Minimum Spanning Trees

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In this lecture, we will study another classic problem: finding a minimum spanning tree of an undirected weighted graph.
Undirected Weighted Graphs

Let $G = (V, E)$ be an undirected graph. Let $w$ be a function that maps each edge of $G$ to a positive integer value. Specifically, for each edge $e$, $w(e)$ is a positive integer value, which we call the weight of $e$.

A **undirected weighted graph** is defined as the pair $(G, w)$.

We consider that $G$ is **connected**, namely, there is a path between any two vertices in $V$. 
The integer on each edge indicates its weight. For example, the weight of \( \{g, h\} \) is 9, and that of \( \{d, g\} \) is 11.
Spanning Trees

Remember that a tree is defined as a connected undirected graph with no cycles.

Given a connected undirected weighted graph \((G, w)\) with \(G = (V, E)\), a spanning tree \(T\) is a tree satisfying the following conditions:

- The vertex set of \(T\) is \(V\).
- Every edge of \(T\) is an edge in \(G\).

The cost of \(T\) is the sum of the weights of all the edges in \(T\) (note that \(T\) must have \(|V| - 1\) edges).
The second row shows three spanning trees (of the graph in the first row). The cost of the first two trees is 37, and that of the right tree is 48.
The Minimum Spanning Tree Problem

Given a connected undirected weighted graph \((G, w)\) with \(G = (V, E)\), the goal of the minimum spanning tree (MST) problem is to find a spanning tree of the smallest cost.

Such a tree is called an MST of \((G, w)\).
Example

Both trees in the second row are MSTs. This means that MSTs may not be unique.
Next, we will discuss an algorithm—called Prim’s algorithm—for solving the MST problem.

We assume that $G$ is stored in the adjacency list format. Recall that an edge $\{u, v\}$ is represented twice: once by placing $u$ in the adjacency list of $v$, and another time by placing $v$ in the adjacency list of $u$. The weight of $\{u, v\}$ is stored in both places.
Prim’s Algorithm

The algorithm grows a tree $T_{mst}$ by including one vertex at a time. At any moment, it divides the vertex set $V$ into two parts:

- The set $S$ of vertices that are already in $T_{mst}$.
- The set of other vertices: $V \setminus S$.

At the end of the algorithm, $S = V$.

If an edge connects a vertex in $V$ and a vertex in $V \setminus S$, we call it an extension edge.

At all times, the algorithm enforces the following lightest extension principle:

- For every vertex $v \in V \setminus S$, it remembers which extension edge of $v$ has the smallest weight—referred to as the lightest extension edge of $v$, and denoted as $\text{best-ext}(v)$. 

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Prim’s Algorithm

1. Let \( \{u, v\} \) be an edge with the smallest weight among all edges.
2. Set \( S = \{u, v\} \). Initialize a tree \( T_{mst} \) with only one edge \( \{u, v\} \).
3. Enforce the lightest extension principle:
   - For every vertex \( z \) of \( V \setminus S \)
     - If \( z \) is a neighbor of \( u \), but not of \( v \)
       \( best\-ext(z) = \text{edge } \{z, u\} \)
     - If \( z \) is a neighbor of \( v \), but not of \( u \)
       \( best\-ext(z) = \text{edge } \{z, v\} \)
     - Otherwise
       \( best\-ext(z) = \) the lighter edge between \( \{z, u\} \) and \( \{z, v\} \)
Prim’s Algorithm

4. Repeat the following until $S = V$:

5. Get an extension edge $\{u, v\}$ with the smallest weight
/* Without loss of generality, suppose $u \in S$ and $v \notin S$ */

6. Add $v$ into $S$, and add edge $\{u, v\}$ into $T_{mst}$
/* Next, we restore the lightest extension principle. */

7. for every edge $\{v, z\}$ of $v$:
   - If $z \notin S$ then
     - If $\text{best-ext}(z)$ is heavier than edge $\{v, z\}$ then
       - Set $\text{best-ext}(z) = \text{edge} \{v, z\}$
Example

Edge \(\{a, b\}\) is the lightest of all. So, at the beginning \(S = \{a, b\}\). The MST we are growing now has one edge \(\{a, b\}\).

Note: Edges \(\{c, a\}\) and \(\{c, b\}\) have the same weight. Either of them can be \(\text{best-ext}(c)\).
**Example**

Edge \( \{c, a\} \) is the lightest extension edge. So, we add \( c \) to \( S \), which is now \( S = \{a, b, c\} \). Add edge \( \{c, a\} \) into the MST.

![Graph](image)

<table>
<thead>
<tr>
<th>vertex ( v )</th>
<th>best-ext(( v )) and weight</th>
</tr>
</thead>
<tbody>
<tr>
<td>( a )</td>
<td>n/a</td>
</tr>
<tr>
<td>( b )</td>
<td>n/a</td>
</tr>
<tr>
<td>( c )</td>
<td>n/a</td>
</tr>
<tr>
<td>( d )</td>
<td>nil, ( \infty )</td>
</tr>
<tr>
<td>( e )</td>
<td>{e, b}, 10</td>
</tr>
<tr>
<td>( f )</td>
<td>{c, f}, 5</td>
</tr>
<tr>
<td>( g )</td>
<td>{g, b}, 13</td>
</tr>
<tr>
<td>( h )</td>
<td>{c, h}, 6</td>
</tr>
</tbody>
</table>
Example

Edge \{c, f\} is the lightest extension edge. So, we add \(f\) to \(S\), which is now \(S = \{a, b, c, f\}\). Add edge \{c, f\} into the MST.

\[
\begin{array}{|c|c|}
\hline
\text{vertex } v & \text{best-ext}(v) \text{ and weight} \\
\hline
a & n/a \\
b & n/a \\
c & n/a \\
d & \text{nil, } \infty \\
e & \{e, f\}, 2 \\
f & n/a \\
g & \{g, b\}, 13 \\
h & \{c, h\}, 6 \\
\hline
\end{array}
\]
Example

Edge \{e, f\} is the lightest extension edge. So, we add e to S, which is now \(S = \{a, b, c, f, e\}\). Add edge \{e, f\} into the MST.

\[
\begin{array}{|c|c|}
\hline
\text{vertex } v & \text{best-ext}(v) \text{ and weight} \\
\hline
a & \text{n/a} \\
b & \text{n/a} \\
c & \text{n/a} \\
d & \{e, d\}, 12 \\
e & \text{n/a} \\
f & \text{n/a} \\
g & \{g, b\}, 13 \\
h & \{c, h\}, 6 \\
\hline
\end{array}
\]
Example

Edge \{c, h\} is the lightest extension edge. So, we add \(h\) to \(S\), which is now \(S = \{a, b, c, f, e, h\}\). Add edge \{c, h\} into the MST.

vertex \(v\) | \(\text{best-ext}(v)\) and weight
--- | ---
\(a\) | \(\text{n/a}\)
\(b\) | \(\text{n/a}\)
\(c\) | \(\text{n/a}\)
\(d\) | \(\{e, d\}, 12\)
\(e\) | \(\text{n/a}\)
\(f\) | \(\text{n/a}\)
\(g\) | \(\{g, h\}, 9\)
\(h\) | \(\text{n/a}\)
Example

Edge \( \{g, h\} \) is the lightest extension edge. So, we add \( h \) to \( S \), which is now \( S = \{a, b, c, f, e, h\} \). Add edge \( \{g, h\} \) into the MST.
Finally, edge \(\{d, g\}\) is the lightest extension edge. So, we add \(d\) to \(S\), which is now \(S = \{a, b, c, f, e, h, g, d\}\). Add edge \(\{d, g\}\) into the MST.

We have obtained our final MST.
Next we will prove that the algorithm is correct, namely, the tree output is indeed an MST. We will do so by induction on the sequence of the edges added to the tree. Specifically, the claim to be proven is:

**Claim**: For any $i \in [1, |V| - 1]$, there must be an MST containing all the first $i$ edges chosen by our algorithm.

Then the algorithm’s correctness follows from the above claim at $i = |V| - 1$. 

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Minimum Spanning Trees
Correctness Proof

Let us first recall a fundamental property of undirected graphs:

**Lemma:** Let $T$ be a tree of $n$ vertices. Adding an arbitrary edge between two vertices in $T$ introduces a cycle.

**Proof:** Suppose that the edge is added between $u$ and $v$. Before the edge was added, there is already a path allowing us to go from $u$ to $v$ in $T$. Therefore, the edge $\{u, v\}$ allows us to move from $v$ back to $u$, thus witnessing a cycle. 

\[ \square \]
Correctness Proof

Now we proceed to prove the claim on Slide 20.

**Base Case:*** $i = 1$. Let $\{u, v\}$ be an edge with the smallest weight in the graph. We will prove that the edge must exist in some MST.

Take any MST $T$ that does not contain edge $\{u, v\}$. Add the edge to $T$, which creates a cycle. Remove an arbitrary edge $e$ in $T$ such that $e \neq \{u, v\}$, which gives a new tree $T'$. Since $\{u, v\}$ has the smallest weight, the cost of $T'$ is smaller than or equal to that of $T$. This means that $T'$ is also an MST.

Hence, the claim holds for $i = 1$. 

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Minimum Spanning Trees
Correctness Proof

**Inductive Case:** Assuming that the claim holds for \( i \leq k - 1 \) \((k \geq 2)\), next we prove that it also holds for \( i = k \). Let \( \{u, v\} \) be the \( k \)-th edge added by our algorithm, and \( S \) be the set of vertices already in the algorithm’s tree before the addition of \( \{u, v\} \). Without loss of generality, suppose that \( u \in S \), and \( v \notin S \).

By the inductive assumption, we know that there is some MST \( T \) that includes all the first \( k - 1 \) edges. If \( T \) also includes edge \( \{u, v\} \), then the claim already holds.

Next, we consider the case where \( T \) does not have \( \{u, v\} \).
Correctness Proof

We add \( \{u, v\} \) to \( T \), which creates a cycle. Let us walk on this cycle starting from \( v \), cross \( \{u, v\} \) into \( S \), keep walking within \( S \) until traveling out of \( S \) for the first time. Let the edge that brought us out of \( S \) be \( \{u', v'\} \).

Note that both \( \{u, v\} \) and \( \{u', v'\} \) are extension edges right before the moment our algorithm picks the \( k \)-th edge. Since \( \{u, v\} \) has the smallest weight among all the extension edges, we know that the weight of \( \{u, v\} \) is smaller than or equal to that of \( \{u', v'\} \).
Correctness Proof

Now, remove edge \( \{u', v'\} \) from \( T \), which gives another tree \( T' \). The cost of \( T' \) cannot be more than that of \( T \). This means that \( T' \) must also be an MST.

We thus have proved that the claim holds for \( i = k \) as well. \( \square \)
It will be left as an exercise for you to apply the data structures you have learned to implement Prim’s algorithm in $O(|V| + |E| \cdot \log |V|)$ time.