

Point-to-Point Shortest Path

Algorithm gives you shortest paths from a particular given others in a graph.

Are you're only interested in getting to a particular vertex?

If algorithm finds paths in order of length, you *could* and stop when you get to the vertex you want.

It can be really wasteful.

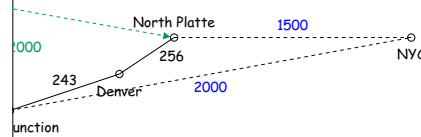
Example: to travel by road from Denver to a destination on lower west in New York City is about 1750 miles (says Google).

Example: from Denver to Berkeley is about 1250 miles.

Example: more much of California, Nevada, Arizona, etc. before reaching destination, even though these are all in the wrong direction.

It can be even worse when graph is infinite, generated on the fly.

Illustration



Lines indicate distances from Denver we have determined.

Algorithm would have us look at Grand Junction next (smallest distance from Denver).

But if we find a better path in the heuristic remaining distance to NYC (our goal), we should go to North Platte instead.

Missible Heuristics for A* Search

If heuristic estimate for the distance to NYC is too high (i.e., greater than the actual path by road), then we may get to NYC without visiting any points along the shortest route.

Example: if our heuristic decided that the midwest was literally empty of cities, and $h(C) = 2000$ for C any city in Michigan or Ohio, we would only find a path that detoured south through Kentucky.

For a *missible* heuristic, $h(C)$ must never overestimate $d(C, NYC)$, the actual distance from C to NYC.

Example: if $h(C) = 0$ will work (what is the result?), but yield a sub-optimal algorithm. This is just Dijkstra's algorithm.

CS61B Lecture #34

Search, Minimum spanning trees, union-find.

A* Search

Algorithm for a path from vertex Denver to the desired NYC.

If we had a simple *heuristic estimate*, $h(V)$, of the length of the shortest path from any vertex V to NYC.

Instead of visiting vertices in the fringe in order of their shortest known path to Denver, we order by the sum of their shortest known path to Denver plus this heuristic estimate of the remaining distance to NYC: $d(Denver, V) + h(V)$.

For example, if we already know the shortest path to Denver and choose to visit Grand Junction next, we look at places that are reachable from places we already know the shortest path to Denver and choose to visit Grand Junction next because they will result in the shortest trip to NYC, given the remaining distance.

Example: if the heuristic estimate is good, then we don't look at, say, Grand Junction next (because it's in the wrong direction).

The algorithm is *A* search*.

For it to work, we must be careful about the heuristic.

Missible Heuristics for A* Search

If heuristic estimate for the distance to NYC is too high (i.e., greater than the actual path by road), then we may get to NYC without visiting any points along the shortest route.

Example: if our heuristic decided that the midwest was literally empty of cities, and $h(C) = 2000$ for C any city in Michigan or Ohio, we would only find a path that detoured south through Kentucky.

For a *missible* heuristic, $h(C)$ must never overestimate $d(C, NYC)$, the actual distance from C to NYC.

Example: if $h(C) = 0$ will work (what is the result?), but yield a sub-optimal algorithm.

Summary of Shortest Paths

Algorithm finds a *shortest-path tree* computing giving shortest paths in a weighted graph from a given starting node to other nodes.

$d =$

remove V nodes from priority queue +

update all neighbors of each of these nodes and add or remove them in queue ($E \lg E$)

$(V + E \lg V) = \Theta((V + E) \lg V)$

Searches for a shortest path to a *particular* target node.

Dijkstra's algorithm, except:

1. We take target from queue.

2. Use by estimated distance to start + heuristic guess of distance ($h(v) = d(v, \text{target})$)

3. Must not overestimate distance and obey triangle inequality ($d(b, c) \geq d(a, c)$).

Minimum Spanning Trees by Prim's Algorithm

Grow a tree starting from an arbitrary node.

1. Add the shortest edge connecting some node already in the tree to one that isn't yet.

2. How does this work?

```

fringe;
{ v.dist() = ∞; v.parent() = null; }
for starting node, s;

```

```

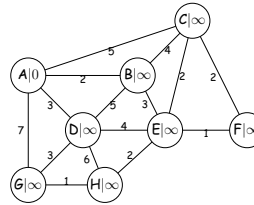
queue ordered by smallest .dist();
fringe;
if !fringe.empty() {
    v = fringe.removeFirst();
}

```

```

for (v, w) {
    if (weight(v, w) < w.dist()) {
        w.dist() = weight(v, w); w.parent() = v; }
}

```



Minimum Spanning Trees by Prim's Algorithm

Grow a tree starting from an arbitrary node.

1. Add the shortest edge connecting some node already in the tree to one that isn't yet.

2. How does this work?

```

fringe;
{ v.dist() = ∞; v.parent() = null; }
for starting node, s;

```

```

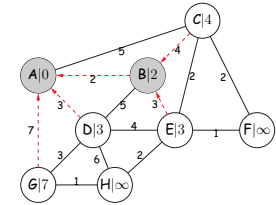
queue ordered by smallest .dist();
fringe;
if !fringe.empty() {
    v = fringe.removeFirst();
}

```

```

for (v, w) {
    if (weight(v, w) < w.dist()) {
        w.dist() = weight(v, w); w.parent() = v; }
}

```



Consistency

If we estimate $h(\text{Chicago}) = 700$, and $h(\text{Springfield, IL}) = 200$, and $d(\text{Chicago, Springfield}) = 200$.

If we guess that we are suddenly 200 miles to Springfield, we guess that we are suddenly closer to NYC.

It's possible, since both estimates are low, but it will mess up our estimate.

It will require that we put processed nodes back into the queue since our estimate was wrong.

Of course, anyway we also require *consistent heuristics*: $h(A) + d(A, B) \leq h(B)$, as for the triangle inequality.

1. Not all heuristics are admissible (why?).

2. "As the crow flies" is a reasonable $h(\cdot)$ in the trip-planning

3. Search (and others) is in [cs61b-software](#) and on the machines as [graph-demo](#).

Minimum Spanning Trees

Given a set of places and distances between them (assume positive), find a set of connecting roads of minimum total length that allows travel between any two.

1. The shortest path you get will not necessarily be shortest paths.

2. That such a set of connecting roads and places must exist because removing one road in a cycle still allows all to

Minimum Spanning Trees by Prim's Algorithm

Grow a tree starting from an arbitrary node.

1. Add the shortest edge connecting some node already in the tree to one that isn't yet.

2. How does this work?

```

fringe;
{ v.dist() = ∞; v.parent() = null; }
for starting node, s;

```

```

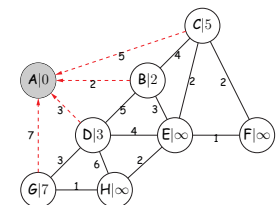
queue ordered by smallest .dist();
fringe;
if !fringe.empty() {
    v = fringe.removeFirst();
}

```

```

for (v, w) {
    if (weight(v, w) < w.dist()) {
        w.dist() = weight(v, w); w.parent() = v; }
}

```



Spanning Trees by Prim's Algorithm

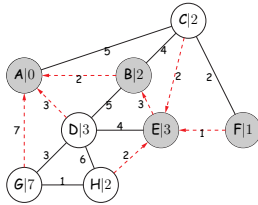
ow a tree starting from an arbitrary node.

, add the shortest edge connecting some node already
o one that isn't yet.

is work?

```
inge;  
{ v.dist() = ∞; v.parent() = null; }  
y starting node, s;
```

```
queue ordered by smallest .dist();  
fringe;  
sEmpty() {  
nge.removeFirst();  
  
v,w) {  
ge && weight(v,w) < w.dist()  
= weight(v, w); w.parent() = v; }
```



59:20 2021

CS61B: Lecture #34 14

Spanning Trees by Prim's Algorithm

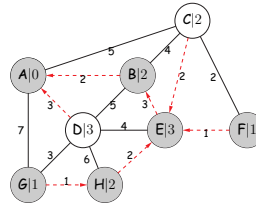
ow a tree starting from an arbitrary node.

, add the shortest edge connecting some node already
o one that isn't yet.

is work?

```
inge;  
{ v.dist() = ∞; v.parent() = null; }  
y starting node, s;
```

```
queue ordered by smallest .dist();  
fringe;  
sEmpty() {  
nge.removeFirst();  
  
v,w) {  
ge && weight(v,w) < w.dist()  
= weight(v, w); w.parent() = v; }
```



59:20 2021

CS61B: Lecture #34 16

Spanning Trees by Prim's Algorithm

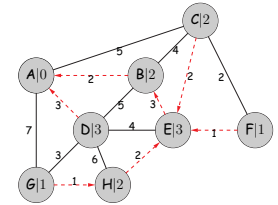
ow a tree starting from an arbitrary node.

, add the shortest edge connecting some node already
o one that isn't yet.

is work?

```
inge;  
{ v.dist() = ∞; v.parent() = null; }  
y starting node, s;
```

```
queue ordered by smallest .dist();  
fringe;  
sEmpty() {  
nge.removeFirst();  
  
v,w) {  
ge && weight(v,w) < w.dist()  
= weight(v, w); w.parent() = v; }
```



59:20 2021

CS61B: Lecture #34 18

Spanning Trees by Prim's Algorithm

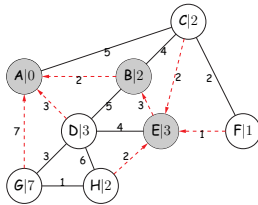
ow a tree starting from an arbitrary node.

, add the shortest edge connecting some node already
o one that isn't yet.

is work?

```
inge;  
{ v.dist() = ∞; v.parent() = null; }  
y starting node, s;
```

```
queue ordered by smallest .dist();  
fringe;  
sEmpty() {  
nge.removeFirst();  
  
v,w) {  
ge && weight(v,w) < w.dist()  
= weight(v, w); w.parent() = v; }
```



59:20 2021

CS61B: Lecture #34 13

Spanning Trees by Prim's Algorithm

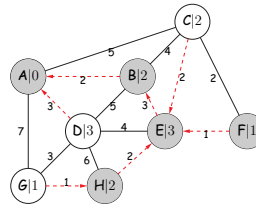
ow a tree starting from an arbitrary node.

, add the shortest edge connecting some node already
o one that isn't yet.

is work?

```
inge;  
{ v.dist() = ∞; v.parent() = null; }  
y starting node, s;
```

```
queue ordered by smallest .dist();  
fringe;  
sEmpty() {  
nge.removeFirst();  
  
v,w) {  
ge && weight(v,w) < w.dist()  
= weight(v, w); w.parent() = v; }
```



59:20 2021

CS61B: Lecture #34 15

Spanning Trees by Prim's Algorithm

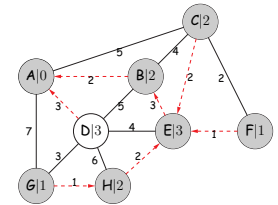
ow a tree starting from an arbitrary node.

, add the shortest edge connecting some node already
o one that isn't yet.

is work?

```
inge;  
{ v.dist() = ∞; v.parent() = null; }  
y starting node, s;
```

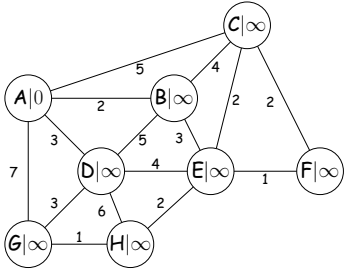
```
queue ordered by smallest .dist();  
fringe;  
sEmpty() {  
nge.removeFirst();  
  
v,w) {  
ge && weight(v,w) < w.dist()  
= weight(v, w); w.parent() = v; }
```



59:20 2021

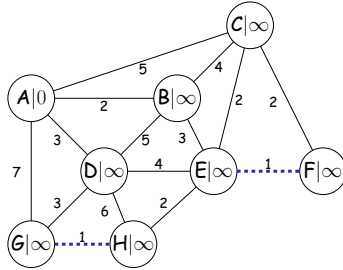
CS61B: Lecture #34 17

Example of Kruskal's Algorithm



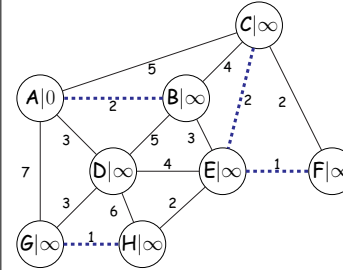
Sort the edges in increasing order of weight (here, alphabetically).
 Edges that would either create a cycle or result in a node with a degree greater than 2 are discarded ('X'ed out here).
 Edges that connect two unconnected groups get added to the tree.
 The result is a minimal spanning tree (a free tree).

Example of Kruskal's Algorithm



Sort the edges in increasing order of weight (here, alphabetically).
 Edges that would either create a cycle or result in a node with a degree greater than 2 are discarded ('X'ed out here).
 Edges that connect two unconnected groups get added to the tree.
 The result is a minimal spanning tree (a free tree).

Example of Kruskal's Algorithm



Sort the edges in increasing order of weight (here, alphabetically).
 Edges that would either create a cycle or result in a node with a degree greater than 2 are discarded ('X'ed out here).
 Edges that connect two unconnected groups get added to the tree.
 The result is a minimal spanning tree (a free tree).

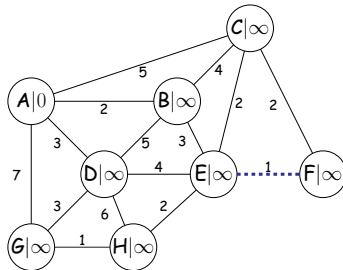
Spanning Trees by Kruskal's Algorithm

The shortest edge in a graph can always be part of a spanning tree.
 If we have a bunch of subtrees of a MST, then the shortest edge that connects two of them can be part of a MST, combining the subtrees into a bigger one.

(trivial) subtree for each node in the graph:

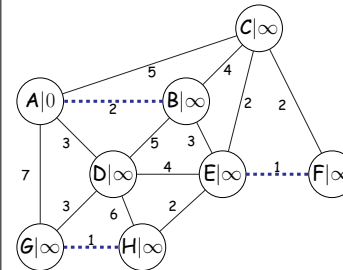
Edge (v,w) , in increasing order of weight {
 such that (v,w) connects two different subtrees } {
 Add (v,w) to MST;
 Merge the two subtrees into one;

Example of Kruskal's Algorithm



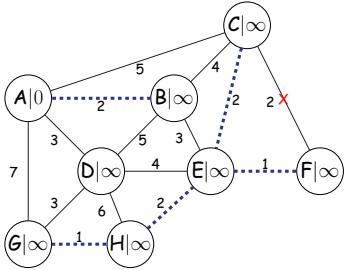
Sort the edges in increasing order of weight (here, alphabetically).
 Edges that would either create a cycle or result in a node with a degree greater than 2 are discarded ('X'ed out here).
 Edges that connect two unconnected groups get added to the tree.
 The result is a minimal spanning tree (a free tree).

Example of Kruskal's Algorithm



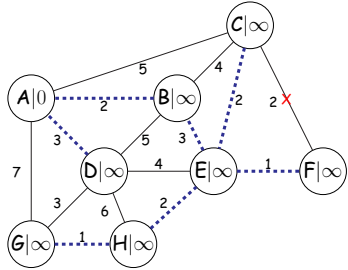
Sort the edges in increasing order of weight (here, alphabetically).
 Edges that would either create a cycle or result in a node with a degree greater than 2 are discarded ('X'ed out here).
 Edges that connect two unconnected groups get added to the tree.
 The result is a minimal spanning tree (a free tree).

Example of Kruskal's Algorithm



Sort the edges in increasing order of weight (here, alphabetically).
 Edges that would either create a cycle or connect two already-connected groups get discarded ('X'ed out here).
 Edges that connect two unconnected groups get added to the tree.
 The result is a minimal spanning tree (a free tree).

Example of Kruskal's Algorithm

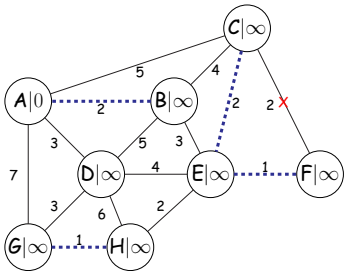


Sort the edges in increasing order of weight (here, alphabetically).
 Edges that would either create a cycle or connect two already-connected groups get discarded ('X'ed out here).
 Edges that connect two unconnected groups get added to the tree.
 The result is a minimal spanning tree (a free tree).

Union Find

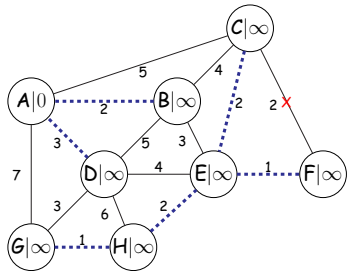
The algorithm required that we have a set of sets of nodes with the following properties:
 1. Each node belongs to exactly one set.
 2. We can find the set a given node belongs to.
 3. We can merge two sets with their union, reassigning all the nodes in the two sets to this union.
 4. The operation to do is to store a set number in each node, making it easy to find the set.
 5. The operation to change the set number in one of the two sets being merged is better choice.
 6. The time for an individual union can take $\Theta(N)$ time.
 7. How fast?

Example of Kruskal's Algorithm



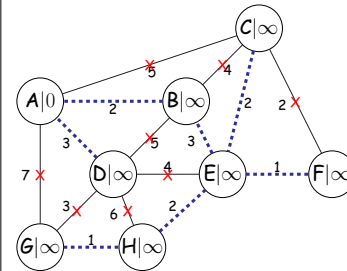
Sort the edges in increasing order of weight (here, alphabetically).
 Edges that would either create a cycle or connect two already-connected groups get discarded ('X'ed out here).
 Edges that connect two unconnected groups get added to the tree.
 The result is a minimal spanning tree (a free tree).

Example of Kruskal's Algorithm



Sort the edges in increasing order of weight (here, alphabetically).
 Edges that would either create a cycle or connect two already-connected groups get discarded ('X'ed out here).
 Edges that connect two unconnected groups get added to the tree.
 The result is a minimal spanning tree (a free tree).

Example of Kruskal's Algorithm



Sort the edges in increasing order of weight (here, alphabetically).
 Edges that would either create a cycle or connect two already-connected groups get discarded ('X'ed out here).
 Edges that connect two unconnected groups get added to the tree.
 The result is a minimal spanning tree (a free tree).

Path Compression

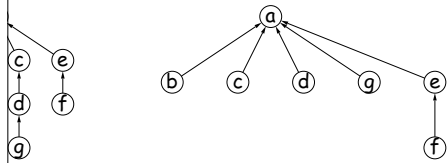
unioning really fast, but the find operation potentially).

Following trick: whenever we do a *find* operation, *compress* the root, so that subsequent finds will be faster.

Each of the nodes in the path point directly to the

very fast, and sequence of unions and finds each have nearly constant amortized time.

What 'g' in last tree (result of compression on right):



A Clever Trick

to represent a set of nodes by *one* arbitrary representative set.

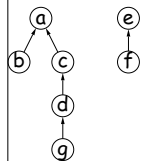
Each node contain a pointer to another node in the same set.

Each pointer to represent the *parent* of a node in a tree representative node as its root.

To find what set a node is in, follow parent pointers.

For such trees, make one root point to the other (choose the larger tree as the union representative).

Two Sets



Their Union

