## CS 4102: Greedy Algorithms

- Topics covered for greedy algorithms
  - General principles
  - Making change
  - Knapsack problems
  - Activity Selection
  - Minimum spanning trees: Prim's, Kruskal's algorithms
  - Single-source shortest path: Dijkstra's algorithm
  - Approximation algorithms

## Greedy Method: Overview

- Optimization problems: terminology
  - Solutions judged on some criteria:

**Objective function** 

- Example: Sum of edge weights in path is smallest
- A solution must meet certain constraints A solution is *feasible*
- Example: All edges in solution are in graph, form a simple path
- One (or more) feasible solutions that scores highest (by the objective function) is the *optimal* solution(s)

# Greedy Method: Overview

#### • Greedy strategy:

- Build solution by stages, adding one item to partial solution found so far
- At each stage, make locally optimal choice based on the <u>greedy rule</u> (sometimes called the selection function)
  - Locally optimal, I.e. best given what info we have now
- Irrevocable, a choice can't be un-done
- Sequence of locally optimal choices leads to globally optimal solution (hopefully)
  - Must prove this for a given problem!
  - Approximation algorithms, heuristics

# Making Change

- Remember? We did this one in class on Day 1
- Inputs:
  - Value N of the change to be returned
  - An unlimited number of coins of values d1, d2,.., dk
- Output: the smallest possible set of coins that sums to N
- Objective function? Smallest set
- Constraints on feasible solutions? Must sum to N
- Greedy rule: choose coin of largest value that is less than N Sum(coins chosen so far)
- Always optimal? Depends on set of coin values

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#### **Algorithm 7.1.1 Greedy Coin Changing**

```
This algorithm makes change for an amount A using coins of
denominations
      denom[1] > denom[2] > \cdots > denom[n] = 1.
Input Parameters: denom, A
Output Parameters: None
greedy_coin_change(denom,A) {
   i = 1
   while (A > 0) {
      c = A/denom[i]
      println("use" + c + "coins of denomination" +
                              denom[i])
      A = A - c * denom[i]
      i = i + 1
   }
}
```

## **Knapsack Problems**

- Section 7.6 in text
- Inputs:
  - n items, each with a weight w\_i and a value v\_i
  - capacity of the knapsack, C
- Output:
  - Fractions for each of the n items, x\_I
  - Chosen to maximize total profit but not to exceed knapsack capacity

## Two Types of Knapsack Problem

- 0/1 knapsack problem
  - Each item is discrete. Must choose all of it or none of it. So each x\_i is 0 or 1
  - Greedy approach does not produce optimal solutions
  - But another approach, dynamic programming, does
- Continuous knapsack problem
  - Can pick up fractions of each item
  - The correct selection function yields a greedy algorithm that produces optimal results

## Greedy Rule for Knapsack?

- Build up a partial solution by choosing x\_i for one item until knapsack is full (or no more items). Which item to choose?
- There are several choices. Pick one and try on this:
  - n = 3, C = 20
  - weights = (18, 15, 10)
  - values = (25, 24, 15)
- What answer do you get?
- The optimal answer is: (0, 1, 0.5), total=31.5 Can you verify this?

#### Possible Greedy Rules for Knapsack

- Build up a partial solution by choosing x\_i for one item until knapsack is full (or no more items). Which item to choose?
  - Maybe this: take as much as possible of the remaining item that has largest value, v\_i
  - Or maybe this: take as much as possible of the remaining items that has smallest weight, w\_i
  - Neither of these produce optimal values! The one that does "combines" these two approaches.

Use ratio of profit-to-weight

#### **Example Knapsack Problem**

- For this example:
  n = 3, C = 20
  weights = (18, 15, 10)
  values = (25, 24, 15)
  Ratios = (25/18, 24/15, 15/10) = (1.39, 1.6, 1.5)
- The optimal answer is: (0, 1, 0.5)

## **Activity-Selection Problem**

- Problem: You and your classmates go on Semester at Sea
  - Many exciting activities each morning
  - Each starting and ending at different times
  - Maximize your "education" by doing as many as possible. (They're all equally good!)
- Welcome to the *activity selection problem*

## The Activities!

Id	Start	End	Activity	
1	9:00	10:45	Fractals, Recursion and Crayolas	
2	9:15	10:15	Tropical Drink Engineering with Prof. Bloomfield	
3	9:30	12:30	Managing Keyboard Fatigue with Swedish Massage	
4	9:45	10:30	Applied ChemE: Suntan Oil or Lotion?	
5	9:45	11:15	Optimization, Greedy Algorithms, and the Buffet Line	
6	10:15	11:00	Hydrodynamics and Surfing	
7	10:15	11:30	Computational Genetics and Infectious Diseases	
8	10:30	11:45	Turing Award Speech Karaoke	
9	11:00	12:00	Pool Tanning for Pale Engineers	
10	11:00	12:15	Mechanics, Dynamics and Shuffleboard Physics	
11	12:00	12:45	Discrete Math Applications in Gambling	

## Generalizing Start, End

ld	Start	End	Len	Activity	
1	0	6	7	Fractals, Recursion and Crayolas	
2	1	4	4	4 Tropical Drink Engineering with Prof. Bloomfield	
3	2	13	12	12 Managing Keyboard Fatigue with Swedish Massage	
4	3	5	3	Applied ChemE: Suntan Oil or Lotion?	
5	3	8	6	Optimization, Greedy Algorithms, and the Buffet Line	
6	5	7	3 Hydrodynamics and Surfing		
7	5	9	5	Computational Genetics and Infectious Diseases	
8	6	10	5	Turing Award Speech Karaoke	
9	8	11	4	Pool Tanning for Pale Engineers	
10	8	12	5	Mechanics, Dynamics and Shuffleboard Physics	
11	12	14	3	Discrete Math Applications in Gambling	

## **Greedy Approach**

- 1. Select a first item.
- 2. Eliminate items that are incompatible with that item. (I.e. they overlap.)
- 3. Apply the *greedy rule* (AKA *selection function*) to pick the next item.
- 4. Go to Step 2

#### What is a good greedy rule for selecting next item?

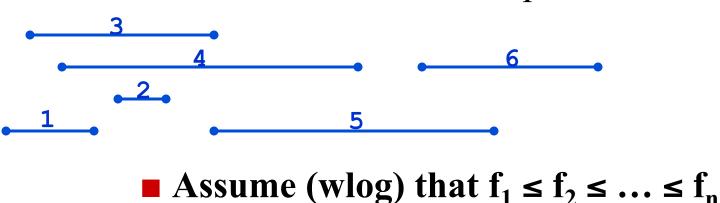
## **Some Possibilities**

- Pick the next compatible one that starts earliest
- Pick the shortest one
- Pick the one that has the least conflicts (i.e. overlaps)

## **Activity-Selection**

#### • Formally:

- Given a set S of n activities
  - $s_i$  = start time of activity *i*
  - $f_i$  = finish time of activity *i*
- Find max-size subset *A* of compatible activities



## Activity Selection: Optimal Substructure

- Let *k* be the minimum activity in *A* (i.e., the one with the earliest finish time). Then *A* {*k*} is an optimal solution to  $S' = \{i \in S: s_i \ge f_k\}$ 
  - In words: once activity #1 is selected, the problem reduces to finding an optimal solution for activityselection over activities in *S* compatible with #1
  - Proof: if we could find optimal solution *B*' to *S*' with  $|B| > |A \{k\}|$ ,
    - Then  $B \cup \{k\}$  is compatible
    - And  $|B \cup \{k\}| > |A|$

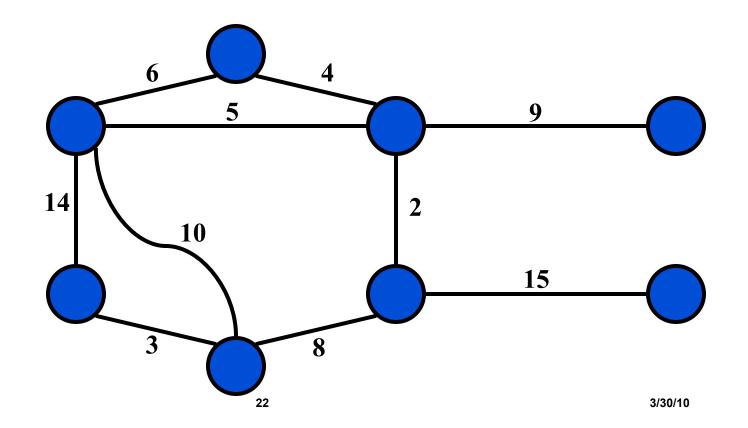
Activity Selection: A Greedy Algorithm

- So actual algorithm is simple:
  - Sort the activities by finish time
  - Schedule the first activity
  - Then schedule the next activity in sorted list which starts after previous activity finishes
  - Repeat until no more activities
- Intuition is even more simple:
  - Always pick next activity that finishes earliest

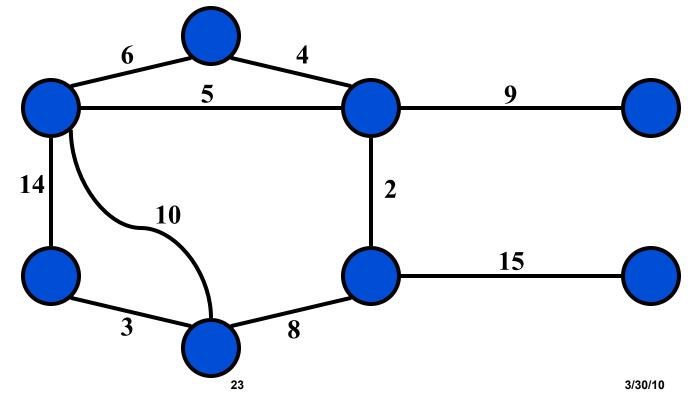
#### Back to Semester at Sea...

ld	Start	End	Len	Activity
2	1	4	4	Tropical Drink Engineering with Prof. Bloomfield
4	3	5	3	Applied ChemE: Suntan Oil or Lotion?
1	0	6	7	Fractals, Recursion and Crayolas
6	5	7	3	Hydrodynamics and Surfing
5	3	8	6	Optimization, Greedy Algorithms, and the Buffet Line
7	5	9	5	Computational Genetics and Infectious Diseases
8	6	10	5	Turing Award Speech Karaoke
9	8	11	4	Pool Tanning for Pale Engineers
10	8	12	5	Mechanics, Dynamics and Shuffleboard Physics
3	2	13	12	Managing Keyboard Fatigue with Swedish Massage
11	12	14	3	Discrete Math Applications in Gambling

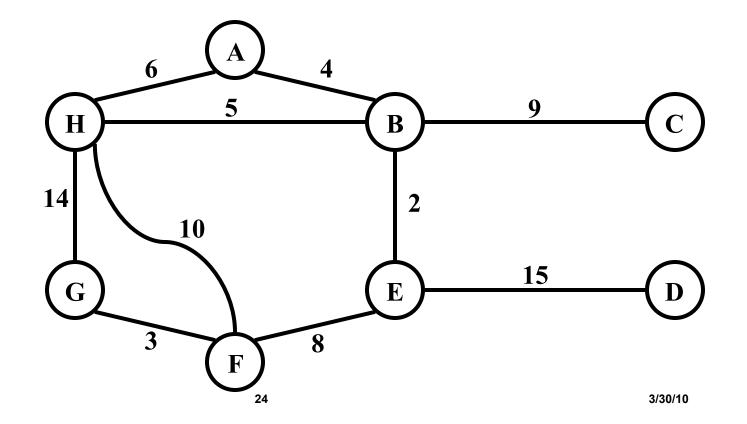
• Problem: given a connected, undirected, weighted graph:



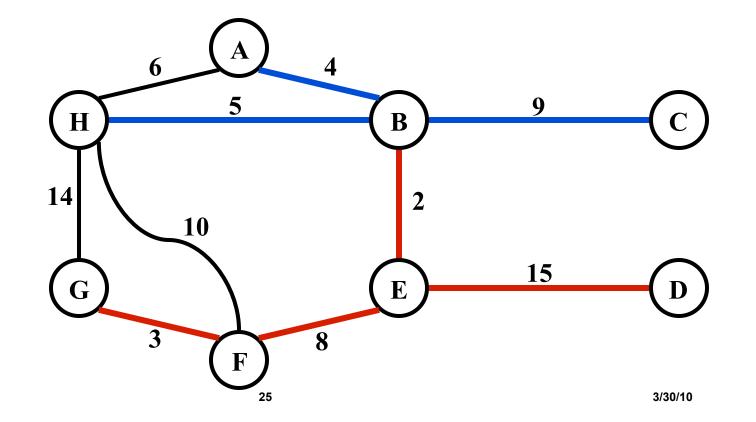
 Problem: given a connected, undirected, weighted graph, find a *spanning tree* using edges that minimize the total weight



• Which edges form the minimum spanning tree *(MST) of the below graph?* 



• Answer:



- MSTs satisfy the *optimal substructure* property. (More on this in Chapter 8.) Here: an optimal tree is composed of optimal subtrees
  - Let T be an MST of G with an edge (u, v) in the middle
  - Removing (u, v) partitions T into two trees  $T_1$  and  $T_2$
  - Claim:  $T_1$  is an MST of  $G_1 = (V_1, E_1)$ , and  $T_2$  is an MST of  $G_2 = (V_2, E_2)$
  - Proof: w(T) = w(u,v) + w(T<sub>1</sub>) + w(T<sub>2</sub>)
     (There can't be a better tree than T<sub>1</sub> or T<sub>2</sub>, or T would be suboptimal)

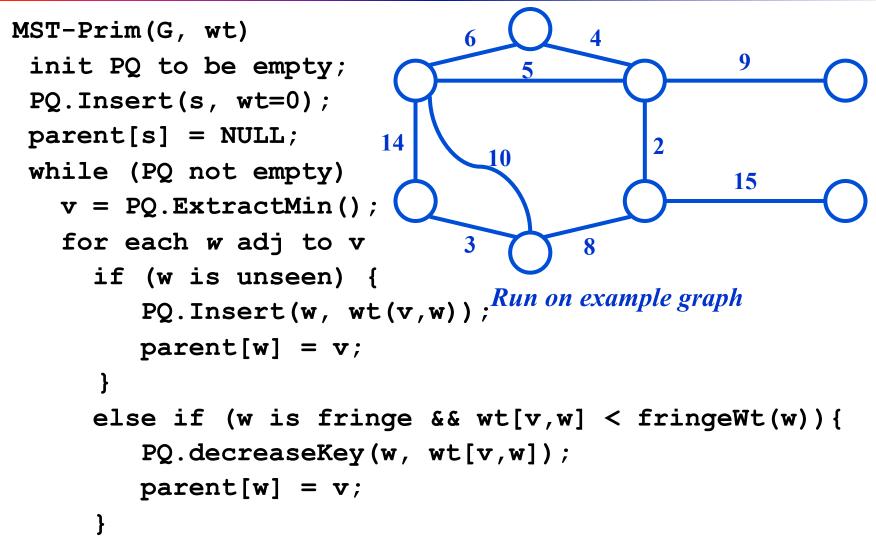
## Prim's MST Algorithm

#### • Greedy strategy:

- Choose some start vertex as current-tree
- Greedy rule: Add edge from graph to current-tree that
  - has the lowest weight of edges that...
  - have one vertex in the tree and one not in the tree.
- Thus builds-up one tree by adding a new edge to it
- Can this lead to an infeasible solution? (Tell me why not.)
- Is it optimal? (Yes. Need a proof.)

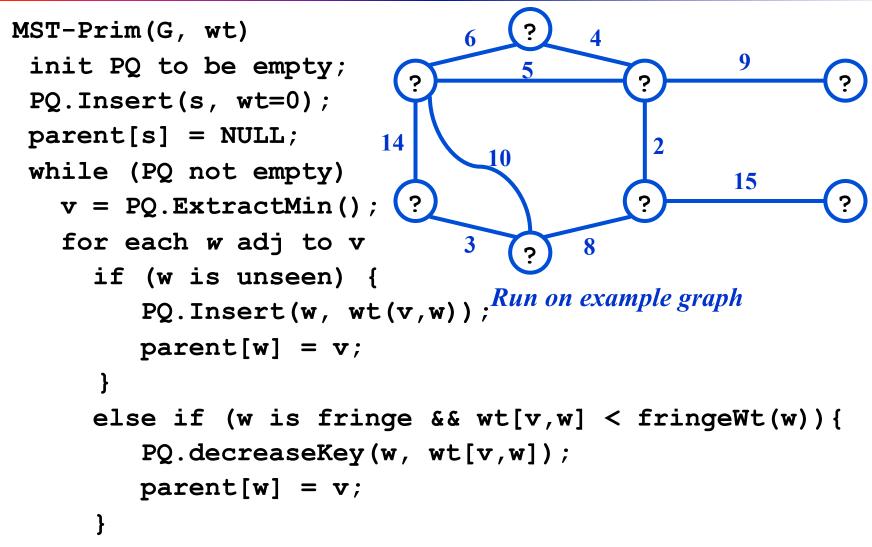
## Tracking Edges for Prim's MST

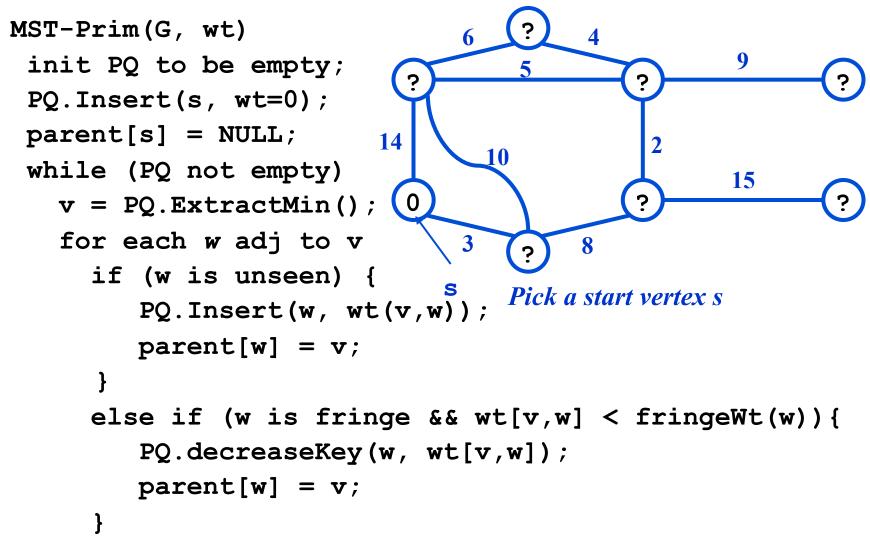
- Candidates edges: edge from a tree-node to a nontree node
  - Since we'll choose smallest, keep only one candidate edge for each non-tree node
  - But, may need to make sure we always have the smallest edge for each non-tree node
- Fringe-nodes: non-trees nodes adjacent to the tree
- Need data structure to hold fringe-nodes
  - Priority queue, ordered by min-edge weight
  - May need to update priorities!

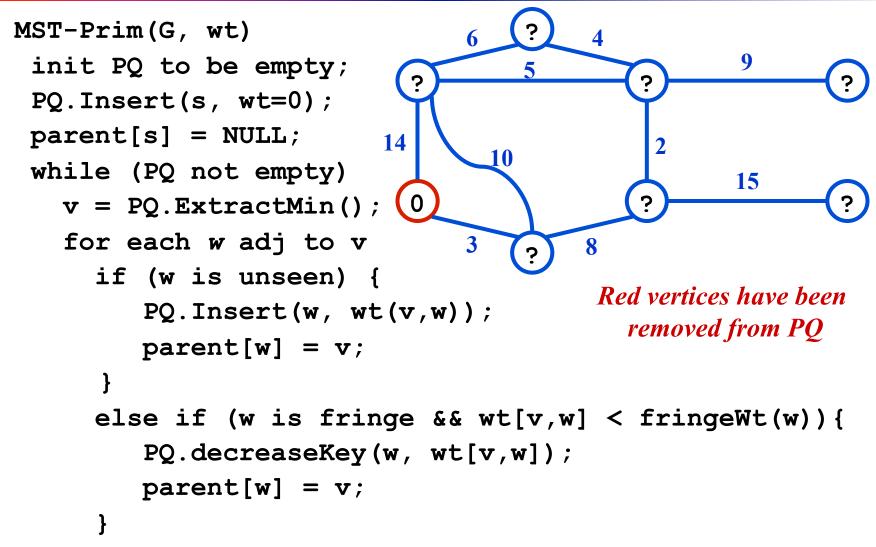


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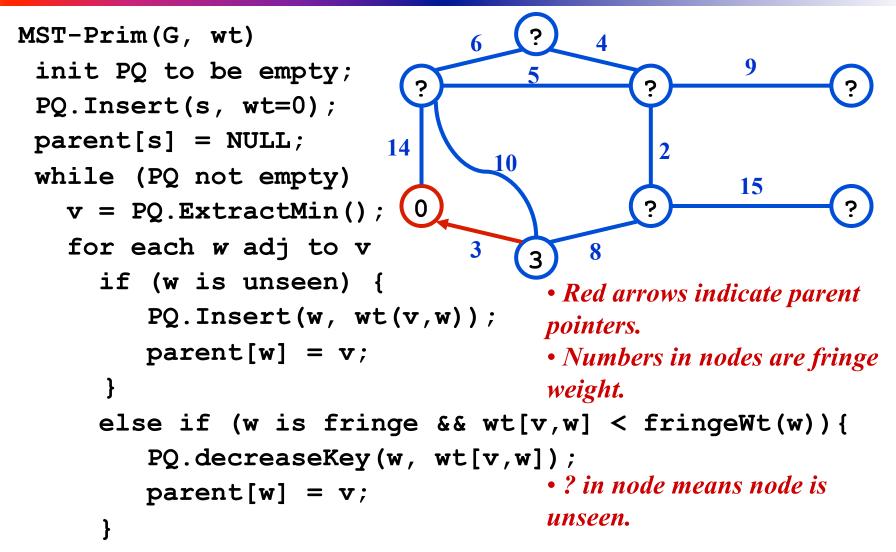
```
prim(adj,start,parent) { // Textbook's code - compare!
   n = adj.last
   for i = 1 to n
      key[i] = \infty // key is a local array
   key[start] = 0
   parent[start] = 0
   // the following statement initializes the
   // container h to the values in the array key
   h.init(key,n)
   for i = 1 to n {
      v = h.del()
       ref = adj[v]
      while (ref != null) {
          w = ref.ver
          if (h.isin(w) && ref.weight < h.keyval(w)) {</pre>
              parent[w] = v
             h.decrease(w,ref.weight)
          }
          ref = ref.next
       }
   }
}
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```



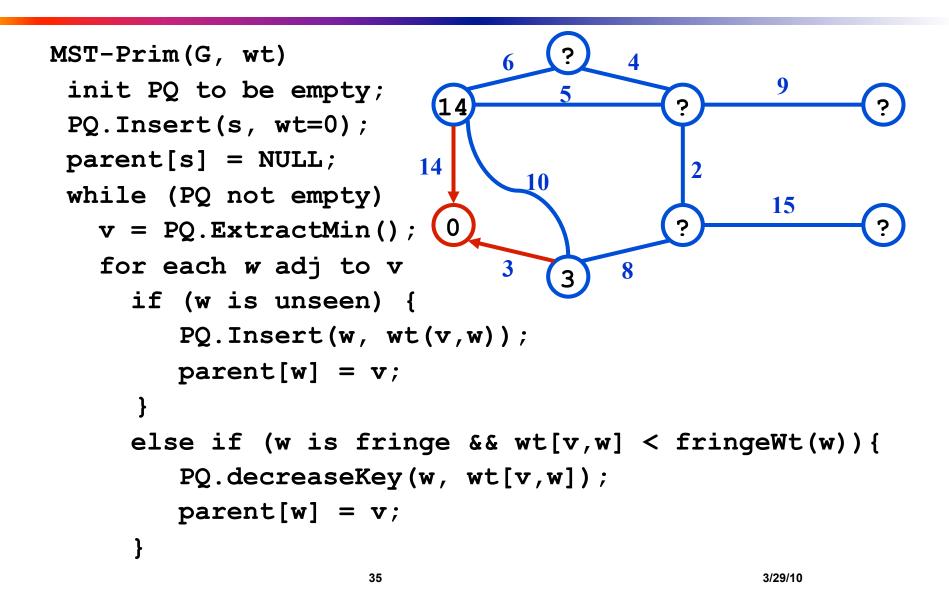


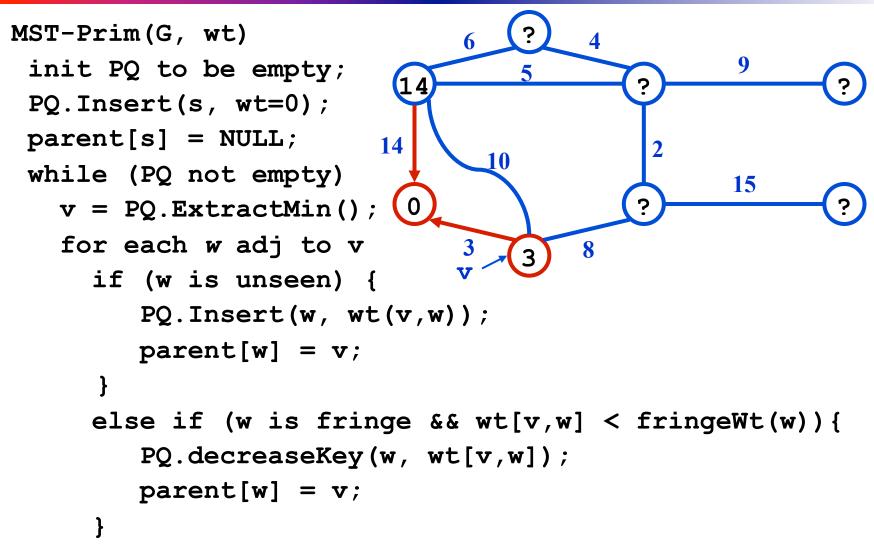


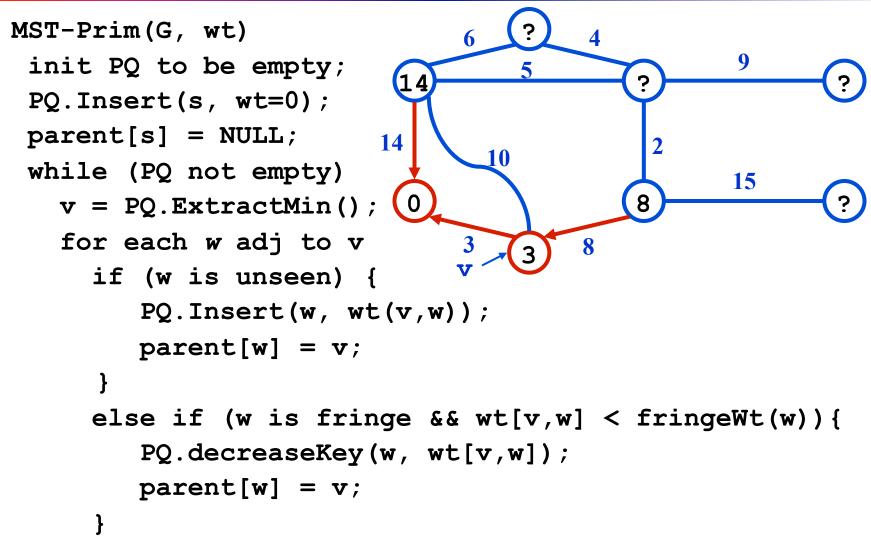
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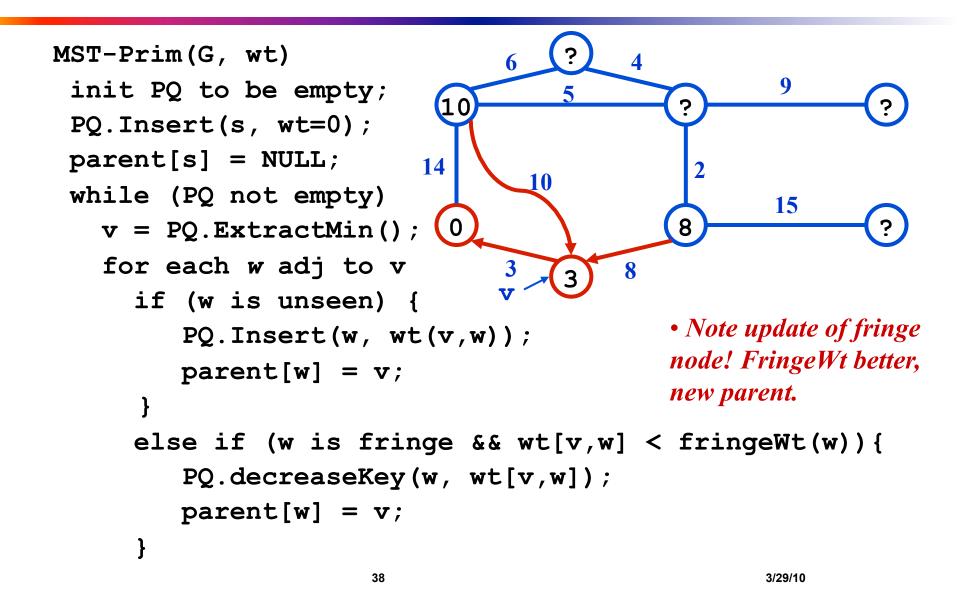


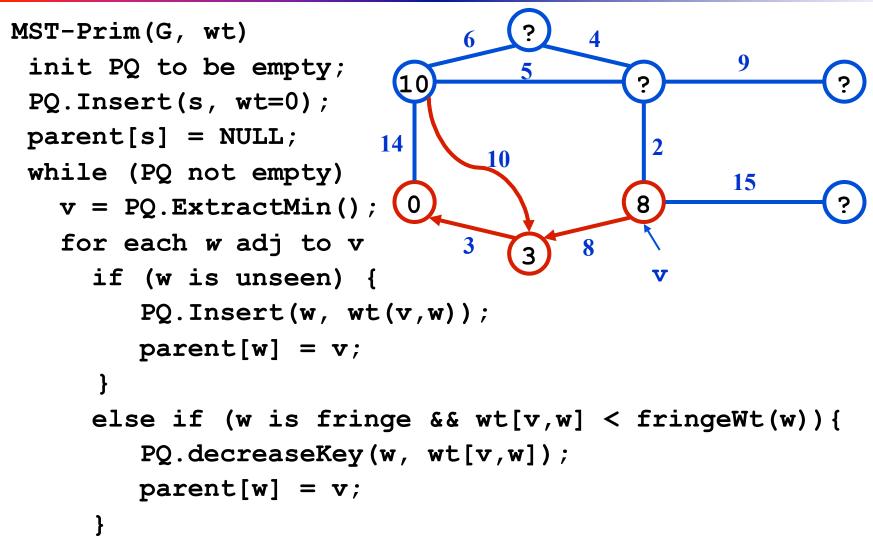
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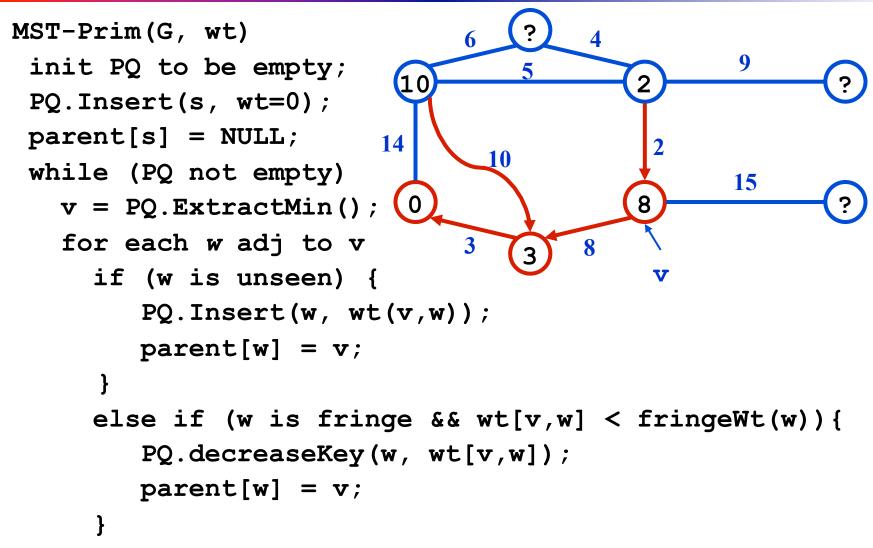


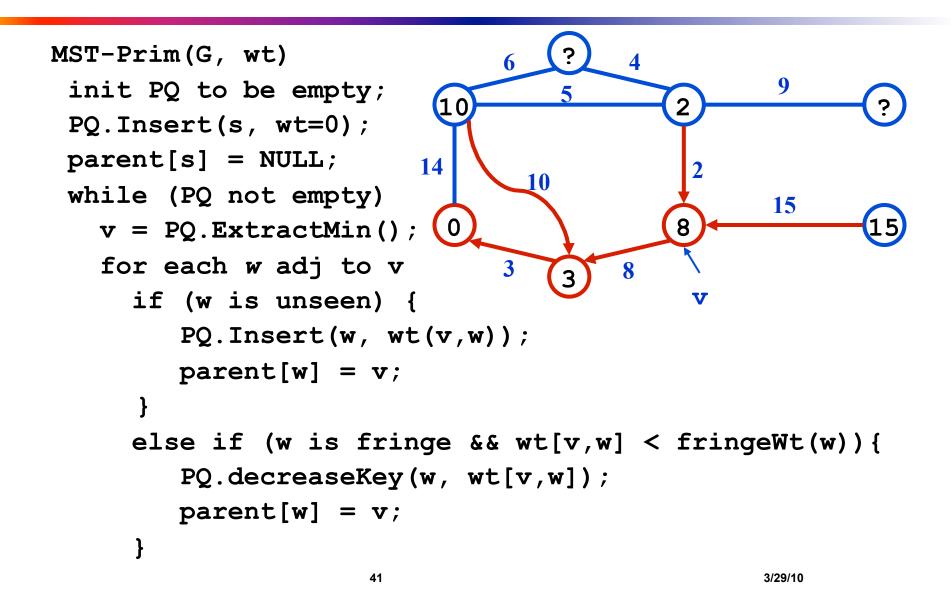


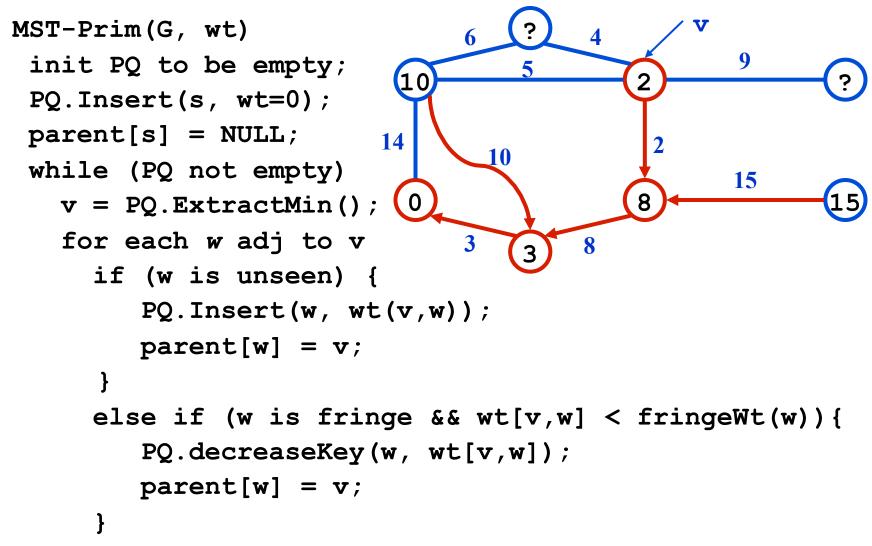


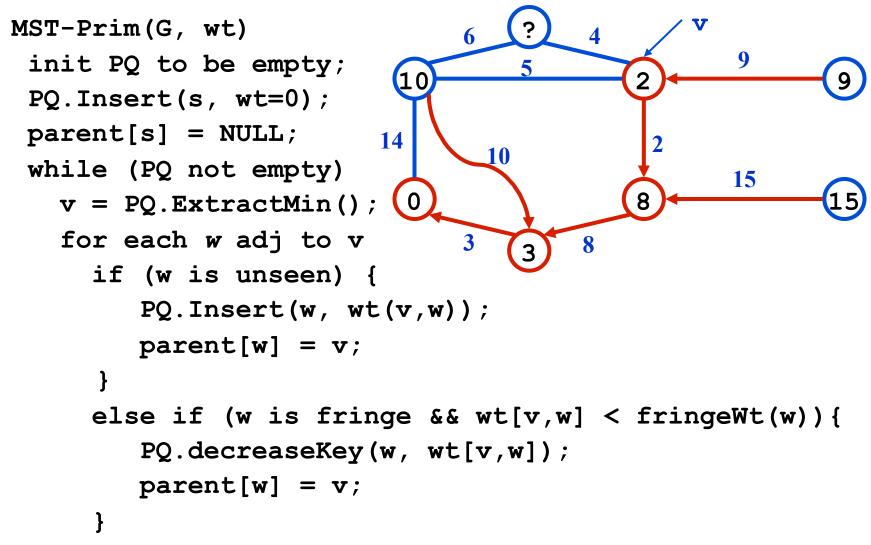


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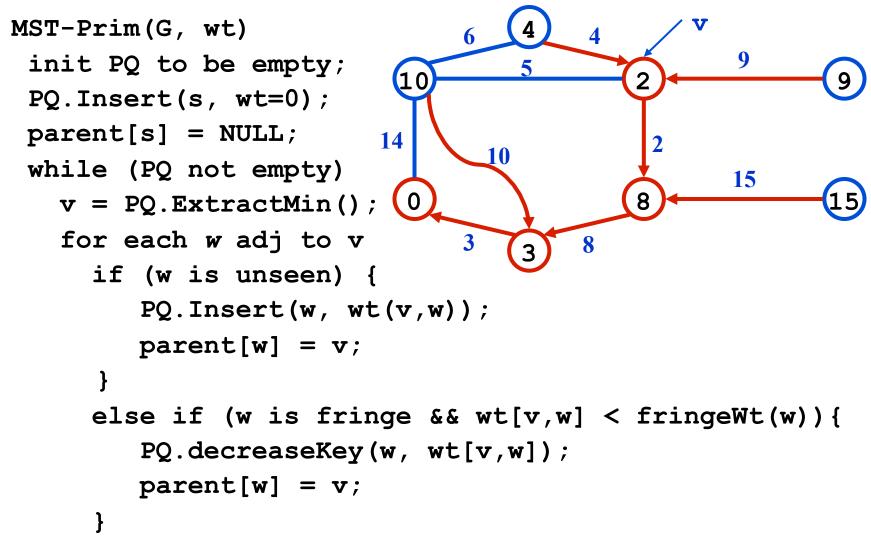


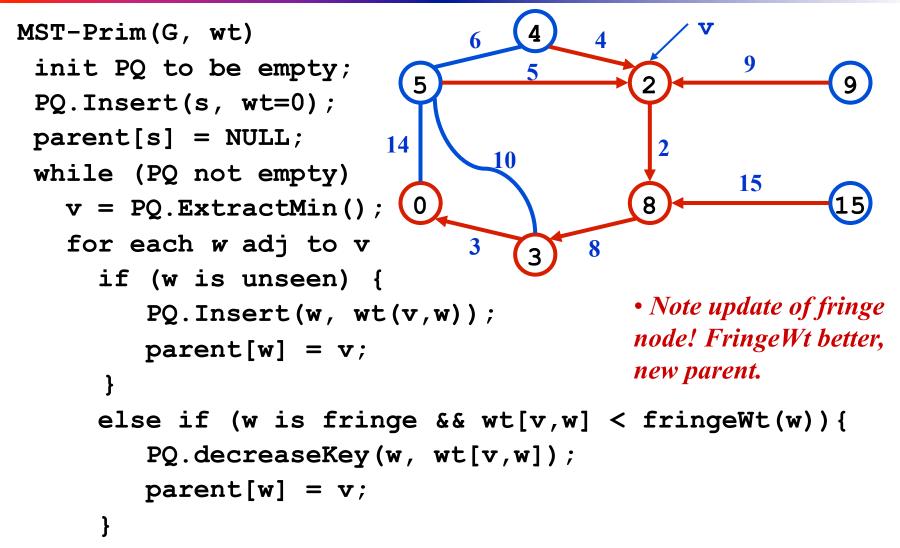




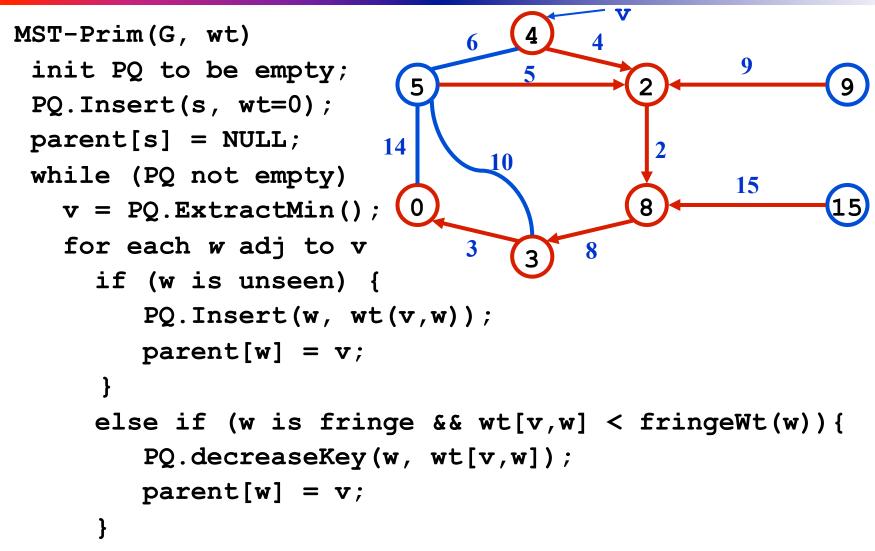


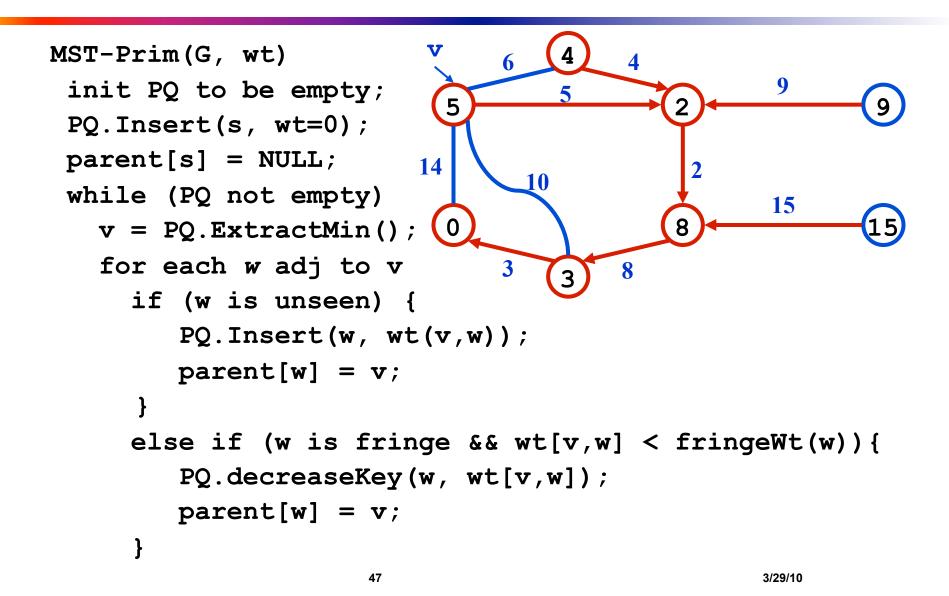
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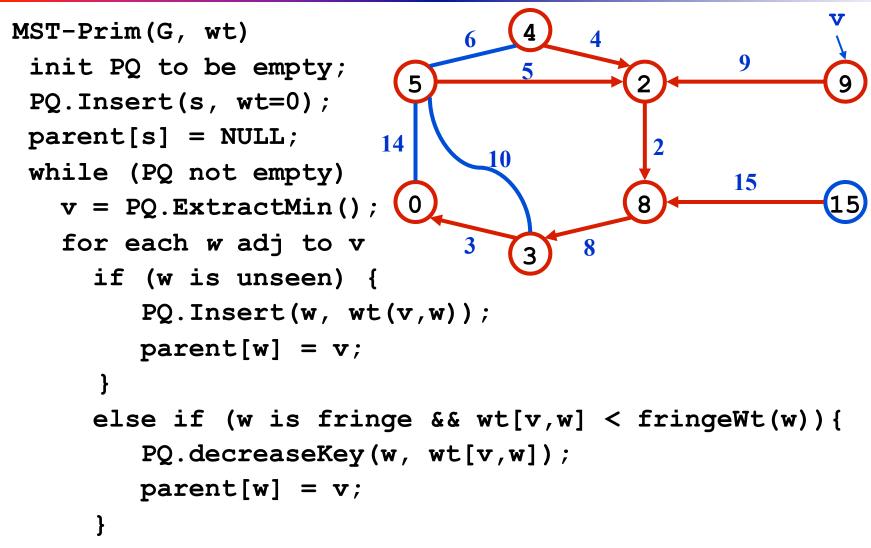


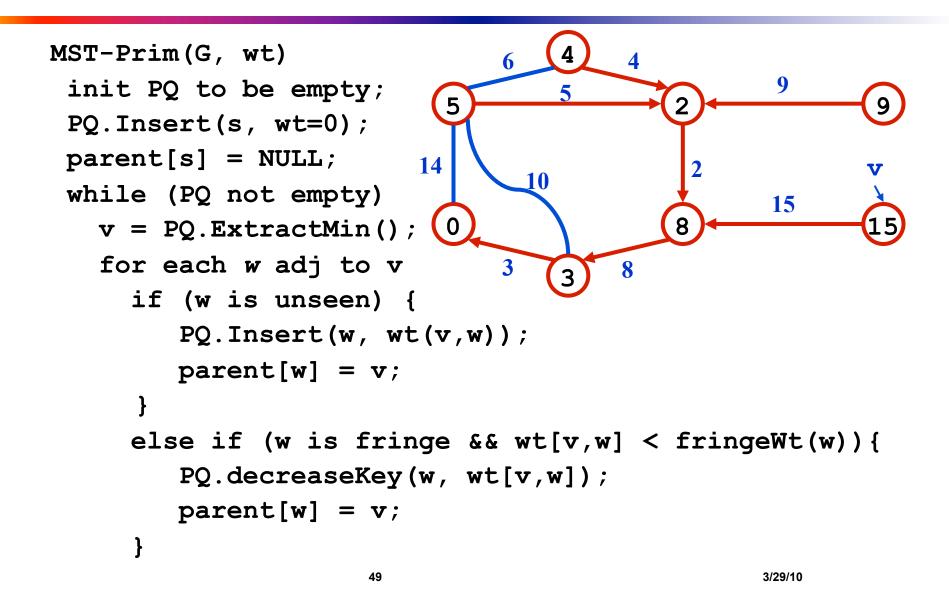


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# Cost of Prim's Algorithm

- (Assume connected graph)
- Clearly it looks at every edge, so  $\Omega(n+m)$
- Is there more?
  - Yes, priority queue operations
  - ExtractMin called n times
    - How expensive? Depends on the size of the PQ
  - descreaseKey could be called for each edge
    - How expensive is each call?

# Worst Case

- If all nodes connected to start, then size of PQ is n-1 right away.
  - Decreases by 1 for each node selected
  - Total cost is O(cost of extractMin for size n-1)
    - Note use of Big-Oh (not Big-Theta)
- Could descreaseKey be called a lot?
  - Yes! Imagine an input that adds all nodes to the PQ at the first step, and then after that calls descreaseKey every possible time. (For you to do.)

### Priority Queue Costs and Prim's

- Simplest choice: unordered list
  - PQ.ExtractMin() is just a "findMin"
    - Cost for one call is  $\Theta(n)$
    - Total cost for all n calls is  $\Theta(n^2)$
  - PQ.decreaseKey() on a node finds it, changes it.
    - Cost for one call is  $\Theta(n)$
    - But, if we can index an array by vertex number, the cost would be Θ(1).
      - If so, worst-case total cost is  $\Theta(m)$
- Conclusion: Easy to get  $\Theta(n^2)$

# **Better PQ Implementations**

- Consider using a min-heap for the Priority Queue
  - PQ.ExtractMin() is O(lg n) each time
    - Called n times, so like Heap's Construct: efficient!
  - What about PQ.decreaseKey()?
- Our need: given a vertex-ID, change the value stored
  - But our basic heap implementation does not allow look-ups based on vertex-ID!
- Solution: Indirect heaps (see pages 142-145)
  - Heap structure stores indices to data in an array that doesn't change
  - Can increase or decrease key in O(lg n) after O(1) lookup

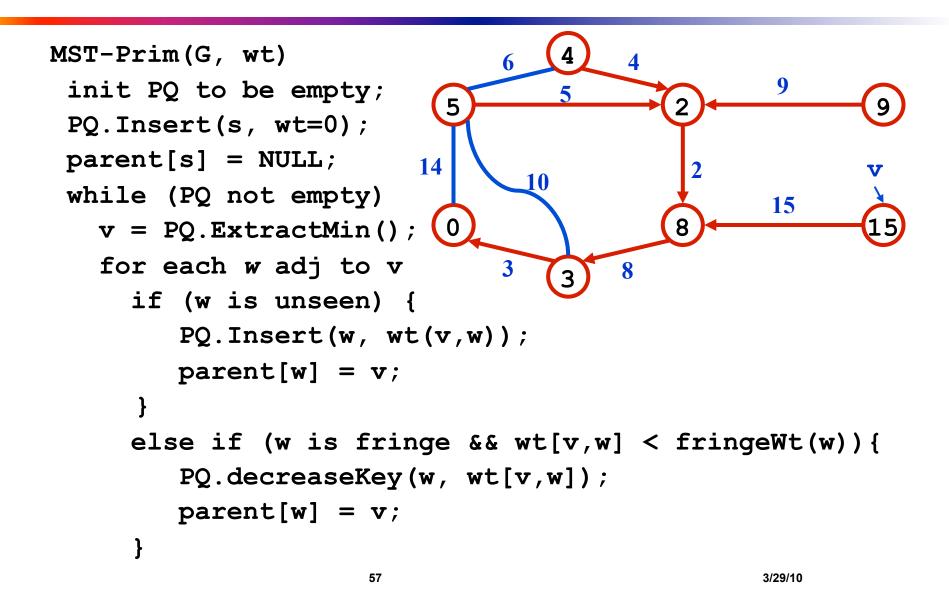
# **Better PQ Implementations (2)**

- Use Indirect Heaps for the PQ
  - PQ.decreaseKey() is O(lg n) also
    - Called for each edge encountered in MST algorithm
    - So O(m x lg n)
    - Overall: Might be better  $\Theta(n^2)$  than if m <<  $n^2$
- Fibonacci heaps: an even more efficient PQ implementation. We won't cover these.
  - $\blacksquare \Theta(m + n \lg n)$

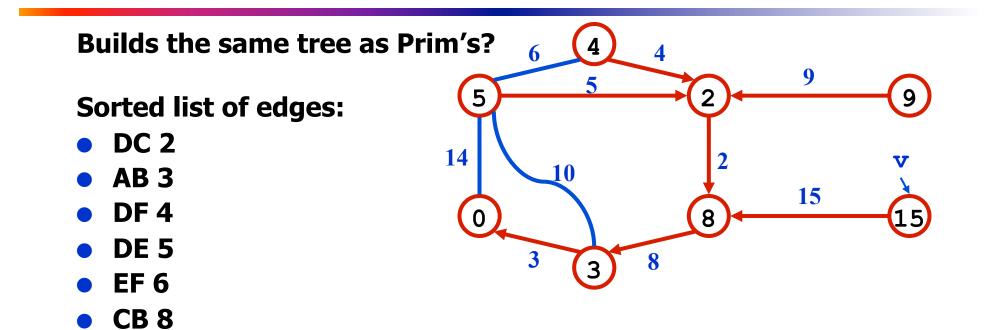
# Kruskal's MST Algorithm

#### • Prim's approach:

- Build one tree. Make the one tree bigger and as good as it can be.
- Kruskal's approach
  - Choose the best edge possible: smallest weight
  - Not one tree maintain a forest!
  - Each edge added will connect two trees. Can't form a cycle in a tree!
  - After adding n-1 edges, you have one tree, the MST



### Kruskal's Algorithm



**DG 9** 

**BE 10** 

**AE 14** 

CH 15

# Strategy for Kruskal's

- EL = sorted set of edges ascending by weight
- Foreach edge e in EL
  - T1 = tree for head(e)
  - T2 = tree for tail(e)
  - If (T1 != T2)
    - add e to the output (the MST)
    - Combine trees T1 and T2
- Seems simple, no?
  - But, how do you keep track of what trees a node is in?
  - Trees are sets. Need to findset(v) and "union" two sets

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```
kruskal(edgelist,n) {
   sort(edgelist)
   for i = 1 to n
      makeset(i)
   count = 0
   i = 1
  while (count < n - 1) {
      if (findset(edgelist[i].v) !=
                 findset(edgelist[i].w)) {
         println(edgelist[i].v + "
                  + edgelist[i].w)
            count = count + 1
            union(edgelist[i].v,edgelist[i].w)
      }
i = i + 1
   }
}
```

# **Union/Find and Disjoint Sets**

- See Section 3.6, page 150-161
- Sets stored as a parent array (see bottom of p. 151)
  - findset(v): trace upward in parent array
  - union(i,j): make one tree a child of a node it the other
- Improvements! E.g. path compression
   O(lg m)

### Complexity for Kruskal's

• Overall:  $\Theta(m \lg m)$ 

# Single Source Shortest Path

- Problem: Given a node v, find the minimum distance from v to *either* another node w *or* to all other nodes, where distance is the sum of the edge-weights on the path
- A solution: Dijkstra's algorithm
  - Who's Dijkstra? See class wall-of-fame!

### Dijkstra's Shortest Path Algorithm

• Identical *in structure* to <u>Prim's</u> MST algorithm

- Of course it solves a different problem!
- Same time complexity
- Additional input parameter(s)
  - Start node v
  - Destination node w (if needed)
- Different output: a path from v to w and a cost (or sets of paths and costs)

■ The tree is the sets of shortest paths to nodes

- Different greedy strategy:
  - Store shortest paths to fringe-nodes in priority queue
  - Store path-distance to node, not just the one edge-weight

### Reminder: Prim's Algorithm

```
MST-Prim(G, wt)
 init PQ to be empty;
 PQ.Insert(s, wt=0);
 parent[s] = NULL;
 while (PQ not empty) {
   v = PQ.ExtractMin();
   for each w adj to v
     if (w is unseen) {
        PQ.Insert(w, wt(v,w));
        parent[w] = v;
     }
     else if (w is fringe && wt[v,w] < fringeWt(w)) {</pre>
        PQ.decreaseKey(w, wt[v,w]);
        parent[w] = v;
     }
```

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# Dijkstra' Algorithm

```
dijkstra(G, wt, s)
 init PQ to be empty;
 PQ.Insert(s, dist=0);
 parent[s] = NULL; dist[s] = 0;
 while (PQ not empty)
   v = PQ.ExtractMin();
   for each w adj to v
     if (w is unseen) {
        dist[w] = dist[v] + wt(v,w)
        PQ.Insert(w, dist[w] );
        parent[w] = v;
      }
     else if (w is fringe && dist[v] + wt(v,w) < dist[w] )</pre>
   {
        dist[w] = dist[v] + wt(v,w)
        PQ.decreaseKey(w, dist[w]);
        parent[w] = v;
     }
```

# Notes on Dijkstra's Algorithm

- Use dist[] to store distances from start to any fringe or tree node
- Store and calculate using distances instead of edge-weights (like in Kruskal's MST)
- What's the output?
  - Tree captured in the parent[] array
  - Shortest distance to each node in dist[] array
  - Trace shortest path in reverse by using parent[] to move from target back to start node, s

```
dijkstra(adj, start, parent) {
   n = adj.last
   for i = 1 to n { key[i] = \infty } // key is a local array
   key[start] = 0; predecessor[start] = 0
   // the following statement initializes the
   // container h to the values in the array key
   h.init(key,n)
   for i = 1 to n {
         v = h.min weight index()
         min_cost = h.keyval(v)
         v = h.del()
         ref = adj[v]
         while (ref != null) {
                  w = ref.ver
                  if (h.isin(w) && min_cost + ref.weight < h.keyval(w)) {</pre>
                     predecessor[w] = v
                     h.decrease(w, min_cost+ref.weight)
                  } // end if
                  ref = ref.next
         } // end while
   } // end for
}
```

### Correctness of These Greedy Algorithms

- Recall that the greedy approach may or may not guarantee an optimal result
- Do these produce optimal solutions?
  - The min weight spanning tree? Kruskal's, Prim's
  - The shortest path from s? Dijkstra's
- Answer: Yes, they do.
  - Proofs in the text
  - Proofs by induction, also using proof by contradiction