Dijkstra's algorithm: summary

• Finds the shortest path between two specified vertices, or between one source vertex and every other.

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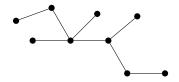
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Dijkstra's algorithm: summary

- Finds the shortest path between two specified vertices, or between one source vertex and every other.
- Does so by exploring vertices in order of their distance from the source.
- Running time is $O((n + e) \log n)$ for a graph with n vertices and e edges.

Trees as graphs

- We have seen trees as data structures.
- A tree is also a graph: a connected graph with no cycles:

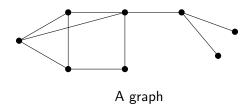


• Every tree with n vertices has n-1 edges (prove by induction).

Subgraphs

A **subgraph** of a graph G is a graph that can be made from G by deleting vertices and/or edges.

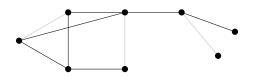
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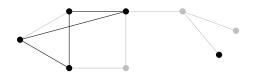


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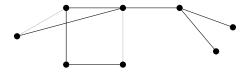
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A non-spanning subgraph

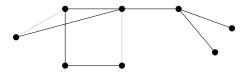
Spanning trees

A **spanning tree** is a spanning subgraph that is a tree.



Spanning trees

A spanning tree is a spanning subgraph that is a tree.



Spanning trees connect all the vertices using the least possible number of edges.

Every connected graph has at least one spanning tree.

Minimum spanning trees

The **minimum spanning tree** (MST) of a weighted graph is the spanning tree that has the least total edge weight.

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E.g., you're building a wind farm and you want to connect your turbines using the least amount of cable.

(Assuming cables must run from turbine to turbine with no junctions in the middle.)

Minimum spanning trees

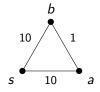
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Strictly, we should say " \mathbf{An} MST is \mathbf{a} spanning tree that has least possible weight" – a graph may have more than one MST.

Dijkstra's algorithm computes a spanning tree, but not necessarily a minimal one.

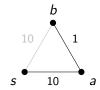


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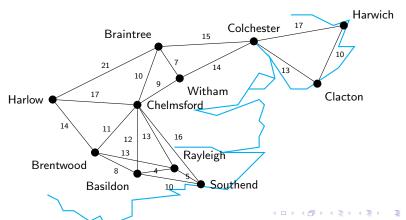
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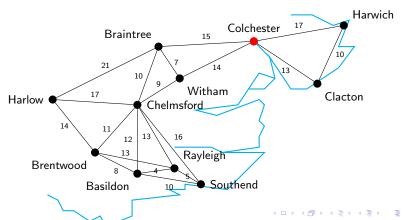


The other minimum spanning tree

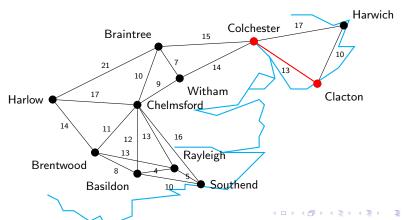
- Add an arbitrary vertex to the tree.
- Add to the tree the least-weight edge in G that connects a vertex in the tree to one not yet in it.
- Repeat 2. until the tree is spanning.



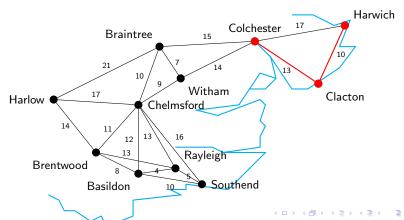
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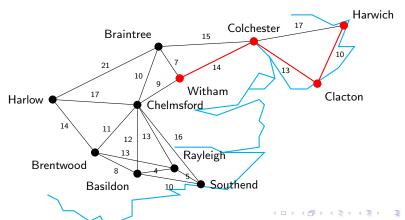
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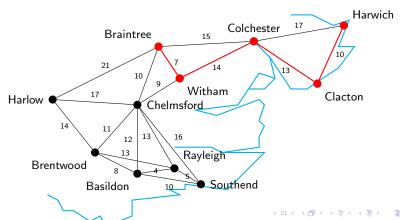
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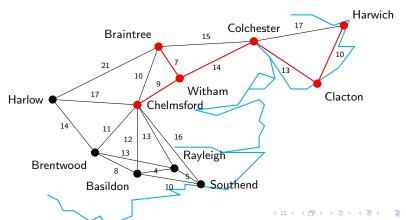
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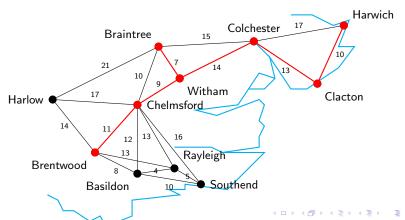
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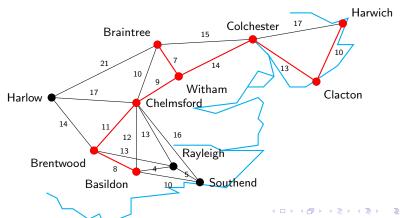
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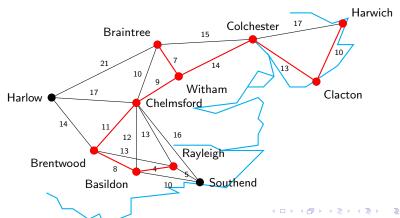
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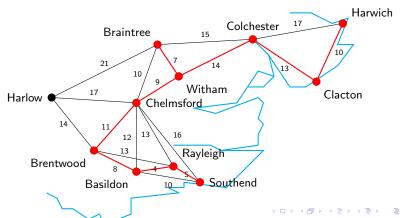
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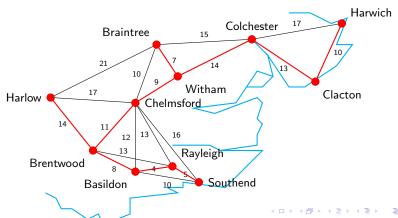
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- A priority queue of vertices that are adjacent to the tree built so far, with the weight of the shortest connecting edge as priority.
- int[] treeNbr stores, for each vertex x in the queue, its nearest neighbour in the tree built so far.

Prim's algorithm: pseudocode (1)

Prim's algorithm for a graph with n vertices.

```
// Initialization
inTree = {true, false, false, ..., false};
treeNbr = {0, ..., 0};
Q = new empty priority queue;
mst = new graph with n vertices;

for each nedge (0,y) in G
    Q.insert (weight (0,y), y);
...
```

Prim's algorithm: pseudocode (2)

```
// Main loop
while (Q not empty)
   // Add next vertex to tree
   x = Q.next();
   mst.addEdge (x, treeNbr[x], weight (x, treeNbr[x]));
   inTree[x] = true:
   // Process neighbours
   for each edge (x,y) in G
       if (!inTree[y])
           if (!Q.contains(y))
              Q.insert (weight(x, y), y);
              treeNbr[y] = x;
           else if (weight(x,y) < Q.priorityOf(y))
              Q.setPriority (weight(x,y), y);
              treeNbr[v] = x;
return mst;
```

A trick with priority queues

- Graph algorithms (Dijkstra, Prim, etc.) often need to find or update the priority of a specific item in a priority queue.
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- We can find items in time O(1) by maintaining an array that tells us the index of each vertex in the priority queue's internal array.
- Allows find priority in time O(1) (just look it up) and update priority in time $O(\log n)$ (because of bubble up/down).

Prim's algorithm: running time

For an input graph with n vertices and e edges, using adjacency lists.

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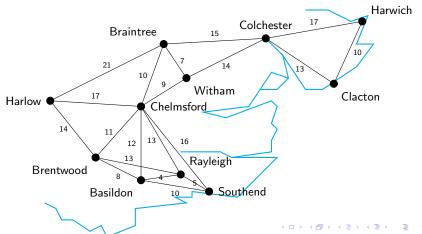
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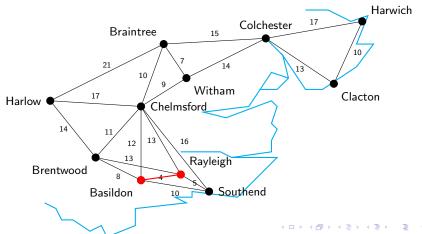
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- For each edge, it can call insert() or setPriority(): each call takes time $O(\log n)$.
- Total running time is $O((n + e) \log n)$.
- This is $O(e \log n)$ if the input is connected (a connected graph must have at least n-1 edges).

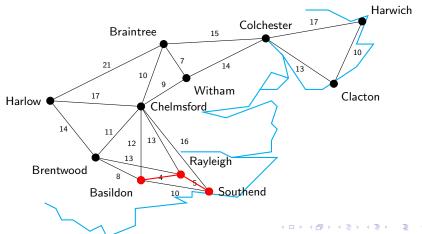
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- Repeat 2. until we have a spanning tree.



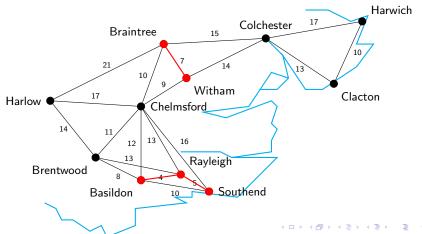
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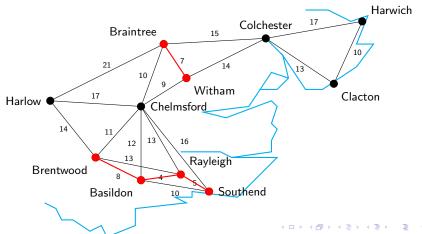
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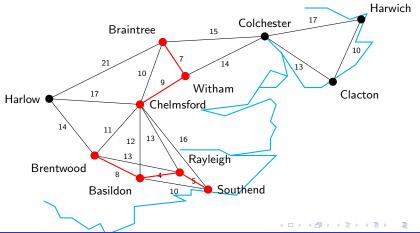
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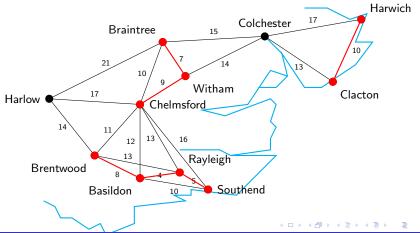
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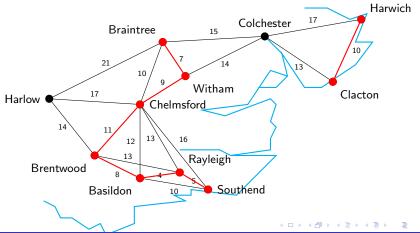
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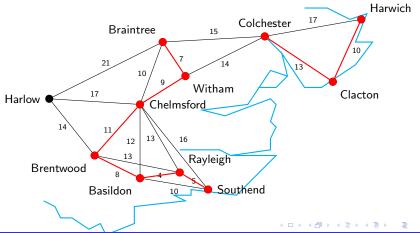
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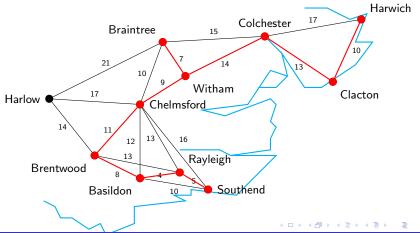
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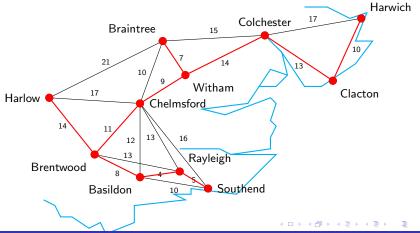
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Each time we merge, the shorter list doubles (or more) in length, so a vertex can't be moved more than $\log n$ times.

Kruskal's algorithm: pseudocode

For an input graph G with n edges.

```
mst = new graph with n vertices;
L = list of G's edges sorted by increasing weight;
for each edge (x, y, w) in L, in order
   if (componentOf(x) != componentOf(y))
        mst.addEdge (x, y, w);
        mergeComponents (x, y);

return mst;
```

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Conclusion:

- If there are only O(e) possible edge-weights, use Kruskal.
- Otherwise, if e > O(n) (usually true), use Prim.
- Otherwise, use either.

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- Both run in time $O(e \log n)$.
- Prim can be improved to $O(e + n \log n)$ for "most" graphs.
- Kruskal can be improved to almost O(e) if few possible edge weights.