

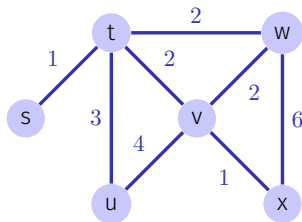
14. Minimum Spanning Trees

Motivation, Greedy, Algorithm Kruskal, General Rules, ADT Union-Find, Algorithm Jarnik, Prim, Dijkstra [Ottman/Widmayer, Kap. 9.6, 6.2, 6.1, Cormen et al, Kap. 23, 19]

Problem

Given: Undirected, weighted, connected graph $G = (V, E, c)$.

Wanted: Minimum Spanning Tree $T = (V, E')$: connected, cycle-free subgraph $E' \subset E$, such that $\sum_{e \in E'} c(e)$ minimal.



Application Examples

- Network-Design: find the cheapest / shortest network that connects all nodes.
- Approximation of a solution of the travelling salesman problem: find a round-trip, as short as possible, that visits each node once.¹⁸

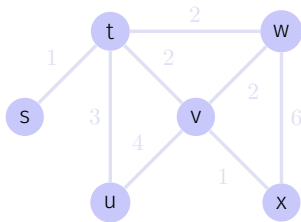
¹⁸The best known algorithm to solve the TS problem exactly has exponential running time.

Greedy Procedure

- Greedy algorithms compute the solution stepwise choosing locally optimal solutions.
- Most problems cannot be solved with a greedy algorithm.
- The Minimum Spanning Tree problem can be solved with a greedy strategy.

Greedy Idea (Kruskal, 1956)

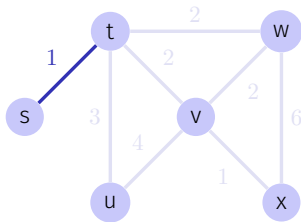
Construct T by adding the cheapest edge that does not generate a cycle.



(Solution is not unique.)

Greedy Idea (Kruskal, 1956)

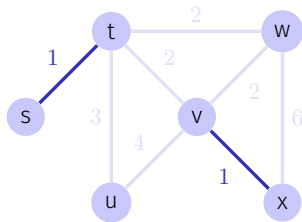
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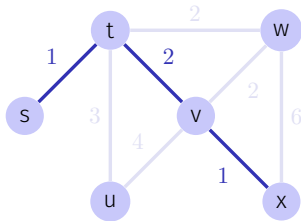
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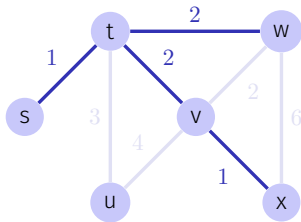
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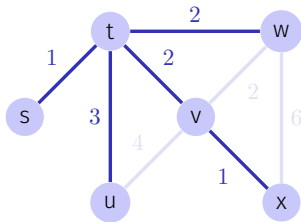
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Algorithm MST-Kruskal(G)

Input: Weighted Graph $G = (V, E, c)$

Output: Minimum spanning tree with edges A .

Sort edges by weight $c(e_1) \leq \dots \leq c(e_m)$

$A \leftarrow \emptyset$

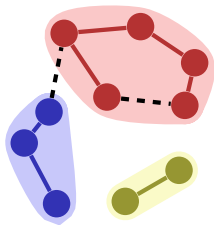
for $k = 1$ **to** $|E|$ **do**

if $(V, A \cup \{e_k\})$ acyclic **then**
 $A \leftarrow A \cup \{e_k\}$

return (V, A, c)

Implementation Issues

Consider a set of sets $i \equiv A_i \subset V$. To identify cuts and cycles: membership of the both ends of an edge to sets?



Implementation Issues

General problem: partition (set of subsets) .e.g.

$\{\{1, 2, 3, 9\}, \{7, 6, 4\}, \{5, 8\}, \{10\}\}$

Required: Abstract data type “Union-Find” with the following operations

- $\text{Make-Set}(i)$: create a new set represented by i .
- $\text{Find}(e)$: name of the set i that contains e .
- $\text{Union}(i, j)$: union of the sets with names i and j .

Union-Find Algorithm MST-Kruskal(G)

Input: Weighted Graph $G = (V, E, c)$

Output: Minimum spanning tree with edges A .

Sort edges by weight $c(e_1) \leq \dots \leq c(e_m)$

$A \leftarrow \emptyset$

for $k = 1$ **to** $|V|$ **do**

\lfloor MakeSet(k)

for $k = 1$ **to** m **do**

$(u, v) \leftarrow e_k$

if Find(u) \neq Find(v) **then**

 Union(Find(u), Find(v))

$A \leftarrow A \cup e_k$

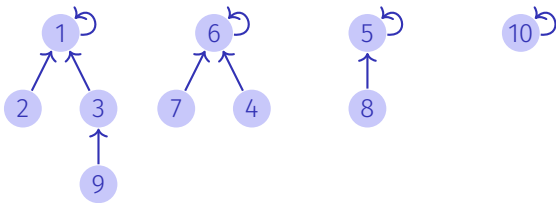
else

// conceptual: $R \leftarrow R \cup e_k$

return (V, A, c)

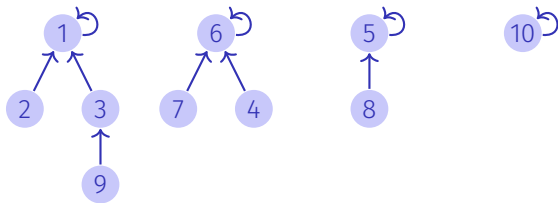
Implementation Union-Find

Idea: tree for each subset in the partition, e.g.
 $\{\{1, 2, 3, 9\}, \{7, 6, 4\}, \{5, 8\}, \{10\}\}$



roots = names (representatives) of the sets,
trees = elements of the sets

Implementation Union-Find



Representation as array:

| | | | | | | | | | | |
|--------|---|---|---|---|---|---|---|---|---|----|
| Index | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 |
| Parent | 1 | 1 | 1 | 6 | 5 | 6 | 5 | 5 | 3 | 10 |

Implementation Union-Find

| | | | | | | | | | | |
|--------|---|---|---|---|---|---|---|---|---|----|
| Index | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 |
| Parent | 1 | 1 | 1 | 6 | 5 | 6 | 5 | 5 | 3 | 10 |

Make-Set(i) $p[i] \leftarrow i$; **return** i

Find(i) **while** ($p[i] \neq i$) **do** $i \leftarrow p[i]$
 return i

Union(i, j)¹⁹ $p[j] \leftarrow i$;

¹⁹ i and j need to be names (roots) of the sets. Otherwise use Union(Find(i),Find(j))

Optimisation of the runtime for Find

Tree may degenerate. Example: Union(8, 7), Union(7, 6), Union(6, 5), ...

| | | | | | | | | | |
|--------|---|---|---|---|---|---|---|---|----|
| Index | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | .. |
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Worst-case running time of Find in $\Theta(n)$.

Optimisation of the runtime for Find

Idea: always append smaller tree to larger tree. Requires additional size information (array) g

Make-Set(i) $p[i] \leftarrow i; g[i] \leftarrow 1; \mathbf{return} \ i$

Union(i, j) **if** $g[j] > g[i]$ **then** swap(i, j)
 $p[j] \leftarrow i$
 if $g[i] = g[j]$ **then** $g[i] \leftarrow g[i] + 1$

⇒ Tree depth (and worst-case running time for Find) in $\Theta(\log n)$

Further improvement

Link all nodes to the root when Find is called.

Find(i):

$j \leftarrow i$

while ($p[i] \neq i$) **do** $i \leftarrow p[i]$

while ($j \neq i$) **do**

$t \leftarrow j$
 $j \leftarrow p[j]$
 $p[t] \leftarrow i$

return i

Cost: amortised *nearly* constant (inverse of the Ackermann-function).²⁰

²⁰We do not go into details here.

Running time of Kruskal's Algorithm

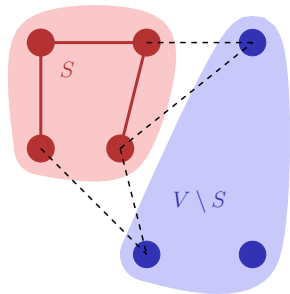
- Sorting of the edges: $\Theta(|E| \log |E|) = \Theta(|E| \log |V|)$.²¹
 - Initialisation of the Union-Find data structure $\Theta(|V|)$
 - $|E| \times \text{Union}(\text{Find}(x), \text{Find}(y))$: $\mathcal{O}(|E| \log |E|) = \mathcal{O}(|E| \log |V|)$.
- Overall $\Theta(|E| \log |V|)$.

²¹because G is connected: $|V| \leq |E| \leq |V|^2$

Algorithm of Jarnik (1930), Prim, Dijkstra (1959)

Idea: start with some $v \in V$ and grow the spanning tree from here by the acceptance rule.

```
A ← ∅  
S ← {v0}  
for  $i \leftarrow 1$  to  $|V|$  do  
    Choose cheapest  $(u, v)$  mit  $u \in S, v \notin S$   
    A ← A ∪ {(u, v)}  
    S ← S ∪ {v} // (Coloring)
```



Remark: a union-Find data structure is not required. It suffices to color nodes when they are added to S .

Running time

Trivially $\mathcal{O}(|V| \cdot |E|)$.

Improvement (like with Dijkstra's ShortestPath)

■ With Min-Heap: costs

- Initialization (node coloring) $\mathcal{O}(|V|)$
- $|V| \times \text{ExtractMin} = \mathcal{O}(|V| \log |V|)$,
- $|E| \times \text{Insert or DecreaseKey} = \mathcal{O}(|E| \log |V|)$,

$\mathcal{O}(|E| \cdot \log |V|)$