Algorithm strategies: Greedy algorithms

Optimization problems

- An optimization problem is one in which you want to find, not just a solution, but the best solution
- A "greedy algorithm" sometimes works well for optimization problems
- A greedy algorithm works in phases. At each phase:
 - You take the **best you can get right now**, without regard for future consequences
 - You hope that by choosing a local optimum at each step, you will end up at a global optimum

• The greedy approach suggests constructing a solution through a sequence of steps, each expanding a partially constructed solution obtained so far, until a complete solution to the problem is reached.

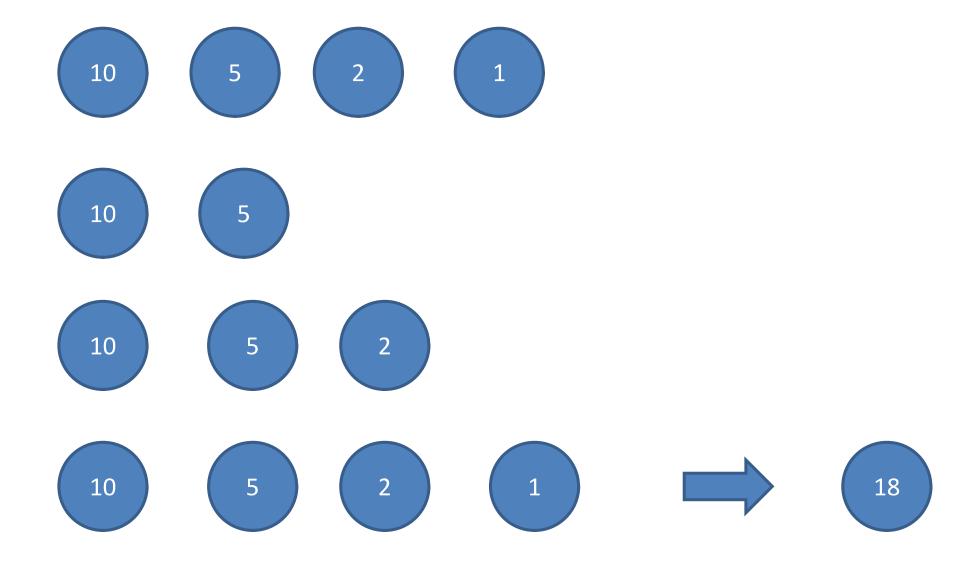
- The central point of this technique—the choice made must be:
- Feasible, i.e., it has to satisfy the problem's constraints
- Locally optimal, i.e., it has to be the best local choice among all feasible choices available on that step
- Irrevocable, i.e., once made, it cannot be changed on subsequent steps of the algorithm

Change-making problem

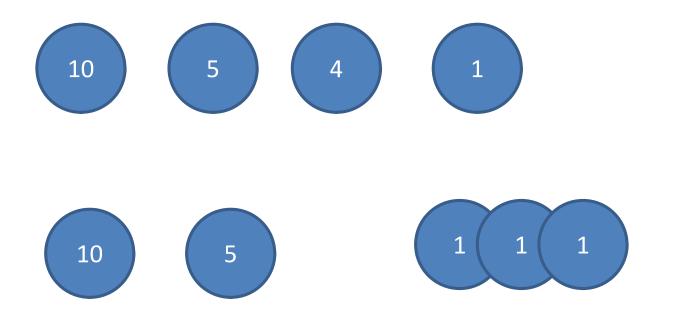
• Give change for a specific amount **n** with the least number of coins of the denominations



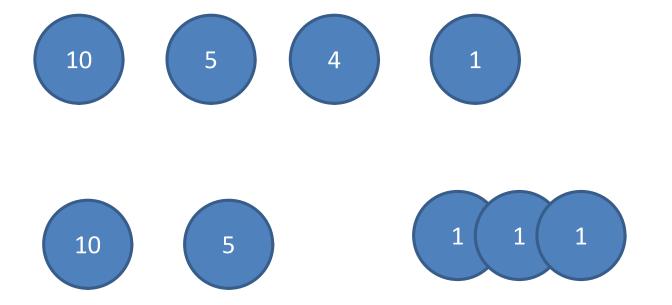
• Get 18 with least number of coins







Is this optimal?



Not optimal



Knapsack Problem

- Given n items of known weights w1,..., wn and values v1,..., vn and a knapsack of capacity W,
- find the most valuable subset of the items that fit into the knapsack.
- Given a set of items, each with a weight and a value, determine a subset of items to include in a collection so that the total weight is less than or equal to a given limit and the total value is as large as possible.
- The knapsack problem is in combinatorial optimization problem.

Problem Scenario

- A thief is robbing a store and can carry a maximal weight of W into his knapsack.
- There are **n** items available in the store and weight of **i**th item is **w**_i and its profit is **p**_i. What items should the thief take?
- In this context, the items should be selected in such a way that the thief will carry those items for which he will gain maximum profit.
- Based on the nature of the items, Knapsack problems are categorized as
 - Fractional Knapsack
 - 0-1 Knapsack

0-1 Knapsack

Let us consider that the capacity of the knapsack W = 60 and the list of provided items are shown in the following table.

Item	A	В	С	D
Profit (pi)	280	100	120	50
Weight(wi)	40	10	20	10
Ratio (pi/wi)	7	10	6	5

0-1 Knapsack

After sorting, the items are as shown in the following table.

ltem	В	Α	С	D
Profit (pi)	100	280	120	50
Weight(wi)	10	40	20	10
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The total weight of the selected items is 10 + 40 + 10 = 60And the total profit is 100 + 280 + 50 = 380 + 50 = 430

Fractional Knapsack

After sorting, the items are as shown in the following table.

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Profit (pi)	100	280	120	50
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The total weight of the selected items is 10 + 40 + 20 * (10/20) = 60And the total profit is 100 + 280 + 120 * (10/20) = 380 + 60 = 440

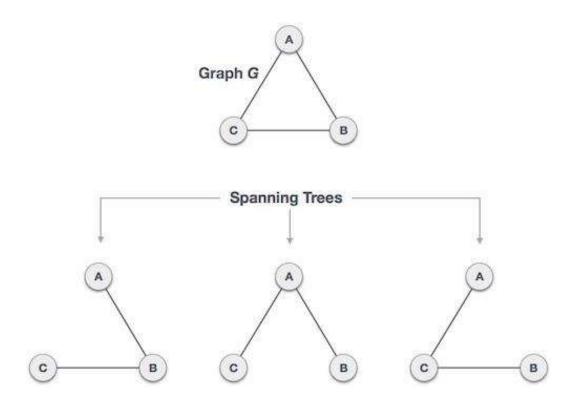
Fractional Knapsack

O(n)

```
Algorithm: Greedy-Fractional-Knapsack (w[1..n], p[1..n], W)
for i = 1 to n do
       x[i] = 0
weight = 0
for i = 1 to n
                                                                O(n logn)
       if weight + w[i] \le W then
                x[i] = 1
                weight = weight + w[i]
        else
                x[i] = (W - weight) / w[i]
                weight = W
                break
return x
```

spanning tree

• A spanning tree of an undirected connected graph is its connected acyclic sub graph (i.e., a tree) that contains all the vertices of the graph.

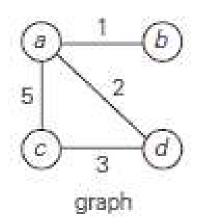


minimum spanning tree problem

• If a undirected connected graph has weights assigned to its edges, a minimum spanning tree is its spanning tree of the smallest weight, where the weight of a tree is defined as the sum of the weights on all its edges.

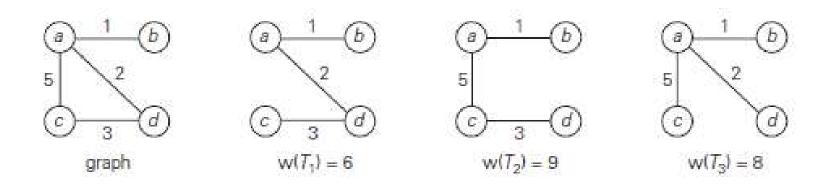
• The minimum spanning tree problem is the problem of finding a minimum spanning tree for a given weighted connected graph.

minimum spanning tree problem



Find minimum spanning tree using brute force approach.

minimum spanning tree problem



T₁ is the minimum spanning tree.

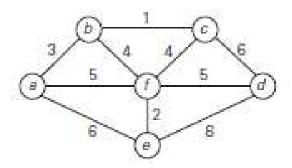
- Prim's algorithm constructs a minimum spanning tree through a sequence of expanding sub-trees.
- The initial sub-tree in such a sequence consists of a single vertex selected arbitrarily from the set V of the graph's vertices.
- On each iteration, the algorithm expands the current tree in the greedy manner by simply attaching to it the nearest vertex not in that tree. (By the nearest vertex, we mean a vertex not in the tree connected to a vertex in the tree by an edge of the smallest weight.)
- The algorithm stops after all the graph's vertices have been included in the tree being constructed.

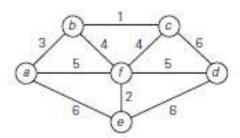
ALGORITHM Prim(G)

```
//Prim's algorithm for constructing a minimum spanning tree //Input: A weighted connected graph G = \langle V, E \rangle //Output: E_T, the set of edges composing a minimum spanning tree of G V_T \leftarrow \{v_0\} //the set of tree vertices can be initialized with any vertex E_T \leftarrow \varnothing for i \leftarrow 1 to |V| - 1 do find a minimum-weight edge e^* = (v^*, u^*) among all the edges (v, u) such that v is in V_T and u is in V - V_T V_T \leftarrow V_T \cup \{u^*\} E_T \leftarrow E_T \cup \{e^*\} return E_T
```

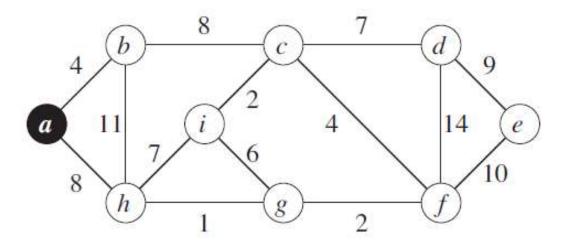
After we have identified a vertex u^* to be added to the tree, we need to perform two operations:

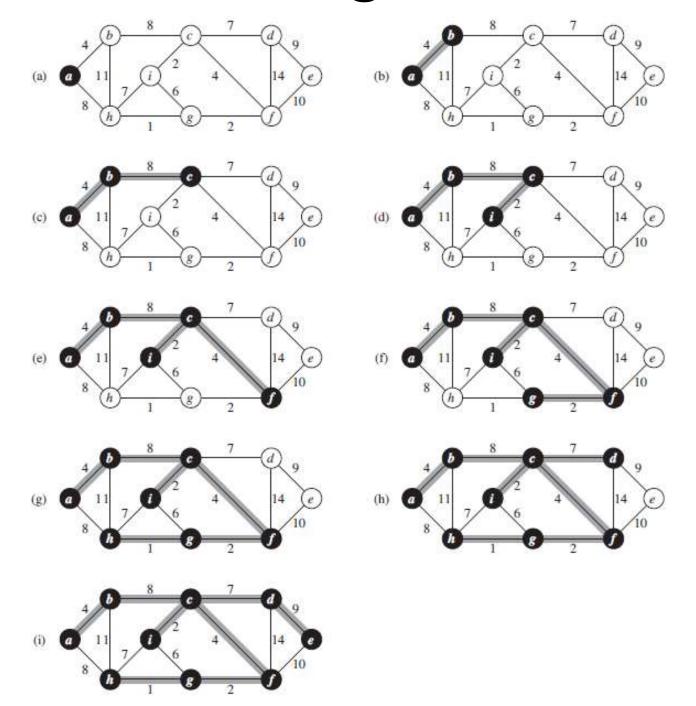
- 1. Move \mathbf{u}^* from the set $\mathbf{V} \mathbf{V}\mathbf{T}$ to the set of tree vertices $\mathbf{V}\mathbf{T}$.
- 2. For each remaining vertex u in V VT that is connected to u* by a shorter edge than the u's current distance label, update its labels by u* and the weight of the edge between u* and u, respectively.





Tree vertices	Remaining vertices	Illustration
a(-, -)	$b(a, 3) \ c(-, \infty) \ d(-, \infty)$ e(a, 6) f(a, 5)	3 5 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6
b(a, 3)	$c(b, 1) d(-, \infty) e(a, 6)$ f(b, 4)	3 5 f 5 d 6 d 6 d
c(b, 1)	d(c, 6) e(a, 6) f(b, 4)	3 5 f 5 d 5 d
f(b, 4)	d(f, 5) e(f, 2)	3 5 f 5 d 5 d
e(f, 2)	d(f, 5)	3 5 1 C 6 6 6 D 8
d(f, 5)		<u>~</u>





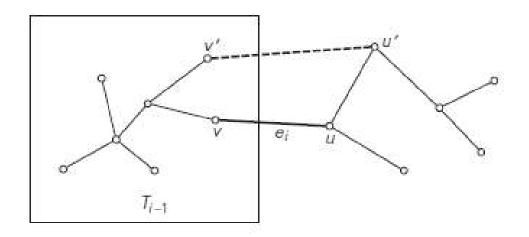
• Does Prim's algorithm always yield a minimum spanning tree?

• Does Prim's algorithm always yield a minimum spanning tree?

The answer to this question is yes.

- Let us prove by induction that each of the sub-trees T_i,
- i = 0, ..., n 1, generated by Prim's algorithm is a part (i.e., a sub-graph) of some minimum spanning tree.
- To consists of a single vertex and hence must be a part of any minimum spanning tree.
- For the inductive step, let us assume that T_{i-1} is part of some minimum spanning tree T.
- We need to prove that T_i, generated from T_{i-1} by Prim's algorithm, is also a part of a minimum spanning tree

• By contradiction by assuming that no minimum spanning tree of the graph can contain T_i .



```
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E_T \leftarrow \varnothing

for i \leftarrow 1 to |V| - 1 do

find a minimum-weight edge e^* = (v^*, u^*) among all the edges (v, u)

such that v is in V_T and u is in V - V_T

V_T \leftarrow V_T \cup \{u^*\}

E_T \leftarrow E_T \cup \{e^*\}

return E_T

MST-PRIM(G, w, r)
```

Compute the efficiency of the algorithm

```
1 for each u \in G.V

2 u.key = \infty

3 u.\pi = \text{NIL}

4 r.key = 0

5 Q = G.V

6 while Q \neq \emptyset

7 u = \text{EXTRACT-MIN}(Q)

8 for each v \in G.Adj[u]

9 if v \in Q and w(u, v) < v.key

10 v.\pi = u

11 v.key = w(u, v)
```

```
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```

Compute the efficiency of the algorithm

The answer depends on the data structures chosen for the graph itself and for the priority queue of the set V-VT whose vertex priorities are the distances to the nearest tree vertices.

```
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```

If a graph is represented by its weight matrix and the priority queue is implemented as an unordered array

The algorithm's running time will be in $O(|V|^2)$

On each of the |V| - 1 iterations, the array implementing the priority queue is traversed to find and delete the minimum and then to update, if necessary, the priorities of the remaining vertices.

```
MST-KRUSKAL(G, w)

1 A = \emptyset

2 for each vertex v \in G.V

3 MAKE-SET(v)

4 sort the edges of G.E into nondecreasing order by weight w

5 for each edge (u, v) \in G.E, taken in nondecreasing order by weight

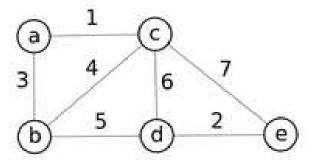
6 if FIND-SET(u) \neq FIND-SET(v)

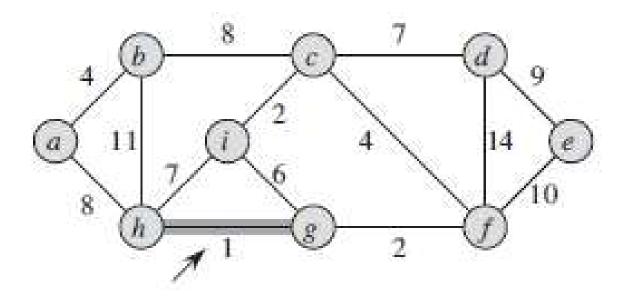
7 A = A \cup \{(u, v)\}

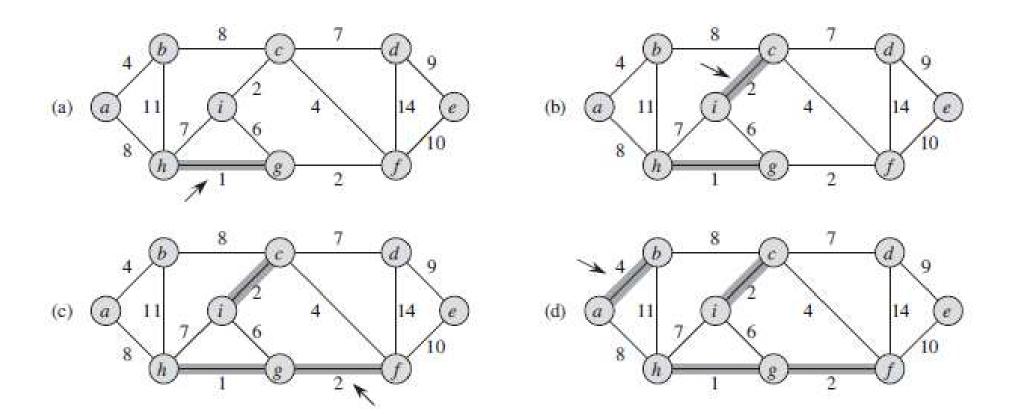
UNION(u, v)

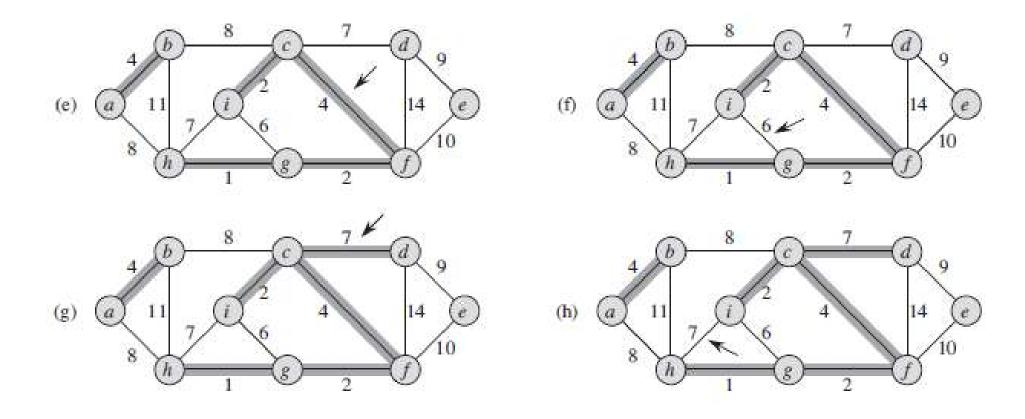
9 return A
```

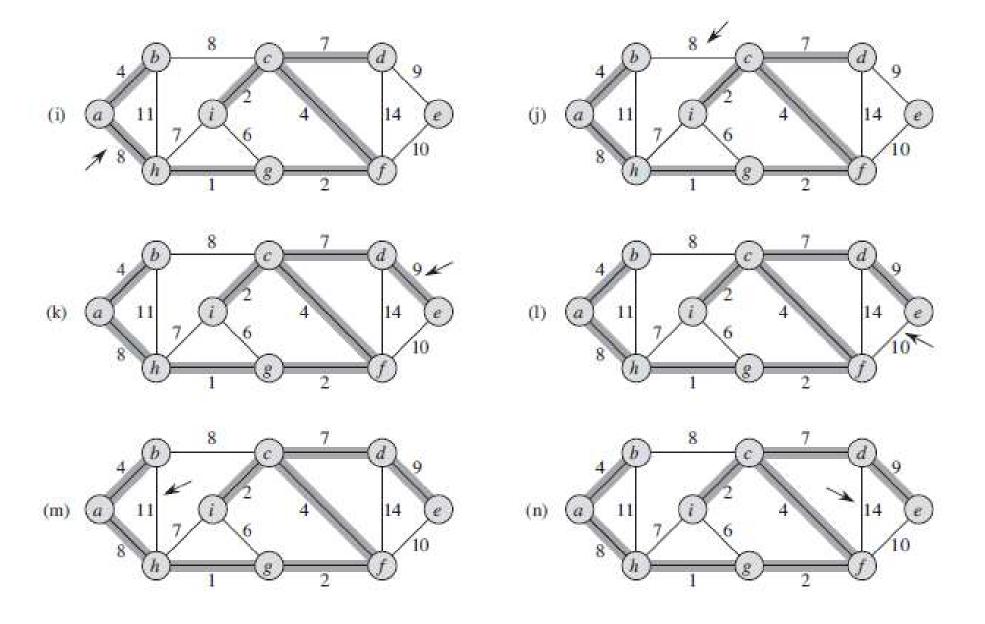
 $O(E \lg E)$.











Thank you