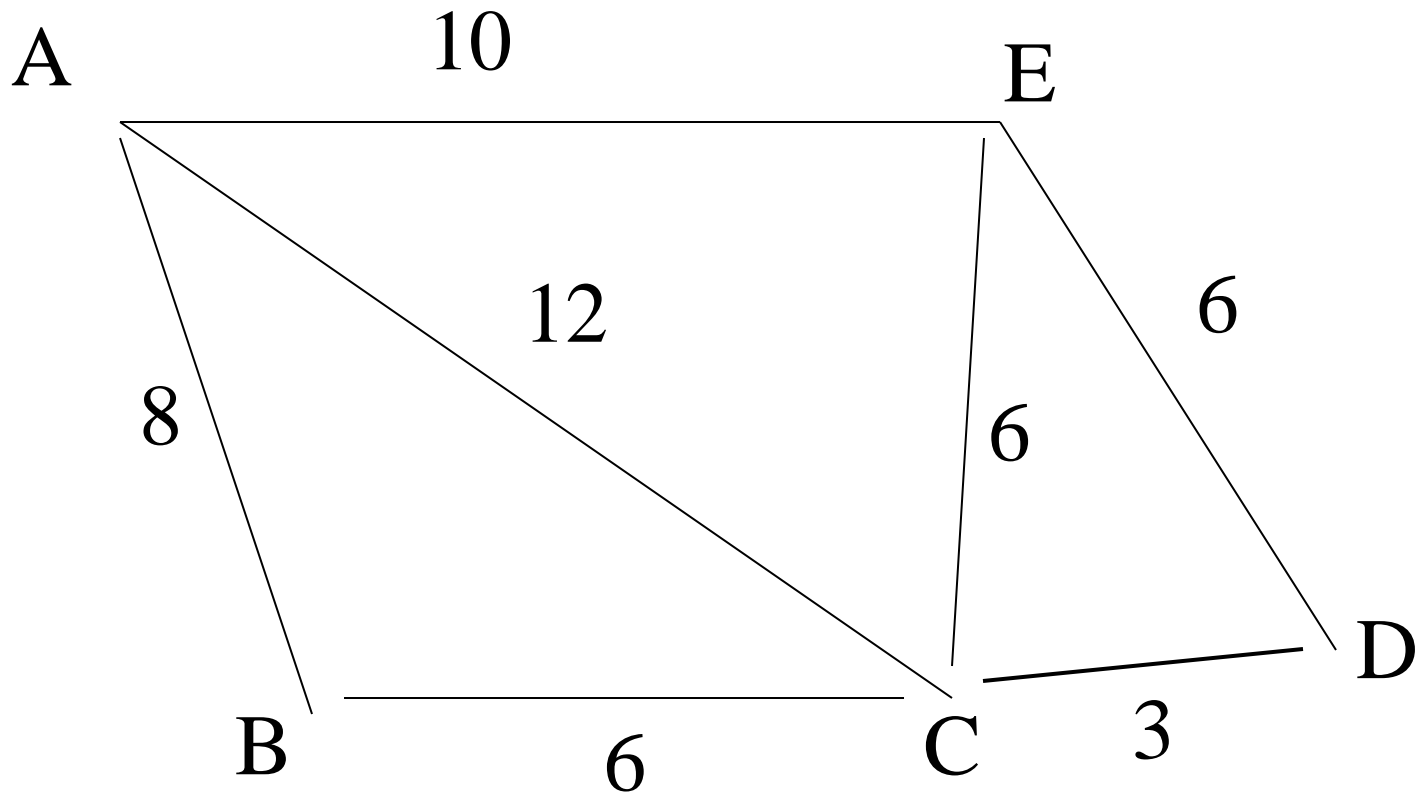
生成树就是从原图中挑出一些边，既能保持图的连通性，又不形成环。

Minimal Spanning Tree 最小生成树

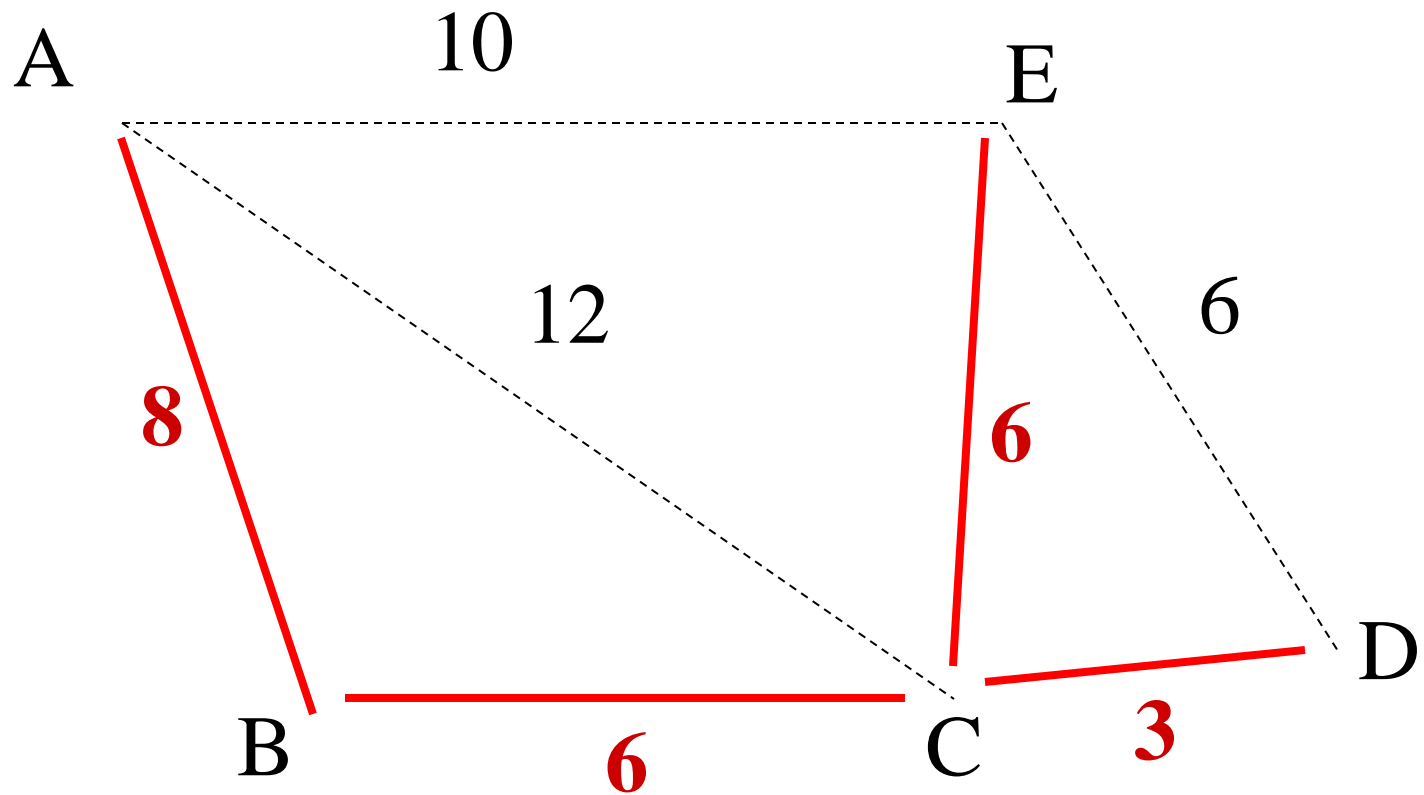
- *Input*: connected, undirected, weighted graph 一个连通的、无向的、有权图
- *Output*: a spanning tree
 - (connects all vertices in the graph using only the edges present in the graph)
 - and is *minimal* in the sense that the sum of weights of the edges is the smallest possible

一棵生成树，满足以下两个条件：
连接图中所有顶点
权重之和最小

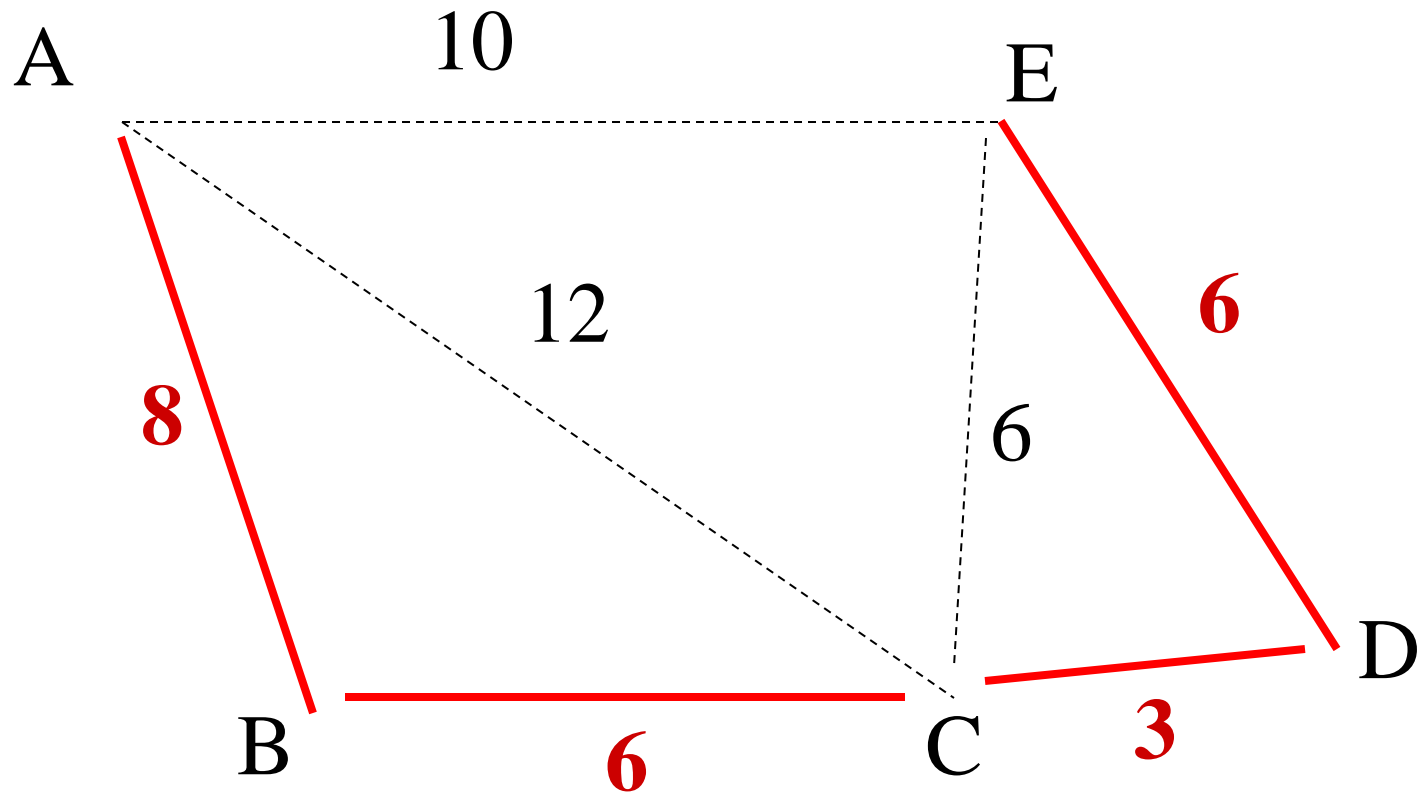
Example: graph



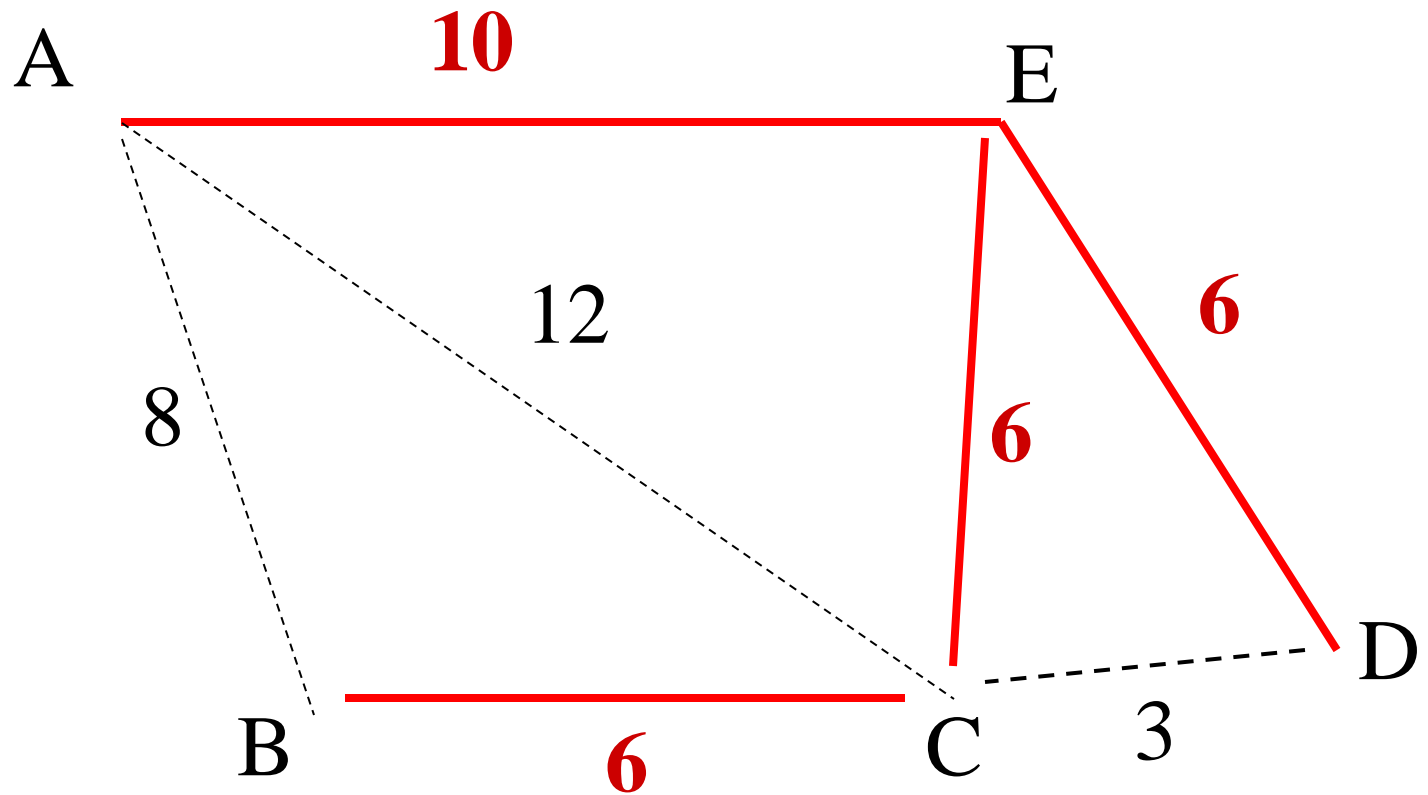
Example: MST (cost 23)



Example: another MST (cost 23)



Example: not MST (cost 28)



我们需要的是最小生成子图（Minimum Spanning Subgraph）：即一个连接所有节点的边的子集。
假设所有权重都是非负的（non-negative）：如果生成的子图中包含环（cycle）：可以删除这个环上的一条边而不破坏连通性；这样会减少总权重，所以原来的图就不是最小的；所以MST 不可能含有环。
如果一个图是连通的、且无环，那么它就是一棵树。

Why MST is a tree

目标：找到一个连通子图，使用图中的边，使得：包含所有顶点；总权重最小；无环（即为树结构）。

- We want a minimum spanning sub-graph
 - a subset of the edges that is connected and that contains every node
- (Assuming all weights are non-negative)
 - If the graph has a cycle, then we can remove an edge of the cycle, and the graph will still be connected and will have a smaller weight.
- If a graph is connected and acyclic, then it is a tree.

选择任意一个顶点 M 作为起点。

从当前生成树中的顶点 M ，选取一条权重最小的边 (M, N) ，使得： M 在当前生成树中； N 不在生成树中。将边 (M, N) 加入到生成树中。

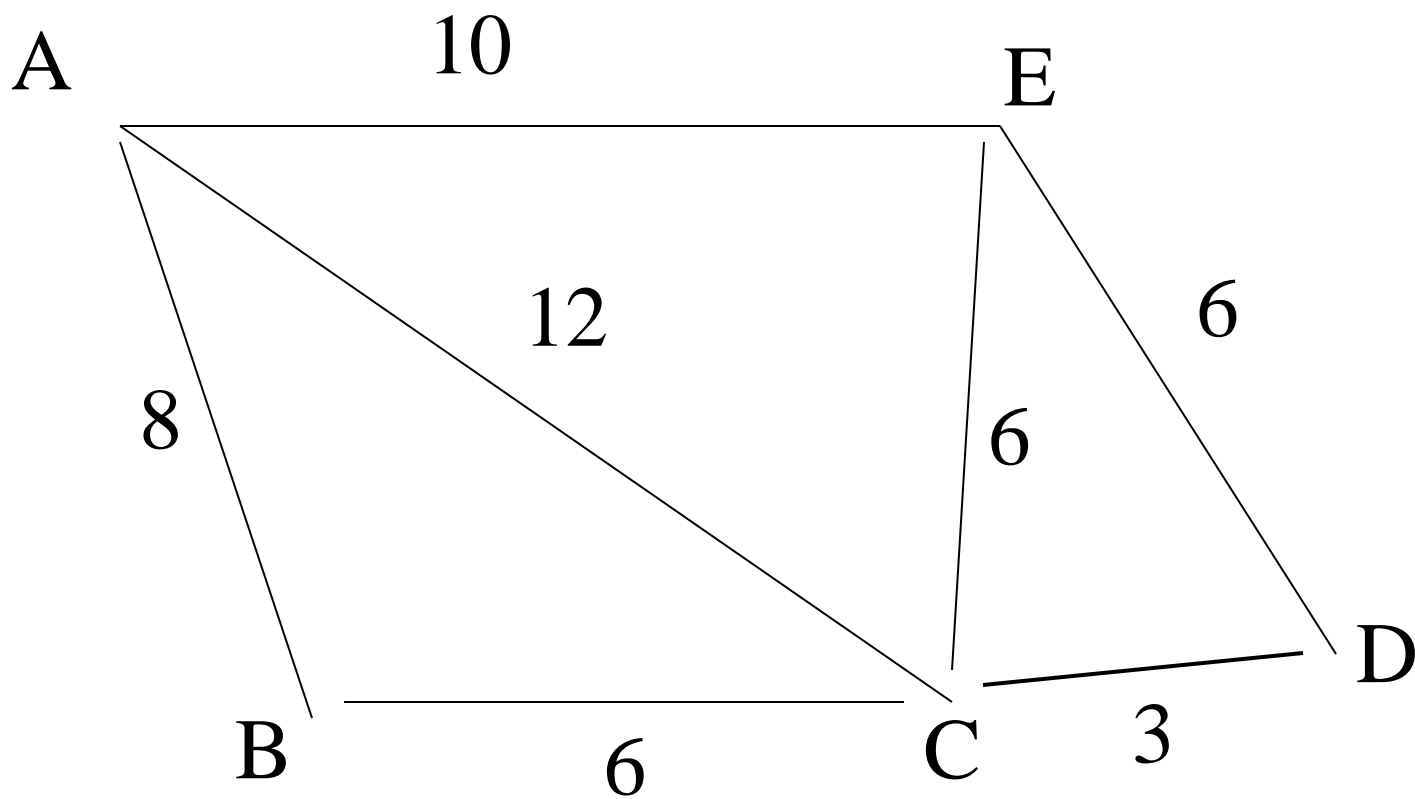
重复步骤 2：每次从生成树中的任意顶点出发；选择连接“树内”和“树外”的最短边；直到所有顶点都加入生成树为止。

Prim's algorithm

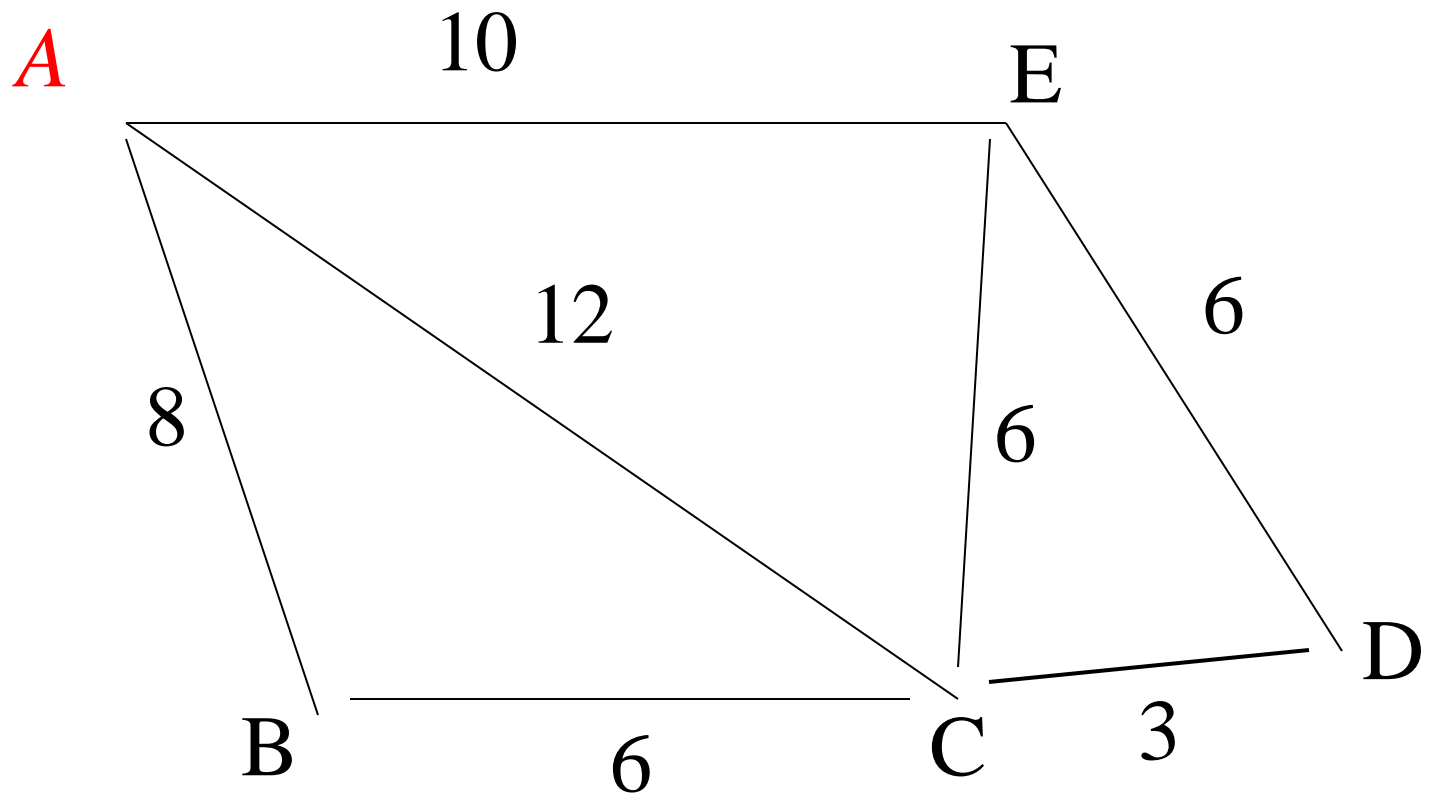
To construct an MST:

- Pick any vertex M
- Choose the shortest edge from M to any other vertex N
- Add the edge (M, N) to the MST
- Continue to add at every step the shortest edge from a vertex in MST to a vertex outside, until all vertices are in MST

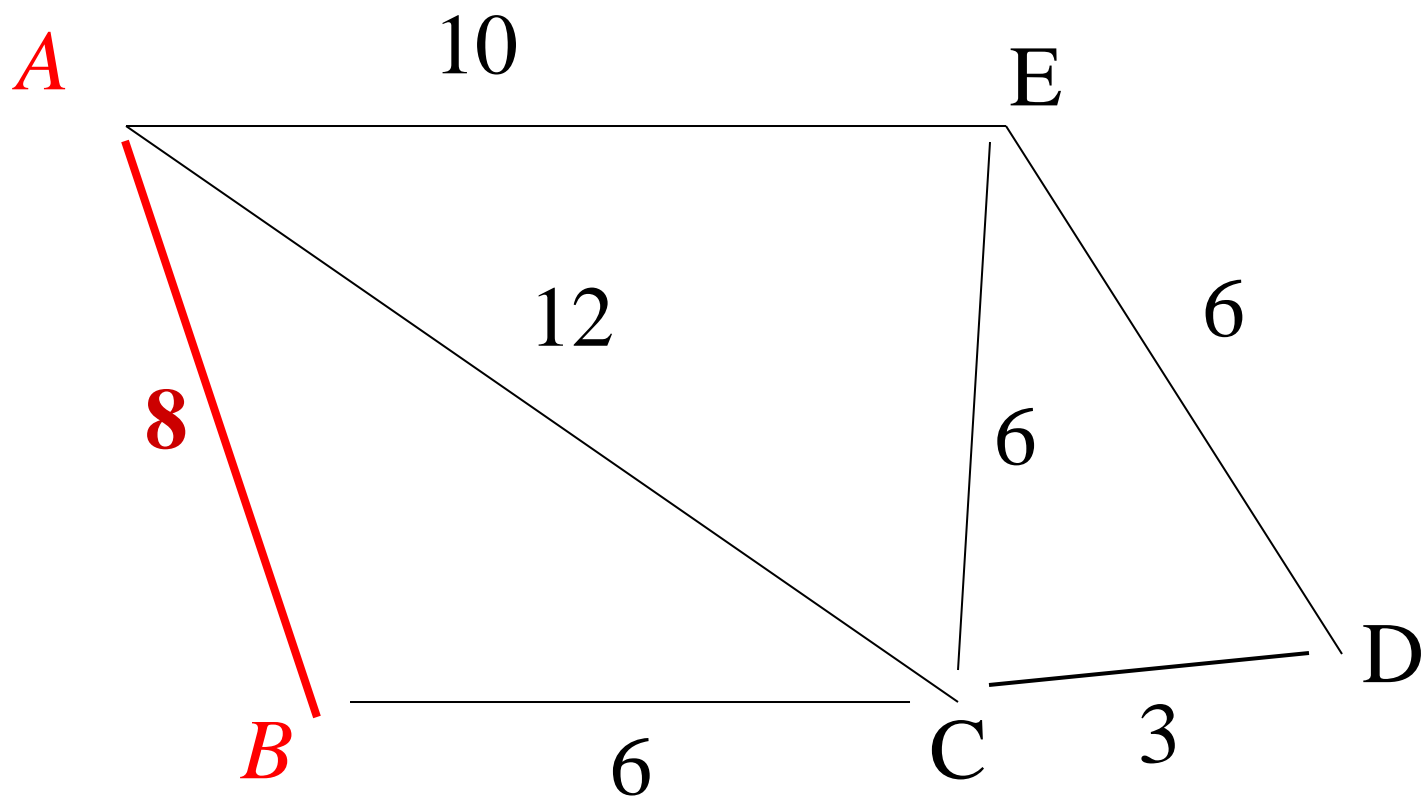
Example



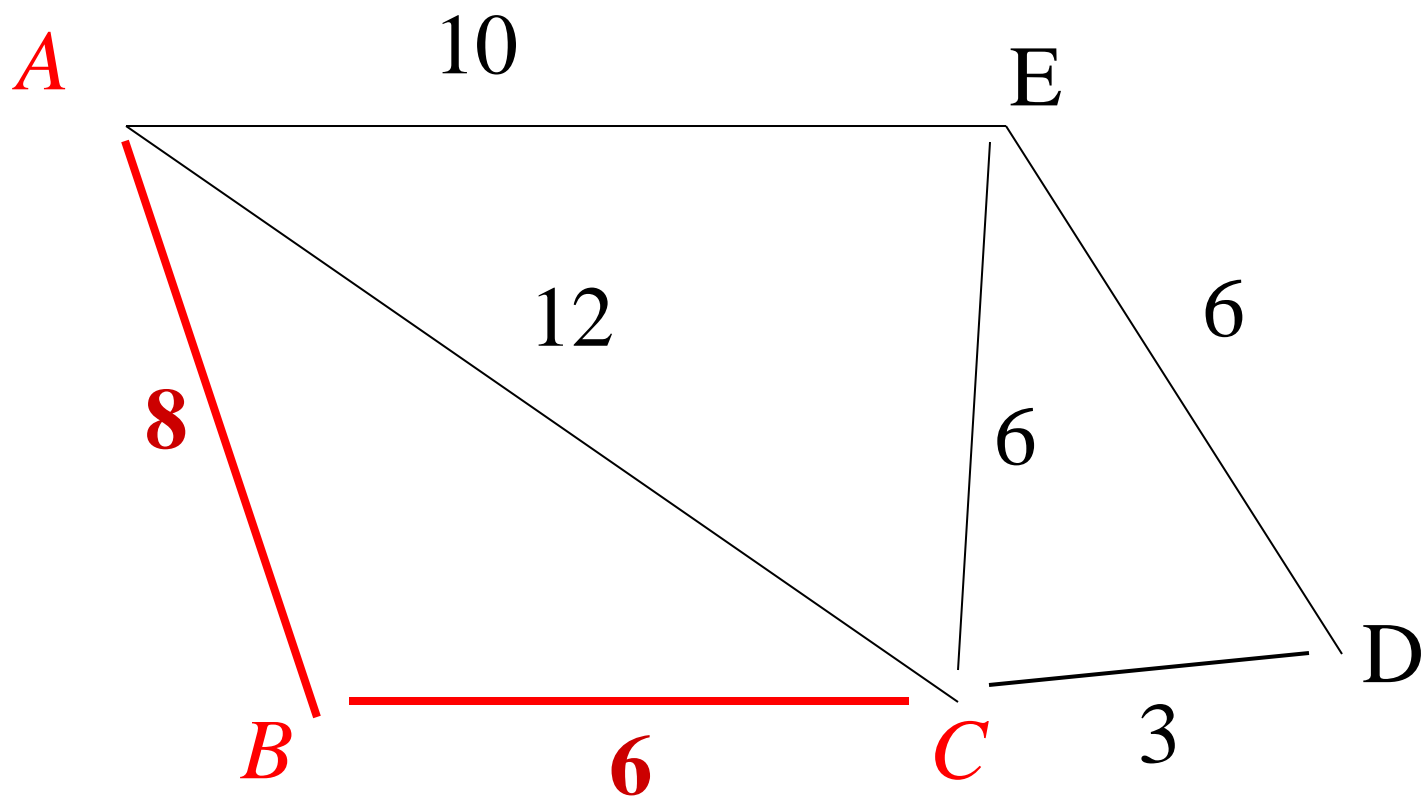
Example



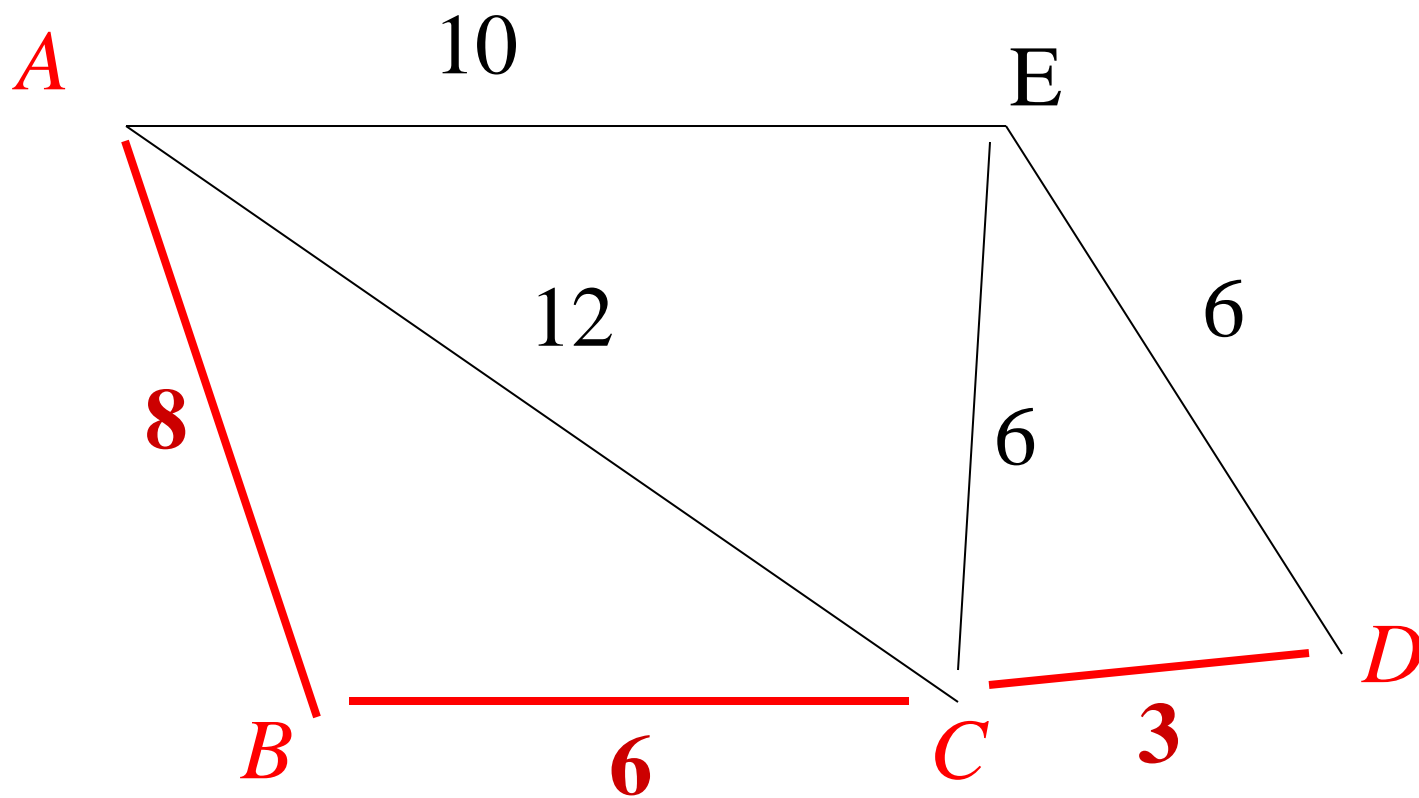
Example



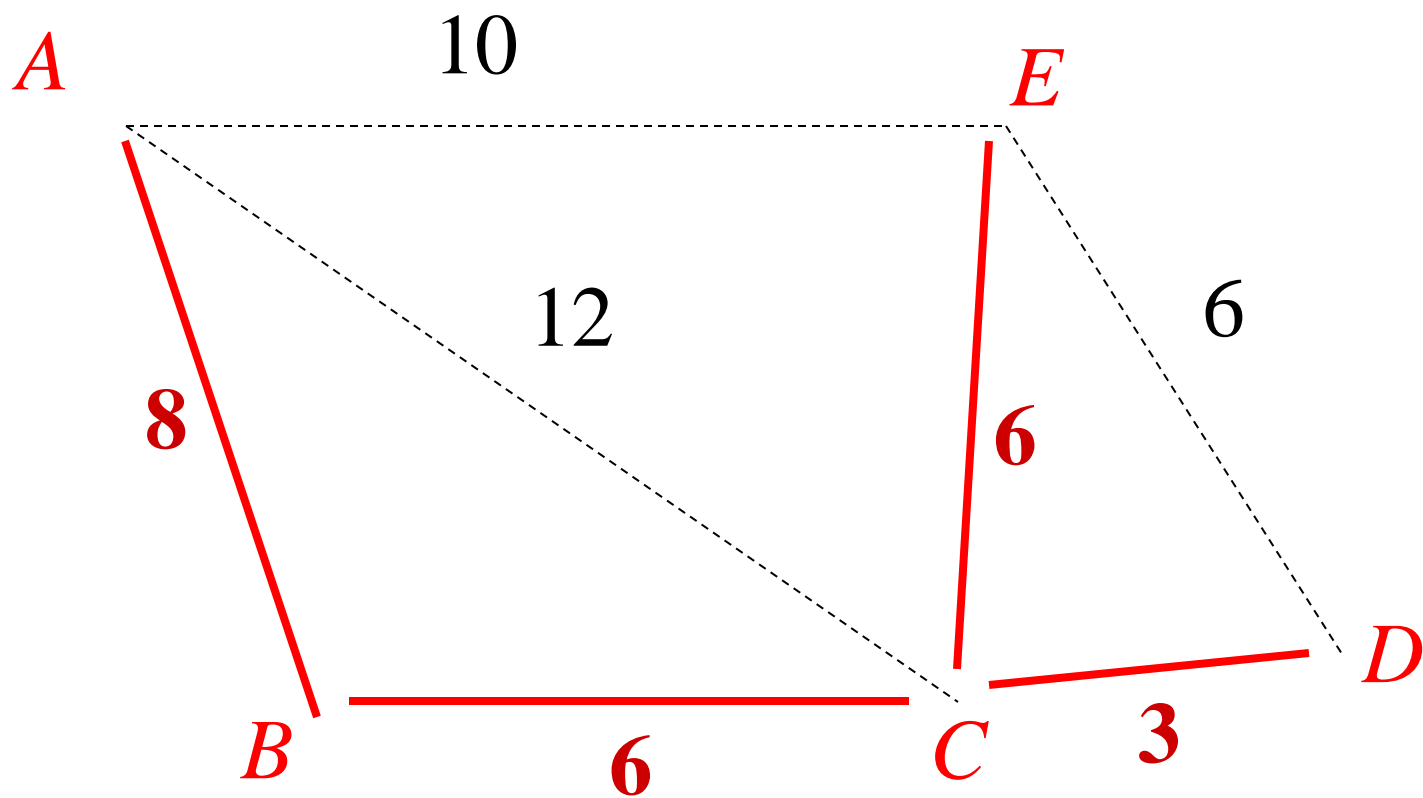
Example



Example



Example



对于一个加权连通图 G ，将其顶点集合划分为两个不相交非空子集 V_1 和 V_2 。如果边 e 是所有连接 V_1 和 V_2 的边中权重最小的一条边，那么：存在一个包含该边 e 的最小生成树 T 。

Correctness of Prim's algorithm

Proposition 1: Let G be a weighted connected graph, and let V_1 and V_2 be a partition of the vertices of G into two disjoint nonempty sets. Furthermore, let e be an edge in G with minimum weight from among those with one endpoint in V_1 and the other in V_2 .
There is a minimum spanning tree T that has e as one of its edges.

Reading Section 14.7 Minimum Spanning Trees

对于图 G 的任意切分 (partition) V_1 和 V_2 , 若 e 是连接 V_1 和 V_2 的最小权重边, 则 存在一个最小生成树 T , 包含这条边 e 。

假设有一个 MST T , 但 不包含这条边 e 。把边 e 加入到 T 中 (这会产生一个环, 因为原来是树, 多加一条边一定成环)。这个环中一定有另一条边 f 连接 V_1 和 V_2 。由于 e 是跨切分的最小边, 说明 $w(e) \leq w(f)$ 。把环中的边 f 移除 (也就是从 $T \cup \{e\}$ 中去掉 f) : 得到的新图仍然是一个生成树; 总权重不超过原来的树。结论: 你得到了另一个权重不更大的生成树, 说明它也是 MST, 且包含了边 e 。

Justification of Proposition 1

Let T be a minimum spanning tree of G . If T does *not* contain edge e , the addition of e to T must create a cycle. Therefore, there is some edge $f \neq e$ of this cycle that has one endpoint in V_1 and the other in V_2 . Moreover, by the choice of e , $w(e) \leq w(f)$. If we remove f from $T \cup \{e\}$, we obtain a spanning tree whose total weight is no more than before. Since T was a minimum spanning tree, this new tree must also be a minimum spanning tree.

Self-Study

Let G be a weighted connected graph, if the weights in G are distinct, then the minimum spanning tree is unique. *Why?*

Reading Section 14.7 Minimum Spanning Trees

Greedy algorithm

Prim's algorithm for constructing a Minimal Spanning Tree is a *greedy algorithm*:

- it just adds the shortest edge
- without worrying about the overall structure, without looking ahead
- It makes a locally optimal choice at each step.

每一步只选当前最短的边，连接一个未被包含的顶点；
不回溯、不重考虑是否未来更优；
局部最优选择最后构建出一个全局最优解（MST）。

Dijkstra 算法的贪心策略：

每次从未访问的节点中选取当前路径最短的点；

这个点被访问后，其路径就是最终最短路径（无法再被更新）；

然后更新其邻居节点的最短距离，继续迭代。

Greedy Algorithms

为什么它是全局最优？

因为每个节点一旦出队（被选中）后就不再更新，说明从起点到它的路径已经是最短路径；

可严格证明：一旦你选出某个点，它的路径不可能再被某条边“绕路改进”。

- Dijkstra's algorithm: pick the vertex to which there is the shortest path currently known at the moment.
- For Dijkstra's algorithm, this also turns out to be globally optimal: can show that a shorter path to the vertex can never be discovered.
- There are also greedy strategies which are not globally optimal.

但并不是所有贪心算法都这样！有些贪心策略看起来聪明，但其实会错过全局最优；

比如在背包问题、调度问题中，一些贪心策略只能得到近似解（不是最优解）。

Example: non-optimal greedy algorithm

- Problem: given a number of coins, count the change in as few coins as possible.
- Greedy strategy: start with the largest coin which is available; for the remaining change, again pick the largest coin; and so on.
- e.g., coins of values 1, 3, 4, 5; change is 7.

贪心算法解法（错误的）：目标：7 先选最大硬币：5 剩 2 剩 2，最大能用的硬币是 1 选两个 1

总共用了：5 + 1 + 1 = 3 枚硬币

正确的最优解：选 3 + 4 恰好 7 总共用了：2 枚硬币

Shortest path

- Find the shortest route between two vertices u and v .
- It turns out that we can just as well compute shortest routes to ALL vertices reachable from u (including v). This is called *single-source shortest path problem* for weighted graphs, and u is the source.

Dijkstra 算法是用于解决 加权图中起点到其他所有点最短路径 的算法，前提是边权重不能为负数。

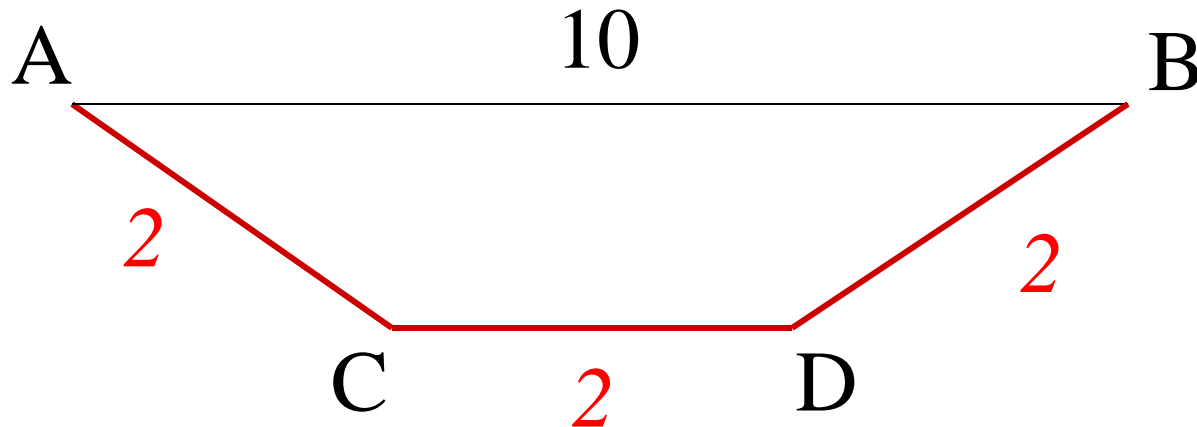
从起点开始，初始化起点到自己的距离为 0，其他为 ∞ 。每次从 未访问的节点中选出当前距离起点最近的节点（即“已知最短路径最小”的点）。以该节点为中心，尝试更新它的邻接点的距离（relax）。重复步骤 2 和 3，直到所有点都被访问。这就是典型的“局部最优 全局最优”的 贪心策略。

Dijkstra's Algorithm

- An algorithm for solving the single-source shortest path problem. Greedy algorithm.
- The first version of the Dijkstra's algorithm (traditionally given in textbooks) returns not the actual path, but a number - the shortest distance between u and v .
- (Assume that weights are distances, and the length of the path is the sum of the lengths of edges.)

Example

- Dijkstra's algorithm should return 6 for the shortest path between A and B:



优先队列 PQ：存放所有待处理的节点，按照当前已知的最短距离排序（最短距离最小的节点在前）。实现方式：可以用 最小堆（Min Heap）。
距离数组 dist[]：用来记录从起点 s 到图中每个节点的当前最短距离。初始化：所有点为 ∞ ，dist[s] = 0。

Dijkstra's algorithm

To find the shortest paths (distances) from the start vertex s :

- keep a priority queue PQ of vertices to be processed
- keep an array with current known shortest distances from s to every vertex (initially set to be infinity for all but s , and 0 for s)
- order the queue so that the vertex with the shortest distance is at the front.

当优先队列 PQ 非空时，重复执行以下步骤：

1. 取出当前最短路径的节点 u (从 PQ 中 dequeue)
2. 对于所有与 u 相邻的点 v (仍在 PQ 中):
 - 如果:

$$\text{distance}(s, v) > \text{distance}(s, u) + \text{weight}(u, v)$$

那么更新:

$$\text{distance}(s, v) = \text{distance}(s, u) + \text{weight}(u, v)$$

并更新 v 在 PQ 中的优先级

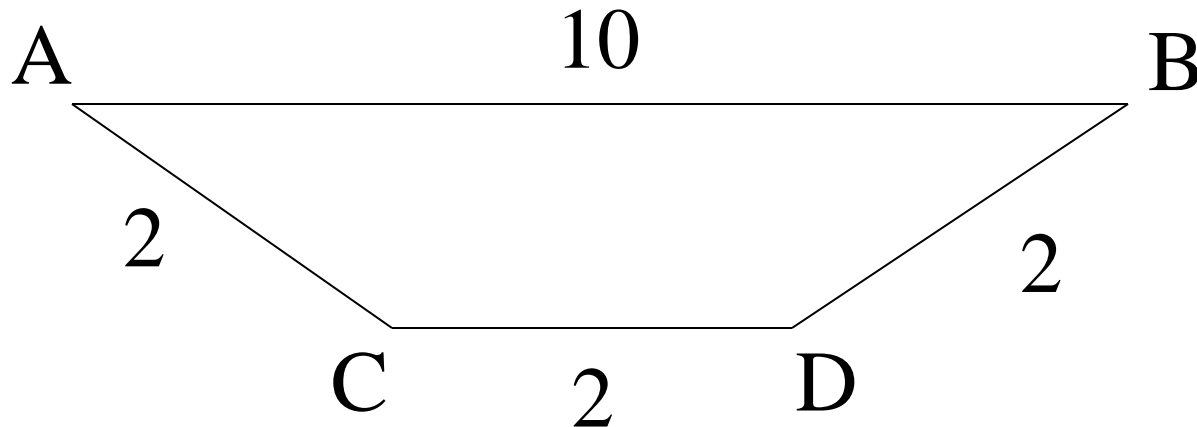
Dijkstra's algorithm

Loop while there are vertices in the queue PQ:

- dequeue a vertex u
- recompute shortest distances for all vertices in the queue as follows: if there is an edge from u to a vertex v in PQ and the current shortest distance to v is greater than $\text{distance}(s, u) + \text{weight}(u, v)$ then replace $\text{distance}(s, v)$ with $\text{distance}(s, u) + \text{weight}(u, v)$.

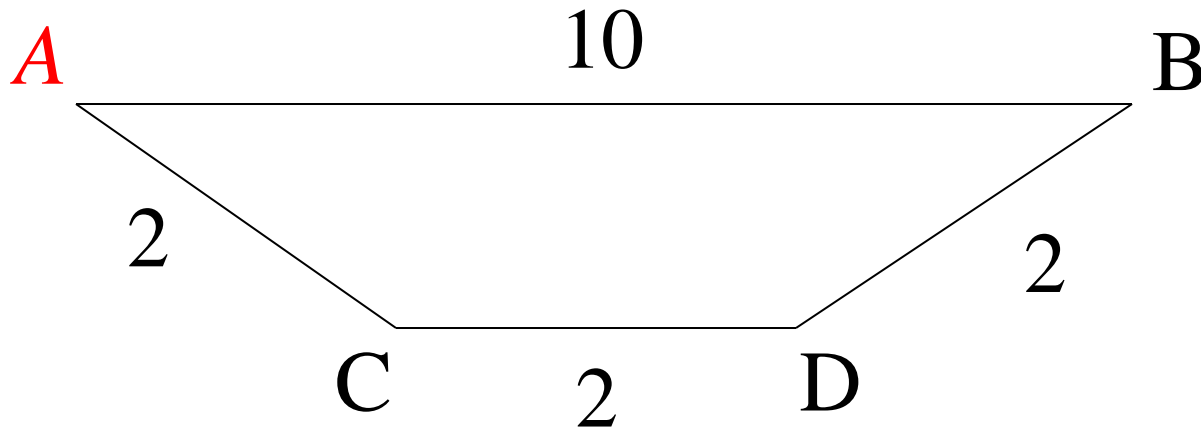
Example

- Distances: (A,0), (B,INF), (C,INF), (D,INF)
- $PQ = \{A,B,C,D\}$



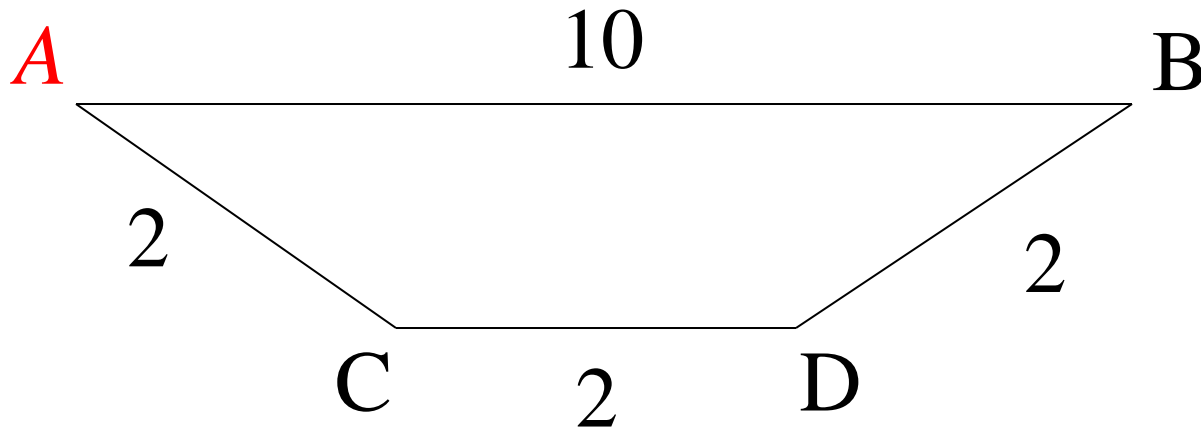
Example (dequeue A)

- Distances: (A,0), (B,INF), (C,INF), (D,INF)
- $PQ = \{B, C, D\}$



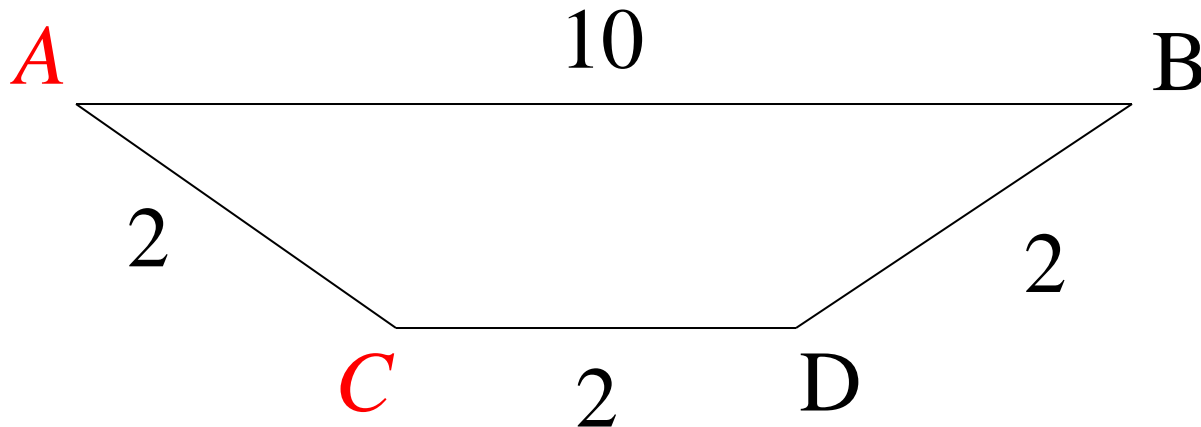
Example (recompute distances)

- Distances: (A,0), (B,10), (C,2), (D,INF)
- PQ= {C,B,D}



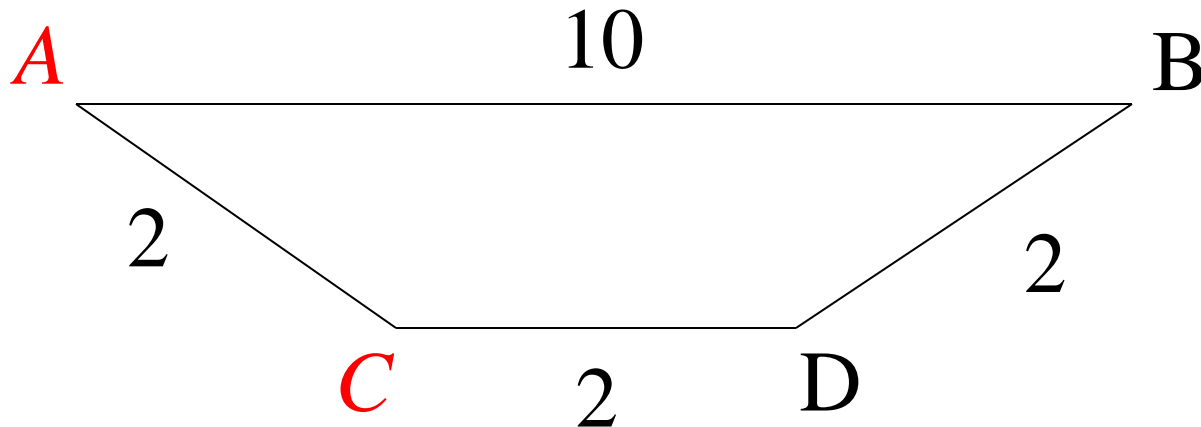
Example (dequeue C)

- Distances: (A,0), (B,10), (C,2), (D,INF)
- $PQ = \{B,D\}$



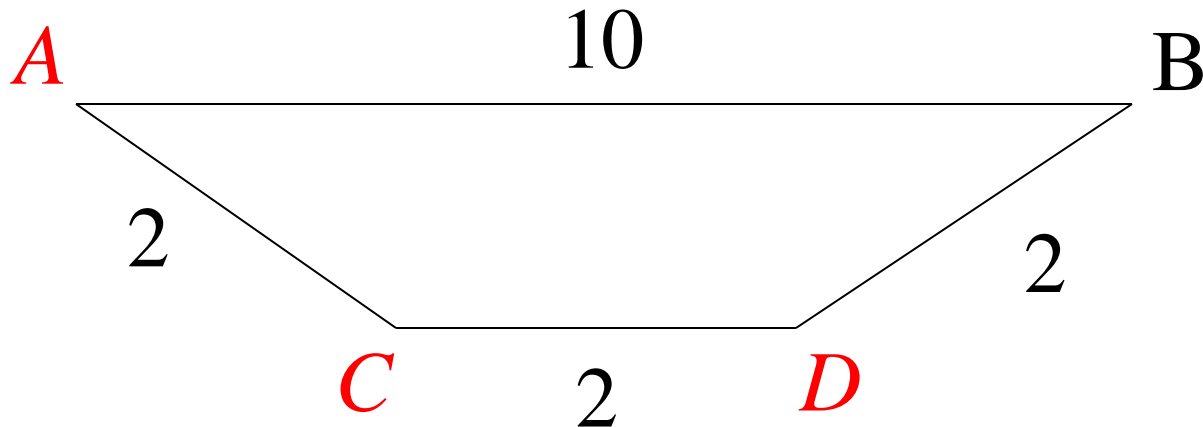
Example (recompute distances)

- Distances: (A,0), (B,10), (C,2), (D,4)
- $PQ = \{D, B\}$



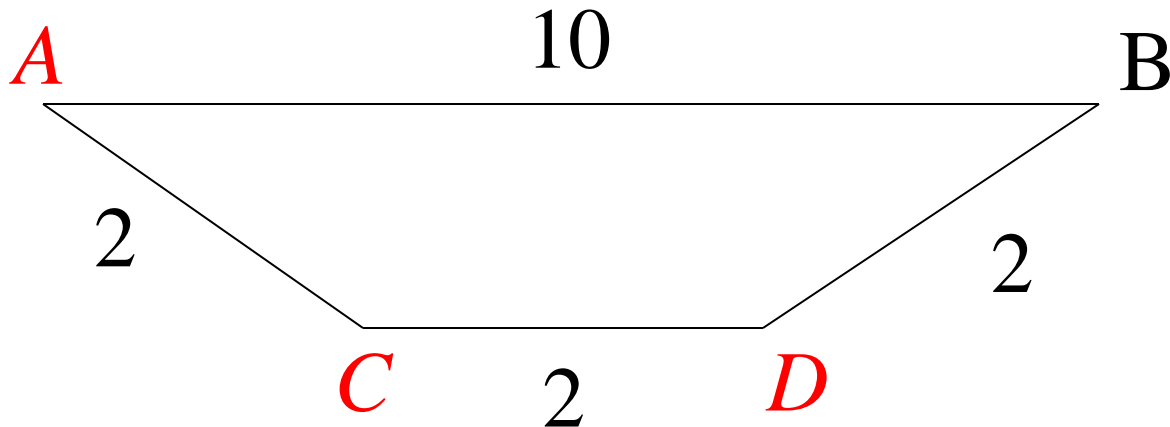
Example (dequeue D)

- Distances: (A,0), (B,10), (C,2), (D,4)
- $PQ = \{B\}$



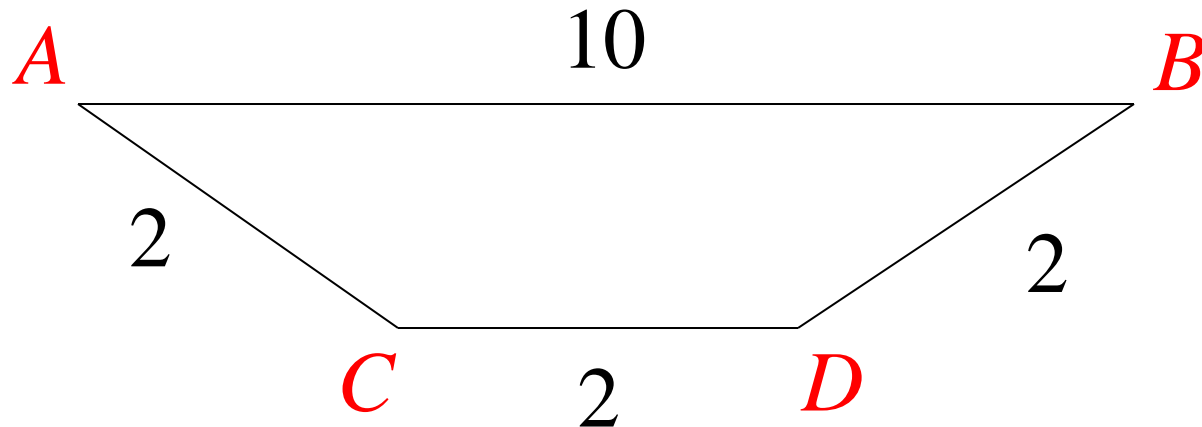
Example (recompute distances)

- Distances: (A,0), (B,6), (C,2), (D,4)
- $PQ = \{B\}$



Example (dequeue B)

- Distances: (A,0), (B,6), (C,2), (D,4)
- $PQ = \{\}$



Pseudocode for D's Algorithm

表示“无穷大”，比图中任何可能的边权重都大。用于初始化起点以外所有点的最短路径值（*dist*），表示尚未可达。

- *INF* is supposed to be greater than any number
- *dist* : array holding shortest distances from source *S* 是一个数组，记录从起点 *s* 到图中每个顶点的当前最短距离。
- *PQ* : priority queue of unvisited vertices prioritised by shortest recorded distance from source 优先队列，保存“尚未确定最短路径”的所有顶点。以当前 *dist* 值为优先级，小的先出队。每次取出的是“当前距离最小”的顶点，进行扩展（relaxation）。
- *PQ.reorder()* reorders *PQ* if the values in *dist* change. 如果某个顶点的最短距离（*dist*[*v*]）在过程中更新了，优先队列需要重新排列，确保新的最短距离的顶点能在下一轮被优先处理。

Pseudocode for Dijkstra's Algorithm

```
for (each v in V) {  
    dist[v] = INF;  
    dist[s] = 0;  
}  
PriorityQueue PQ = new PriorityQueue();  
// insert all vertices in PQ,  
// in reverse order of dist[]  
// values
```

Pseudocode for D's Algorithm

```
while (! PQ.isEmpty()) {  
    u = PQ.dequeue();  
    for (each v in PQ adjacent to u) {  
        if (dist[v] > (dist[u] + weight(u, v))) {  
            dist[v] = (dist[u] + weight(u, v));  
        }  
    }  
    PQ.reorder();  
}  
return dist;
```

Modified algorithm

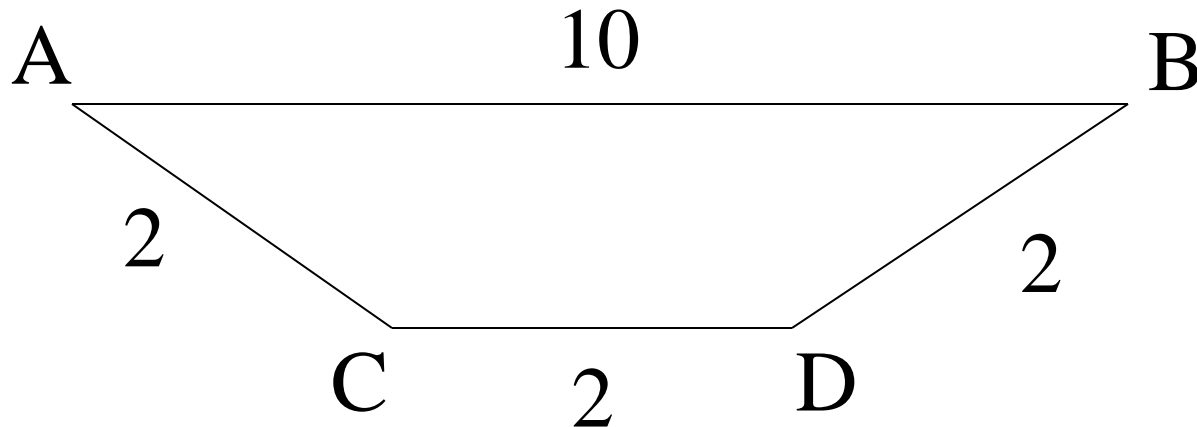
To make Dijkstra's algorithm to return the path itself, not just the distance:

- In addition to distances, maintain a path (list of vertices) for every vertex.
- At the beginning, paths are empty.
- When assigning $dist(s, v) = dist(s, u) + weight(u, v)$, also assign $path(v) = path(u)$.
- When dequeuing a vertex, add it to its path.

Example

Distances and paths:

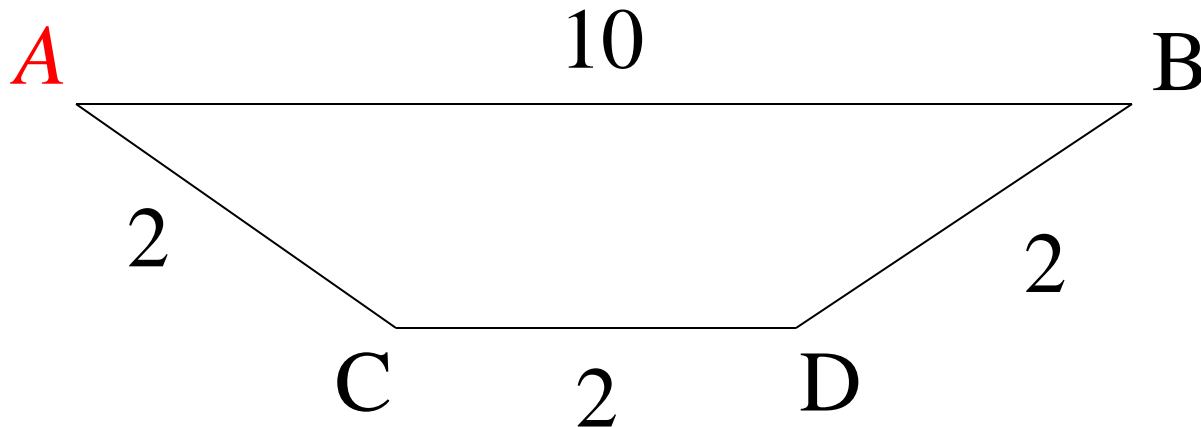
$(A, 0, \{\})$, $(B, \text{INF}, \{\})$, $(C, \text{INF}, \{\})$, $(D, \text{INF}, \{\})$



Deque A, recompute paths

Distances and paths:

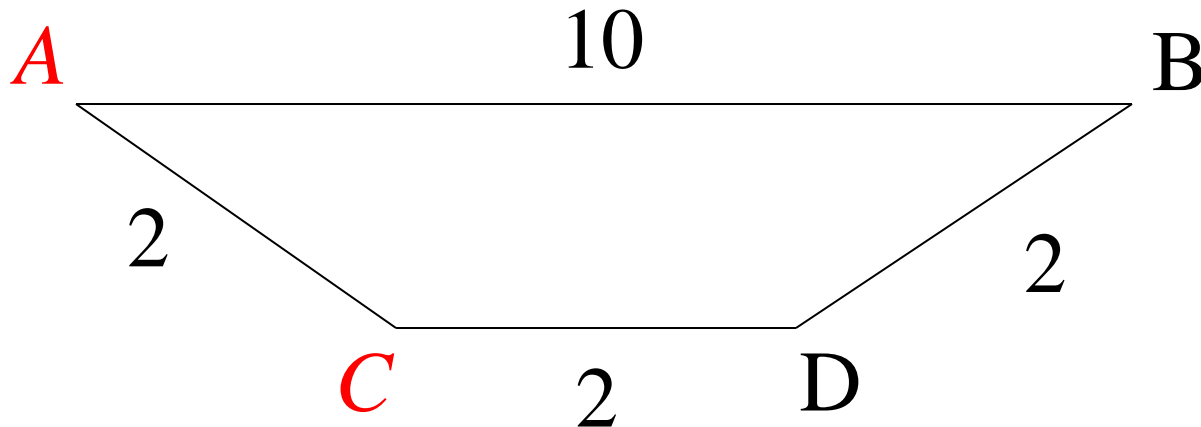
$(A, 0, \{A\})$, $(B, 10, \{A\})$, $(C, 2, \{A\})$, $(D, \text{INF}, \{\})$



Deque C, recompute paths

Distances and paths:

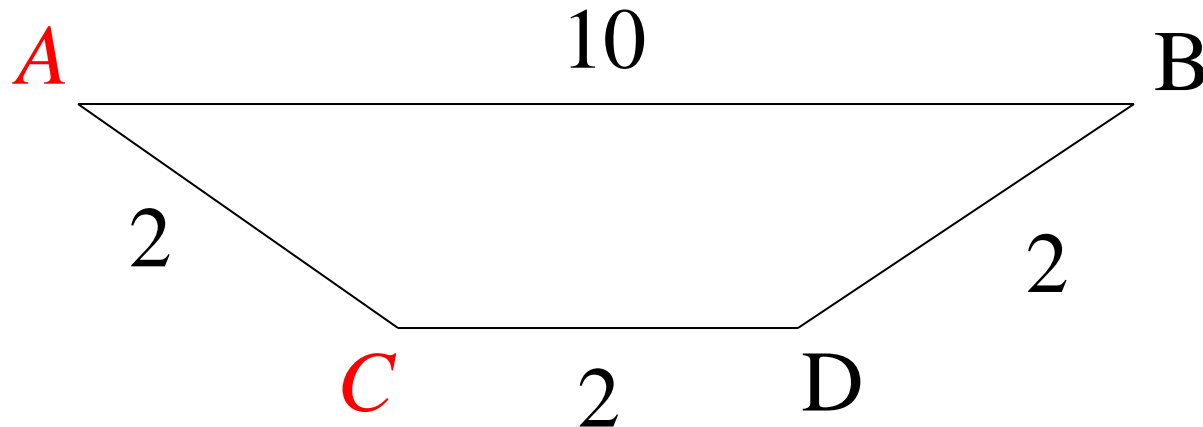
$(A, 0, \{A\})$, $(B, 10, \{A\})$, $(C, 2, \{A, C\})$, $(D, \text{INF}, \{\})$



Deque C, recompute paths

Distances and paths:

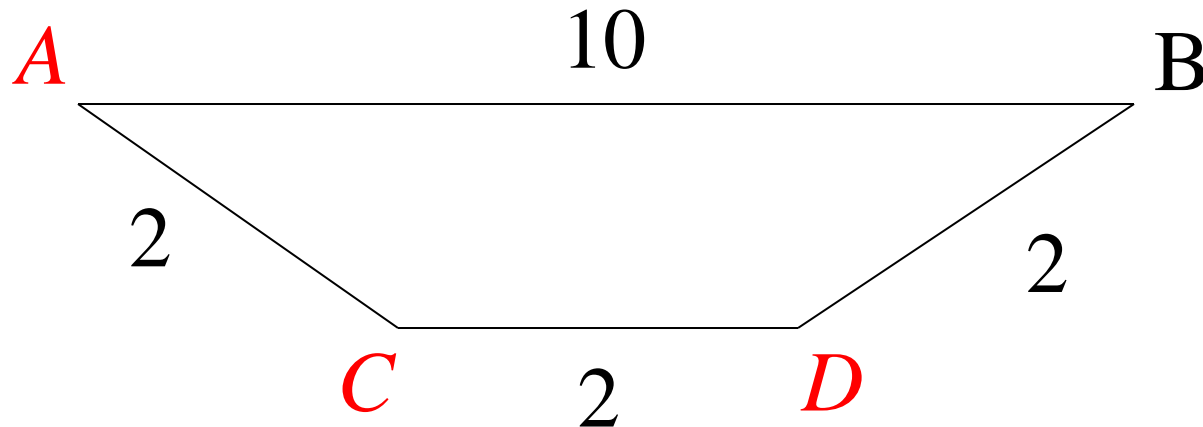
$(A, 0, \{A\})$, $(B, 10, \{A\})$, $(C, 2, \{A, C\})$, $(D, 4, \{A, C\})$



Deque D, recompute paths

Distances and paths:

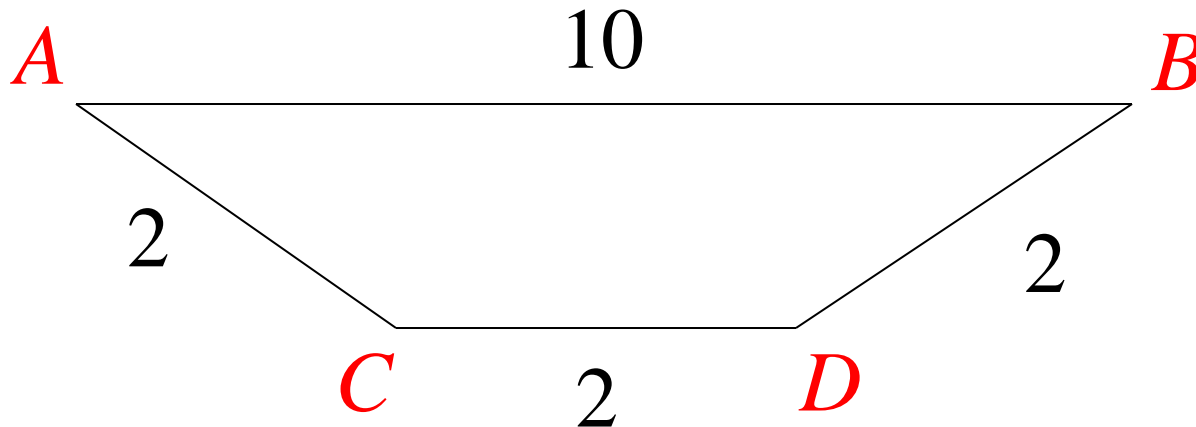
$(A, 0, \{A\})$, $(B, 6, \{A, C, D\})$, $(C, 2, \{A, C\})$,
 $(D, 4, \{A, C, D\})$



Deque B, recompute paths

Distances and paths:

$(A, 0, \{A\})$, $(B, 6, \{A, C, D, B\})$, $(C, 2, \{A, C\})$,
 $(D, 4, \{A, C, D\})$



Optimality of Dijkstra's algorithm

So, why is Dijkstra's algorithm optimal
(gives the shortest path)?

Let us first see where it *could* go wrong.

What the algorithm does

- For every vertex in the priority queue, we keep updating the current distance downwards, until we remove the vertex from the queue.
- After that the shortest distance for the vertex is set.
- What if a shorter path can be discovered later?

Optimality proof

基础情况 (Base Case) :

起点 s 的最短距离是 0 ($\text{dist}(s) = 0$) , 这是由算法初始化设置的, 显然正确。

- Base case: the shortest distance to the start node is set correctly (0)
- Inductive step: assume that the shortest distances are set correctly for the first n vertices removed from the queue. Show that it will also be set correctly for the $(n + 1)$ vertex.

归纳步骤 (Inductive Step) :

假设已经正确找到了前 n 个被移除顶点的最短距离 (即, 它们出队时, 其最短距离已经固定) ;

现在我们要证明: 第 n

$+1$ 个出队顶点 u 的距离也是正确的。

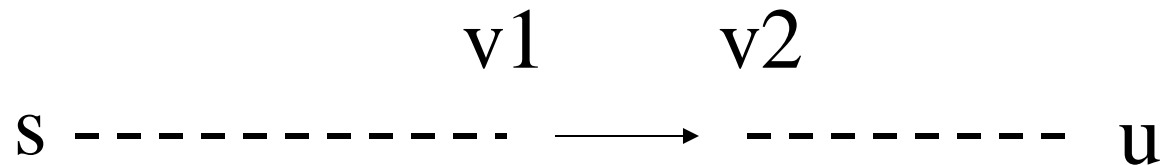
Optimality proof

Assume that the $(n + 1)$ vertex is u . It is at the front of the priority queue and its current known shortest distance is $\text{dist}(s, u)$. We need to show that there is no path in the graph from s to u with the length smaller than $\text{dist}(s, u)$.

我们假设当前是第 $n+1$ 个顶点 u ，它现在位于优先队列最前面，我们已经计算得出当前最短距离为：
 $\text{dist}(s, u)$
我们需要证明：不存在任何从源点 s 到 u 的路径，其路径长度小于 $\text{dist}(s, u)$ 。

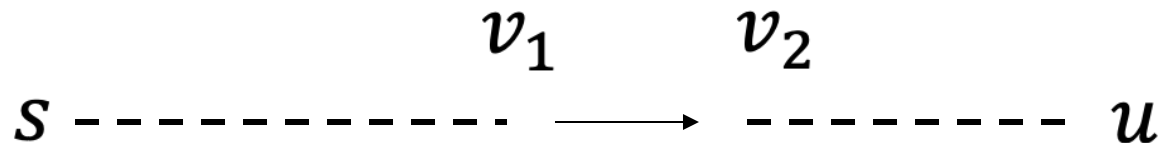
Optimality proof

Proof by contradiction: assume there is such a (shorter) path:



Optimality proof

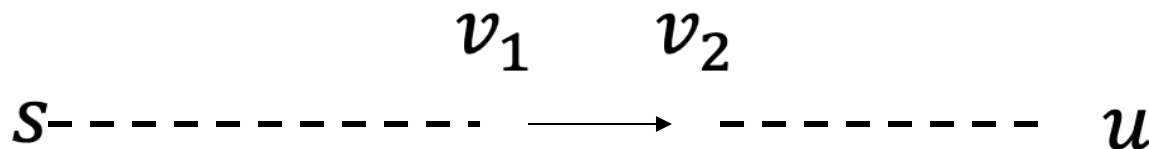
Here the vertices from s to v_1 have correct shortest distances (inductive hypothesis) and v_2 is still in the priority queue.



Optimality proof

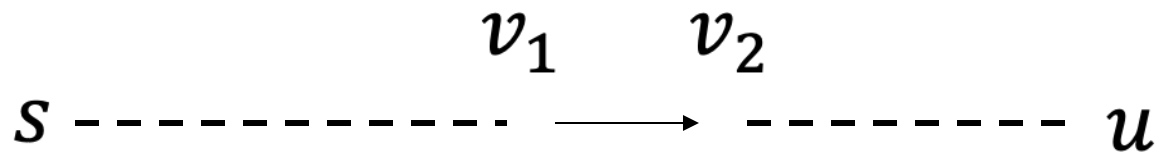
So $\text{dist}(s, v_1)$ is indeed the shortest path from s to v_1 . Current distance to v_2 is

$$\text{dist}(s, v_2) = \text{dist}(s, v_1) + \text{weight}(v_1, v_2).$$



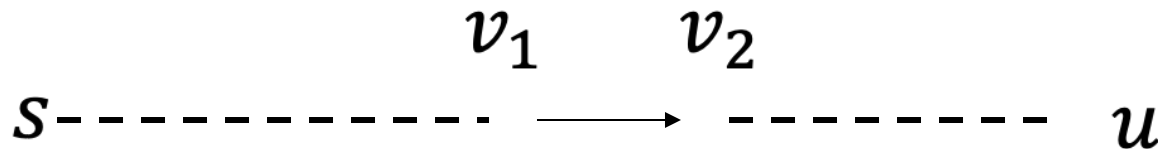
Optimality proof

If v_2 is still in the priority queue, then
 $\text{dist}(s, v_1) + \text{weight}(v_1, v_2) \geq \text{dist}(s, u)$.



Optimality proof

But then the path going through v_1 and v_2 cannot be shorter than $\text{dist}(s, u)$. QED




对于每一个出队的顶点 u ，我们需要：

- 访问从 u 出发的所有边 E_u
- 并对每一条边进行松弛 (relax) 操作
- 每次操作涉及堆的更新：代价为 $\log |V|$

所以单步开销是：

$$O(|E_u| \cdot \log |V|)$$

Complexity

 总体复杂度：

- 所有边的总访问次数是 $\sum_u |E_u| = |E|$
- 每个顶点最多出队一次，最多执行 $|V|$ 次堆操作

所以总复杂度是：

$$O((|V| + |E|) \cdot \log |V|)$$

- Assume that the priority queue is implemented as a heap;
- At each step (dequeuing a vertex u and recomputing distances) we do $O(|E_u| \cdot \log(|V|))$ work, where E_u is the set of edges with source u .
- We do this for every vertex, so total complexity is $O((|V| + |E|) \cdot \log(|V|))$.
- Really similar to BFS and DFS, but instead of choosing some successor, we re-order a priority queue at each step, hence the $\cdot \log(|V|)$ factor.