Algorithms Design Chap04-Greedy Algorithms

College of Computer Science

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Chap04-Greedy Algorithms Outline

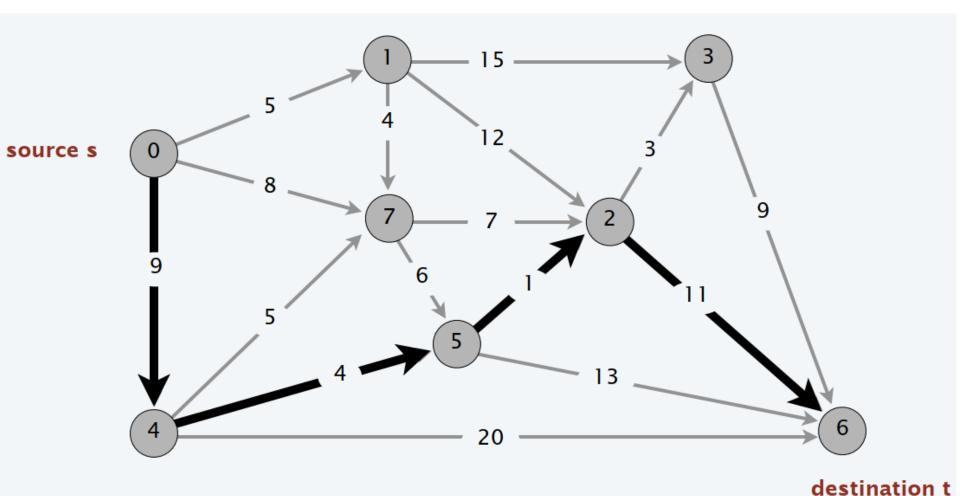
- 4.1 Interval Scheduling and Interval Partitioning
- 4.2 Scheduling to Minimize Lateness
- 4.3 Optimal Caching
- 4.4 Shortest Paths in a Graph
- 4.5 Minimum Spanning Tree
- 4.7 Clustering
- 4.8 Huffman Codes

Shortest path network.

- Directed graph G = (V, E).
- Source $s \in V$, destination $t \in V$.
- •Length l_e = length of edge $e.(l_e \ge 0)$

Goal

- find a shortest path from s to t.
- Cost of path = sum of edge costs in path



Suppose that you change the length of every edge of G as follows. For which is every shortest path in G a shortest path in G'?

- A. Add 17.
- B. Multiply by 17.
- C. Either A or B.
- D. Neither A nor B

Shortest path applications

- Map routing.
- Robot navigation.
- Texture mapping.
- Urban traffic planning.
- Network routing protocols (OSPF, BGP, RIP).

• . . .

Which variant in car GPS?

- A. Single source: from one node *s* to every other node.
- B. Single target: from every node to one node *t*.
- C. Source—target: from one node *s* to another node *t*.
- D. All pairs: between all pairs of nodes.

Dijkstra's algorithm

• single-source shortest paths problem

Greedy approach

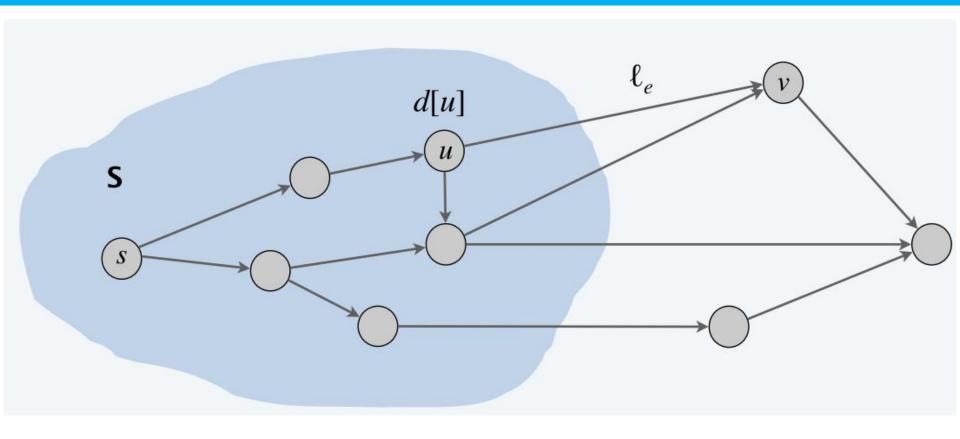
- Maintain a set of explored nodes S for which algorithm has determined the shortest path distance d[u] from s to u.
- Initialize $S = \{s\}, d[s] = 0$.

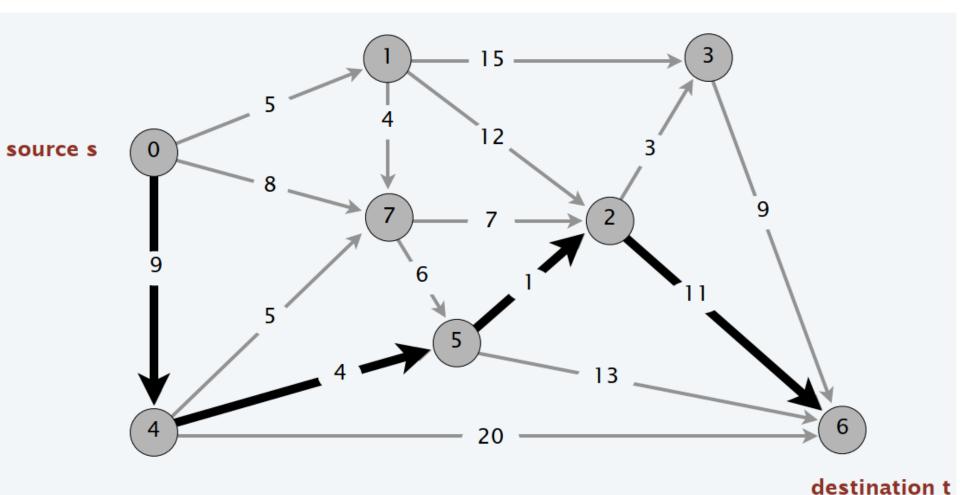
- [continue]
- Repeatedly choose unexplored node $v \notin S$ which minimizes

$$\pi(v) = \min_{e=(u,v): u \in S} d[u] + l_e$$

Add v to S, and set $d[v] = \pi(v)$

• To recover path, set pred[v] ← e that achieves min.





```
S = \{0\}, d[0] = 0
\pi[1]=5, \pi[7]=8, \pi[4]=9, S=\{0,1\}, d[1]=5, pred[1]=0
\pi[7] = \min(8,9), \pi[3] = 15 + 5 = 20, \pi[2] = 17, |\pi[4] = 9,
S = \{0,1,7\}, d[7] = 8, pred[7] = 0
\pi[5] = 8+6=14, \pi[2]=\min(8+7, 5+12), |\pi[4]=9.
\pi[3]=20, S={0,1,4,7}, d[4]=9, pred[4]=0
\pi[5] = \min(9+4,14), \pi[6] = 9+20 = 29, |\pi[2]=15,
\pi[3]=20, S={0,1,4,5,7},d[5]=13, pred[5]=4
```

```
\pi[2]=\min(13+1,8+7), \pi[6]=\min(13+13,29), |
\pi[3]=20, S=\{0,1,2,4,5,7\}, d[2]=14, pred[2]=5
\pi[3]=\min(20, 14+3), \pi[6]=\min(26, 14+11),
S=\{0,1,2,3,4,5,7\}, d[3]=17, pred[3]=2
\pi[6]=\min(17+9, 14+11)=25,
S=\{0,1,2,3,4,5,6,7\}, d[6]=25, pred[6]=2
```

	V1	V2	V3	V4	V5	V6	V7	S
R1	5= 5 (<i>v</i> ₀)	∞	∞	9 (v ₀)	∞	∞	(v_0)	{0,1}
R2		$5+12=17$ (v_1)	5+15=20 (v ₁)	9 (v ₀)	∞	∞	$min(9,8)$ =8(v_0)	{0,1,7}
R3		min(15,17) =15(v_7)	20 (v ₁)	9= 9 (v ₀)	8+6=14 (<i>v</i> ₇)	∞		{0,1,4,7 }
R4		15 (v ₇)	20 (v ₁)		$min(13, 14)=13$ (v_4)	20+9=29 (v ₄)		{0,1,4,5, 7}
R5		min(14,15) =14(v_5)	20 (v ₁)			min(26,29) =26(v_5)		{0,1,2,4, 5,7}
R6			min(17,20) $=17(v_2)$			min(25,26) =25(v_2)		{0,1,2,3, 4,5,7}
R7						min(26,25) =25(v_2)		{0,1,2,3, 4,5,6,7}

Invariant. For each node $u \in S$, d[u] is the length of the shortest s-u path.

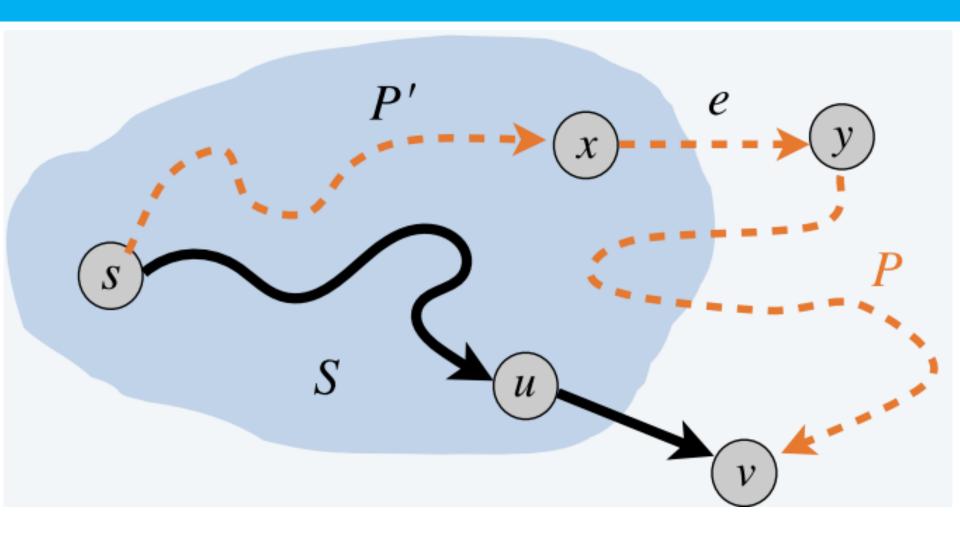
Pf. (by induction on |S|)

Base case: |S| = 1 is easy since $S = \{s\}$ and d[s] = 0.

Inductive hypothesis:

Assume true for $|S| = k \ge 1$.

- Let v be next node added to S, and let u-v be the chosen edge.
- The shortest s-u path plus (u, v) is an s-v path of length π (v).
- Consider any s-v path P. We'll see that it's no shorter than $\pi(v)$.
- Let x-y be the first edge in P that leaves S, and let P' be the subpath to x.
- *P* is already too long as soon as it leaves *S*.



Efficient implementation

• For each unexplored node $v \notin S$: explicitly maintain $\pi[v]$ instead of computing directly from definition

$$\pi(v) = \min_{e=(u,v): u \in S} d[u] + l_e$$

- The set of unexplored nodes can only decrease because set *S* increases.
- Specifically, it suffices to update: $\pi[v] = \min(\pi[v], \pi[u] + l_e)$

Efficient implementation

Use a min-oriented priority queue (PQ) to choose an unexplored node that minimizes $\pi[v]$.

Efficient Implementation.

- Algorithm maintains $\pi[v]$ for each node v.
- Priority Queue(PQ) stores unexplored nodes, using $\pi[]$ as priorities.
- •Once u is deleted from the PQ, $\pi[u] =$ length of a shortest s-u path.

DIJKSTRA (V, E, ℓ, s)

FOREACH $v \neq s$: $\pi[v] \leftarrow \infty$, $pred[v] \leftarrow null$; $\pi[s] \leftarrow 0$.

Create an empty priority queue pq.

FOREACH $v \in V$: INSERT $(pq, v, \pi[v])$.

WHILE (IS-NOT-EMPTY(pq))

 $u \leftarrow \text{DEL-MIN}(pq)$.

FOREACH edge $e = (u, v) \in E$ leaving u:

IF $(\pi[v] > \pi[u] + \ell_e)$

DECREASE-KEY(pq, v, $\pi[u] + \ell_e$).

 $\pi[v] \leftarrow \pi[u] + \ell_e$; $pred[v] \leftarrow e$.

Performance. n INSERT, n DELETE-MIN, $\leq m$ DECREASE-KEY.

priority queue	Insert	DELETE-MIN	Decrease-Key	total
node-indexed array (A[i] = priority of i)	<i>O</i> (1)	O(n)	<i>O</i> (1)	$O(n^2)$
binary heap	$O(\log n)$	$O(\log n)$	$O(\log n)$	$O(m \log n)$
d-way heap (Johnson 1975)	$O(d \log_d n)$	$O(d \log_d n)$	$O(\log_d n)$	$O(m \log_{m/n} n)$
Fibonacci heap (Fredman-Tarjan 1984)	<i>O</i> (1)	$O(\log n)^{\dagger}$	O(1) †	$O(m + n \log n)$
integer priority queue (Thorup 2004)	<i>O</i> (1)	$O(\log \log n)$	<i>O</i> (1)	$O(m + n \log \log n)$

Dijkstra's algorithm and proof extend to several related problems

- Shortest paths in undirected graphs: $\pi[v] \leq \pi[u] + \ell(u, v)$.
- Maximum capacity paths: $\pi[v] \ge \min(\pi[u], c(u, v))$.
- Maximum reliability paths: $\pi[v] \ge \pi[u] \times \gamma(u, v)$.

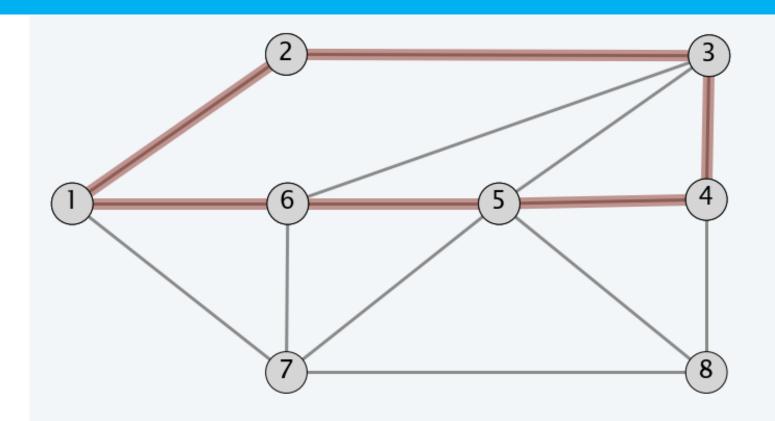
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Def.(4.5.2) A path is a sequence of edges which connects a sequence of nodes.

Def.(4.5.3) A cycle is a path with no repeated nodes or edges other than the starting and ending nodes.

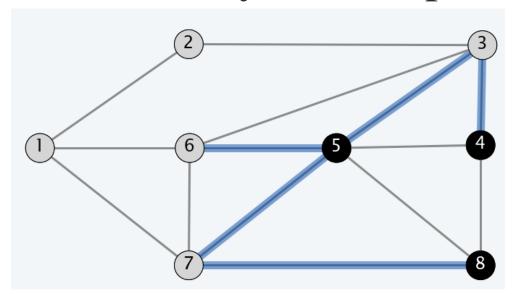


```
path P = { (1, 2), (2, 3), (3, 4), (4, 5), (5, 6) }

cycle C = { (1, 2), (2, 3), (3, 4), (4, 5), (5, 6), (6, 1) }
```

Def. A cut is a partition of the nodes into two nonempty subsets S and V-S.

Def. The cutset of a cut S is the set of edges with exactly one endpoint in S.



Quiz(4.5.1)

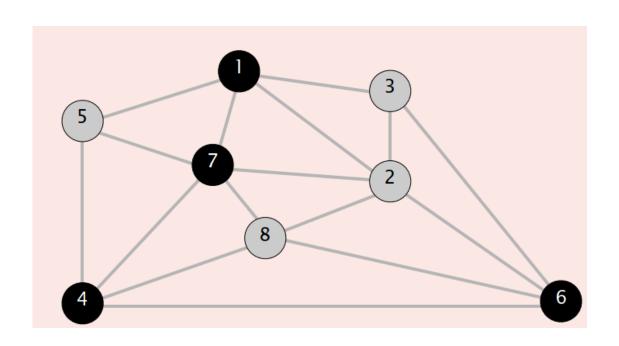
Consider the cut $S = \{1, 4, 6, 7\}$. Which edge is in the cutset of S?

A. S is not a cut (not connected)

B. 1–7

C. 5–7

D. 2-3



Quiz(4.5.2)

Let *C* be a cycle and let *D* be a cutset. How many edges do *C* and *D* have in common? Choose the best answer.

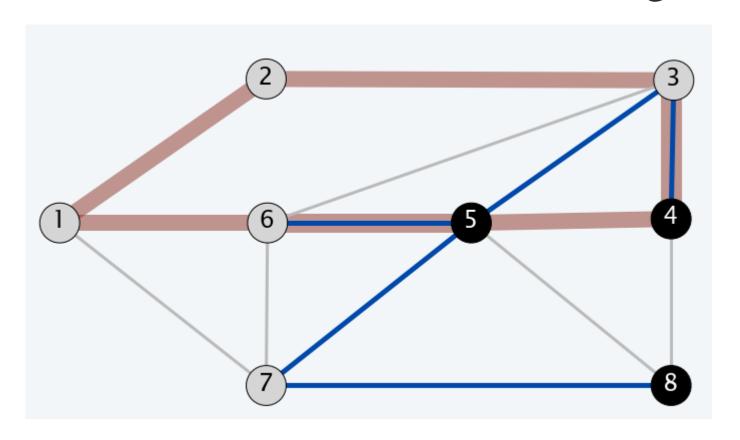
A. 0

B. 2

C. not 1

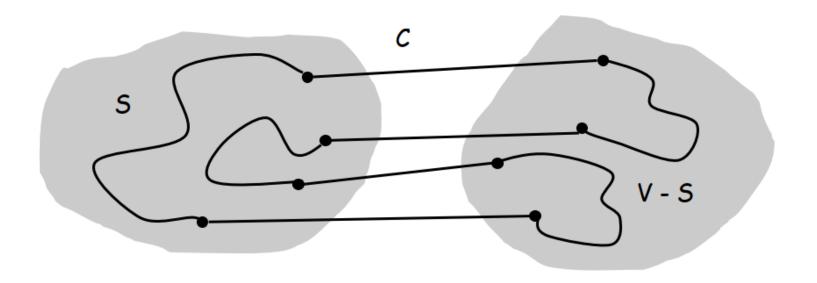
D. an even number

Proposition. A cycle and a cutset intersect in an even number of edges.



Proposition. A cycle and a cutset intersect in an even number of edges.

Pf. [by picture]



Def. Let H = (V, T) be a subgraph of an undirected graph G = (V, E). H is a spanning tree of G if H is both acyclic and connected.

graph G = (V, E)
spanning tree H = (V, T)

Proposition.

Let H = (V, T) be a subgraph of an undirected graph G = (V, E).

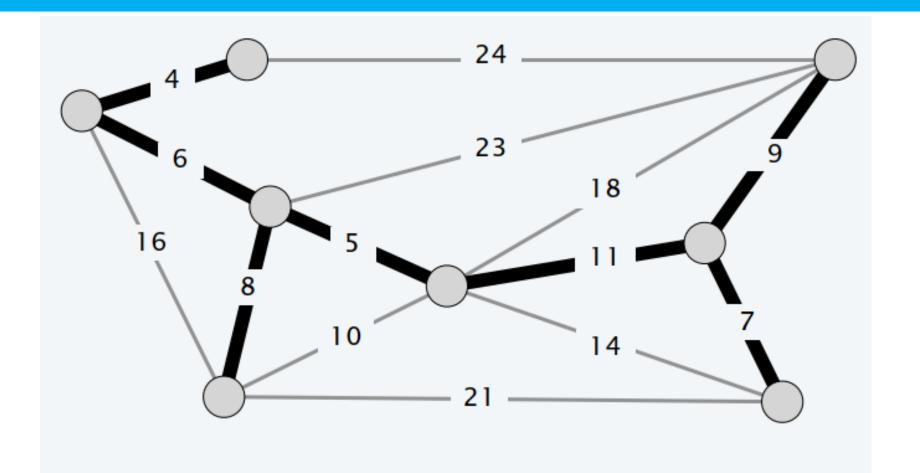
Then, the following are equivalent:

- *H* is a spanning tree of *G*.
- *H* is acyclic and connected.
- H is connected and has |V| 1 edges.
- H is acyclic and has |V| 1 edges.
- *H* is minimally connected: removal of any edge disconnects it.
- *H* is maximally acyclic: addition of any edge creates a cycle.

Minimum Spanning Tree(MST).

Def. Given a connected, undirected graph G = (V, E) with edge weights c_e , a MST is a spanning tree of G such that the sum of edge weights of tree is minimized.

• Note that the set of edges of a MST is a subset of the edges *E*.



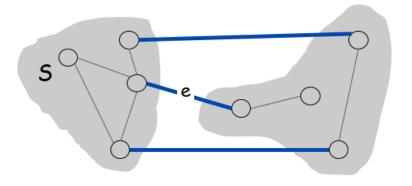
MST cost = 50 = 4 + 6 + 8 + 5 + 11 + 9 + 7

Simplifying assumption.

• All edge costs c_e are distinct.

Cut property.

• Let S be any subset of nodes, and let e be the min-cost edge in the cutset of S. Then the MST contains e.



e is in the MST

Simplifying assumption.

• All edge costs c_e are distinct.

Cycle property.

•Let *C* be any cycle, and let *f* be the max cost edge belonging to *C*. Then the MST does not contain *f*.

f is not in the MST

Cut property. Let S be any subset of nodes, and let e be the min cost edge with exactly one endpoint in S. Then the MST contains e.

Pf.

- Suppose e does not belong to T^* , and let's see what happens.
- Adding e to T^* creates a cycle C in T^* .
- Edge *e* is both in the cycle *C* and in the cutset *D* corresponding to $S \Rightarrow$ there exists another edge, say f, that is in both C and D.
- $T' = T^* \cup \{e\} \{f\}$ is also a spanning tree.
- Since $c_e < c_f$, $cost(T') < cost(T^*)$.
- This is a contradiction.

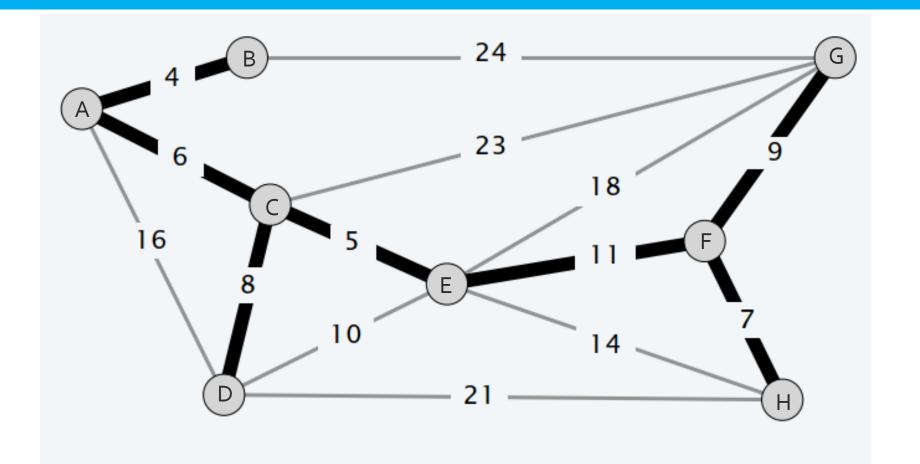
Cycle property. Let C be any cycle, and let f be the max cost edge belonging to C. Then the MST does not contain f.

Pf.

- Suppose f belongs to T^* , and let's see what happens.
- Deleting f from T^* creates a cut S in T^* .
- Edge f is both in the cycle C and in the cutset D corresponding to $S \Rightarrow$ there exists another edge, say e, that is in both C and D.
- $T' = T^* \cup \{e\} \{f\}$ is also a spanning tree.
- Since $c_e < c_f$, $cost(T') < cost(T^*)$.
- This is a contradiction.

Prim's algorithm. [Jarn & 1930, Dijkstra 1957, Prim 1959]

- •Initialize $S = \{s\}$ for any node s, $T = \emptyset$.
- Repeat n-1 times:
 - Add to T a min-cost edge in the cutset of S.
 - Add the other endpoint to *S*.



MST cost = 50 = 4 + 6 + 8 + 5 + 11 + 9 + 7

Prim Algorithm:(*TBC*)

$$5:S = \{A,B,C,E,D,F\},T = \{A-B,A-C,C-E,C-D,E-F\}$$

 $6:S = \{A,B,C,E,D,F,H\},T = \{A-B,A-C,C-E,C-D,E-F,F-H\}$

7:S ={A,B,C,E,D,F,H,G},T={A-B,A-C,C-E,C-D,E-F,F-H,F-G}

Start with a root node and grow greedily a tree outward.

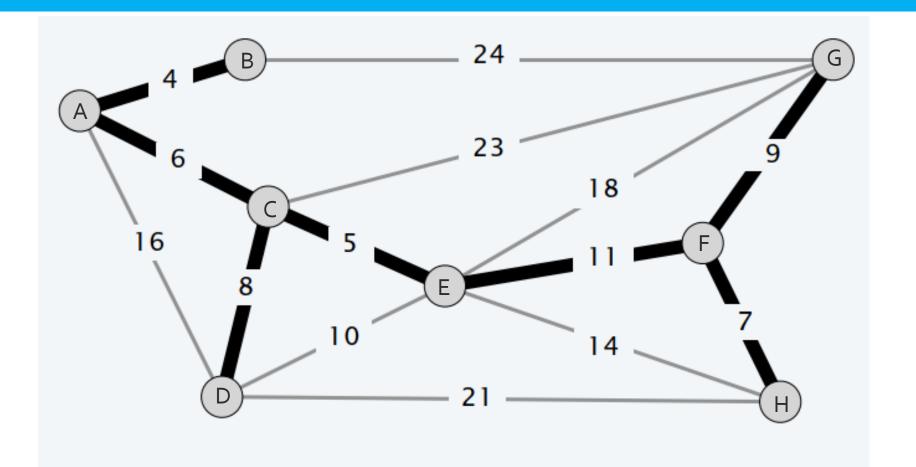
Theorem. Prim's algorithm can be implemented to run in $O(m \log n)$ time.

Pf. Implementation almost identical to Dijkstra's algorithm.

```
PRIM (V, E, c)
S \leftarrow \emptyset, T \leftarrow \emptyset.
s \leftarrow any node in V.
FOREACH v \neq s: \pi[v] \leftarrow \infty, pred[v] \leftarrow null; \pi[s] \leftarrow 0.
Create an empty priority queue pq.
FOREACH v \in V: INSERT(pq, v, \pi[v]).
WHILE (IS-NOT-EMPTY(pq))
                                                           \pi[v] = \text{cost of cheapest}
                                                        known edge between v and S
   u \leftarrow \text{DEL-MIN}(pq).
   S \leftarrow S \cup \{u\}, T \leftarrow T \cup \{pred[u]\}.
   FOREACH edge e = (u, v) \in E with v \notin S:
       IF (c_e < \pi[v])
           DECREASE-KEY(pq, v, c_e).
           \pi[v] \leftarrow c_e; pred[v] \leftarrow e.
```

Kruskal's algorithm. [Kruskal, 1956]

- Sort edges in ascending order of cost
- Repeat *m* times:
 - Select the min-cost edge e so far
 - If adding *e* to *T* creates a cycle, discard *e* according to cycle property.
 - Otherwise, insert e = (u, v) into T according to cut property where S = set of nodes in u's connected component.

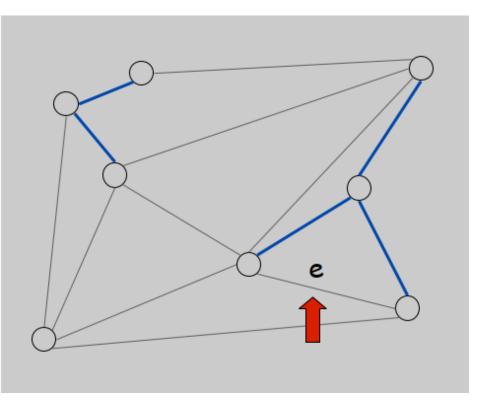


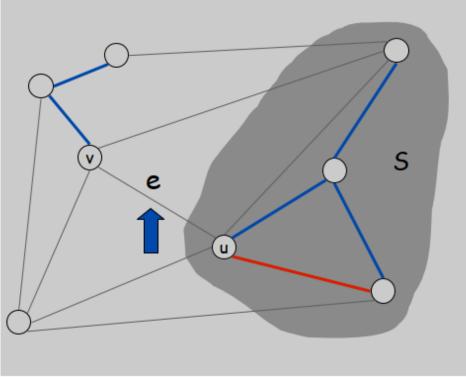
MST cost = 50 = 4 + 6 + 8 + 5 + 11 + 9 + 7

```
Kruskal's algorithm:(n=8,m=13)
e=(A,B)[4],T=\{A-B\}
e=(C,E)[5],T=\{A-B,C-E\}
e=(A,C)[6],T=\{A-B,C-E,A-C\}
e=(F,H)[7],T={A-B,C-E,A-C,F-H}
e=(C,D)[8],T={A-B,C-E,A-C,F-H,C-}
D, }
```

In brief, start without any edges at all and build a SMT by inserting edges in the ascending order of cost.

E.g.





Theorem. Kruskal's algorithm can be implemented to run in $O(m \log m)$ time.

- Sort edges by cost.
- Use union—find data structure to dynamically maintain connected components.

```
Kruskal(G, c) {
    Sort edges weights so that c_1 \le c_2 \le \ldots \le c_m.
    T ← b
    foreach (u ∈ V) make a set containing singleton u
    for i = 1 to m are u and v in different connected components?
        (\mathbf{u},\mathbf{v}) = \mathbf{e},
       if (u and v are in different sets) {
           T \leftarrow T \cup \{e_i\}
           merge the sets containing u and v
                          merge two components
    return T
```

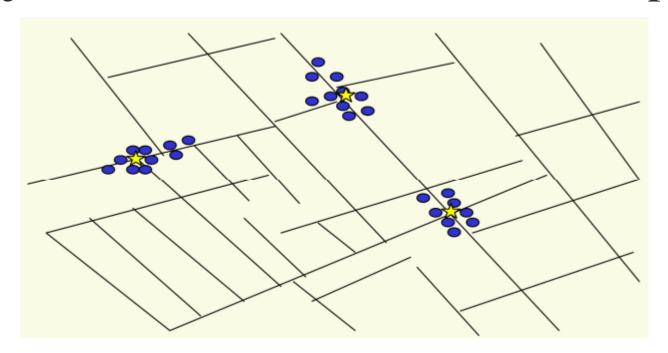
Reverse-delete algorithm

- Start with all edges in *T* and consider them in descending order of cost
- Delete edge from T unless it would disconnect T

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Goal. Given a set U of n objects labeled p_1, \dots, p_n , partition into clusters so that objects in different clusters are far apart.



Applications

- Routing in mobile ad-hoc networks.
- Document categorization for web search.
- Similarity searching in medical image databases
- Identify patterns in gene expression

k-clustering.

• Divide objects into *k* non-empty groups.

Distance function.

- Numeric value specifying "closeness" of two objects.
- Assume it satisfies several properties.
 - $d(p_i, p_j) = 0$ iff $p_i = p_j$ [identity]
 - $d(p_i, p_i) \ge 0$ [non-negativity]
 - $d(p_i, p_i) = d(p_i, p_i)$ [symmetry]

Single-linkage *k*-clustering algorithm

- Form a graph on the node set *U*, corresponding to *n* clusters.
- Repeat n-k times until there are exactly k clusters.
 - Find the closest pair of objects such that each object is in a different cluster, and add an edge between them.

Key observation.

• This procedure is precisely Kruskal's algorithm (except we stop when there are *k* connected components)

Theorem. Let C^* denote the clustering C_1^*, \dots, C_k^* formed by deleting the k-1 longest edges of an MST. Then, C^* is a k-clustering of max spacing.

Pf.

- Let C denote any other clustering C_1, \dots, C_k .
- Let p_i and p_j be in the same cluster in C^* , say C_r^* , but different clusters in C, say C_s and C_t .
- Some edge (p, q) on $p_i p_j$ path in C_r^* spans two different clusters in C.
- Spacing of $C^* = \text{length } d^*$ of the $(k-1)^{St}$ longest edge in MST.
- Edge (p, q) has length $\leq d^*$ since it was added by Kruskal.
- Spacing of C is $\leq d^*$ since p and q are in different clusters.

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- Q. Given a text that uses 32 symbols (26 different letters, space, and some punctuation characters), how can we encode this text in bits?
- A. We can encode 2⁵ different symbols using a fixed length of 5 bits per symbol. This is called fixed length encoding.
- Q. Some symbols (i.e., e, t, a, o, i, n) are used far more often than others. How can we use this to reduce our encoding?
- A. Encode these characters with fewer bits, and the others with more bits.

- Q. How do we know when the next symbol begins?
- A. Use a separation symbol (like the pause in Morse), or make sure that there is no ambiguity by ensuring that no code is a prefix of another one.

E.g., c(a)=01, c(b)=010, c(e)=1

What is 0101?

Definition. A prefix code for a set S is a function c that maps each $x \in S$ to 1s and 0s in such a way that for $x, y \in S$, $x \neq y$, c(x) is not a prefix of c(y).

E.g.,

- c(a) = 11
- c(e) = 01
- c(k) = 001
- c(l) = 10
- c(u) = 000
- Q. What is the meaning of 1001000001?

Definition. The average bits per letter of a prefix code *c* is the sum over all symbols of its frequency times the number of bits of its encoding:

$$ABL(c) = \sum_{x \in S} f_c |c(x)|$$

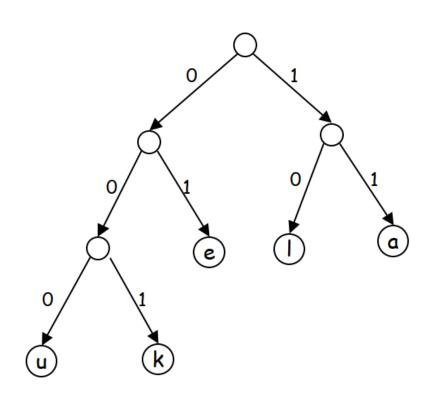
We would like to find a prefix code that is has the lowest possible average bits per letter.

Representing Prefix Codes using

Binary Trees

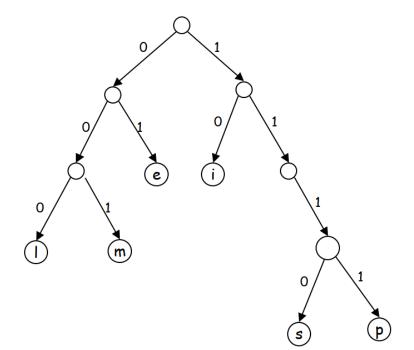
E.g.,

- c(a) = 11
- c(e) = 01
- c(k) = 001
- c(l) = 10
- c(u) = 000



- Q. How does the tree of a prefix code look?
- A. Only the leaves have a label.
- Pf. An encoding of x is a prefix of an encoding of y if and only if the path of x is a prefix of the path of y.

- Q. How can this prefix code be made more efficient?
- A. Change encoding of *p* and *s* to a shorter one. This tree is now full.



Definition. A tree is full if every node that is not a leaf has two children.

Claim. The binary tree corresponding to the optimal prefix code is full.

Pf. [by contradiction]

- Suppose *T* is binary tree of optimal prefix code and is not full.
- This means there is a node u with only one child v.
- Case 1: *u* is the root; delete *u* and use *v* as the root

Pf. [TBC]

- Case 2: *u* is not the root
 - let w be the parent of u
 - delete u and make v be a child of w in place of u
- In both cases the number of bits needed to encode any leaf in the subtree of v is decreased. The rest of the tree is not affected.
- Clearly this new tree T' has a smaller ABL than T. Contradiction.

- Q. Where in the tree of an optimal prefix code should letters be placed with a high frequency?
- A. Near the top.

Greedy choice.

- Create tree top-down, split S into two sets S_1 and S_2 with (almost) equal frequencies.
- Recursively build tree for S_1 and S_2 .

Optimal Prefix Codes: Huffman Encoding

Observation 1. Lowest frequency items should be at the lowest level in tree of optimal prefix code.

Observation 2. For n > 1, the lowest level always contains at least two leaves.

Observation 3. The order in which items appear in a level does not matter.

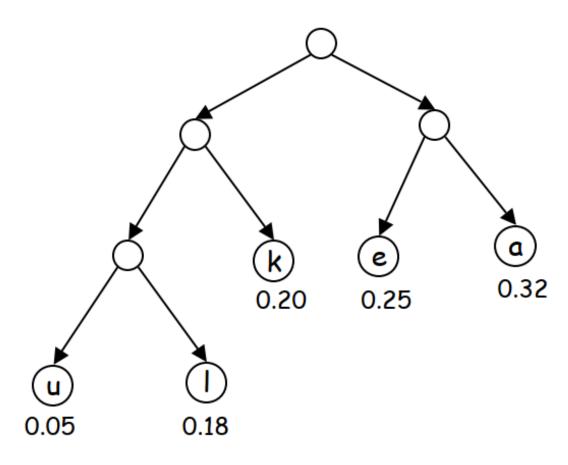
Claim. There is an optimal prefix code with tree T^* where the two lowest-frequency letters are assigned to leaves that are siblings in T^* .

Greedy Choice. [Huffman, 1952]

- Create tree bottom-up.
- Make two leaves for two lowest-frequency letters y and z.
- Recursively build tree for the rest using a meta-letter for yz.

```
Huffman(S) {
   if |S|=2 {
       return tree with root and 2 leaves
   } else {
       let y and z be lowest-frequency letters in S
      S' = S
       remove y and z from S'
       insert new letter \omega in S' with f_{\omega} = f_{v} + f_{z}
       T' = Huffman(S')
       T = add two children y and z to leaf \omega from T'
      return T
```

 f_a =0.32, f_e =0.25, f_k =0.20, f_l =0.18, f_u =0.05



Thanks for Listening

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