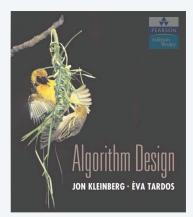


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http://www.cs.princeton.edu/~wayne/kleinberg-tardos

4. GREEDY ALGORITHMS II

- Dijkstra's algorithm
- minimum spanning trees
- → Prim, Kruskal, Boruvka
- ▶ single-link clustering
- ▶ min-cost arborescences

Last updated on Sep 8, 2013 6:30 AM



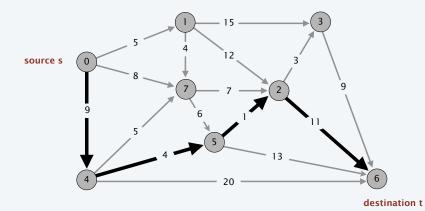
SECTION 4.4

4. GREEDY ALGORITHMS II

- Dijkstra's algorithm
- ▶ minimum spanning trees
- ▶ Prim, Kruskal, Boruvka
- ▶ single-link clustering
- ▶ min-cost arborescences

Shortest-paths problem

Problem. Given a digraph G = (V, E), edge lengths $\ell_e \ge 0$, source $s \in V$, and destination $t \in V$, find the shortest directed path from s to t.



length of path = 9 + 4 + 1 + 11 = 25

Car navigation



3

Shortest path applications

- PERT/CPM.
- Map routing.
- · Seam carving.
- · Robot navigation.
- · Texture mapping.
- Typesetting in LaTeX.
- · Urban traffic planning.
- Telemarketer operator scheduling.
- Routing of telecommunications messages.
- Network routing protocols (OSPF, BGP, RIP).
- Optimal truck routing through given traffic congestion pattern.

Reference: Network Flows: Theory, Algorithms, and Applications, R. K. Ahuja, T. L. Magnanti, and J. B. Orlin, Prentice Hall, 1993.

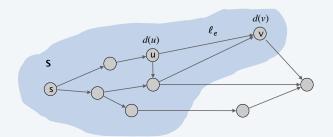
Dijkstra's algorithm

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 $S = \{s\}, d(s) = 0.$
- Repeatedly choose unexplored node v which minimizes

$$\pi(v) = \min_{e = (u,v) : u \in S} d(u) + \ell_e,$$
 add v to S , and set $d(v) = \pi(v)$. Shortest path to some node u in explored part, followed by a single edge (u,v)



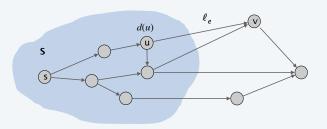
Dijkstra's algorithm

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 $S = \{ s \}, d(s) = 0.$
- Repeatedly choose unexplored node v which minimizes

$$\pi(v) = \min_{e = (u, v) : u \in S} d(u) + \ell_e,$$
shortest path to some node u in explored part, followed by a single edge (u, v)



Dijkstra's algorithm: proof of correctness

Invariant. For each node $u \in S$, d(u) is the length of the shortest $s \rightarrow u$ path.

Pf. [by induction on |S|]

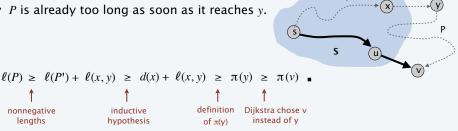
nonnegative

lengths

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 final edge.
- The shortest $s \rightarrow u$ path plus (u, v) is an $s \rightarrow v$ path of length $\pi(v)$.
- Consider any $s \rightarrow v$ path P. We show that it is 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 reaches y.



Dijkstra's algorithm: efficient implementation

Critical optimization 1. For each unexplored node ν , explicitly maintain $\pi(\nu)$ instead of computing directly from formula:



$$\pi(v) = \min_{e = (u,v): u \in S} d(u) + \ell_e.$$

- For each $v \notin S$, $\pi(v)$ can only decrease (because S only increases).
- More specifically, suppose *u* is added to *S* and there is an edge (*u*, *v*) leaving *u*. Then, it suffices to update:

$$\pi(v) = \min \{ \pi(v), d(u) + \ell(u, v) \}$$

Critical optimization 2. Use a priority queue to choose the unexplored node that minimizes $\pi(v)$.

9

Dijkstra's algorithm: efficient implementation

Implementation.

- Algorithm stores d(v) for each explored node v.
- Priority queue stores $\pi(v)$ for each unexplored node v.
- Recall: $d(u) = \pi(u)$ when u is deleted from priority queue.

DIJKSTRA (V, E, s)

Create an empty priority queue.

FOR EACH $v \neq s$: $d(v) \leftarrow \infty$; $d(s) \leftarrow 0$.

FOR EACH $v \in V$: *insert* v with key d(v) into priority queue.

WHILE (the priority queue *is not empty*)

 $u \leftarrow delete-min$ from priority queue.

FOR EACH edge $(u, v) \in E$ leaving u:

IF
$$d(v) > d(u) + \ell(u, v)$$

decrease-key of v to $d(u) + \ell(u, v)$ in priority queue.

$$d(v) \leftarrow d(u) + \ell(u, v)$$
.

10

Dijkstra's algorithm: which priority queue?

Performance. Depends on PQ: *n* insert, *n* delete-min, *m* decrease-key.

- Array implementation optimal for dense graphs.
- · Binary heap much faster for sparse graphs.
- 4-way heap worth the trouble in performance-critical situations.
- Fibonacci/Brodal best in theory, but not worth implementing.

PQ implementation	insert	delete-min	decrease-key	total
unordered array	<i>O</i> (1)	O(n)	O(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) †	$O(m + n \log n)$
Brodal queue (Brodal 1996)	<i>O</i> (1)	$O(\log n)$	<i>O</i> (1)	$O(m + n \log n)$

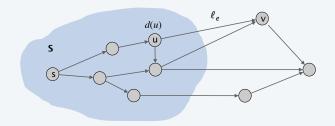
† amortized

Extensions of Dijkstra's algorithm

Dijkstra's algorithm and proof extend to several related problems:

- Shortest paths in undirected graphs: $d(v) \le d(u) + \ell(u, v)$.
- Maximum capacity paths: $d(v) \ge \min \{ \pi(u), c(u, v) \}$.
- Maximum reliability paths: $d(v) \ge d(u) \times \gamma(u, v)$.
- ...

Key algebraic structure. Closed semiring (tropical, bottleneck, Viterbi).





SECTION 6.1

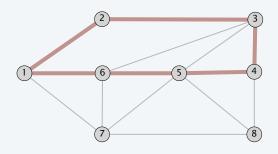
4. GREEDY ALGORITHMS II

- ▶ Dijkstra's algorithm
- minimum spanning trees
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- min-cost arborescences

Cycles and cuts

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

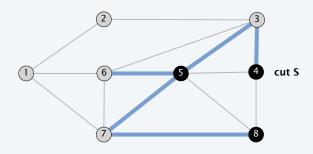
Def. A cycle is a path with no repeated nodes or edges other than the starting and ending nodes.



Cycles and cuts

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

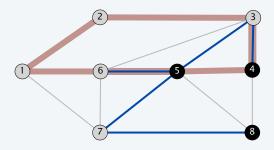
Def. The cutset of a cut S is the set of edges with exactly one endpoint in S.



cutset D = { (3, 4), (3, 5), (5, 6), (5, 7), (8, 7) }

Cycle-cut intersection

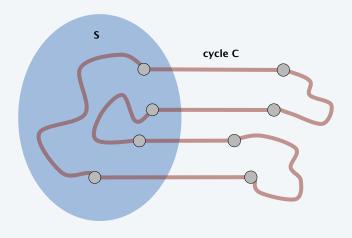
Proposition. A cycle and a cutset intersect in an even number of edges.



 $cutset \ D = \{ \ (3, \, 4), \, (3, \, 5), \, (5, \, 6), \, (5, \, 7), \, (8, \, 7) \, \}$ $cycle \ C = \{ \ (1, \, 2), \, (2, \, 3), \, (3, \, 4), \, (4, \, 5), \, (5, \, 6), \, (6, \, 1) \, \}$ $intersection \ C \cap D = \{ \ (3, \, 4), \, (5, \, 6) \, \}$

Cycle-cut intersection

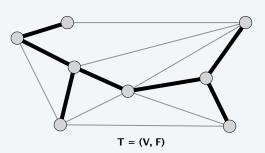
Proposition. A cycle and a cutset intersect in an even number of edges. Pf. [by picture]



Spanning tree properties

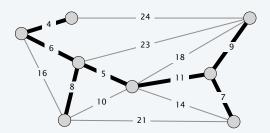
Proposition. Let T = (V, F) be a subgraph of G = (V, E). TFAE:

- *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

Given a connected graph G = (V, E) with edge costs c_e , an MST is a subset of the edges $T \subseteq E$ such that T is a spanning tree whose sum of edge costs is minimized.



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

Applications

MST is fundamental problem with diverse applications.

- Dithering.
- · Cluster analysis.
- · Max bottleneck paths.
- · Real-time face verification.
- LDPC codes for error correction.
- Image registration with Renyi entropy.
- · Find road networks in satellite and aerial imagery.
- Reducing data storage in sequencing amino acids in a protein.
- · Model locality of particle interactions in turbulent fluid flows.
- Autoconfig protocol for Ethernet bridging to avoid cycles in a network.
- $\bullet \ \, \text{Approximation algorithms for NP-hard problems (e.g., TSP, Steiner tree)}.$

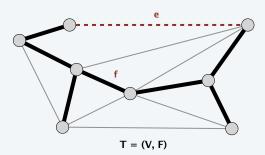
• Network design (communication, electrical, hydraulic, computer, road).

Cayley's theorem. There are n^{n-2} spanning trees of K_n . \leftarrow can't solve by brute force

Fundamental cycle

Fundamental cycle.

- Adding any non-tree edge e to a spanning tree T forms unique cycle C.
- Deleting any edge $f \in C$ from $T \cup \{e\}$ results in new spanning tree.

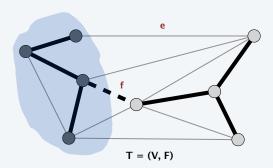


Observation. If $c_e < c_f$, then T is not an MST.

Fundamental cutset

Fundamental cutset.

- Deleting any tree edge f from a spanning tree T divide nodes into two connected components. Let D be cutset.
- Adding any edge $e \in D$ to $T \{f\}$ results in new spanning tree.



Observation. If $c_e < c_f$, then T is not an MST.

The greedy algorithm

Red rule.

- Let *C* be a cycle with no red edges.
- Select an uncolored edge of C of max weight and color it red.

Blue rule.

- Let *D* be a cutset with no blue edges.
- Select an uncolored edge in *D* of min weight and color it blue.

Greedy algorithm.

- Apply the red and blue rules (non-deterministically!) until all edges are colored. The blue edges form an MST.
- Note: can stop once n-1 edges colored blue.

Greedy algorithm: proof of correctness

Color invariant. There exists an MST T^* containing all of the blue edges and none of the red edges.

Pf. [by induction on number of iterations]

Base case. No edges colored ⇒ every MST satisfies invariant.

22

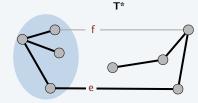
Greedy algorithm: proof of correctness

Color invariant. There exists an MST T^* containing all of the blue edges and none of the red edges.

Pf. [by induction on number of iterations]

Induction step (blue rule). Suppose color invariant true before blue rule.

- let *D* be chosen cutset, and let *f* be edge colored blue.
- if $f \in T^*$, T^* still satisfies invariant.
- Otherwise, consider fundamental cycle C by adding f to T^* .
- let $e \in C$ be another edge in D.
- e is uncolored and $c_e \ge c_f$ since
- $-e \in T^* \Rightarrow e \text{ not red}$
- blue rule \Rightarrow *e* not blue and $c_e \ge c_f$
- Thus, $T^* \cup \{f\} \{e\}$ satisfies invariant.



25

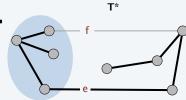
Greedy algorithm: proof of correctness

Color invariant. There exists an MST T^* containing all of the blue edges and none of the red edges.

Pf. [by induction on number of iterations]

Induction step (red rule). Suppose color invariant true before red rule.

- let *C* be chosen cycle, and let *e* be edge colored red.
- if $e \notin T^*$, T^* still satisfies invariant.
- Otherwise, consider fundamental cutset D by deleting e from T^* .
- let $f \in D$ be another edge in C.
- f is uncolored and $c_e \ge c_f$ since
- $f \notin T^* \Rightarrow f$ not blue
- red rule \Rightarrow f not red and $c_e \ge c_f$
- Thus, $T^* \cup \{f\} \{e\}$ satisfies invariant.

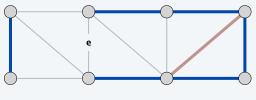


Greedy algorithm: proof of correctness

Theorem. The greedy algorithm terminates. Blue edges form an MST.

Pf. We need to show that either the red or blue rule (or both) applies.

- Suppose edge e is left uncolored.
- Blue edges form a forest.
- Case 1: both endpoints of \emph{e} are in same blue tree.
 - \Rightarrow apply red rule to cycle formed by adding e to blue forest.



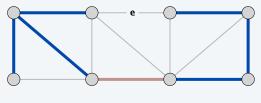
Case 1

Greedy algorithm: proof of correctness

Theorem. The greedy algorithm terminates. Blue edges form an MST.

Pf. We need to show that either the red or blue rule (or both) applies.

- Suppose edge \emph{e} is left uncolored.
- Blue edges form a forest.
- ullet Case 1: both endpoints of e are in same blue tree.
 - \Rightarrow apply red rule to cycle formed by adding e to blue forest.
- ullet Case 2: both endpoints of e are in different blue trees.
 - \Rightarrow apply blue rule to cutset induced by either of two blue trees. ullet



Case 2



SECTION 6.2

4. GREEDY ALGORITHMS II

- ▶ Dijkstra's algorithm
- ▶ minimum spanning trees
- Prim, Kruskal, Boruvka
- ▶ single-link clustering
- min-cost arborescences

D::| . . . | .:|

 $\bullet\,$ Add to tree the min weight edge with one endpoint in S.

• Add new node to *S*.

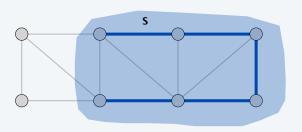
Prim's algorithm

Repeat n-1 times:

Initialize S =any node.

Theorem. Prim's algorithm computes the MST.

Pf. Special case of greedy algorithm (blue rule repeatedly applied to S). •



Prim's algorithm: implementation

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

Pf. Implementation almost identical to Dijkstra's algorithm.

[d(v) = weight of cheapest known edge between v and S]

PRIM(V, E, c)

Create an empty priority queue.

 $s \leftarrow \text{any node in } V$.

FOR EACH $v \neq s$: $d(v) \leftarrow \infty$; $d(s) \leftarrow 0$.

FOR EACH v: *insert* v with key d(v) into priority queue.

WHILE (the priority queue *is not empty*)

 $u \leftarrow delete-min$ from priority queue.

FOR EACH edge $(u, v) \in E$ incident to u:

IF d(v) > c(u, v)

decrease-key of v to c(u, v) in priority queue.

 $d(v) \leftarrow c(u, v)$.

Kruskal's algorithm

Consider edges in ascending order of weight:

· Add to tree unless it would create a cycle.



Theorem. Kruskal's algorithm computes the MST.

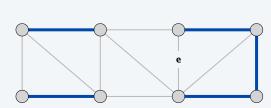
Pf. Special case of greedy algorithm.

• Case 1: both endpoints of e in same blue tree.

⇒ color red by applying red rule to unique cycle.

ullet Case 2. If both endpoints of e are in different blue trees.

⇒ color blue by applying blue rule to cutset defined by either tree. •



no edge in cutset has smaller weight (since Kruskal chose it first)

all other edges in cycle are blue

2.5

Kruskal's algorithm: implementation

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

- · Sort edges by weight.
- · Use union-find data structure to dynamically maintain connected components.

```
KRUSKAL (V, E, c)
SORT m edges by weight so that c(e_1) \le c(e_2) \le ... \le c(e_m)
S \leftarrow \phi
FOREACH v \in V: MAKESET(v).
FOR i = 1 TO m
   (u, v) \leftarrow e_i
   IF FINDSET(u) \neq FINDSET(v) \leftarrow are u ariu v iii same component?
       S \leftarrow S \cup \{e_i\}
       UNION(u, v). \longleftarrow make u and v in same component
RETURN S
```

Consider edges in descending order of weight:

Reverse-delete algorithm

· Remove edge unless it would disconnect the graph.

Theorem. The reverse-delete algorithm computes the MST. Pf. Special case of greedy algorithm.

- Case 1: removing edge *e* does not disconnect graph.
 - \Rightarrow apply red rule to cycle C formed by adding e to existing path between its two endpoints any edge in C with larger weight would have been deleted when considered
- Case 2: removing edge *e* disconnects graph.
 - \Rightarrow apply blue rule to cutset *D* induced by either component.

e is the only edge in the cutset (any other edges must have been colored red / deleted)

Fact. [Thorup 2000] Can be implemented in $O(m \log n (\log \log n)^3)$ time.

Review: the greedy MST algorithm

Red rule.

- Let C be a cycle with no red edges.
- Select an uncolored edge of C of max weight and color it red.

Blue rule.

- Let D be a cutset with no blue edges.
- Select an uncolored edge in D of min weight and color it blue.

Greedy algorithm.

- Apply the red and blue rules (non-deterministically!) until all edges are colored. The blue edges form an MST.
- Note: can stop once n-1 edges colored blue.

Theorem. The greedy algorithm is correct.

Special cases. Prim, Kruskal, reverse-delete, ...

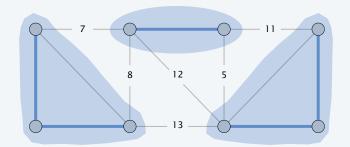
Borůvka's algorithm

Repeat until only one tree.

- Apply blue rule to cutset corresponding to each blue tree.
- · Color all selected edges blue.

Theorem. Borůvka's algorithm computes the MST. ← assume edge costs are distinct

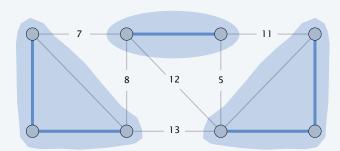
Pf. Special case of greedy algorithm (repeatedly apply blue rule). •



Borůvka's algorithm: implementation

Theorem. Borůvka's algorithm can be implemented in $O(m \log n)$ time. Pf.

- To implement a phase in O(m) time:
 - compute connected components of blue edges
- for each edge $(u, v) \in E$, check if u and v are in different components; if so, update each component's best edge in cutset
- At most $\log_2 n$ phases since each phase (at least) halves total # trees.



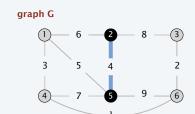
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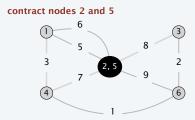
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Borůvka's algorithm: implementation

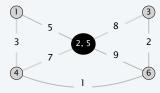
Node contraction version.

- After each phase, contract each blue tree to a single supernode.
- Delete parallel edges (keeping only cheapest one) and self loops.
- Borůvka phase becomes: take cheapest edge incident to each node.





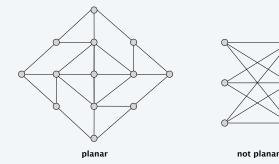
delete parallel edges and self loops



Borůvka's algorithm on planar graphs

Theorem. Borůvka's algorithm runs in O(n) time on planar graphs. Pf.

- To implement a Borůvka phase in O(n) time:
 - use contraction version of algorithm
- in planar graphs, $m \le 3n 6$.
- graph stays planar when we contract a blue tree
- Number of nodes (at least) halves.
- At most $\log_2 n$ phases: $cn + cn/2 + cn/4 + cn/8 + \dots = O(n)$.



Borůvka-Prim algorithm

Borůvka-Prim algorithm.

- Run Borůvka (contraction version) for $\log_2\log_2 n$ phases.
- Run Prim on resulting, contracted graph.

Theorem. The Borůvka-Prim algorithm computes an MST and can be implemented in $O(m \log \log n)$ time.

Pf.

- Correctness: special case of the greedy algorithm.
- The $\log_2 \log_2 n$ phases of Borůvka's algorithm take $O(m \log \log n)$ time; resulting graph has at most $n / \log_2 n$ nodes and m edges.
- Prim's algorithm (using Fibonacci heaps) takes O(m+n) time on a graph with $n/\log_2 n$ nodes and m edges. •

$$O\left(m + \frac{n}{\log n} \log \left(\frac{n}{\log n}\right)\right)$$

Does a linear-time MST algorithm exist?

deterministic compare-based MST algorithms

worst case	discovered by	
$O(m \log \log n)$	Yao	
$O(m \log \log n)$	Cheriton-Tarjan	
$O(m \log^* n) \ O(m + n \log n)$	Fredman-Tarjan	
$O(m \log (\log^* n))$	Gabow-Galil-Spencer-Tarjan	
$O(m \alpha(n) \log \alpha(n))$	Chazelle	
$O(m \alpha(n))$	Chazelle	
optimal	Pettie-Ramachandran	
O(m)	???	
	$O(m \log \log n)$ $O(m \log \log n)$ $O(m \log^* n) O(m + n \log n)$ $O(m \log (\log^* n))$ $O(m \alpha(n) \log \alpha(n))$ $O(m \alpha(n))$ $optimal$	



Remark 1. O(m) randomized MST algorithm. [Karger-Klein-Tarjan 1995] Remark 2. O(m) MST verification algorithm. [Dixon-Rauch-Tarjan 1992]

Algorithm Design

Jon Kleinberg - Éva tardos

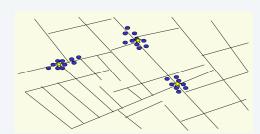
SECTION 4.7

4. GREEDY ALGORITHMS II

- Dijkstra's algorithm
- ▶ minimum spanning trees
- Prim, Kruskal, Boruvka
- ▶ single-link clustering
- min-cost arborescences

Clustering

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



outbreak of cholera deaths in London in 1850s (Nina Mishra)

Applications.

- · Routing in mobile ad hoc networks.
- · Document categorization for web search.
- · Similarity searching in medical image databases
- Skycat: cluster 109 sky objects into stars, quasars, galaxies.

• ...

Clustering of maximum spacing

k-clustering. Divide objects into k non-empty groups.

Distance function. Numeric value specifying "closeness" of two objects.

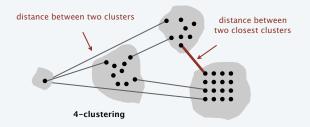
• $d(p_i, p_j) = 0$ iff $p_i = p_j$ [identity of indiscernibles]

• $d(p_i, p_j) \ge 0$ [nonnegativity]

• $d(p_i, p_j) = d(p_j, p_i)$ [symmetry]

Spacing. Min distance between any pair of points in different clusters.

Goal. Given an integer k, find a k-clustering of maximum spacing.



Greedy clustering algorithm

"Well-known" algorithm in science literature for single-linkage k-clustering:

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



Key observation. This procedure is precisely Kruskal's algorithm (except we stop when there are k connected components).

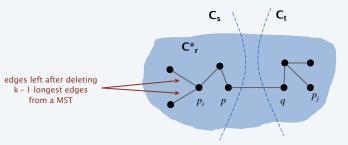
Alternative. Find an MST and delete the k-1 longest edges.

Greedy clustering algorithm: analysis

Theorem. Let C^* denote the clustering $C^*_1, ..., C^*_k$ formed by deleting the k-1 longest edges of an MST. Then, C^* is a k-clustering of max spacing.

Pf. Let C denote some other clustering $C_1, ..., C_k$.

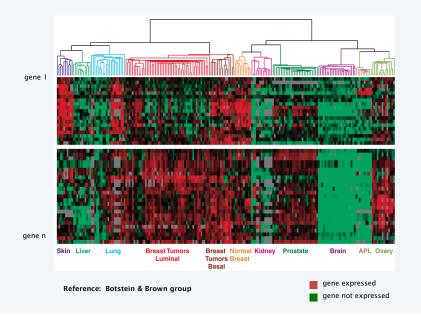
- The spacing of C^* is the length d^* of the $(k-1)^{st}$ longest edge in MST.
- Let p_i and p_j be in the same cluster in C^* , say C^*_r , but different clusters in C, say C_s and C_r .
- Some edge (p,q) on $p_i p_i$ path in C^*_r spans two different clusters in C.
- Edge (p, q) has length $\leq d^*$ since it wasn't deleted.
- Spacing of C is $\leq d^*$ since p and q are in different clusters. •

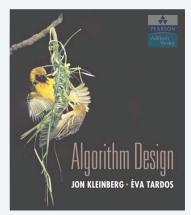


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Dendrogram of cancers in human

Tumors in similar tissues cluster together.





SECTION 4.9

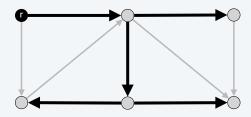
4. GREEDY ALGORITHMS II

- Dijkstra's algorithm
- minimum spanning trees
- Prim, Kruskal, Boruvka
- ▶ single-link clustering
- min-cost arborescences

Arborescences

Def. Given a digraph G = (V, E) and a root $r \in V$, an arborescence (rooted at r) is a subgraph T = (V, F) such that

- *T* is a spanning tree of *G* if we ignore the direction of edges.
- There is a directed path in *T* from *r* to each other node $v \in V$.



Warmup. Given a digraph G, find an arborescence rooted at r (if one exists). Algorithm. BFS or DFS from r is an arborescence (iff all nodes reachable).

Arborescences

Def. Given a digraph G = (V, E) and a root $r \in V$, an arborescence (rooted at r) is a subgraph T = (V, F) such that

- *T* is a spanning tree of *G* if we ignore the direction of edges.
- There is a directed path in *T* from *r* to each other node $v \in V$.

Proposition. A subgraph T = (V, F) of G is an arborescence rooted at r iff T has no directed cycles and each node $v \ne r$ has exactly one entering edge. Pf.

- ⇒ If T is an arborescence, then no (directed) cycles and every node $v \neq r$ has exactly one entering edge—the last edge on the unique $r \rightarrow v$ path.
- \Leftarrow Suppose *T* has no cycles and each node $v \neq r$ has one entering edge.
 - To construct an $r \rightarrow v$ path, start at v and repeatedly follow edges in the backward direction.
- Since *T* has no directed cycles, the process must terminate.
- It must terminate at *r* since *r* is the only node with no entering edge. •

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Min-cost arborescence problem

Problem. Given a digraph G with a root node r and with a nonnegative cost $c_e \ge 0$ on each edge e, compute an arborescence rooted at r of minimum cost.

5 1 3 8 9

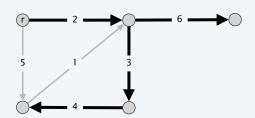
Assumption 1. G has an arborescence rooted at r.

Assumption 2. No edge enters r (safe to delete since they won't help).

Simple greedy approaches do not work

Observations. A min-cost arborescence need not:

- · Be a shortest-paths tree.
- Include the cheapest edge (in some cut).
- Exclude the most expensive edge (in some cycle).



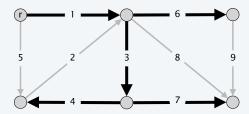
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A sufficient optimality condition

Property. For each node $v \neq r$, choose one cheapest edge entering v and let F^* denote this set of n-1 edges. If (V,F^*) is an arborescence, then it is a min-cost arborescence.

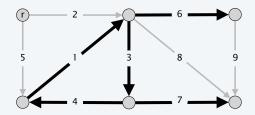
Pf. An arborescence needs exactly one edge entering each node $v \neq r$ and (V, F^*) is the cheapest way to make these choices. \blacksquare



A sufficient optimality condition

Property. For each node $v \neq r$, choose one cheapest edge entering v and let F^* denote this set of n-1 edges. If (V,F^*) is an arborescence, then it is a min-cost arborescence.

Note. F^* may not be an arborescence (since it may have directed cycles).



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Reduced costs

Def. For each $v \ne r$, let y(v) denote the min cost of any edge entering v. The reduced cost of an edge (u, v) is $c'(u, v) = c(u, v) - y(v) \ge 0$.

Observation. T is a min-cost arborescence in G using costs c iff T is a min-cost arborescence in G using reduced costs c'. Pf. Each arborescence has exactly one edge entering v.

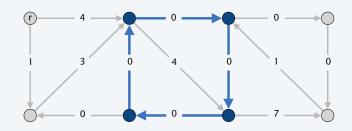
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reduced costs c'

Edmonds branching algorithm: intuition

Intuition. Recall F^* = set of cheapest edges entering v for each $v \neq r$.

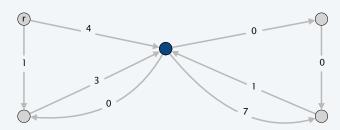
- Now, all edges in F^* have 0 cost with respect to costs c'(u, v).
- If F^* does not contain a cycle, then it is a min-cost arborescence.
- If F^* contains a cycle C, can afford to use as many edges in C as desired.
- Contract nodes in C to a supernode.
- Recursively solve problem in contracted network G' with costs c'(u, v).



Edmonds branching algorithm: intuition

Intuition. Recall F^* = set of cheapest edges entering v for each $v \neq r$.

- Now, all edges in F^* have 0 cost with respect to costs c'(u, v).
- If F^* does not contain a cycle, then it is a min-cost arborescence.
- If F^* contains a cycle C, can afford to use as many edges in C as desired.
- Contract nodes in C to a supernode (removing any self-loops).
- Recursively solve problem in contracted network G' with costs c'(u, v).



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Edmonds branching algorithm

EDMONDSBRANCHING(G, r, c)

FOREACH $v \neq r$

 $y(v) \leftarrow \min \text{ cost of an edge entering } v.$

 $c'(u, v) \leftarrow c'(u, v) - y(v)$ for each edge (u, v) entering v.

FOREACH $v \neq r$: choose one 0-cost edge entering v and let F^* be the resulting set of edges.

IF F^* forms an arborescence, RETURN $T = (V, F^*)$.

ELSE

 $C \leftarrow$ directed cycle in F^* .

Contract C to a single supernode, yielding G' = (V', E').

 $T' \leftarrow \text{EDMONDSBRANCHING}(G', r, c')$

Extend T' to an arborescence T in G by adding all but one edge of C.

RETURN T.

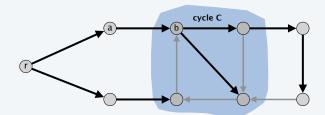
Edmonds branching algorithm

Q. What could go wrong?

Α

- Min-cost arborescence in G' has exactly one edge entering a node in C (since C is contracted to a single node)
- But min-cost arborescence in G might have more edges entering C.

min-cost arborescence in G



Edmonds branching algorithm: key lemma

Lemma. Let C be a cycle in G consisting of 0-cost edges. There exists a mincost arborescence rooted at r that has exactly one edge entering C.

Pf. Let T be a min-cost arborescence rooted at r.

Case 0. *T* has no edges entering *C*.

Since *T* is an arborescence, there is an $r \rightarrow v$ path fore each node $v \Rightarrow$ at least one edge enters *C*.

Case 1. T has exactly one edge entering C.

T satisfies the lemma.

Case 2. T has more than one edge that enters C.

We construct another min-cost arborescence $T^{\scriptscriptstyle \dagger}$ that has exactly one edge entering C.

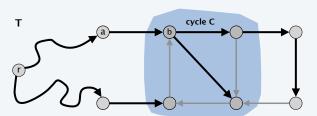
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Edmonds branching algorithm: key lemma

Case 2 construction of T'.

- Let (a, b) be an edge in T entering C that lies on a shortest path from r.
- We delete all edges of *T* that enter a node in *C* except (*a*, *b*).
- We add in all edges of *C* except the one that enters *b*.

path from r to C uses only one node in C



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Edmonds branching algorithm: key lemma

Case 2 construction of T'.

- Let (a, b) be an edge in T entering C that lies on a shortest path from r.
- We delete all edges of T that enter a node in C except (a, b).
- We add in all edges of C except the one that enters b.

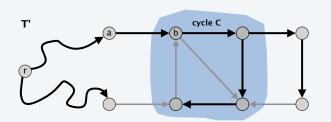
path from r to C uses only one node in C

Claim. T' is a min-cost arborescence.

- The cost of T is at most that of T since we add only 0-cost edges.
- T' has exactly one edge entering each node $v \neq r$.

• T is an arborescence rooted at r

(T had no cycles before; no cycles within C; now only (a, b) enters C)



and the only path in T' to a is the path from r to a (since any path must follow unique entering edge back to r)

Edmonds branching algorithm: analysis

Theorem. [Chu-Liu 1965, Edmonds 1967] The greedy algorithm finds a min-cost arborescence.

Pf. [by induction on number of nodes in *G*]

- If the edges of F^* form an arborescence, then min-cost arborescence.
- Otherwise, we use reduced costs, which is equivalent.
- After contracting a 0-cost cycle C to obtain a smaller graph G', the algorithm finds a min-cost arborescence T' in G' (by induction).
- Key lemma: there exists a min-cost arborescence *T* in *G* that corresponds to *T*'. ■

Theorem. The greedy algorithm can be implemented in O(mn) time. Pf.

- At most *n* contractions (since each reduces the number of nodes).
- Finding and contracting the cycle C takes O(m) time.
- Transforming T' into T takes O(m) time. •

Min-cost arborescence

Theorem. [Gabow-Galil-Spencer-Tarjan 1985] There exists an $O(m + n \log n)$ time algorithm to compute a min-cost arborescence.

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EFFICIENT ALGORITHMS FOR FINDING MINIMUM SPANNING TREES IN UNDIRECTED AND DIRECTED GRAPHS

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Received 23 January 1985 Revised 1 December 1985

Recently, Fredman and Tarjan invented a new, especially efficient form of heap (priority queue). Their data structure, the Fibonacci heap (or F-heap) supports arbitrary deletion in $O(\log n)$ amortized time and other heap operations in O(1) amortized time. In this paper we use F-heaps to obtain fast algorithms for finding minimum spanning trees in undirected and directed graphs. For an undirected graph containing n vertices and m edges, our minimum spanning tree algorithm runs in $O(m \log p(m, n))$ time, improved from $O(m \log n, m)$ time, where $\beta(m, n) = \min_{n} \{r \log n, m \geq n, m\}$. Our minimum spanning tree algorithm for directed graphs runs in $O(n \log n + m)$ line, improved from $O(n \log n + m)$ log $\log \log_{m(m+1)} n$). Both algorithms can be extended to allow a degree constraint at one vertex.