

## Point-to-Point Shortest Path

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

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

Algorithm finds paths in order of length, you *could* simply 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 Manhattan in New York City is about 1750 miles (says Google).

Example: going from Denver to the Gourmet Ghetto in Berkeley is 1750 miles.

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

Example: even worse when graph is infinite, generated on the fly.

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## Admissible Heuristics for A\* Search

Example: 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 all points along the shortest route.

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

Example: *admissible*,  $h(C)$  must never overestimate  $d(C, NYC)$ , the actual distance from  $C$  to NYC.

Example: on the other hand,  $h(C) = 0$  will work (what is the result?), but yield a naive algorithm.

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## Summary of Shortest Paths

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

Algorithm:

1. Remove  $V$  nodes from priority queue +

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

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

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

Algorithm: Dijkstra's algorithm, except:

1. When we take target from queue.

2. Update by estimated distance to start + heuristic guess of distance ( $h(v) = d(v, target)$ )

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

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## CS61B Lecture #34

Search, Minimum spanning trees, union-find.

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## A\* Search

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

Example: if we had a *heuristic guess*,  $h(V)$ , of the length of a path from vertex  $V$  to NYC.

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

Example: if we already know the shortest path to Denver and choose to visit places that look like they will result in the shortest trip to NYC, we look at places that are reachable from places we already know the shortest path to Denver and choose to visit places that look like they will result in the shortest trip to NYC, the remaining distance.

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

Example: the algorithm is *A\* search*.

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

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## Consistency

Example: if we estimate  $h(\text{Chicago}) = 700$ , and  $h(\text{Springfield, IL}) = 200$ .

Example: if we are 200 miles to Springfield, we guess that we are suddenly closer to NYC.

Example: it is possible, since both estimates are low, but it will mess up the algorithm.

Example: it will require that we put processed nodes back into the queue if our estimate was wrong.

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

Example: if the heuristics are admissible (why?).

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

Example: the search (and others) is in *cs61b-software* and on the machines as *graph-demo*.

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## Minimum Spanning Trees by Prim's Algorithm

Show a tree starting from an arbitrary node.  
 Add the shortest edge connecting some node already in the tree to one that isn't yet.

How does this work?

```

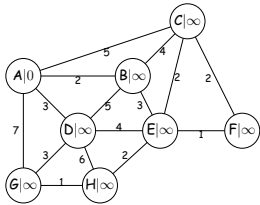
Initialize;
{ v.dist() = ∞; v.parent() = null; }
for each starting node, s;

```

```

queue ordered by smallest .dist();
fringe;
while !fringe.empty() {
    v = fringe.removeFirst();
    for each (w) {
        if weight(v,w) < w.dist()
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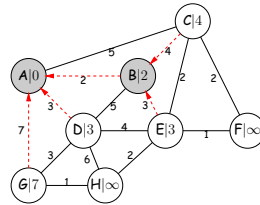
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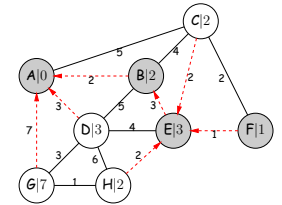
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## Minimum Spanning Trees

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.

Shortest paths are not necessarily the best.

Such a set of connecting roads and places must be chosen because removing one road in a cycle still allows all to be connected.

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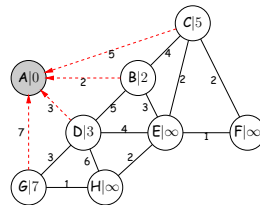
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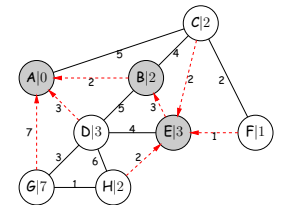
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## Spanning Trees by Prim's Algorithm

How to 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.

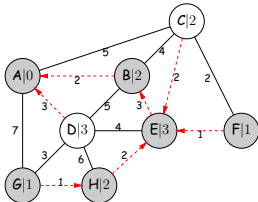
2) Repeat.

```

Initialize:
    v: starting node, s;
    v.dist() = ∞; v.parent() = null;
    
```

```

while (fringe not empty) {
    v = fringe.removeFirst();
    for (w in neighbors of v) {
        if (weight(v,w) < w.dist()) {
            w.dist() = weight(v,w); w.parent() = v;
        }
    }
}
    
```



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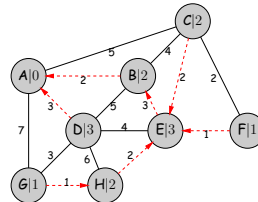
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```



## Union Find

Prim's algorithm required that we have a set of sets of nodes with disjoint nodes:

1) For each node, find the set of nodes it belongs to.

2) For each edge, if the two nodes belong to different sets, union the two sets with their union, reassigning all the nodes in the old sets to this union.

3) The only thing to do is to store a set number in each node, making it easy to find the set a node belongs to.

4) When considering an edge, if the two nodes belong to the same set, skip it. Otherwise, union the two sets. The smaller set is the better choice.

5) The time for an individual union can take  $\Theta(N)$  time.

6) How fast?

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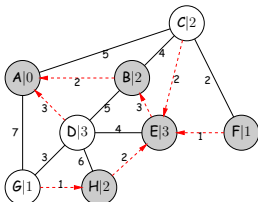
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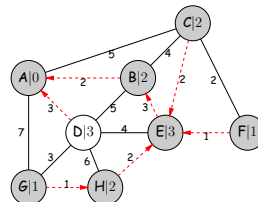
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```



## Spanning Trees by Kruskal's Algorithm

1) Sort the edges in the graph by weight. The shortest edge in a graph can always be part of a spanning tree.

2) If the edge connects two different subtrees of a MST, then the shortest edge connects two of them can be part of a MST, combining the subtrees into a bigger one.

3) A subtree for each node in the graph:

```

Initialize:
    v: starting node, s;
    v.dist() = ∞; v.parent() = null;
    
```

## Path Compression

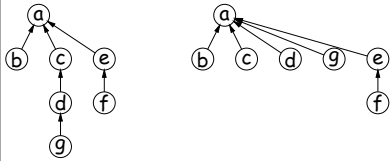
unioning really fast, but the find operation potentially ).

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

Make each of the nodes in the path point directly to the root.

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

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



## A Clever Trick

How to represent a set of nodes by *one* arbitrary representative node in that set.

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

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

To find which 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).

