

CSE 431/531: Algorithm Analysis and Design (Spring 2019)

## Greedy Algorithms

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## Main Goal of Algorithm Design

- Design fast algorithms to solve problems
- Design **more efficient** algorithms to solve problems

**Def.** The goal of an **optimization problem** is to find a valid solution with the minimum (or maximum) cost (or value).

## Trivial Algorithm for an Optimization Problem

Enumerate all valid solutions, compare them and output the best one.

- However, trivial algorithm often runs in **exponential** time, as the number of potential solutions is often exponentially large.
- $f(n)$  is polynomial if  $f(n) = O(n^k)$  for some constant  $k > 0$ .
- convention: polynomial time = efficient

# Common Paradigms for Algorithm Design

- Greedy Algorithms
- Divide and Conquer
- Dynamic Programming

## Greedy Algorithm

- Build up the solutions in steps
- At each step, make an **irrevocable** decision using a “reasonable” strategy

## Analysis of Greedy Algorithm

- Prove that the reasonable strategy is “safe” (**key**)
- Show that the remaining task after applying the strategy is to solve a (many) smaller instance(s) of the same problem (**usually trivial**)

# Outline

- 1 Toy Examples
- 2 Interval Scheduling
- 3 Minimum Spanning Tree
  - Kruskal's Algorithm
  - Reverse-Kruskal's Algorithm
  - Prim's Algorithm
- 4 Single Source Shortest Paths
  - Dijkstra's Algorithm
- 5 Data Compression and Huffman Code
- 6 Summary

# Toy Problem 1: Bill Changing

**Input:** Integer  $A \geq 0$

Currency denominations: \$1, \$2, \$5, \$10, \$20

**Output:** A way to pay  $A$  dollars using **fewest** number of bills

Example:

- Input: 48
- Output: 5 bills,  $\$48 = \$20 \times 2 + \$5 + \$2 + \$1$

Cashier's Algorithm

- 1 while  $A \geq 0$  do
- 2      $a \leftarrow \max\{t \in \{1, 2, 5, 10, 20\} : t \leq A\}$
- 3     pay a  $\$a$  bill
- 4      $A \leftarrow A - a$

## Greedy Algorithm

- Build up the solutions in steps
  - At each step, make an **irrevocable** decision using a “reasonable” strategy
- 
- strategy: choose the largest bill that does not exceed  $A$
  - the strategy is “reasonable”: choosing a larger bill help us in minimizing the number of bills
  - The decision is irrevocable : once we choose a  $\$a$  bill, we let  $A \leftarrow A - a$  and proceed to the next

## Analysis of Greedy Algorithm

- Prove that the reasonable strategy is “safe”
- Show that the remaining task after applying the strategy is to solve a (many) smaller instance(s) of the same problem
- $n_1, n_2, n_5, n_{10}, n_{20}$ : number of \$1, \$2, \$5, \$10, \$20 bills paid
- minimize  $n_1 + n_2 + n_5 + n_{10} + n_{20}$  subject to  $n_1 + 2n_2 + 5n_5 + 10n_{10} + 20n_{20} = A$

### Obs.

- $n_1 < 2$   $2 \leq A < 5$ : pay a \$2 bill
- $n_1 + 2n_2 < 5$   $5 \leq A < 10$ : pay a \$5 bill
- $n_1 + 2n_2 + 5n_5 < 10$   $10 \leq A < 20$ : pay a \$10 bill
- $n_1 + 2n_2 + 5n_5 + 10n_{10} < 20$   $20 \leq A < \infty$ : pay a \$20 bill



## Analysis of Greedy Algorithm

- Prove that the reasonable strategy is “safe”
- Show that the remaining task after applying the strategy is to solve a (many) smaller instance(s) of the same problem
- Trivial: in residual problem, we need to pay  $A' = A - a$  dollars, using the fewest number of bills

# Toy Example 2: Box Packing

## Box Packing

**Input:**  $n$  boxes of capacities  $c_1, c_2, \dots, c_n$

$m$  items of sizes  $s_1, s_2, \dots, s_m$

Can put **at most 1** item in a box

Item  $j$  can be put into box  $i$  if  $s_j \leq c_i$

**Output:** A way to put as many items as possible in the boxes.

## Example:

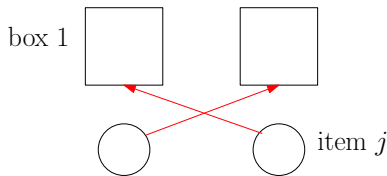
- Box capacities: 60, 40, 25, 15, 12
- Item sizes: 45, 42, 20, 19, 16
- Can put 3 items in boxes: 45  $\rightarrow$  60, 20  $\rightarrow$  40, 19  $\rightarrow$  25

# Box Packing: Design a Safe Strategy

**Q:** Take box 1 (with capacity  $c_1$ ). Which item should we put in box 1?

**A:** The item of the largest size that can be put into the box.

- putting the item gives us the **easiest residual problem**.
- formal proof via **exchanging argument**:  $j =$  largest item that can be put into box 1.



- Residual task: solve the instance obtained by removing box 1 and item  $j$

### Greedy Algorithm for Box Packing

- 1  $T \leftarrow \{1, 2, 3, \dots, m\}$
- 2 for  $i \leftarrow 1$  to  $n$  do
- 3     if some item in  $T$  can be put into box  $i$ , then
- 4          $j \leftarrow$  the largest item in  $T$  that can be put into box  $i$
- 5         print("put item  $j$  in box  $i$ ")
- 6          $T \leftarrow T \setminus \{j\}$

## Steps of Designing A Greedy Algorithm

- Design a “reasonable” strategy
- Prove that the reasonable strategy is “safe” (key, usually done by “exchanging argument”)
- Show that the remaining task after applying the strategy is to solve a (many) smaller instance(s) of the same problem (usually trivial)

**Def.** A choice is “safe” if there is an optimum solution that is “consistent” with the choice

Exchanging argument: let  $S$  be an arbitrary optimum solution. If  $S$  is consistent with the greedy choice, we are done. Otherwise, modify it to another optimum solution  $S'$  such that  $S'$  is consistent with the greedy choice.

## Generic Greedy Algorithm

- 1 **while** the instance is non-trivial
- 2 make the choice using the greedy strategy
- 3 reduce the instance

Algorithm is correct **if and only if** the greedy strategy is safe.

- Greedy strategy is safe: we will not miss the optimum solution
- Greedy strategy is not safe: we will miss the optimum solution for some instance, since the choices we made are irrevocable.

# Outline

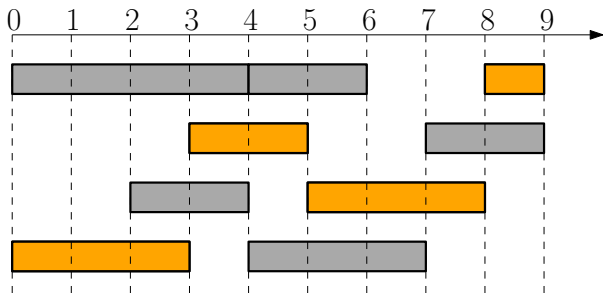
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## Interval Scheduling

**Input:**  $n$  jobs, job  $i$  with start time  $s_i$  and finish time  $f_i$

$i$  and  $j$  are **compatible** if  $[s_i, f_i)$  and  $[s_j, f_j)$  are disjoint

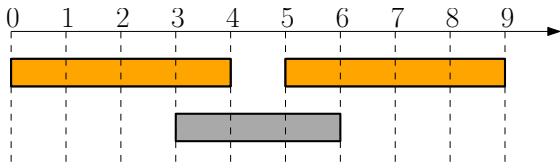
**Output:** A maximum-size subset of mutually compatible jobs





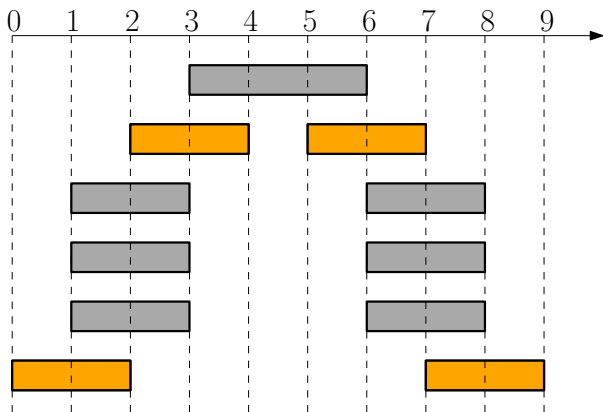
# Greedy Algorithm for Interval Scheduling

- Which of the following decisions are safe?
- Schedule the job with the smallest size? **No!**



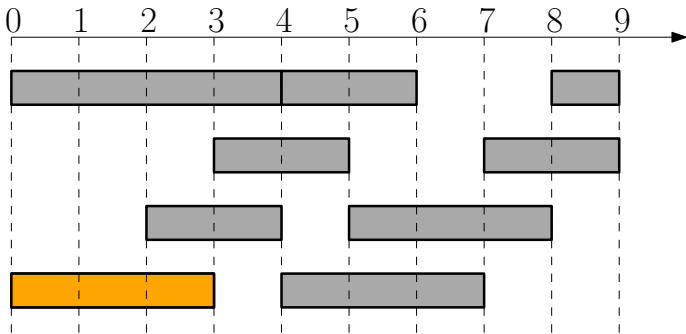
# Greedy Algorithm for Interval Scheduling

- Which of the following decisions are safe?
- Schedule the job with the smallest size? No!
- Schedule the job conflicting with smallest number of other jobs? **No!**



# Greedy Algorithm for Interval Scheduling

- Which of the following decisions are safe?
- Schedule the job with the smallest size? No!
- Schedule the job conflicting with smallest number of other jobs? No!
- Schedule the job with the earliest finish time? **Yes!**



# Greedy Algorithm for Interval Scheduling

**Lemma** It is safe to schedule the job  $j$  with the earliest finish time: there is an optimum solution where  $j$  is scheduled.

Proof.

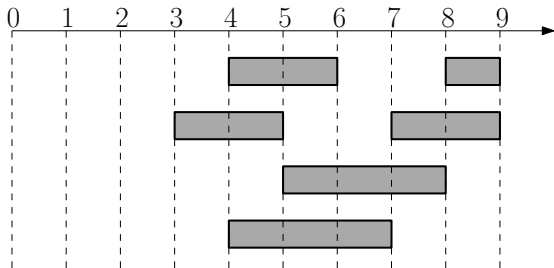
- Take an arbitrary optimum solution  $S$
- If it contains  $j$ , done
- Otherwise, replace the first job in  $S$  with  $j$  to obtain a new optimum schedule  $S'$ . □



# Greedy Algorithm for Interval Scheduling

**Lemma** It is safe to schedule the job  $j$  with the earliest finish time: there is an optimum solution where  $j$  is scheduled.

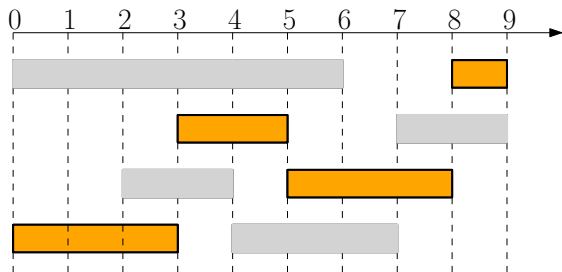
- What is the remaining task after we decided to schedule  $j$ ?
- Is it another instance of interval scheduling problem? **Yes!**



# Greedy Algorithm for Interval Scheduling

## Schedule( $s, f, n$ )

- 1  $A \leftarrow \{1, 2, \dots, n\}, S \leftarrow \emptyset$
- 2 while  $A \neq \emptyset$
- 3     $j \leftarrow \arg \min_{j' \in A} f_{j'}$
- 4     $S \leftarrow S \cup \{j\}; A \leftarrow \{j' \in A : s_{j'} \geq f_j\}$
- 5 return  $S$



# Greedy Algorithm for Interval Scheduling

## Schedule( $s, f, n$ )

- 1  $A \leftarrow \{1, 2, \dots, n\}, S \leftarrow \emptyset$
- 2 while  $A \neq \emptyset$
- 3      $j \leftarrow \arg \min_{j' \in A} f_{j'}$
- 4      $S \leftarrow S \cup \{j\}; A \leftarrow \{j' \in A : s_{j'} \geq f_j\}$
- 5 return  $S$

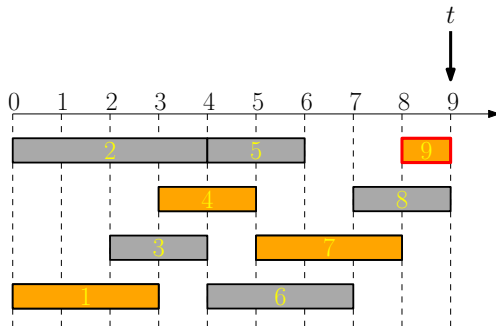
Running time of algorithm?

- Naive implementation:  $O(n^2)$  time
- Clever implementation:  $O(n \lg n)$  time

# Clever Implementation of Greedy Algorithm

## Schedule( $s, f, n$ )

- 1 sort jobs according to  $f$  values
- 2  $t \leftarrow 0, S \leftarrow \emptyset$
- 3 for every  $j \in [n]$  according to non-decreasing order of  $f_j$
- 4 if  $s_j \geq t$  then
- 5      $S \leftarrow S \cup \{j\}$
- 6      $t \leftarrow f_j$
- 7 return  $S$



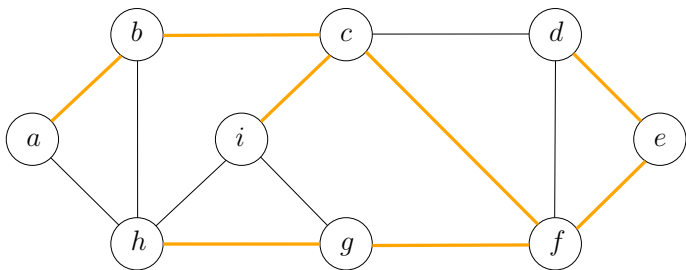


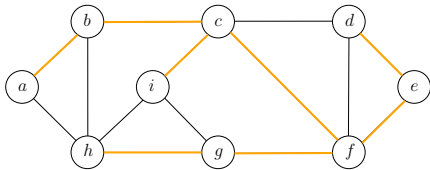
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# Spanning Tree

**Def.** Given a connected graph  $G = (V, E)$ , a **spanning tree**  $T = (V, F)$  of  $G$  is a sub-graph of  $G$  that is a tree including all vertices  $V$ .





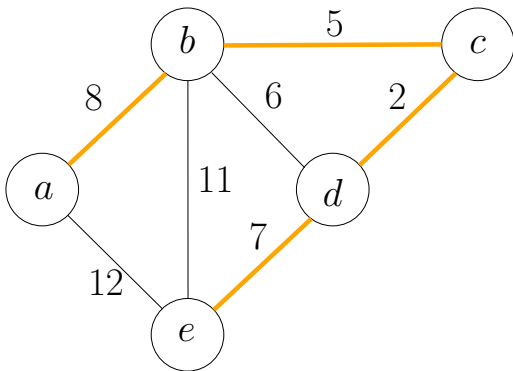
**Lemma** Let  $T = (V, F)$  be a subgraph of  $G = (V, E)$ . The following statements are equivalent:

- $T$  is a spanning tree of  $G$ ;
- $T$  is acyclic and connected;
- $T$  is connected and has  $n - 1$  edges;
- $T$  is acyclic and has  $n - 1$  edges;
- $T$  is minimally connected: removal of any edge disconnects it;
- $T$  is maximally acyclic: addition of any edge creates a cycle;
- $T$  has a unique simple path between every pair of nodes.

## Minimum Spanning Tree (MST) Problem

**Input:** Graph  $G = (V, E)$  and edge weights  $w : E \rightarrow \mathbb{R}$

**Output:** the spanning tree  $T$  of  $G$  with the minimum total weight



## Recall: Steps of Designing A Greedy Algorithm

- Design a “reasonable” strategy
- Prove that the reasonable strategy is “safe” (key, usually done by “exchanging argument”)
- Show that the remaining task after applying the strategy is to solve a (many) smaller instance(s) of the same problem (usually trivial)

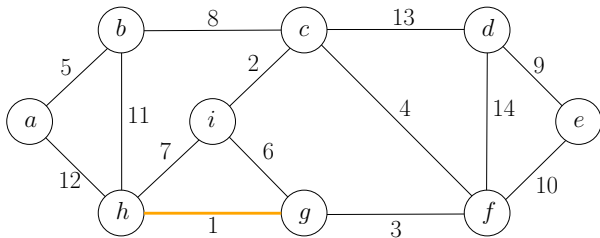
**Def.** A choice is “safe” if there is an optimum solution that is “consistent” with the choice

## Two Classic Greedy Algorithms for MST

- Kruskal's Algorithm
- Prim's Algorithm

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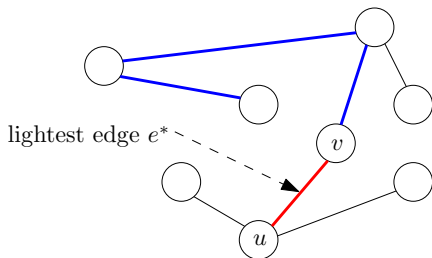
**Q:** Which edge can be safely included in the MST?

**A:** The edge with the smallest weight (lightest edge).

**Lemma** It is safe to include the lightest edge: there is a minimum spanning tree, that contains the lightest edge.

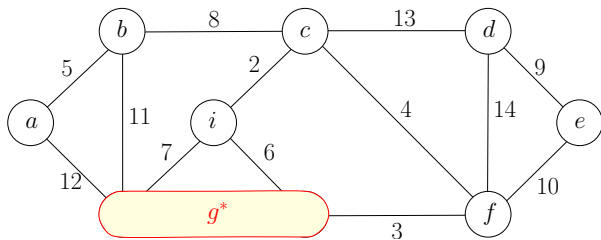
Proof.

- Take a minimum spanning tree  $T$
- Assume the lightest edge  $e^*$  is not in  $T$
- There is a unique path in  $T$  connecting  $u$  and  $v$
- Remove any edge  $e$  in the path to obtain tree  $T'$
- $w(e^*) \leq w(e) \implies w(T') \leq w(T)$ :  $T'$  is also a MST □



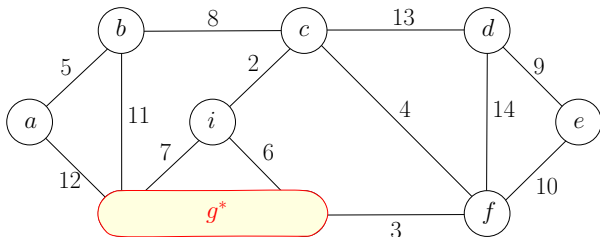


# Is the Residual Problem Still a MST Problem?



- Residual problem: find the minimum spanning tree that contains edge  $(g, h)$
- **Contract** the edge  $(g, h)$
- Residual problem: find the minimum spanning tree in the contracted graph

# Contraction of an Edge $(u, v)$



- Remove  $u$  and  $v$  from the graph, and add a new vertex  $u^*$
- Remove all edges parallel connecting  $u$  to  $v$  from  $E$
- For every edge  $(u, w) \in E, w \neq v$ , change it to  $(u^*, w)$
- For every edge  $(v, w) \in E, w \neq u$ , change it to  $(u^*, w)$
- **May create parallel edges!** E.g. : two edges  $(i, g^*)$

# Greedy Algorithm

Repeat the following step until  $G$  contains only one vertex:

- 1 Choose the lightest edge  $e^*$ , add  $e^*$  to the spanning tree
- 2 Contract  $e^*$  and update  $G$  be the contracted graph

**Q:** What edges are removed due to contractions?

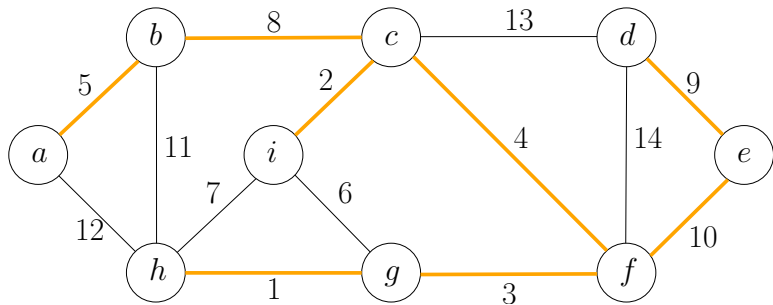
**A:** Edge  $(u, v)$  is removed if and only if there is a path connecting  $u$  and  $v$  formed by edges we selected

# Greedy Algorithm

## MST-Greedy( $G, w$ )

- 1  $F = \emptyset$
- 2 sort edges in  $E$  in non-decreasing order of weights  $w$
- 3 for each edge  $(u, v)$  in the order
- 4     if  $u$  and  $v$  are not connected by a path of edges in  $F$
- 5          $F = F \cup \{(u, v)\}$
- 6 return  $(V, F)$

# Kruskal's Algorithm: Example



Sets:  $\{a, b, c, i, f, g, h, d, e\}$

# Kruskal's Algorithm: Efficient Implementation of Greedy Algorithm

## MST-Kruskal( $G, w$ )

- 1  $F \leftarrow \emptyset$
- 2  $\mathcal{S} \leftarrow \{\{v\} : v \in V\}$
- 3 sort the edges of  $E$  in non-decreasing order of weights  $w$
- 4 for each edge  $(u, v) \in E$  in the order
- 5      $S_u \leftarrow$  the set in  $\mathcal{S}$  containing  $u$
- 6      $S_v \leftarrow$  the set in  $\mathcal{S}$  containing  $v$
- 7     if  $S_u \neq S_v$
- 8          $F \leftarrow F \cup \{(u, v)\}$
- 9          $\mathcal{S} \leftarrow \mathcal{S} \setminus \{S_u\} \setminus \{S_v\} \cup \{S_u \cup S_v\}$
- 10 return  $(V, F)$

# Running Time of Kruskal's Algorithm

## MST-Kruskal( $G, w$ )

- 1  $F \leftarrow \emptyset$
- 2  $\mathcal{S} \leftarrow \{\{v\} : v \in V\}$
- 3 sort the edges of  $E$  in non-decreasing order of weights  $w$
- 4 for each edge  $(u, v) \in E$  in the order
  - 5  $S_u \leftarrow$  the set in  $\mathcal{S}$  containing  $u$
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  - 7 if  $S_u \neq S_v$ 
    - 8  $F \leftarrow F \cup \{(u, v)\}$
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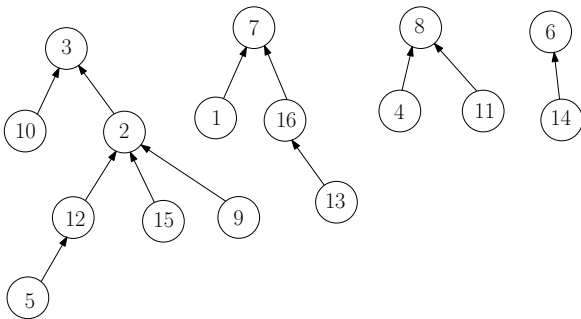
Use **union-find** data structure to support 2, 5, 6, 7, 9.

# Union-Find Data Structure

- $V$ : ground set
- We need to maintain a partition of  $V$  and support following operations:
  - Check if  $u$  and  $v$  are in the same set of the partition
  - Merge two sets in partition

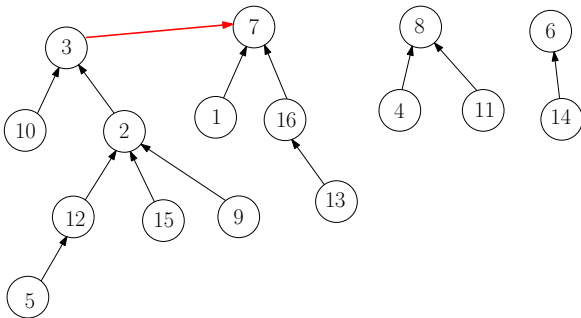


- $V = \{1, 2, 3, \dots, 16\}$
- Partition:
  - $\{2, 3, 5, 9, 10, 12, 15\}$ ,  $\{1, 7, 13, 16\}$ ,  $\{4, 8, 11\}$ ,  $\{6, 14\}$



- $par[i]$ : parent of  $i$ , ( $par[i] = \text{nil}$  if  $i$  is a root).

# Union-Find Data Structure



- Q: how can we check if  $u$  and  $v$  are in the same set?
- A: Check if  $\text{root}(u) = \text{root}(v)$ .
- $\text{root}(u)$ : the root of the tree containing  $u$
- Merge the trees with root  $r$  and  $r'$ :  $\text{par}[r] \leftarrow r'$ .

# Union-Find Data Structure

## $root(v)$

- 1 if  $par[v] = nil$  then
- 2     return  $v$
- 3 else
- 4     return  $root(par[v])$

## $root(v)$

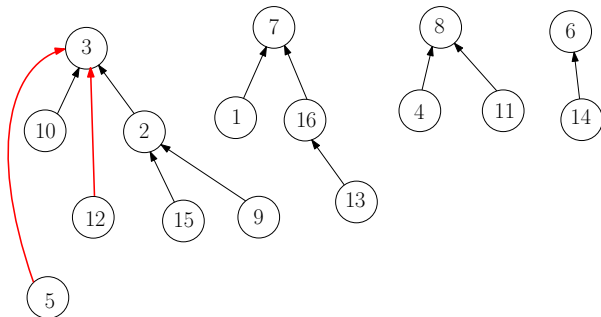
- 1 if  $par[v] = nil$  then
- 2     return  $v$
- 3 else
- 4      $par[v] \leftarrow root(par[v])$
- 5     return  $par[v]$

- Problem: the tree might too deep; running time might be large
- Improvement: all vertices in the path directly point to the root, saving time in the future.

# Union-Find Data Structure

$\text{root}(v)$

- 1 if  $\text{par}[v] = \text{nil}$  then
- 2     return  $v$
- 3 else
- 4      $\text{par}[v] \leftarrow \text{root}(\text{par}[v])$
- 5     return  $\text{par}[v]$



## MST-Kruskal( $G, w$ )

- 1  $F \leftarrow \emptyset$
- 2  $\mathcal{S} \leftarrow \{\{v\} : v \in V\}$
- 3 sort the edges of  $E$  in non-decreasing order of weights  $w$
- 4 for each edge  $(u, v) \in E$  in the order
- 5      $S_u \leftarrow$  the set in  $\mathcal{S}$  containing  $u$
- 6      $S_v \leftarrow$  the set in  $\mathcal{S}$  containing  $v$
- 7     if  $S_u \neq S_v$
- 8          $F \leftarrow F \cup \{(u, v)\}$
- 9          $\mathcal{S} \leftarrow \mathcal{S} \setminus \{S_u\} \setminus \{S_v\} \cup \{S_u \cup S_v\}$
- 10 return  $(V, F)$

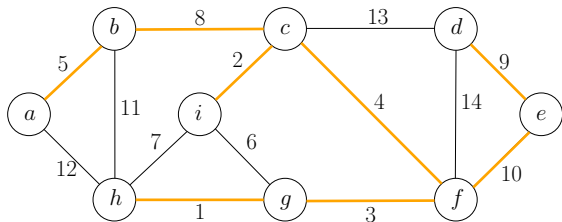
## MST-Kruskal( $G, w$ )

- 1  $F \leftarrow \emptyset$
- 2 for every  $v \in V$ : let  $par[v] \leftarrow nil$
- 3 sort the edges of  $E$  in non-decreasing order of weights  $w$
- 4 for each edge  $(u, v) \in E$  in the order
- 5      $u' \leftarrow root(u)$
- 6      $v' \leftarrow root(v)$
- 7     if  $u' \neq v'$
- 8          $F \leftarrow F \cup \{(u, v)\}$
- 9          $par[u'] \leftarrow v'$
- 10 return  $(V, F)$

- ②, ⑤, ⑥, ⑦, ⑨ takes time  $O(m\alpha(n))$
- $\alpha(n)$  is very slow-growing:  $\alpha(n) \leq 4$  for  $n \leq 10^{80}$ .
- Running time = time for ③ =  $O(m \lg n)$ .

**Assumption** Assume all edge weights are different.

**Lemma** An edge  $e \in E$  is **not** in the MST, if and only if there is cycle  $C$  in  $G$  in which  $e$  is the heaviest edge.



- $(i, g)$  is not in the MST because of cycle  $(i, c, f, g)$
- $(e, f)$  is in the MST because no such cycle exists

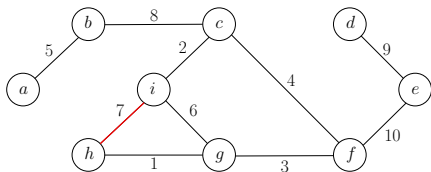
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## Two Methods to Build a MST

- 1 Start from  $F \leftarrow \emptyset$ , and add edges to  $F$  one by one until we obtain a spanning tree
- 2 Start from  $F \leftarrow E$ , and **remove** edges from  $F$  one by one until we obtain a spanning tree



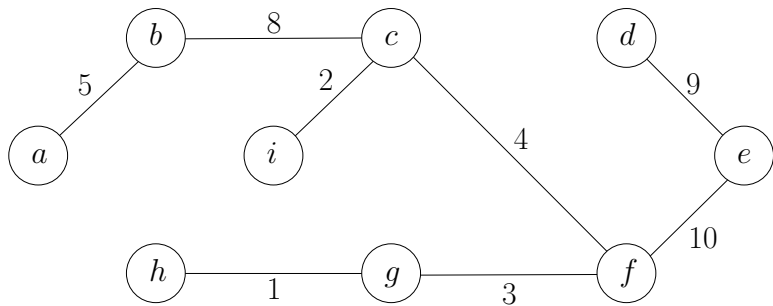
**Lemma** It is safe to exclude the heaviest non-bridge edge: there is a MST that does not contain the heaviest non-bridge edge.

# Reverse Kruskal's Algorithm

## MST-Greedy( $G, w$ )

- 1  $F \leftarrow E$
- 2 sort  $E$  in non-increasing order of weights
- 3 for every  $e$  in this order
- 4     if  $(V, F \setminus \{e\})$  is connected then
- 5          $F \leftarrow F \setminus \{e\}$
- 6 return  $(V, F)$

# Reverse Kruskal's Algorithm: Example

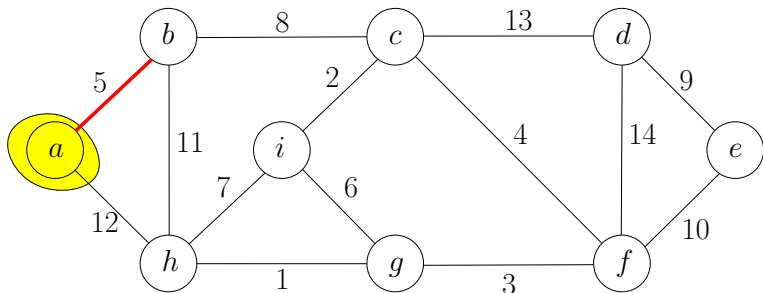


# Outline

- 1 Toy Examples
- 2 Interval Scheduling
- 3 Minimum Spanning Tree**
  - Kruskal's Algorithm
  - Reverse-Kruskal's Algorithm
  - Prim's Algorithm**
- 4 Single Source Shortest Paths
  - Dijkstra's Algorithm
- 5 Data Compression and Huffman Code
- 6 Summary

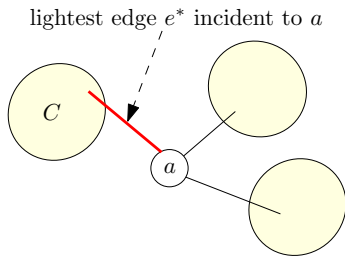
# Design Greedy Strategy for MST

- Recall the greedy strategy for Kruskal's algorithm: choose the edge with the smallest weight.



- Greedy strategy for Prim's algorithm: choose **the lightest edge incident to *a***.

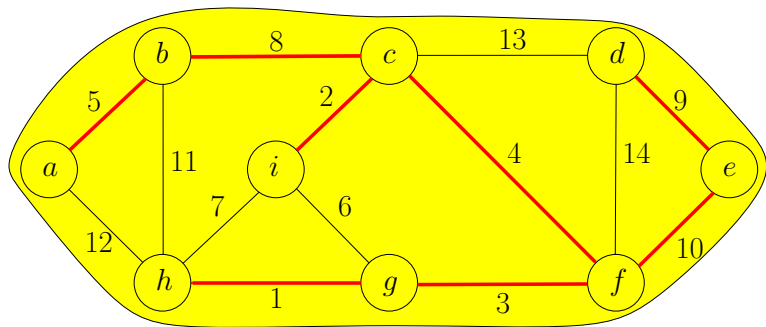
**Lemma** It is safe to include the lightest edge incident to  $a$ .



**Proof.**

- Let  $T$  be a MST
- Consider all components obtained by removing  $a$  from  $T$
- Let  $e^*$  be the lightest edge incident to  $a$  and  $e^*$  connects  $a$  to component  $C$
- Let  $e$  be the edge in  $T$  connecting  $a$  to  $C$
- $T' = T \setminus e \cup \{e^*\}$  is a spanning tree with  $w(T') \leq w(T)$   $\square$

# Prim's Algorithm: Example





# Greedy Algorithm

## MST-Greedy1( $G, w$ )

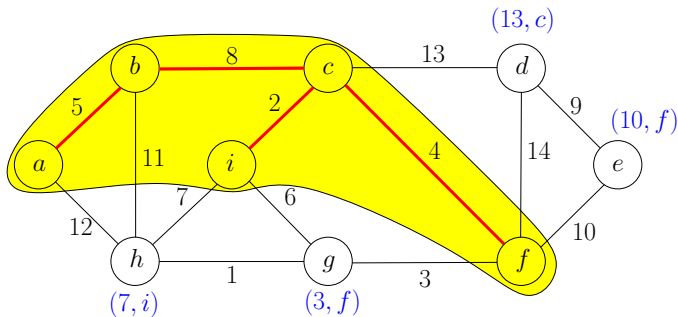
- 1  $S \leftarrow \{s\}$ , where  $s$  is arbitrary vertex in  $V$
- 2  $F \leftarrow \emptyset$
- 3 while  $S \neq V$
- 4  $(u, v) \leftarrow$  lightest edge between  $S$  and  $V \setminus S$ ,  
where  $u \in S$  and  $v \in V \setminus S$
- 5  $S \leftarrow S \cup \{v\}$
- 6  $F \leftarrow F \cup \{(u, v)\}$
- 7 return  $(V, F)$

- Running time of naive implementation:  $O(nm)$

# Prim's Algorithm: Efficient Implementation of Greedy Algorithm

For every  $v \in V \setminus S$  maintain

- $d(v) = \min_{u \in S: (u,v) \in E} w(u, v)$ :  
the weight of the lightest edge between  $v$  and  $S$
- $\pi(v) = \arg \min_{u \in S: (u,v) \in E} w(u, v)$ :  
 $(\pi(v), v)$  is the lightest edge between  $v$  and  $S$



# Prim's Algorithm: Efficient Implementation of Greedy Algorithm

For every  $v \in V \setminus S$  maintain

- $d(v) = \min_{u \in S: (u,v) \in E} w(u, v)$ :  
the weight of the lightest edge between  $v$  and  $S$
- $\pi(v) = \arg \min_{u \in S: (u,v) \in E} w(u, v)$ :  
 $(\pi(v), v)$  is the lightest edge between  $v$  and  $S$

In every iteration

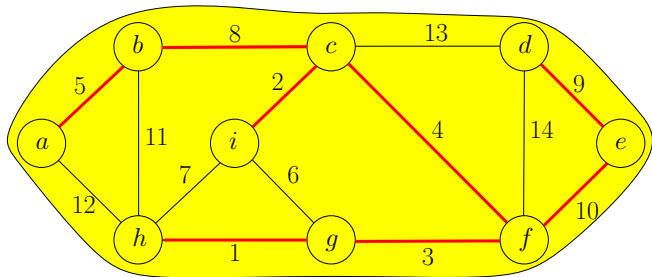
- Pick  $u \in V \setminus S$  with the smallest  $d(u)$  value
- Add  $(\pi(u), u)$  to  $F$
- Add  $u$  to  $S$ , update  $d$  and  $\pi$  values.

# Prim's Algorithm

## MST-Prim( $G, w$ )

- 1  $s \leftarrow$  arbitrary vertex in  $G$
- 2  $S \leftarrow \emptyset$ ,  $d(s) \leftarrow 0$  and  $d(v) \leftarrow \infty$  for every  $v \in V \setminus \{s\}$
- 3 while  $S \neq V$ , do
- 4      $u \leftarrow$  vertex in  $V \setminus S$  with the minimum  $d(u)$
- 5      $S \leftarrow S \cup \{u\}$
- 6     for each  $v \in V \setminus S$  such that  $(u, v) \in E$
- 7         if  $w(u, v) < d(v)$  then
- 8              $d(v) \leftarrow w(u, v)$
- 9              $\pi(v) \leftarrow u$
- 10 return  $\{(u, \pi(u)) \mid u \in V \setminus \{s\}\}$

# Example



# Prim's Algorithm

For every  $v \in V \setminus S$  maintain

- $d(v) = \min_{u \in S: (u,v) \in E} w(u, v)$ :  
the weight of the lightest edge between  $v$  and  $S$
- $\pi(v) = \arg \min_{u \in S: (u,v) \in E} w(u, v)$ :  
 $(\pi(v), v)$  is the lightest edge between  $v$  and  $S$

In every iteration

- Pick  $u \in V \setminus S$  with the smallest  $d(u)$  value extract\_min
- Add  $(\pi(u), u)$  to  $F$
- Add  $u$  to  $S$ , update  $d$  and  $\pi$  values. decrease\_key

Use a priority queue to support the operations

**Def.** A **priority queue** is an **abstract** data structure that maintains a set  $U$  of elements, each with an associated key value, and supports the following operations:

- $\text{insert}(v, \text{key\_value})$ : insert an element  $v$ , whose associated key value is  $\text{key\_value}$ .
- $\text{decrease\_key}(v, \text{new\_key\_value})$ : decrease the key value of an element  $v$  in queue to  $\text{new\_key\_value}$
- $\text{extract\_min}()$ : return and remove the element in queue with the smallest key value
- ...

# Prim's Algorithm

## MST-Prim( $G, w$ )

- 1  $s \leftarrow$  arbitrary vertex in  $G$
- 2  $S \leftarrow \emptyset, d(s) \leftarrow 0$  and  $d(v) \leftarrow \infty$  for every  $v \in V \setminus \{s\}$
- 3
- 4 while  $S \neq V$ , do
- 5      $u \leftarrow$  vertex in  $V \setminus S$  with the minimum  $d(u)$
- 6      $S \leftarrow S \cup \{u\}$
- 7     for each  $v \in V \setminus S$  such that  $(u, v) \in E$
- 8         if  $w(u, v) < d(v)$  then
- 9              $d(v) \leftarrow w(u, v)$
- 10             $\pi(v) \leftarrow u$
- 11 return  $\{(u, \pi(u)) \mid u \in V \setminus \{s\}\}$



# Prim's Algorithm Using Priority Queue

## MST-Prim( $G, w$ )

- 1  $s \leftarrow$  arbitrary vertex in  $G$
- 2  $S \leftarrow \emptyset$ ,  $d(s) \leftarrow 0$  and  $d(v) \leftarrow \infty$  for every  $v \in V \setminus \{s\}$
- 3  $Q \leftarrow$  empty queue, for each  $v \in V$ :  $Q.insert(v, d(v))$
- 4 while  $S \neq V$ , do
  - 5  $u \leftarrow Q.extract\_min()$
  - 6  $S \leftarrow S \cup \{u\}$
  - 7 for each  $v \in V \setminus S$  such that  $(u, v) \in E$ 
    - 8 if  $w(u, v) < d(v)$  then
      - 9  $d(v) \leftarrow w(u, v)$ ,  $Q.decrease\_key(v, d(v))$
      - 10  $\pi(v) \leftarrow u$
- 11 return  $\{(u, \pi(u)) \mid u \in V \setminus \{s\}\}$

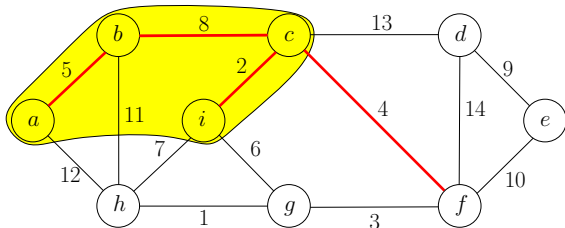
# Running Time of Prim's Algorithm Using Priority Queue

$$O(n) \times (\text{time for extract\_min}) + O(m) \times (\text{time for decrease\_key})$$

concrete DS	extract_min	decrease_key	overall time
heap	$O(\log n)$	$O(\log n)$	$O(m \log n)$
Fibonacci heap	$O(\log n)$	$O(1)$	$O(n \log n + m)$

**Assumption** Assume all edge weights are different.

**Lemma**  $(u, v)$  is in MST, if and only if there exists a **cut**  $(U, V \setminus U)$ , such that  $(u, v)$  is the lightest edge between  $U$  and  $V \setminus U$ .



- $(c, f)$  is in MST because of cut  $(\{a, b, c, i\}, V \setminus \{a, b, c, i\})$
- $(i, g)$  is not in MST because no such cut exists

# “Evidence” for $e \in \text{MST}$ or $e \notin \text{MST}$

**Assumption** Assume all edge weights are different.

- $e \in \text{MST} \leftrightarrow$  there is a cut in which  $e$  is the lightest edge
- $e \notin \text{MST} \leftrightarrow$  there is a cycle in which  $e$  is the heaviest edge

**Exactly one** of the following is true:

- There is a cut in which  $e$  is the lightest edge
- There is a cycle in which  $e$  is the heaviest edge

Thus, the minimum spanning tree is unique with assumption.

# Outline

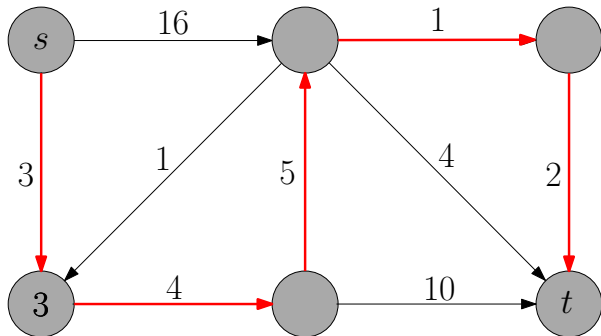
- 1 Toy Examples
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## $s$ - $t$ Shortest Paths

**Input:** (directed or undirected) graph  $G = (V, E)$ ,  $s, t \in V$

$w : E \rightarrow \mathbb{R}_{\geq 0}$

**Output:** shortest path from  $s$  to  $t$



## Single Source Shortest Paths

**Input:** **directed** graph  $G = (V, E)$ ,  $s \in V$

$$w : E \rightarrow \mathbb{R}_{\geq 0}$$

**Output:** shortest paths from  $s$  to **all other vertices**  $v \in V$

## Reason for Considering Single Source Shortest Paths Problem

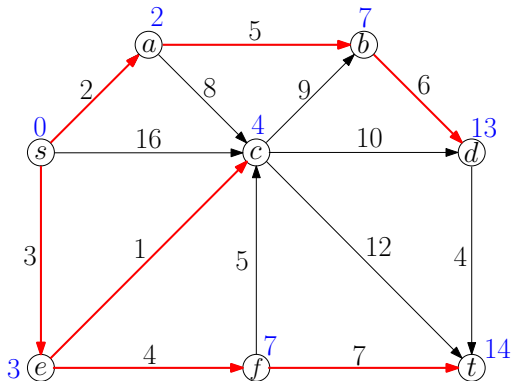
- We do not know how to solve  $s$ - $t$  shortest path problem more efficiently than solving single source shortest path problem
- Shortest paths in directed graphs is more general than in undirected graphs: we can replace every undirected edge with two anti-parallel edges of the same weight

- Shortest path from  $s$  to  $v$  may contain  $\Omega(n)$  edges
- There are  $\Omega(n)$  different vertices  $v$
- Thus, printing out all shortest paths may take time  $\Omega(n^2)$
- Not acceptable if graph is sparse



## Shortest Path Tree

- $O(n)$ -size data structure to represent all shortest paths
- For every vertex  $v$ , we only need to remember the **parent** of  $v$ : second-to-last vertex in the shortest path from  $s$  to  $v$  (why?)



## Single Source Shortest Paths

**Input:** directed graph  $G = (V, E)$ ,  $s \in V$

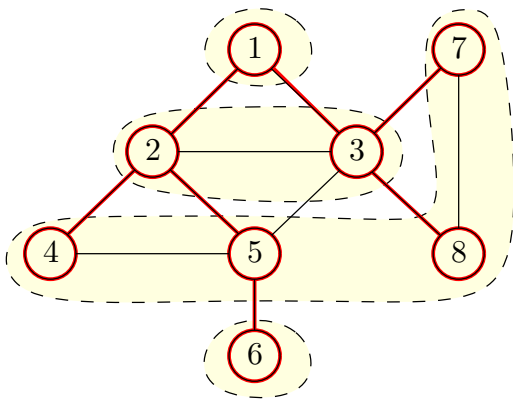
$$w : E \rightarrow \mathbb{R}_{\geq 0}$$

**Output:**  $\pi(v), v \in V \setminus s$ : the parent of  $v$

$d(v), v \in V \setminus s$ : the length of shortest path from  $s$  to  $v$

**Q:** How to compute shortest paths from  $s$  when all edges have weight 1?

**A:** Breadth first search (BFS) from source  $s$



**Assumption** Weights  $w(u, v)$  are integers (w.l.o.g).

- An edge of weight  $w(u, v)$  is equivalent to a path of  $w(u, v)$  unit-weight edges



### Shortest Path Algorithm by Running BFS

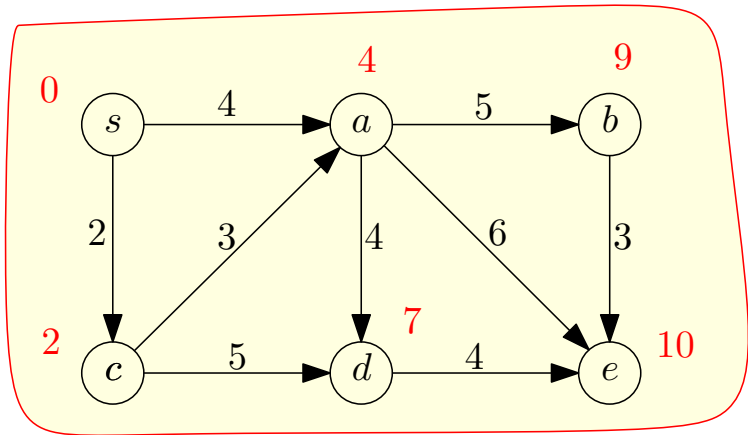
- 1 replace  $(u, v)$  of length  $w(u, v)$  with a path of  $w(u, v)$  unit-weight edges, for every  $(u, v) \in E$
- 2 run BFS **virtually**
- 3  $\pi(v)$  = vertex from which  $v$  is visited
- 4  $d(v)$  = index of the level containing  $v$

- Problem:  $w(u, v)$  may be too large!

## Shortest Path Algorithm by Running BFS Virtually

- 1  $S \leftarrow \{s\}, d(s) \leftarrow 0$
- 2 while  $|S| \leq n$
- 3 find a  $v \notin S$  that minimizes  $\min_{u \in S: (u,v) \in E} \{d(u) + w(u, v)\}$
- 4  $S \leftarrow S \cup \{v\}$
- 5  $d(v) \leftarrow \min_{u \in S: (u,v) \in E} \{d(u) + w(u, v)\}$

# Virtual BFS: Example



Time 10

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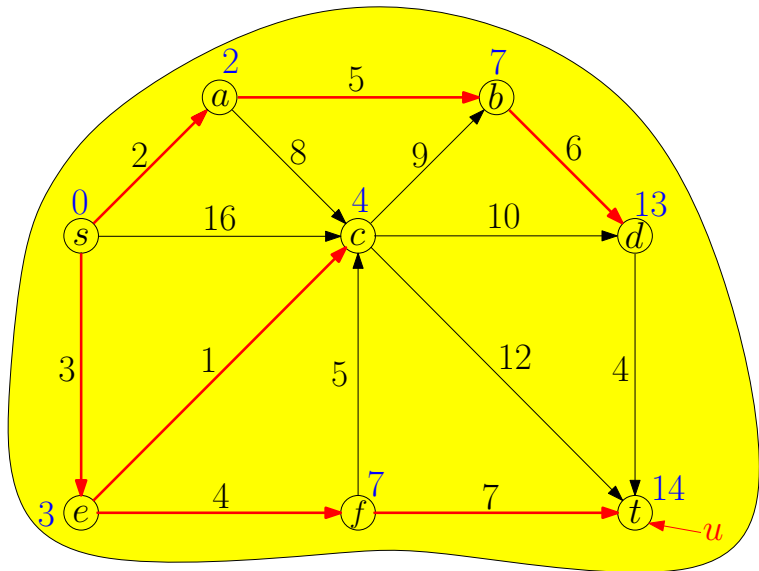
# Dijkstra's Algorithm

## Dijkstra( $G, w, s$ )

- 1  $S \leftarrow \emptyset, d(s) \leftarrow 0$  and  $d(v) \leftarrow \infty$  for every  $v \in V \setminus \{s\}$
- 2 while  $S \neq V$  do
- 3      $u \leftarrow$  vertex in  $V \setminus S$  with the minimum  $d(u)$
- 4     add  $u$  to  $S$
- 5     for each  $v \in V \setminus S$  such that  $(u, v) \in E$
- 6         if  $d(u) + w(u, v) < d(v)$  then
- 7              $d(v) \leftarrow d(u) + w(u, v)$
- 8              $\pi(v) \leftarrow u$
- 9 return  $(d, \pi)$

- Running time =  $O(n^2)$





# Improved Running Time using Priority Queue

## Dijkstra( $G, w, s$ )

- 1
- 2  $S \leftarrow \emptyset, d(s) \leftarrow 0$  and  $d(v) \leftarrow \infty$  for every  $v \in V \setminus \{s\}$
- 3  $Q \leftarrow$  empty queue, for each  $v \in V$ :  $Q.insert(v, d(v))$
- 4 while  $S \neq V$ , do
- 5  $u \leftarrow Q.extract\_min()$
- 6  $S \leftarrow S \cup \{u\}$
- 7 for each  $v \in V \setminus S$  such that  $(u, v) \in E$
- 8 if  $d(u) + w(u, v) < d(v)$  then
- 9  $d(v) \leftarrow d(u) + w(u, v)$ ,  $Q.decrease\_key(v, d(v))$
- 10  $\pi(v) \leftarrow u$
- 11 return  $(\pi, d)$

# Recall: Prim's Algorithm for MST

## MST-Prim( $G, w$ )

- 1  $s \leftarrow$  arbitrary vertex in  $G$
- 2  $S \leftarrow \emptyset$ ,  $d(s) \leftarrow 0$  and  $d(v) \leftarrow \infty$  for every  $v \in V \setminus \{s\}$
- 3  $Q \leftarrow$  empty queue, for each  $v \in V$ :  $Q.insert(v, d(v))$
- 4 while  $S \neq V$ , do
  - 5  $u \leftarrow Q.extract\_min()$
  - 6  $S \leftarrow S \cup \{u\}$
  - 7 for each  $v \in V \setminus S$  such that  $(u, v) \in E$ 
    - 8 if  $w(u, v) < d(v)$  then
      - 9  $d(v) \leftarrow w(u, v)$ ,  $Q.decrease\_key(v, d(v))$
      - 10  $\pi(v) \leftarrow u$
- 11 return  $\{(u, \pi(u)) \mid u \in V \setminus \{s\}\}$

# Improved Running Time

Running time:

$O(n) \times (\text{time for extract\_min}) + O(m) \times (\text{time for decrease\_key})$

Priority-Queue	extract_min	decrease_key	Time
Heap	$O(\log n)$	$O(\log n)$	$O(m \log n)$
Fibonacci Heap	$O(\log n)$	$O(1)$	$O(n \log n + m)$

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# Encoding Symbols Using Bits

- assume: 8 symbols  $a, b, c, d, e, f, g, h$  in a language
- need to encode a message using bits
- idea: use 3 bits per symbol

$a$	$b$	$c$	$d$	$e$	$f$	$g$	$h$
000	001	010	011	100	101	110	111

$deacfg \rightarrow 011100000010101110$

**Q:** Can we have a better encoding scheme?

- Seems unlikely: must use 3 bits per symbol

**Q:** What if some symbols appear more frequently than the others in expectation?

**Q:** If some symbols appear more frequently than the others in expectation, can we have a better encoding scheme?

**A:** Maybe. Using **variable-length encoding scheme**.

Idea

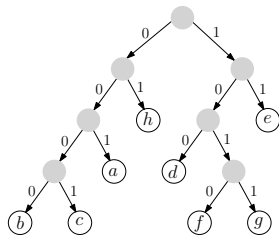
- using fewer bits for symbols that are more frequently used, and more bits for symbols that are less frequently used.

Need to use **prefix codes** to guarantee a unique decoding.

# Prefix Codes

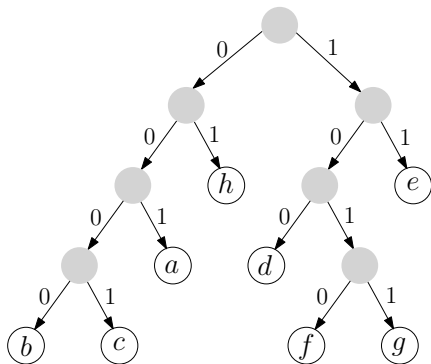
**Def.** A prefix code for a set  $S$  of symbols is a function  $\gamma : S \rightarrow \{0, 1\}^*$  such that for two distinct  $x, y \in S$ ,  $\gamma(x)$  is not a prefix of  $\gamma(y)$ .

$a$	$b$	$c$	$d$
001	0000	0001	100
$e$	$f$	$g$	$h$
11	1010	1011	01



- 0001/001/100/0000/01/01/11/1010/0001/001/
- cadbhhefca





## Properties of Encoding Tree

- Rooted binary tree
- Left edges labelled 0 and right edges labelled 1
- A leaf corresponds to a code for some symbol
- If coding scheme is not wasteful: a non-leaf has exactly two children

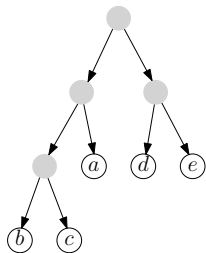
## Best Prefix Codes

**Input:** frequencies of letters in a message

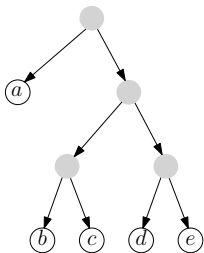
**Output:** prefix coding scheme giving the shortest encoding for the message

## example

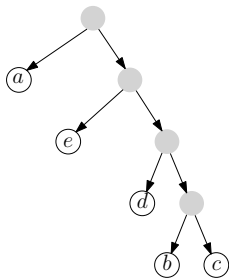
symbols	<i>a</i>	<i>b</i>	<i>c</i>	<i>d</i>	<i>e</i>	
frequencies	18	3	4	6	10	
scheme 1 length	2	3	3	2	2	total = 89
scheme 2 length	1	3	3	3	3	total = 87
scheme 3 length	1	4	4	3	2	<b>total = 84</b>



scheme 1



scheme 2



scheme 3

- Example Input: ( $a$ : 18,  $b$ : 3,  $c$ : 4,  $d$ : 6,  $e$ : 10)

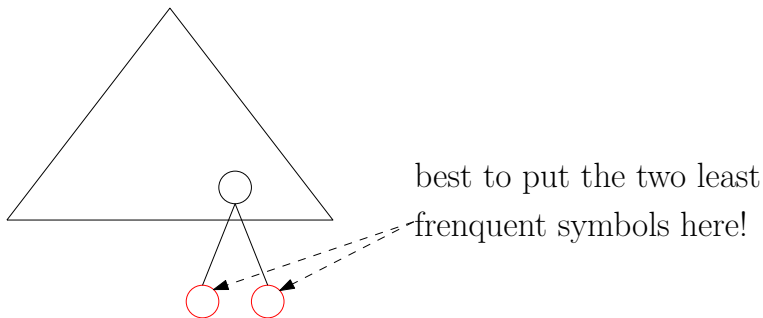
**Q:** What types of decisions should we make?

- the code for some letter?
- hard to design a strategy; residual problem is complicated.
- a partition of letters into left and right sub-trees?
- not clear how to design the greedy algorithm

**A:** Choose two letters and make them brothers in the tree.

# Which Two symbols Can Be Safely Put Together As Brothers?

- Focus a tree structure, without leaf labeling
- There are two deepest leaves that are brothers
- It is safe to make the two least frequent symbols brothers!



- It is safe to make the two least frequent symbols brothers!

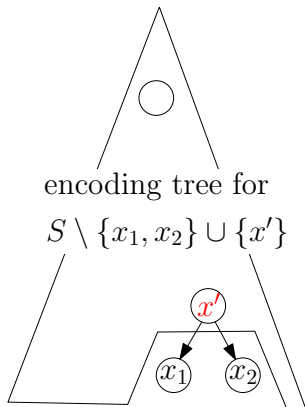
**Lemma** There is an optimum encoding tree, where the two least frequent symbols are brothers.

- So we can make the two least frequent symbols brothers; the decision is irrevocable.

**Q:** Is the residual problem an instance of the best prefix codes problem?

**A:** Yes, although the answer is not immediate.

- $f_x$ : the frequency of the symbol  $x$  in the support.
- $x_1$  and  $x_2$ : the two symbols we decided to put together.
- $d_x$  the depth of symbol  $x$  in our output encoding tree.



$$\begin{aligned}
 & \sum_{x \in S} f_x d_x \\
 = & \sum_{x \in S \setminus \{x_1, x_2\}} f_x d_x + f_{x_1} d_{x_1} + f_{x_2} d_{x_2} \\
 = & \sum_{x \in S \setminus \{x_1, x_2\}} f_x d_x + (f_{x_1} + f_{x_2}) d_{x_1} \\
 = & \sum_{x \in S \setminus \{x_1, x_2\}} f_x d_x + f_{x'} (d_{x'} + 1) \\
 = & \sum_{x \in S \setminus \{x_1, x_2\} \cup \{x'\}} f_x d_x + f_{x'}
 \end{aligned}$$

Def:  $f_{x'} = f_{x_1} + f_{x_2}$

In order to minimize

$$\sum_{x \in S} f_x d_x,$$

we need to minimize

$$\sum_{x \in S \setminus \{x_1, x_2\} \cup \{x'\}} f_x d_x,$$

subject to that  $d$  is the depth function for an encoding tree of  $S \setminus \{x_1, x_2\}$ .

- This is exactly the best prefix codes problem, with symbols  $S \setminus \{x_1, x_2\} \cup \{x'\}$  and frequency vector  $f$ !

# Huffman codes: Recursive Algorithm

## Huffman( $S, f$ )

- 1 **if**  $|S| > 1$  **then**
- 2     let  $x_1, x_2$  be the two symbols with the smallest  $f$  values
- 3     introduce a new symbol  $x'$  and let  $f_{x'} = f_{x_1} + f_{x_2}$
- 4      $S' \leftarrow S \setminus \{x_1, x_2\} \cup \{x'\}$
- 5     call Huffman( $S', f|_{S'}$ ) to build an encoding tree  $T'$
- 6     let  $T$  be obtained from  $T'$  by adding  $x_1, x_2$  as two children of  $x'$
- 7     **return**  $T$
- 8 **else**
- 9     let  $x$  be the symbol in  $S$
- 10    **return** a tree with a single node labeled  $x$

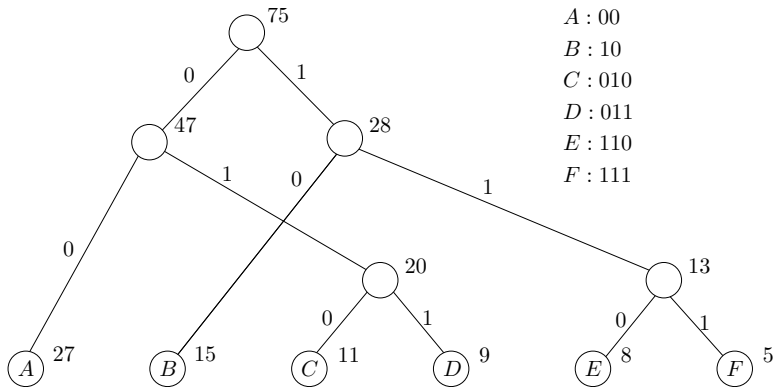


# Huffman codes: Iterative Algorithm

## Huffman( $S, f$ )

- 1 **while**  $|S| > 1$  **do**
- 2     let  $x_1, x_2$  be the two symbols with the smallest  $f$  values
- 3     introduce a new symbol  $x'$  and let  $f_{x'} = f_{x_1} + f_{x_2}$
- 4     let  $x_1$  and  $x_2$  be the two children of  $x'$
- 5      $S \leftarrow S \setminus \{x_1, x_2\} \cup \{x'\}$
- 6 **return** the tree constructed

# Example



# Algorithm using Priority Queue

## Huffman( $S, f$ )

- 1  $Q \leftarrow \text{build-priority-queue}(S)$
- 2 **while**  $Q.\text{size} > 1$  **do**
- 3      $x_1 \leftarrow Q.\text{extract-min}()$
- 4      $x_2 \leftarrow Q.\text{extract-min}()$
- 5     introduce a new symbol  $x'$  and let  $f_{x'} = f_{x_1} + f_{x_2}$
- 6     let  $x_1$  and  $x_2$  be the two children of  $x'$
- 7      $Q.\text{insert}(x')$
- 8 **return** the tree constructed

# Outline

- 1 Toy Examples
- 2 Interval Scheduling
- 3 Minimum Spanning Tree
  - Kruskal's Algorithm
  - Reverse-Kruskal's Algorithm
  - Prim's Algorithm
- 4 Single Source Shortest Paths
  - Dijkstra's Algorithm
- 5 Data Compression and Huffman Code
- 6 Summary**

# Summary for Greedy Algorithms

- 1 Design a “reasonable” strategy
  - Interval scheduling problem: schedule the job  $j^*$  with the earliest deadline
  - Kruskal’s algorithm for MST: select lightest edge  $e^*$
  - Inverse Kruskal’s algorithm for MST: drop the heaviest non-bridge edge  $e^*$
  - Prim’s algorithm for MST: select the lightest edge  $e^*$  incident to a specified vertex  $s$
  - Huffman codes: make the two least frequent symbols brothers

# Summary for Greedy Algorithms

- 1 Design “reasonable” strategy
- 2 Prove that the reasonable strategy is “safe”

**Def.** A choice is “safe” if there is an optimum solution that is “consistent” with the choice

- Usually done by “exchange argument”
- Interval scheduling problem: exchange  $j^*$  with the first job in an optimal solution
- Kruskal’s algorithm: exchange  $e^*$  with some edge  $e$  in the cycle in  $T \cup \{e^*\}$
- Prim’s algorithm: exchange  $e^*$  with some other edge  $e$  incident to  $s$

# Summary for Greedy Algorithms

- 1 Design “reasonable” strategy
- 2 Prove that the reasonable strategy is “safe”
- 3 Show that the remaining task after applying the strategy is to solve a (many) smaller instance(s) of the same problem
  - Interval scheduling problem: remove  $j^*$  and the jobs it conflicts with
  - Kruskal and Prim’s algorithms: contracting  $e^*$
  - Inverse Kruskal’s algorithm: remove  $e^*$
  - Huffman codes: merge two symbols into one

# Summary for Greedy Algorithms

- Dijkstra's algorithm does not quite fit in the framework.
- It combines “greedy algorithm” and “dynamic programming”
- Greedy algorithm: each time select the vertex in  $V \setminus S$  with the smallest  $d$  value and add it to  $S$
- Dynamic programming: remember the  $d$  values of vertices in  $S$  for future use
- Dijkstra's algorithm is very similar to Prim's algorithm for MST