

4. GREEDY ALGORITHMS

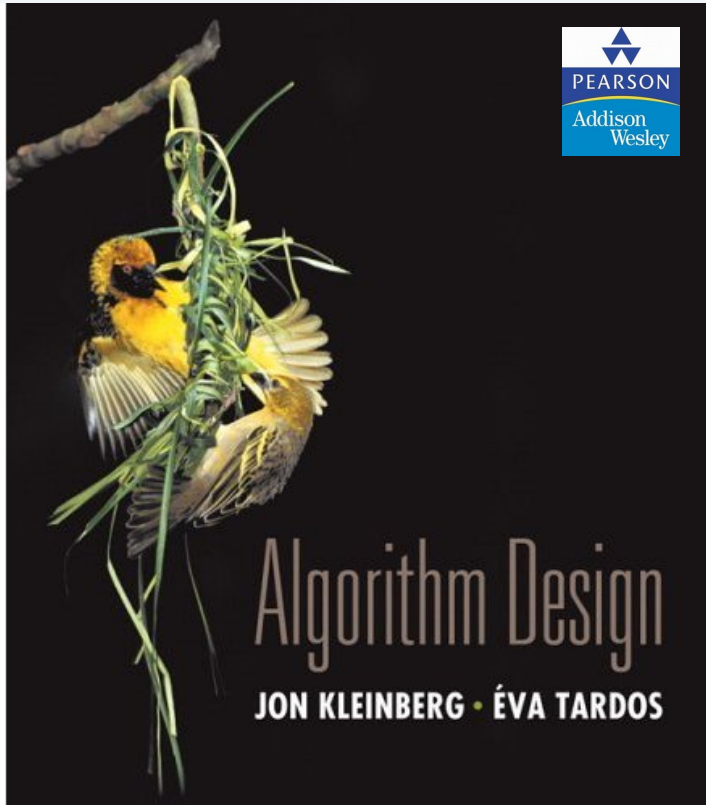
- ▶ *Dijkstra's algorithm*
- ▶ *minimum spanning trees*
- ▶ *Prim, Kruskal, Boruvka*
- ▶ *single-link clustering*
- ▶ *min-cost arborescences*

Lecture slides by Kevin Wayne

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



SECTION 4.4

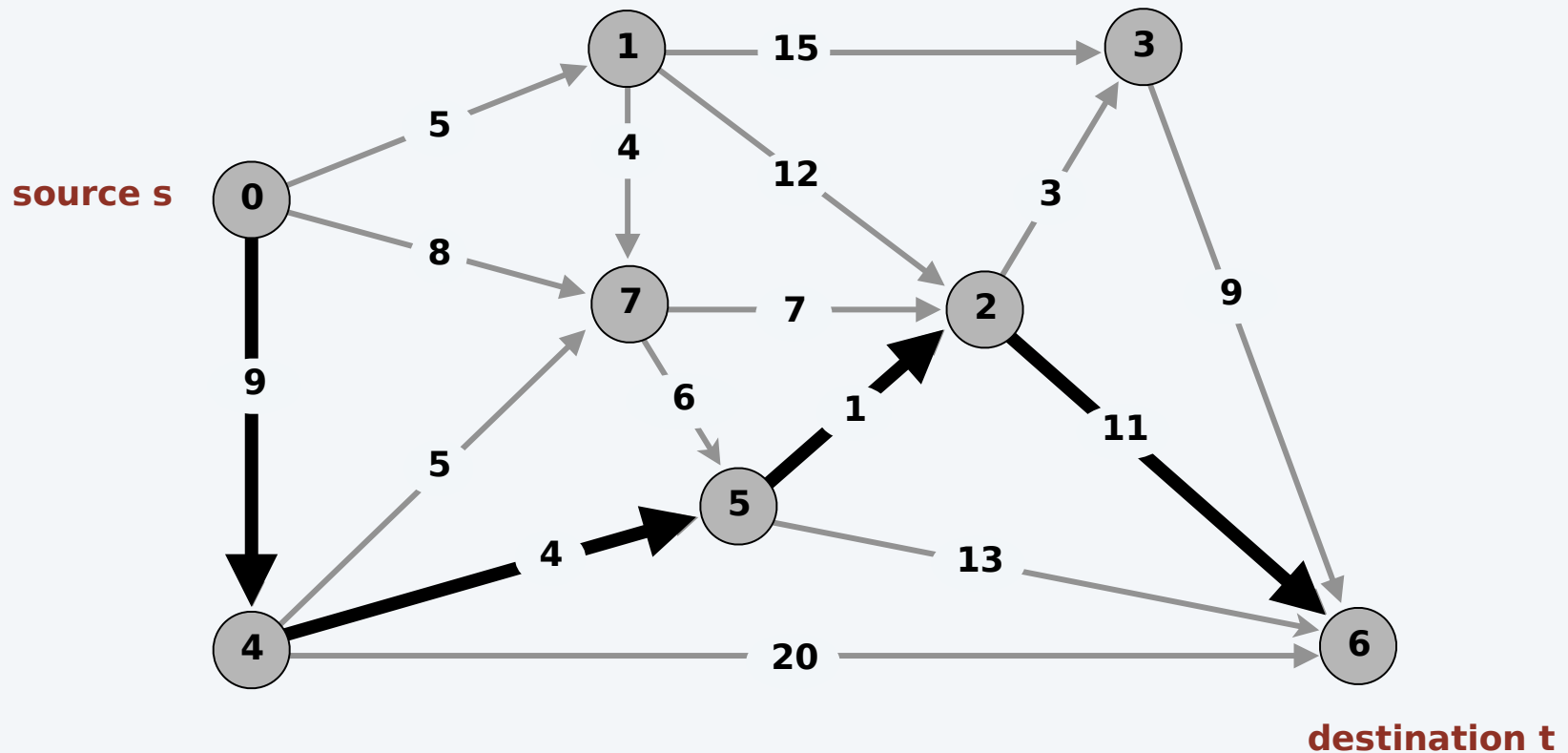
4. GREEDY ALGORITHMS

- ▶ ***Dijkstra's algorithm***
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3/STR/MOT/XPL/1/A

Shortest-paths problem

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



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



5/STR/APP/DEF/1/A

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.

6/STR/ALG/XPL/1/A

Dijkstra's algorithm

Greedy approach. Maintain a set of explored nodes which algorithm has determined the shortest path distance from s to u .

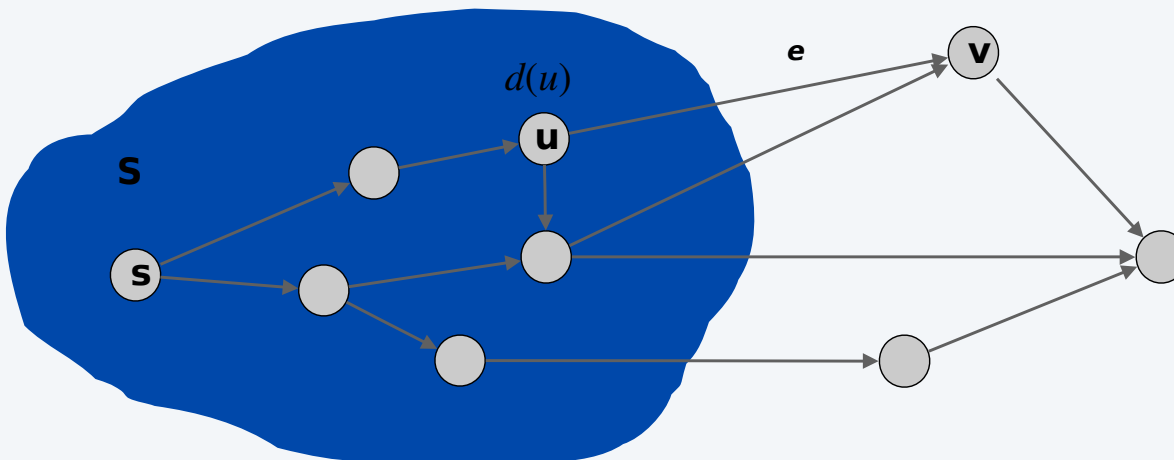


Initialize $S = \{ s \}$, $d(s) = 0$.

Repeatedly choose unexplored node 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)



7/EXP/APP/XPL/1/A

Dijkstra's algorithm

Greedy approach. Maintain a set of explored nodes S which algorithm has determined the shortest path distance from s to u .



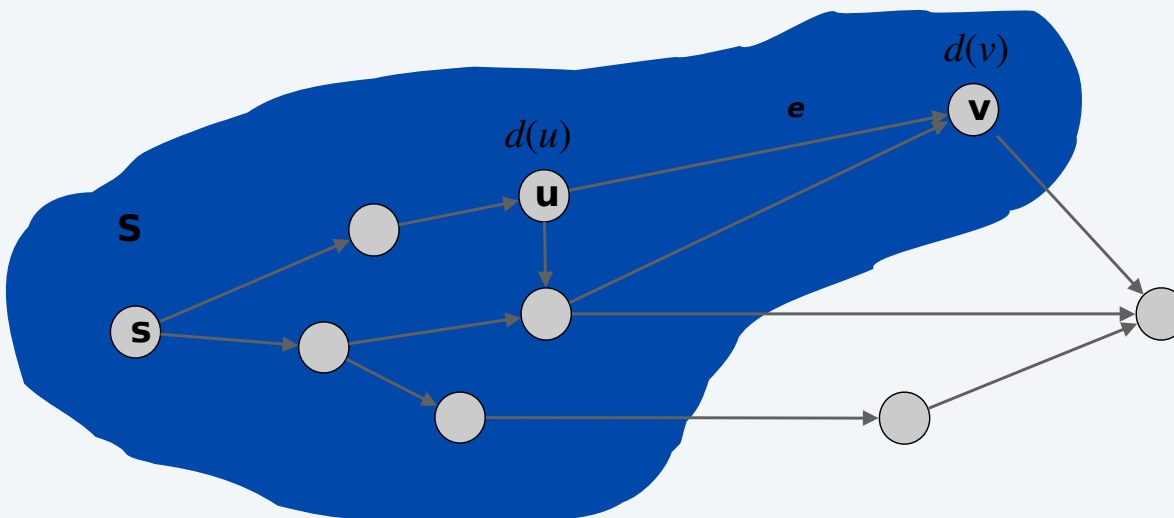
Initialize $S = \{ s \}$, $d(s) = 0$.

Repeatedly choose unexplored node 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: proof of correctness

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

Pf. [by induction on $|S|$]

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

Inductive hypothesis: Assume true for $|S| \geq 1$

Let v be next node added to S , and let (u, v) be the final edge.

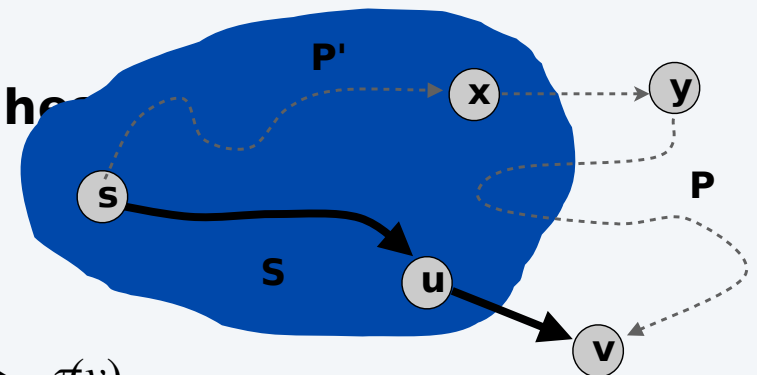
The shortest $s \rightsquigarrow u$ path plus (u, v) is an $s \rightsquigarrow v$ path of length $d(v)$.

Consider any $s \rightsquigarrow v$ path P . We show that it is no shorter than

Let (x, y) be the first edge in P that leaves S

and let P' be the subpath to

P is already too long as soon as it reaches



$$(P) \geq (P') + (x, y) \geq d(x) + (x, y) \geq \pi(y) \geq \pi(v) \quad \blacksquare$$

↑
nonnegative
lengths

↑
inductive
hypothesis

↑ definition of $\pi(y)$
↑ Dijkstra chose v
instead of y

Dijkstra's algorithm: efficient implementation

Critical optimization 1. For each unexplored node v , explicitly maintain $\pi(v)$ 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 $d(u)$ only increases). More specifically, suppose u is added to S and there is an edge $e=(u,v) \in E$ 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 $\pi(v)$.

Implementation.**Algorithm stores $d(v)$ for each explored node****Priority queue stores $d(v)$ for each unexplored node****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 $d(v) > d(u) + c(u, v)$ *decrease-key* of v to $d(u) + c(u, v)$ in priority queue. $d(v) \leftarrow d(u) + c(u, v)$.

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	$O(1)$	$O(n)$	$O(1)$	$O(n^2)$
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)	$O(1)$	$O(\log n)^\dagger$	$O(1)^\dagger$	$O(m + n \log n)$
Brodal queue (Brodal 1996)	$O(1)$	$O(\log n)$	$O(1)$	$O(m + n \log n)$

† amortized

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

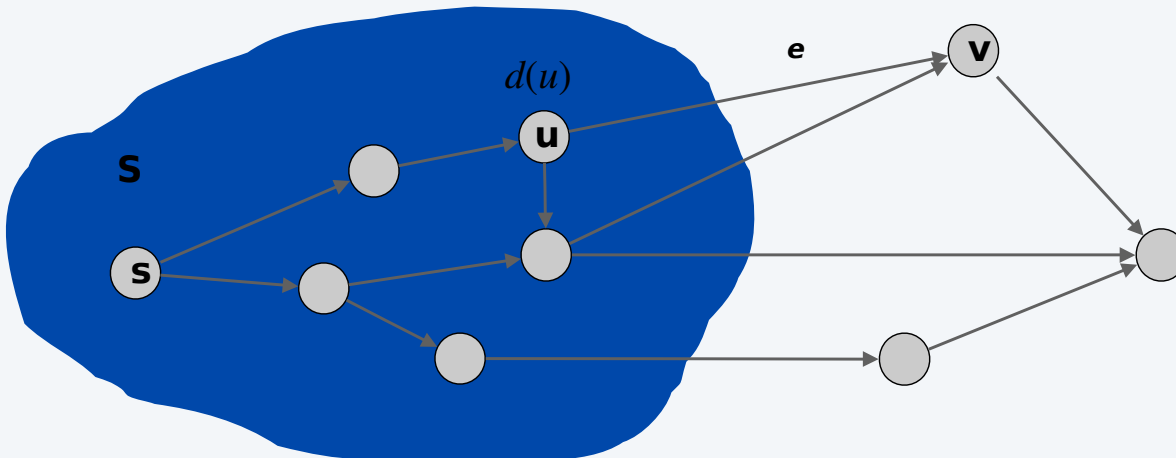
Shortest paths in undirected graphs: $d(u) + c(u, v) \leq d(v)$.

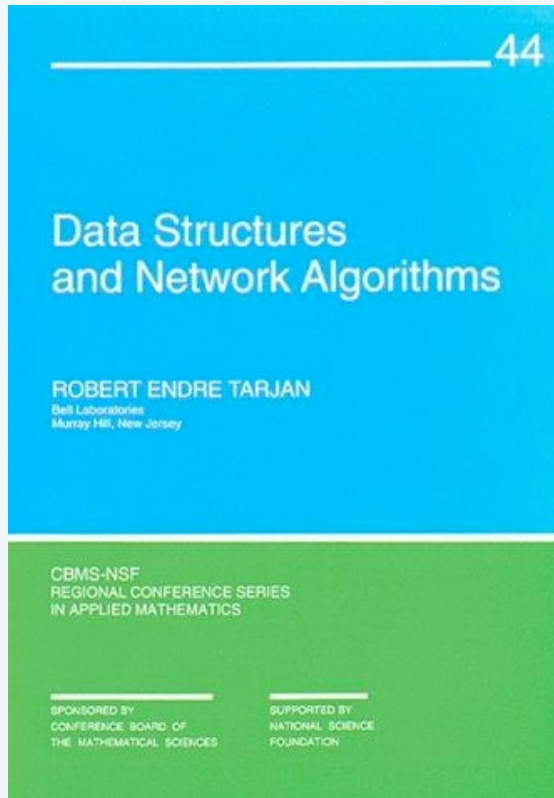
Maximum capacity paths: $\pi(u) \geq \min \{ \pi(v), c(u, v) \}$.

Maximum reliability paths: $\gamma(u) \geq \gamma(v) \square c(u, v)$.

...

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





SECTION 6.1

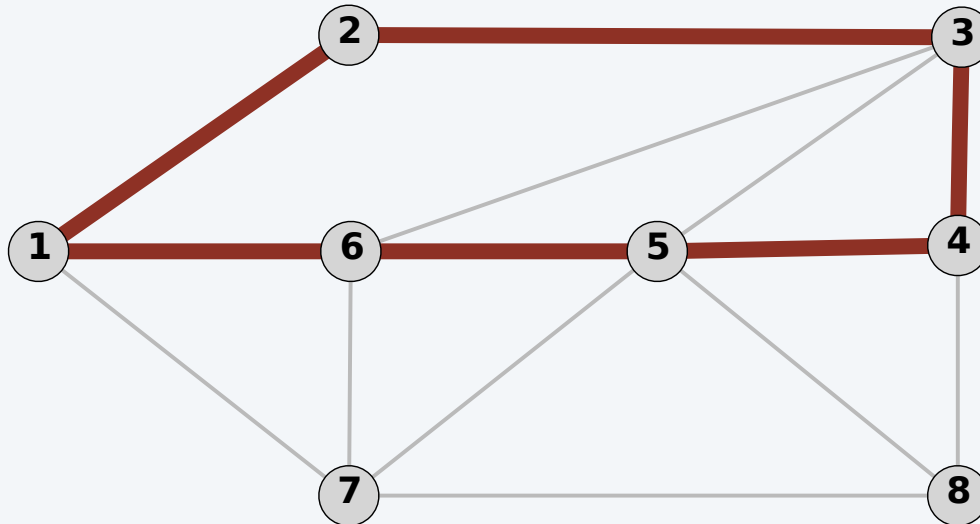
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Cycles and cuts

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.

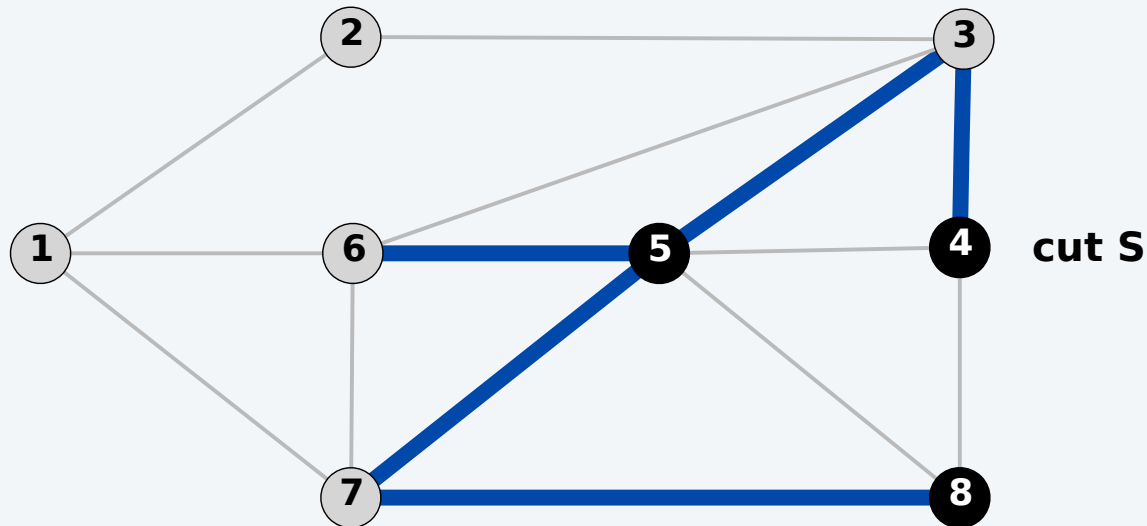


cycle $C = \{ (1, 2), (2, 3), (3, 4), (4, 5), (5, 6), (6, 1) \}$

Cycles and cuts

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

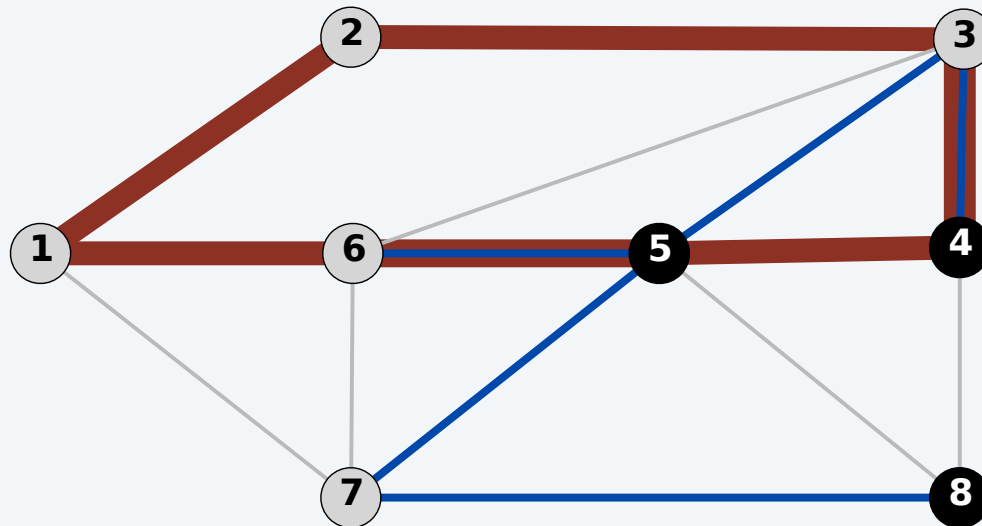
Def. The **cutset** of a cut is the set of edges with exactly one endpoint in S .



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) }

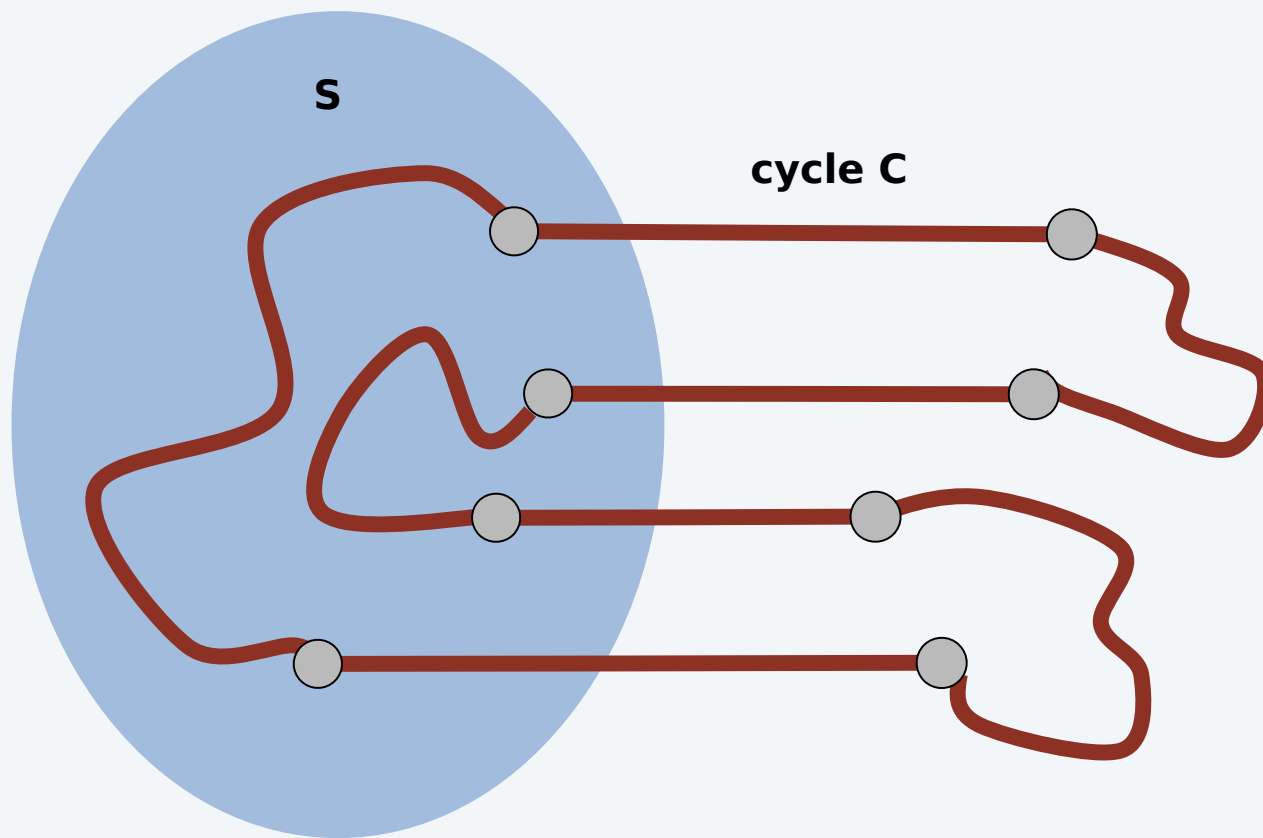
cycle C = { (1, 2), (2, 3), (3, 4), (4, 5), (5, 6), (6, 1) }

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.

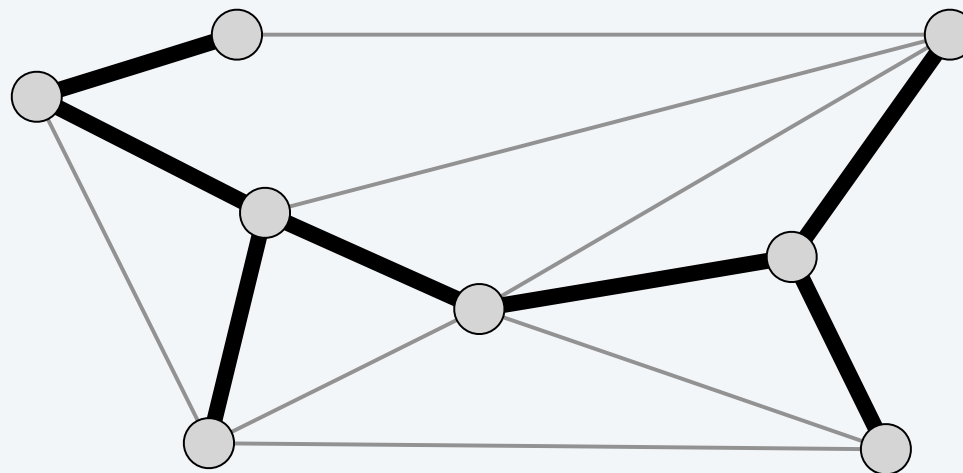
T is connected and has $|V| - 1$ edges.

T is acyclic and has $|V| - 1$ edges.

T is minimally connected: removal of any edge disconnects it.

T is maximally acyclic: addition of any edge creates a cycle.

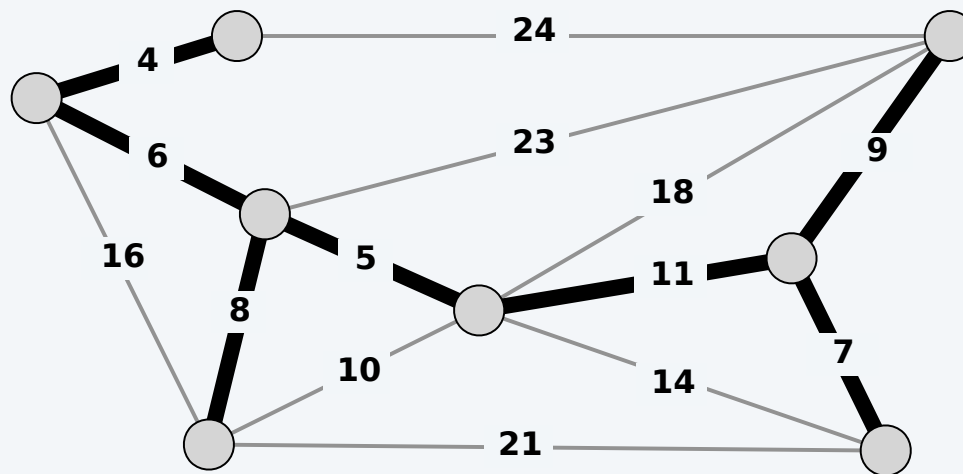
T has a unique simple path between every pair of nodes.



$T = (V, F)$

Minimum spanning tree

Given a connected graph (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.



$$\text{MST cost} = 50 = 4 + 6 + 8 + 5 + 11 + 9 + 7$$

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

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.

Approximation algorithms for NP-hard problems (e.g., TSP, Steiner tree).

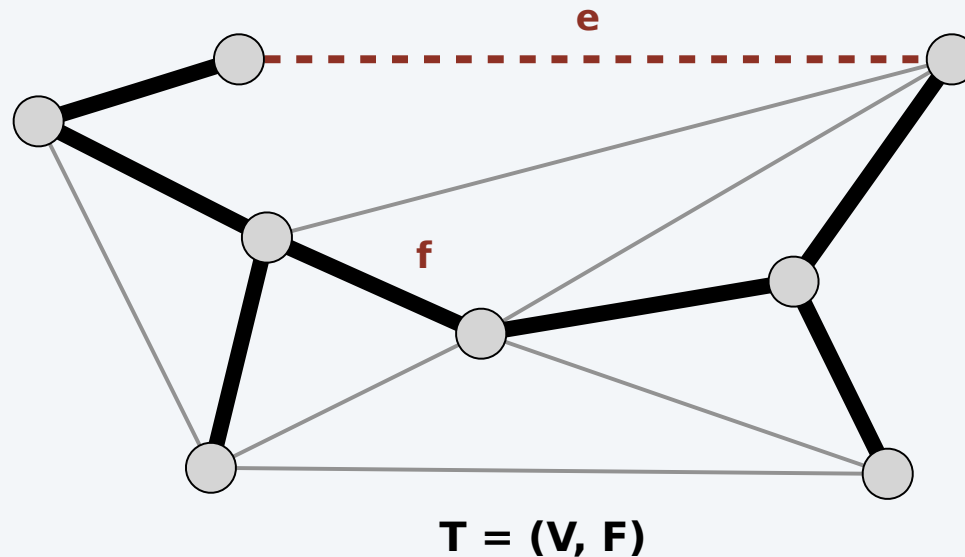
Network design (communication, electrical, hydraulic, computer, road).

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.



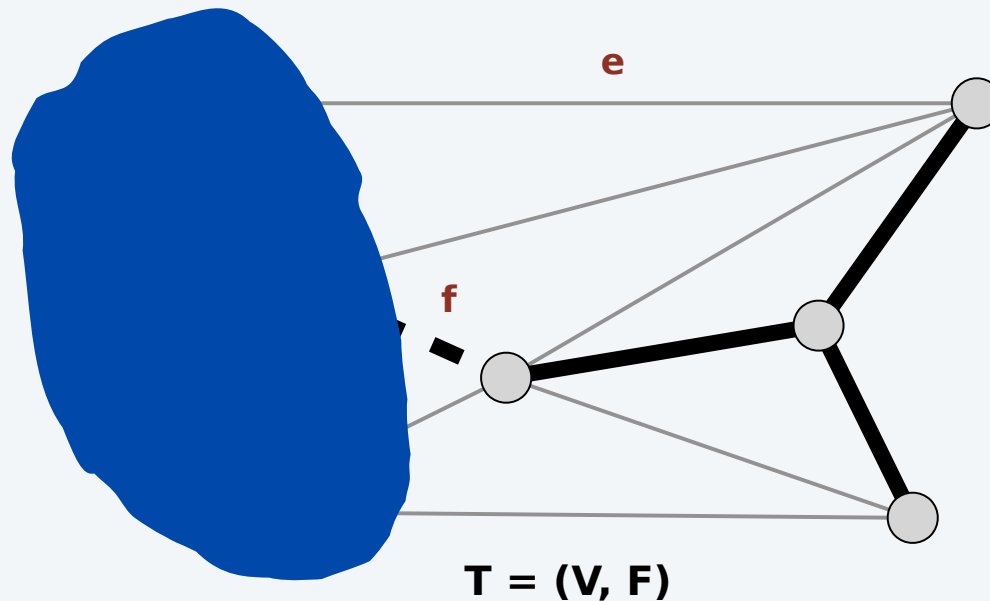
Observation. If $c_e < c_f$, then T is not an MST.

Fundamental cutset

Fundamental cutset.

Deleting any tree edge f from a spanning tree T divide nodes into two connected components. D is cutset.

Adding any edge $e \in D$ to $T - \{f\}$ results in new spanning tree.



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

The greedy algorithm

Red rule.

Let C be a cycle with no red edges.

Select an uncolored edge $e \in C$ of max weight and color it red.



Blue rule.

Let D be a cutset with no blue edges.

Select an uncolored edge $e \in D$ 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 1 edges colored blue.

Greedy algorithm: proof of correctness

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

Pf. [by induction on number of iterations]

Base case. No edges colored \Rightarrow every MST satisfies invariant.

Greedy algorithm: proof of correctness

Color invariant. There exists an MST 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 f be edge colored blue.

if $f \in T^*$, T^* still satisfies invariant.

Otherwise, consider fundamental cycle C by adding f to T^* .

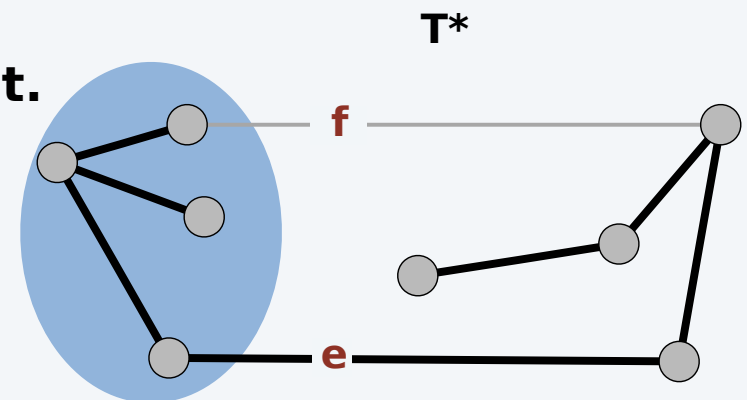
let $e \in C$ be another edge in

e is uncolored and $w_e \geq w_f$ since

$e \in T^* \Rightarrow e$ not red

blue rule $\Rightarrow e$ not blue and $w_e \geq w_f$

Thus, $T^* \cup \{f\} - \{e\}$ satisfies invariant.



Greedy algorithm: proof of correctness

Color invariant. There exists an MST 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 $e \in T^*$, T^* still satisfies invariant.

Otherwise, consider fundamental cut D by deleting from T^* .

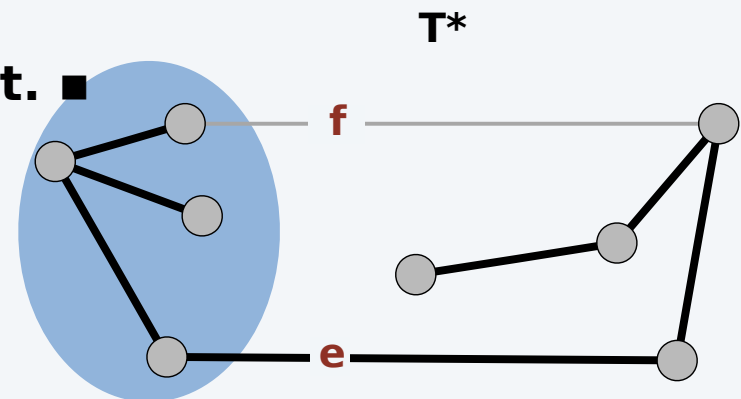
let $f \in D$ be another edge in

f is uncolored and $w_e \geq w_f$ since

$f \notin T^* \Rightarrow f$ not blue

red rule $\Rightarrow f$ not red and $w_e \geq w_f$

Thus, $T^* \cup \{f\} - \{e\}$ satisfies invariant. ■



Greedy algorithm: proof of correctness

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

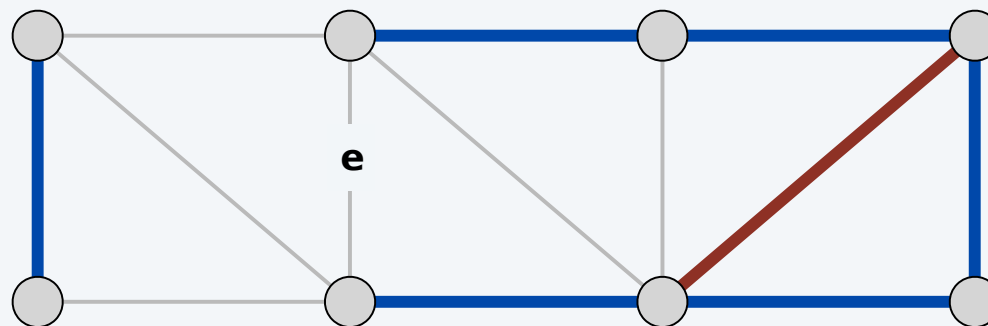
Pf. We need to show that either the red or blue rule (or both) applies.

Suppose edges left uncolored.

Blue edges form a forest.

Case 1: both endpoints are in same blue tree.

⇒ apply red rule to cycle formed by adding 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 edges left uncolored.

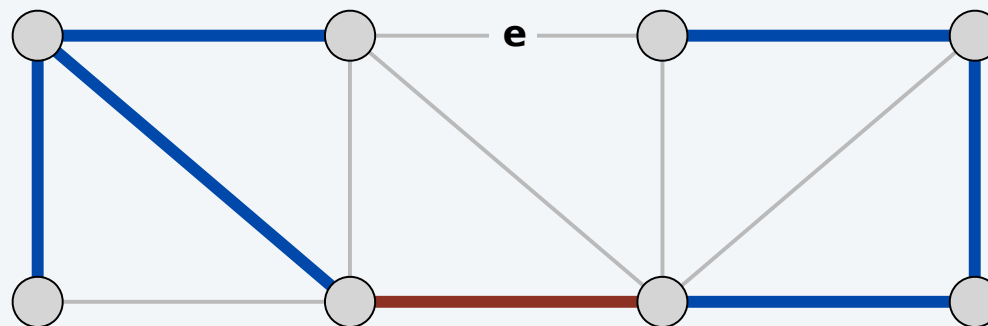
Blue edges form a forest.

Case 1: both endpoints ~~are~~ in same blue tree.

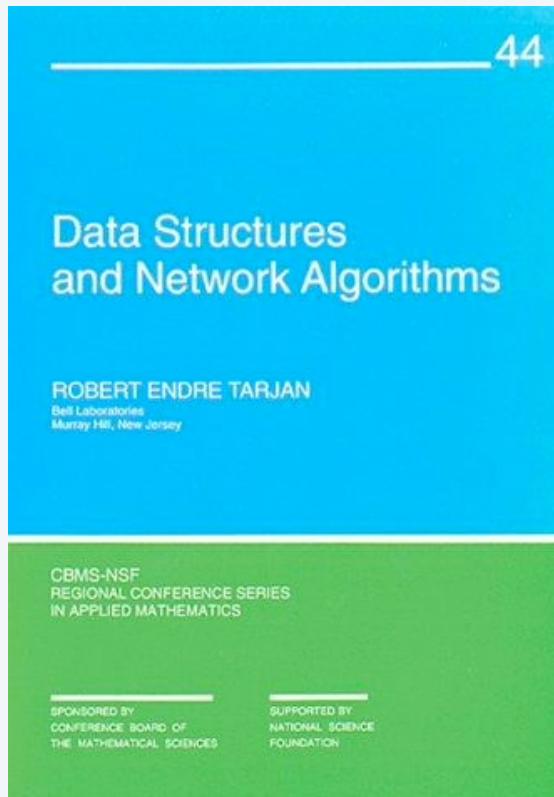
⇒ apply red rule to cycle formed by adding blue forest.

Case 2: both endpoints ~~are~~ in different blue trees.

⇒ apply blue rule to cutset induced by either of two blue trees. ■



Case 2



SECTION 6.2

4. GREEDY ALGORITHMS

- ▶ *Dijkstra's algorithm*
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Prim's algorithm

Initialize $\mathcal{S} = \text{any node.}$

Repeat $n - 1$ **times:**

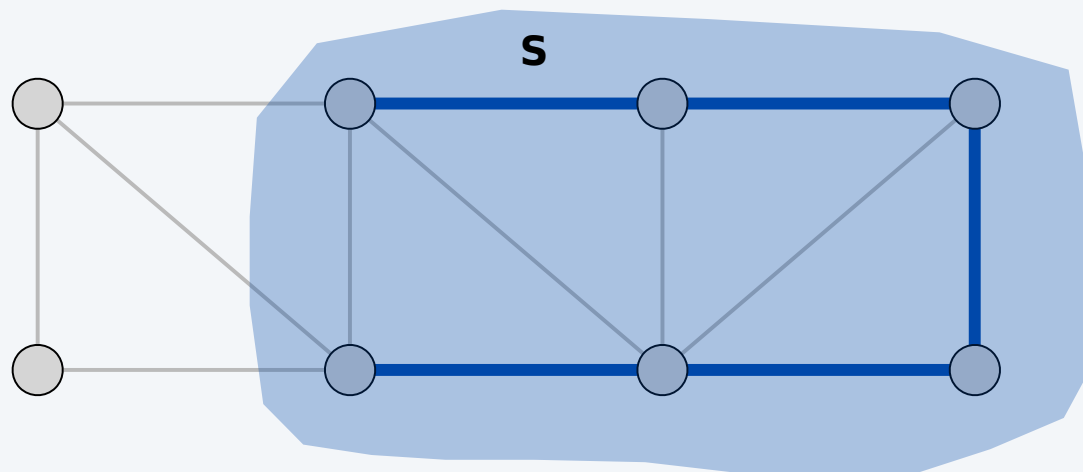
Add to tree the min weight edge with one endpoint in

Add new node to



Theorem. Prim's algorithm computes the MST.

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



Prim's algorithm: implementation

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

Pf. Implementation almost identical to Dijkstra's algorithm.

$[d(v) = \text{weight of cheapest known edge between } v \text{ and } S]$

PRIM (V, E, c)

Create an empty priority queue.

$S \leftarrow$ 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 $(u, v) \in E_{\text{incident to } u}$:

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

decrease-key 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 same blue tree.

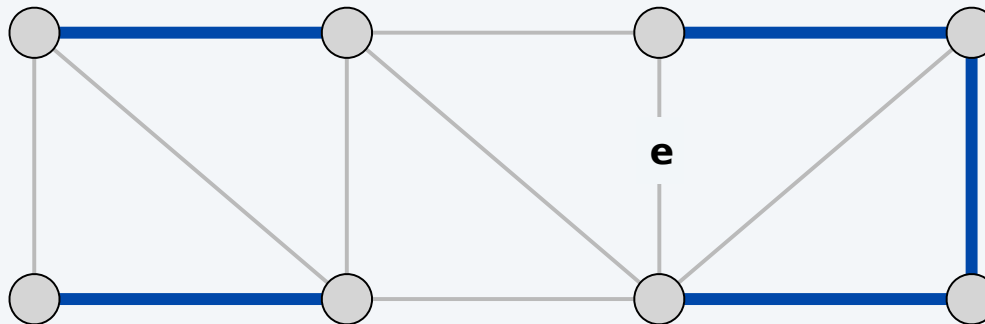
⇒ color red by applying red rule to unique cycle.

all other edges in cycle are blue

Case 2. If both endpoints 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)



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  $c(e_1) \leq c(e_2) \leq \dots \leq c(e_m)$ 
```

```
   $S \leftarrow \emptyset$ 
```

```
  FOREACH  $v \in V$  MAKESET( $v$ ).
```

```
  FOR  $i = 1$  TO  $m$ 
```

```
     $(u, v) \leftarrow e_i$ 
```

```
    IF FINDSET( $u$ )  $\neq$  FINDSET( $v$ )  $\leftarrow$  are  $u$  and  $v$  in  
                                     same component?
```

```
       $S \leftarrow S \cup \{e_i\}$ 
```

```
      UNION( $u, v$ ).  $\leftarrow$  make  $u$  and  $v$  in  
                           same component
```

```
  RETURN  $S$ 
```

Reverse-delete algorithm

Consider edges in descending order of weight:

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.

⇒ 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.

⇒ apply blue rule to cutset S 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 $e \in C$ of max weight and color it red.

Blue rule.

Let D be a cutset with no blue edges.

Select an uncolored edge $e \in D$ 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.

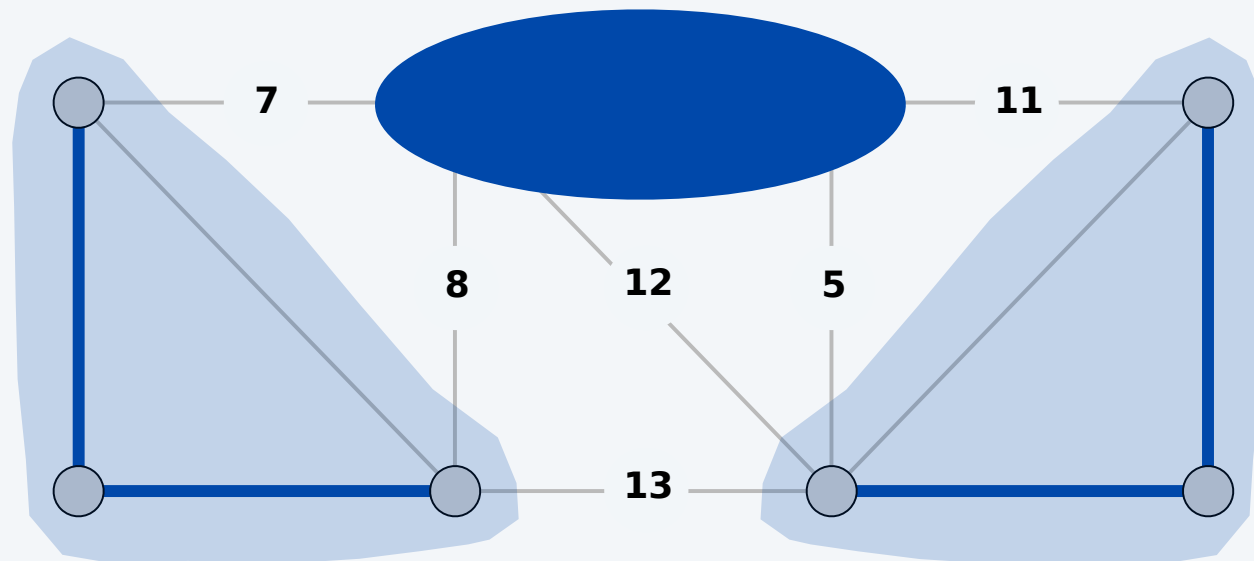
Apply blue rule to cutset corresponding to **each blue tree.**

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 $O(n \log n)$ time.
Pf.

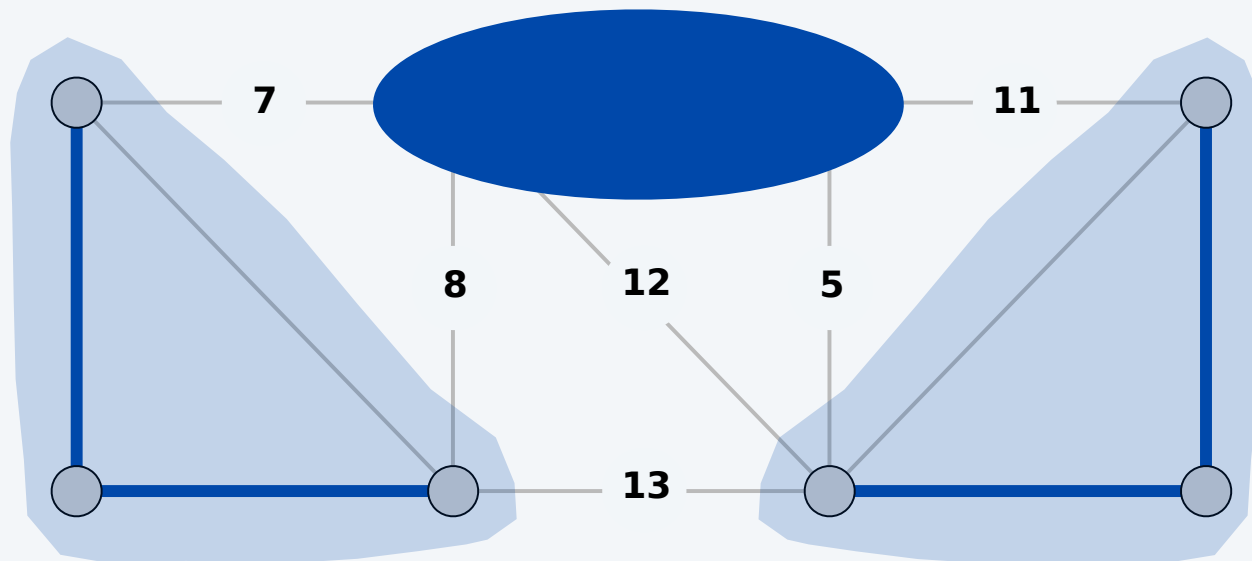
To implement a phase in $O(n)$ time:

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.

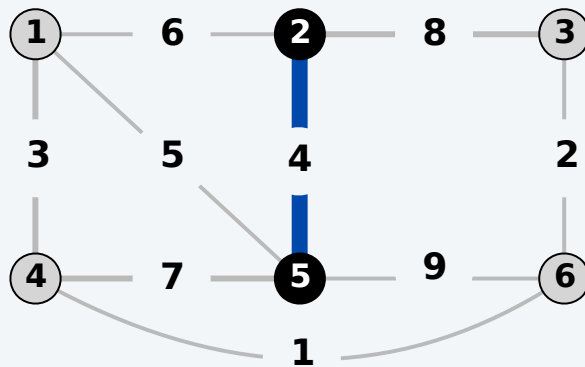


Borůvka's algorithm: implementation

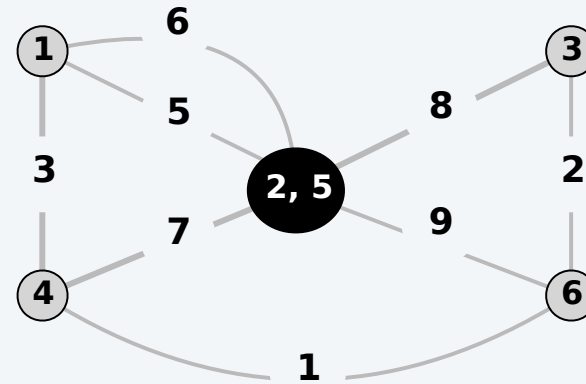
Node contraction version.

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.

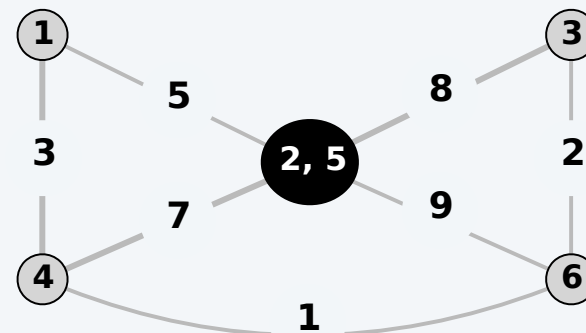
graph G



contract nodes 2 and 5



delete parallel edges and self loops



Borůvka's algorithm on planar graphs

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

To implement a Borůvka phase in time:

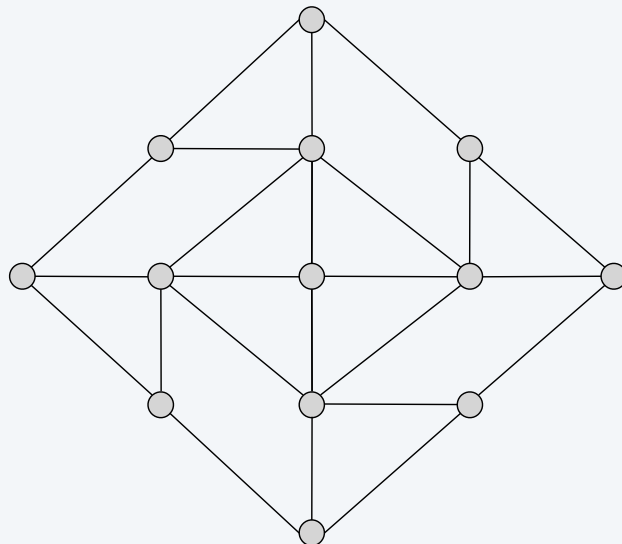
use contraction version of algorithm

in planar graphs, $\leq 3n - 6$

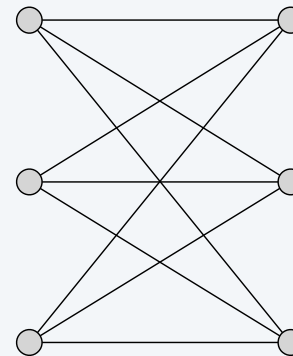
graph stays planar when we contract a blue tree

Number of nodes (at least) halves.

At most $\log_2 n$ phases: $cn + cn/2 + cn/4 + cn/8 + \dots = O(n)$. ■



planar



not planar

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 edges.

Prim's algorithm (using Fibonacci heaps) takes $O(m)$ time on a graph with $n / \log_2 n$ nodes and n edges. ■

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


Does a linear-time MST algorithm exist?

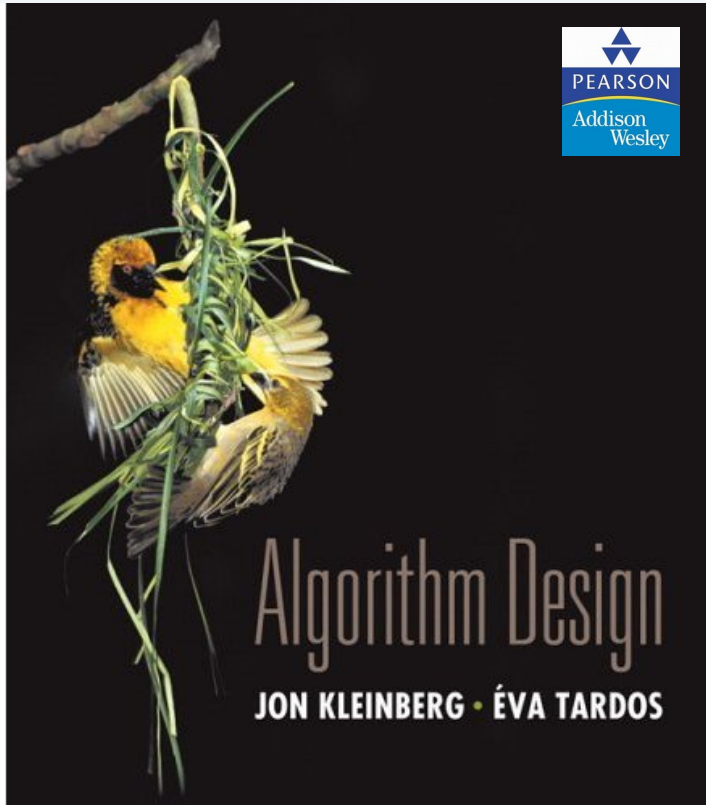
deterministic compare-based MST algorithms

year	worst case	discovered by
1975	$O(m \log \log n)$	Yao
1976	$O(m \log \log n)$	Cheriton-Tarjan
1984	$O(m \log^* n)$ $O(m + n \log n)$	Fredman-Tarjan
1986	$O(m \log (\log^* n))$	Gabow-Galil-Spencer-Tarjan
1997	$O(m \alpha(n) \log \alpha(n))$	Chazelle
2000	$O(m \alpha(n))$	Chazelle
2002	<i>optimal</i>	Pettie-Ramachandran
20xx	$O(m)$???



Remark 1. $O(m)$ randomized MST algorithm. [Karger-Klein-Tarjan 1995]

Remark 2. $O(m)$ MST verification algorithm. [Dixon-Rauch-Tarjan 1992]



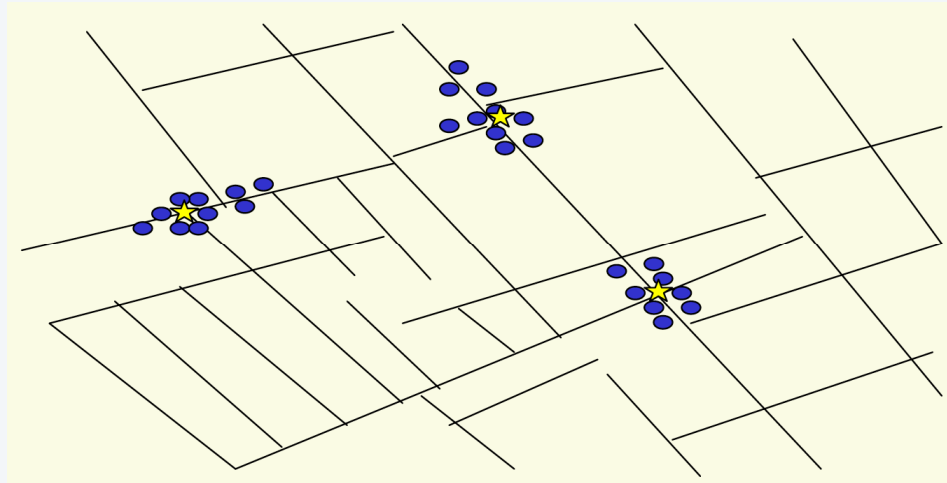
SECTION 4.7

4. GREEDY ALGORITHMS

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

Clustering

Goal. Given a set of n objects labeled p_1, \dots, 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 ¹⁰sky 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.

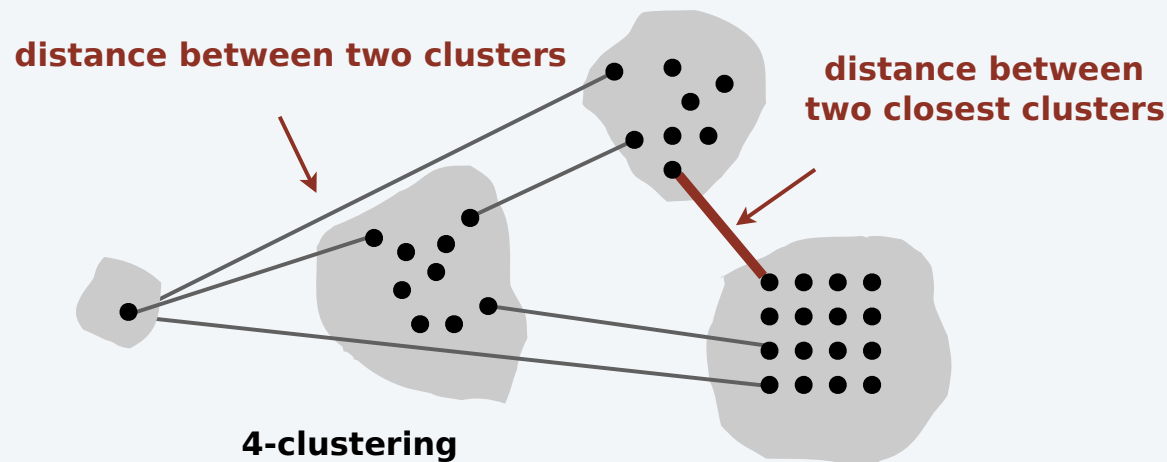
$d(p_i, p_j) = 0$ **iff** $p_i = p_j$ [identity of indiscernibles]

$d(p_i, p_j) \geq 0$ [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 algorithms

“Well-known” algorithm

Form a graph on the data points.

Find the closest pair of points not in the same cluster, and add an edge between them.

Repeat $k - 1$ times until there are k clusters.

Key observation. This algorithm is equivalent to Kruskal's algorithm for finding a minimum spanning tree (except we stop when there are k components).

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

Algorithm for single-linkage k -clustering

Find the $k - 1$ longest edges in the MST.

Remove the $k - 1$ longest edges from the MST.

Result: k clusters.

Algorithm for single-linkage k -clustering

Remove the $k - 1$ longest edges.

Greedy clustering algorithm: analysis

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.

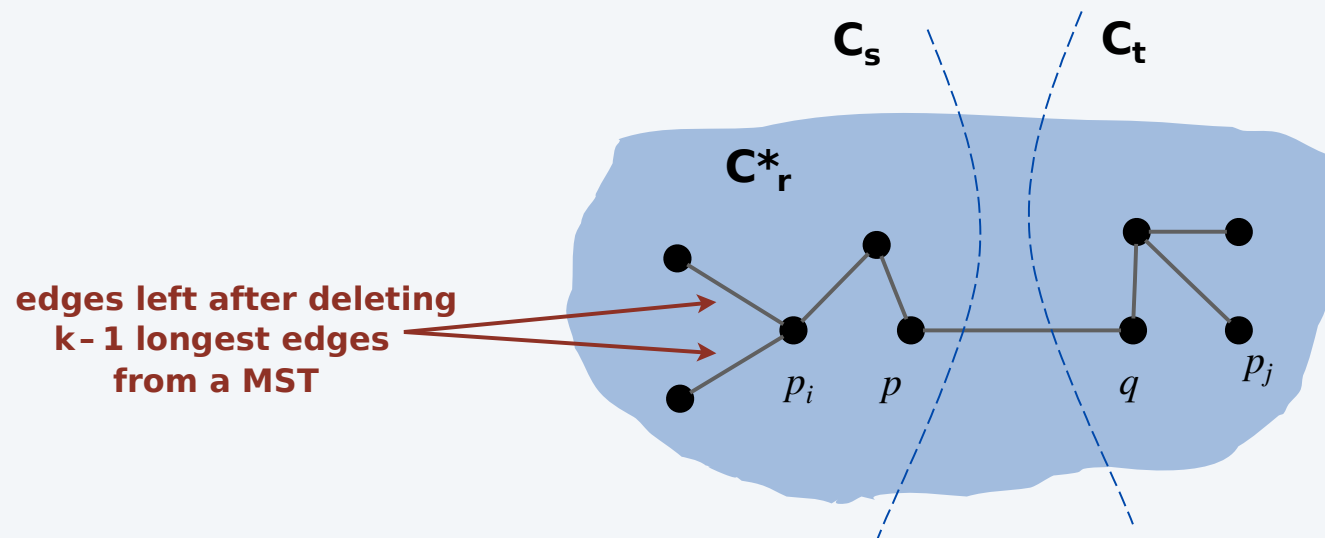
Pf. Let C denote some other clustering C_1, \dots, C_k .

The spacing of C^* is the length d^* of the $(k - 1)^{\text{st}}$ 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_t .

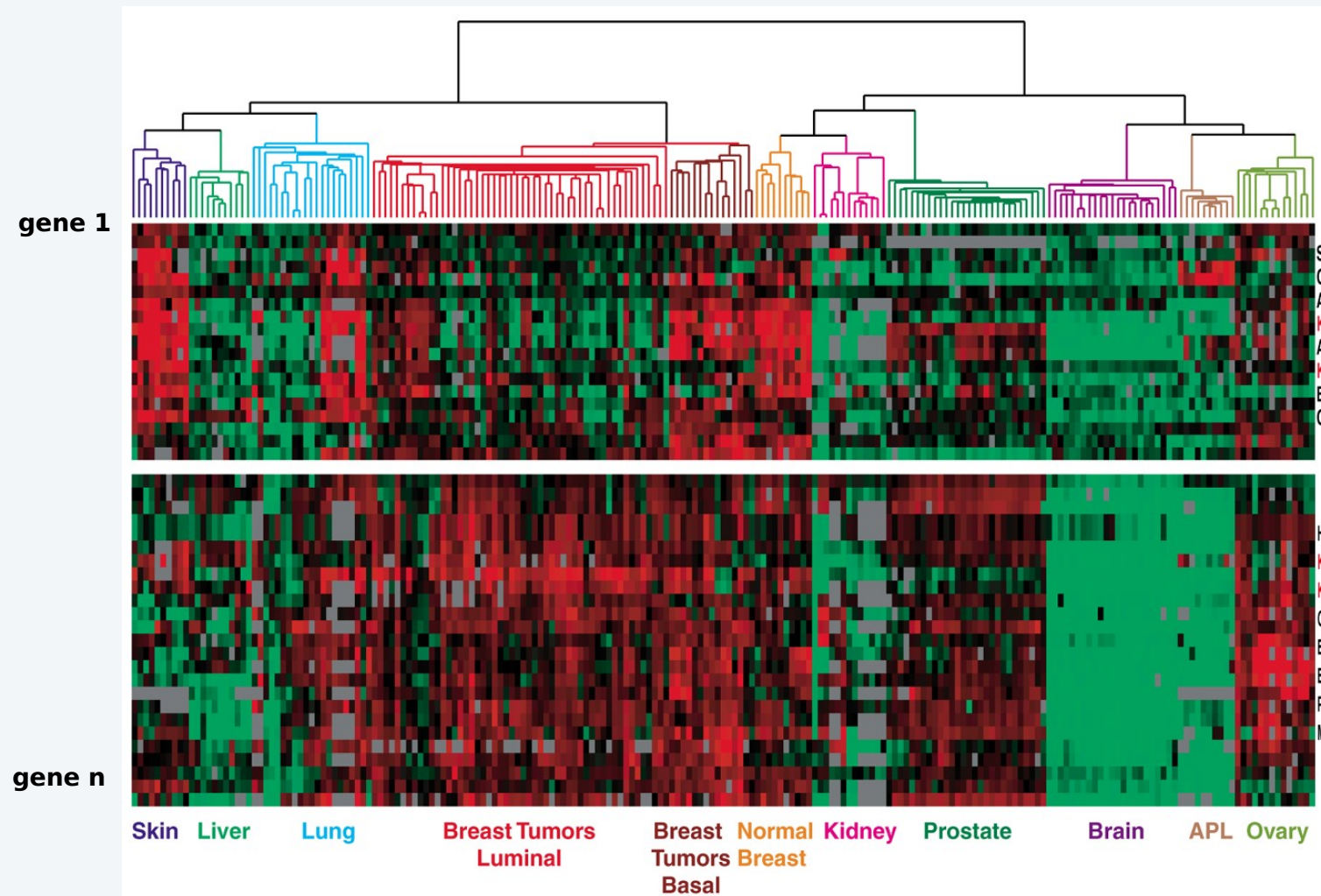
Some edge (p, q) on $p_i - p_j$ path in C^*_r spans two different clusters in C . Edge (p, q) has length $\leq d^*$ since it wasn't deleted.

Spacing of C is $\leq d^*$ since p and q are in different clusters. ■



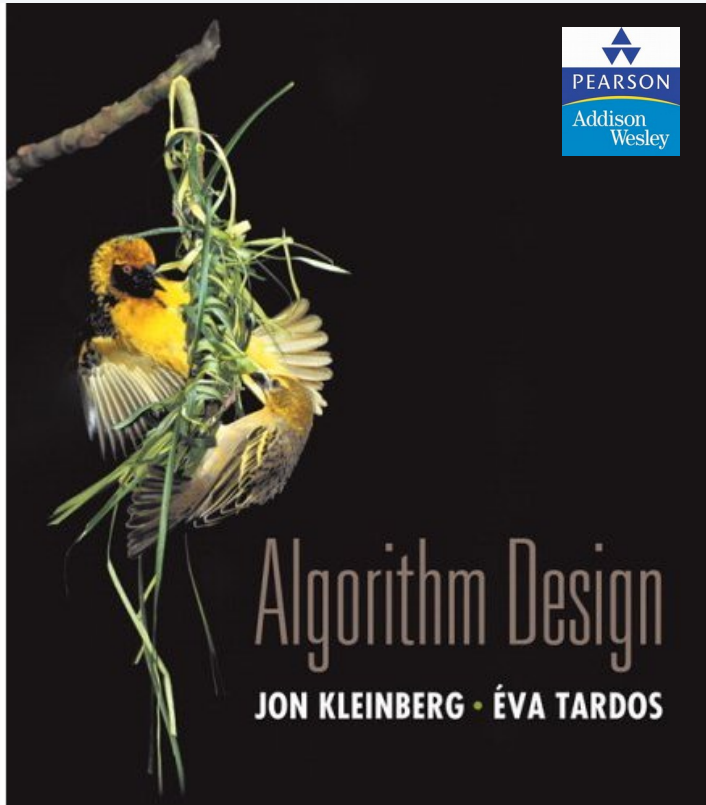
Dendrogram of cancers in human

Tumors in similar tissues cluster together.



Reference: Botstein & Brown group

■ gene expressed
■ gene not expressed

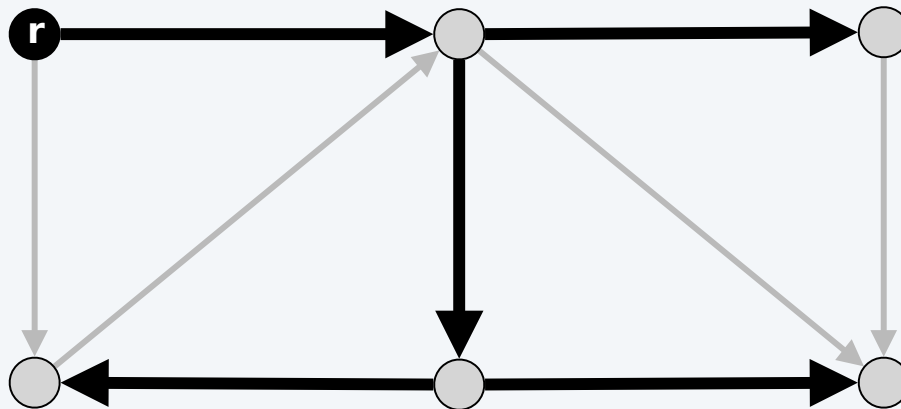


SECTION 4.9

4. GREEDY ALGORITHMS

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

T is a spanning tree. If we ignore the direction of edges, there is a directed path from r to each other node $v \in V$.



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 $H = (V, F)$ such that

T is a spanning tree if we ignore the direction of edges.

There is a directed path from r to each other node $v \in V$.

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

Pf.

\Rightarrow If T is an arborescence, then no (directed) cycles and every node has exactly one entering edge—the last edge on the unique path.

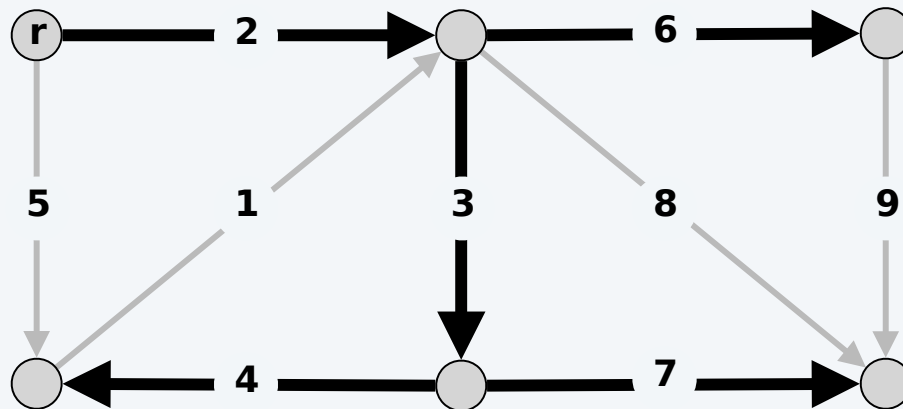
\Leftarrow Suppose T has no cycles and each node has one entering edge. To construct an r -path, start at v and repeatedly follow edges in the backward direction.

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. ■

Min-cost arborescence problem

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

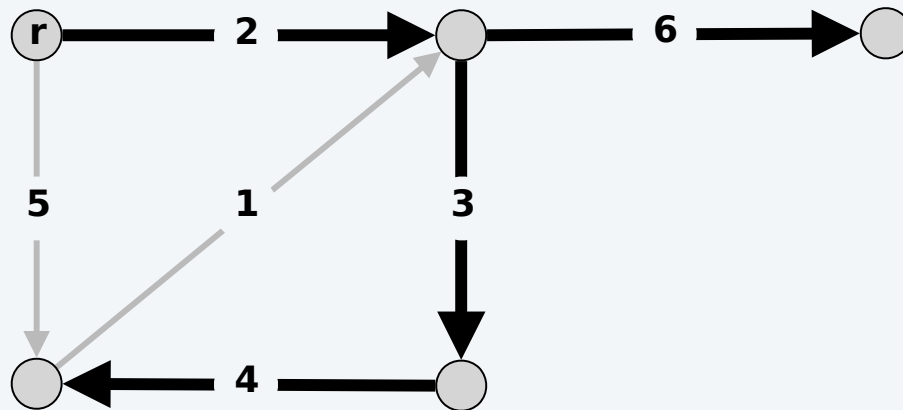


Assumption 1 G has an arborescence rooted at

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

Simple greedy approaches do not work

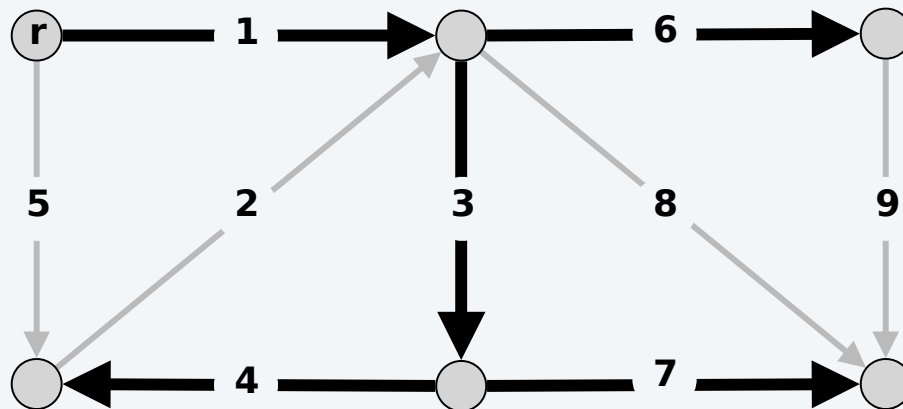
Observations. A min-cost arborescence need not:
Be a shortest-paths tree.
Include the cheapest edge (in some cut).
Exclude the most expensive edge (in some cycle).



A sufficient optimality condition

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

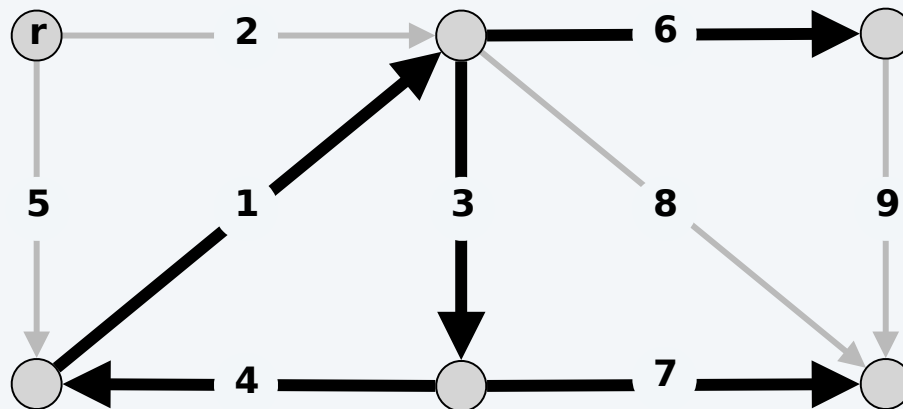
Pf. An arborescence needs exactly one edge entering each node and (V, F^*) is the cheapest way to make these choices. ■



A sufficient optimality condition

Property. For each node r , choose one cheapest edge entering r and let F^* denote this set of n 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)



Reduced costs

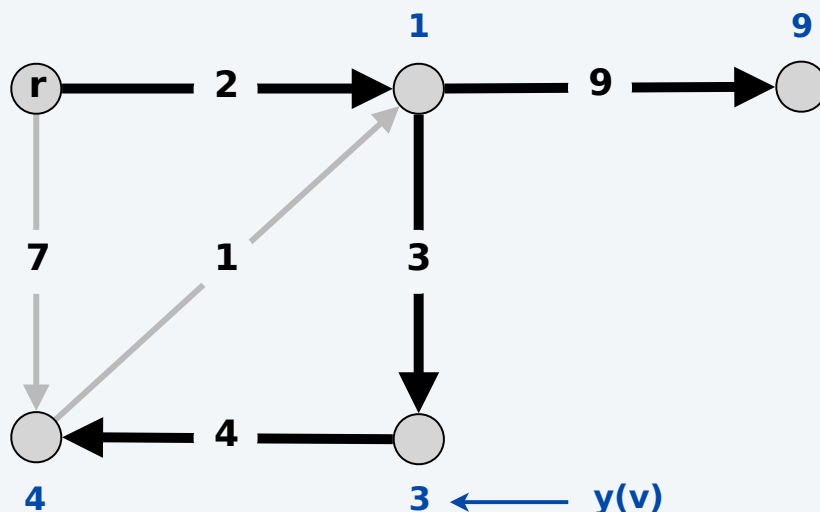
Def. For each $v \neq r$ let $y(v)$ denote the min cost of any edge entering v .
 The **reduced cost** of an edge is $c'(u, v) = c(u, v) - y(v) \geq 0$

Observation T is a min-cost arborescence using costs iff

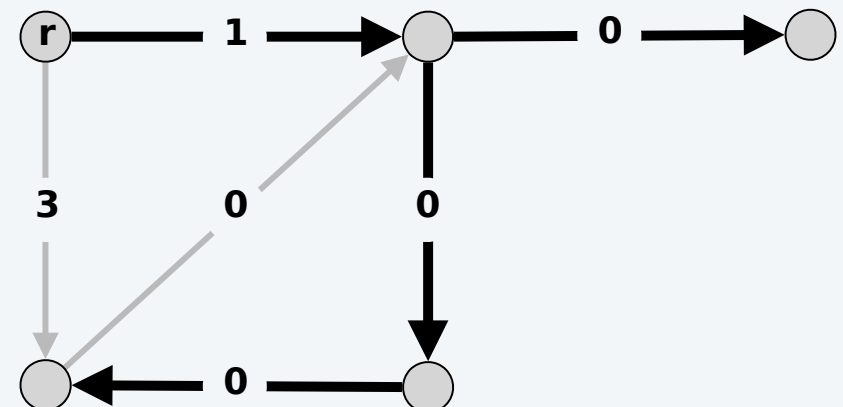
T is a min-cost arborescence using reduced costs

Pf. Each arborescence has exactly one edge entering

costs c



reduced costs c'



Edmonds branching algorithm: intuition

Intuition. Recall $F^* = \{ \text{set of cheapest edges entering } v \mid v \neq r \}$

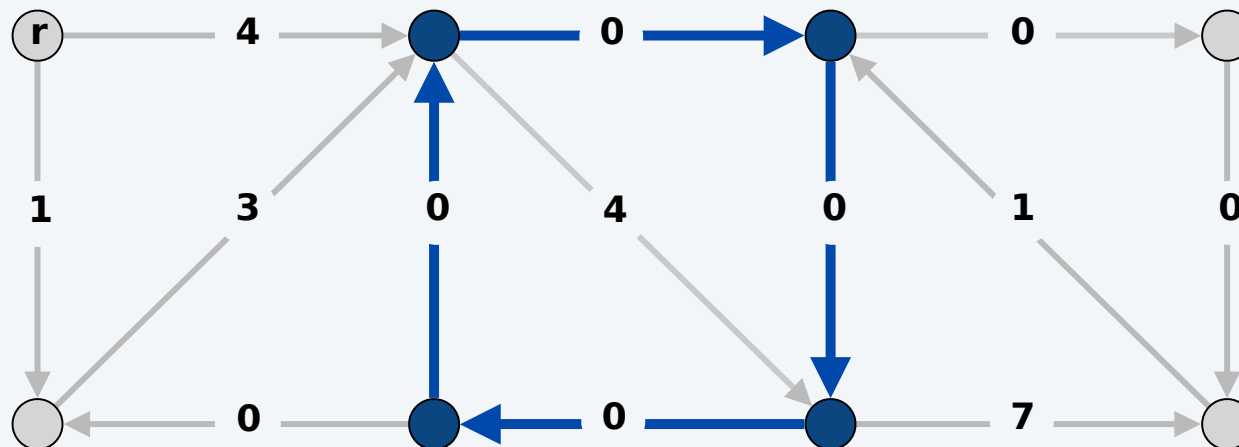
Now, all edges in F^* have 0 cost with respect to costs $c'(u, v)$.

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 into a supernode.

Recursively solve problem in contracted network with costs $c'(u, v)$.



Edmonds branching algorithm: intuition

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

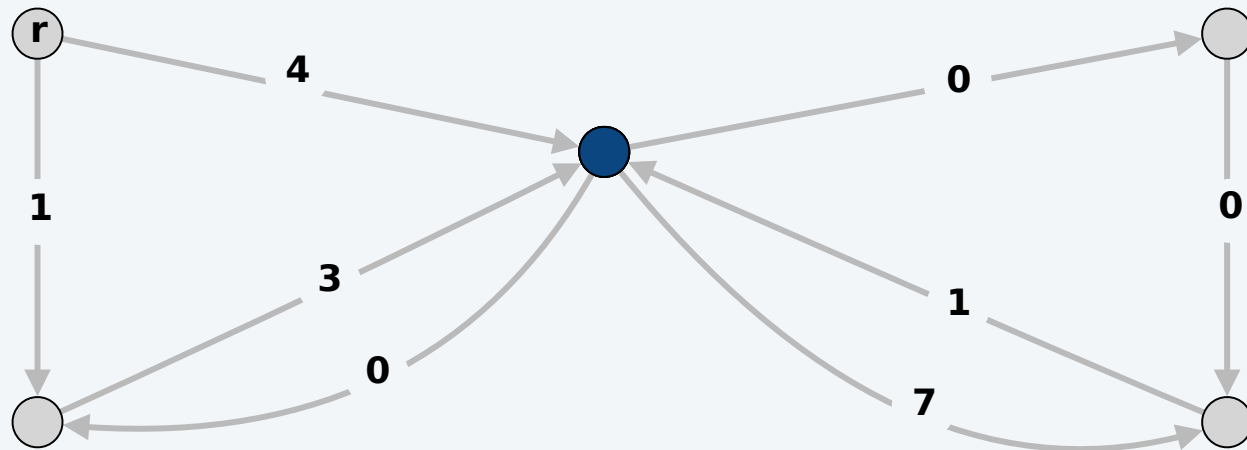
Now, all edges in F^* have 0 cost with respect to costs (c', v) .

If F^* does not contain a cycle, then it is a min-cost arborescence.

If F^* contains a cycle, can afford to use as many edges as desired.

Contract nodes into a supernode (removing any self-loops).

Recursively solve problem in contracted network with costs $c'(u, v)$.



Edmonds branching algorithm



EDMONDSBRANCHING(G, r, c)

FOREACH $v \neq r$

$y(v) \leftarrow$ min 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 $G' = (V', E)$.

$T' \leftarrow$ 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

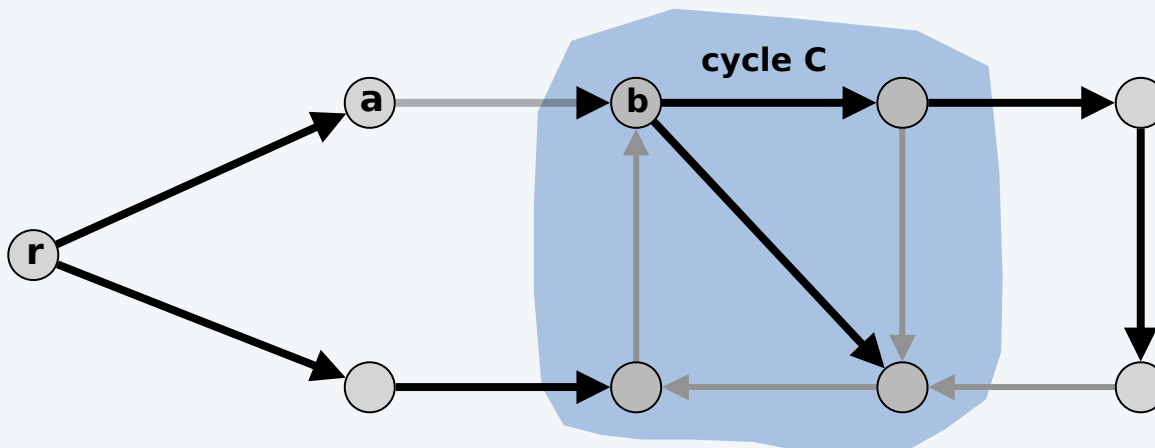
Q. What could go wrong?

A.

Min-cost arborescence G' has exactly one edge entering a node in C (since C is contracted to a single node)

But min-cost arborescence G might have more edges entering C

min-cost arborescence in G



Edmonds branching algorithm: key lemma

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

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

Case 0. T has no edges entering r .

Since T is an arborescence, there is a path from r to each node $v \in V$.
at least one edge enters r .

Case 1. T has exactly one edge entering r .

T satisfies the lemma.

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

We construct another min-cost arborescence T' that has exactly one edge entering r .

Edmonds branching algorithm: key lemma

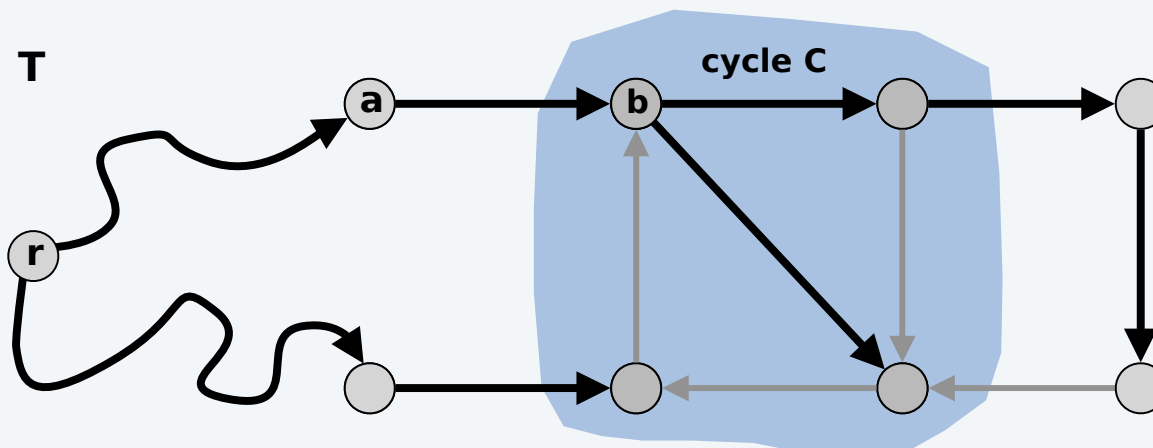
Case 2 construction of T

Let (a, b) be an edge in T entering C that lies on a shortest path from r to C .

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 a .

path from r to C uses only one node in C



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 to C .

We delete all edges T that enter a node in C except (a, b) .

We add in all edges C except the one that enters a .

path from r to C uses only one node in C

Claim. T' is a min-cost arborescence.

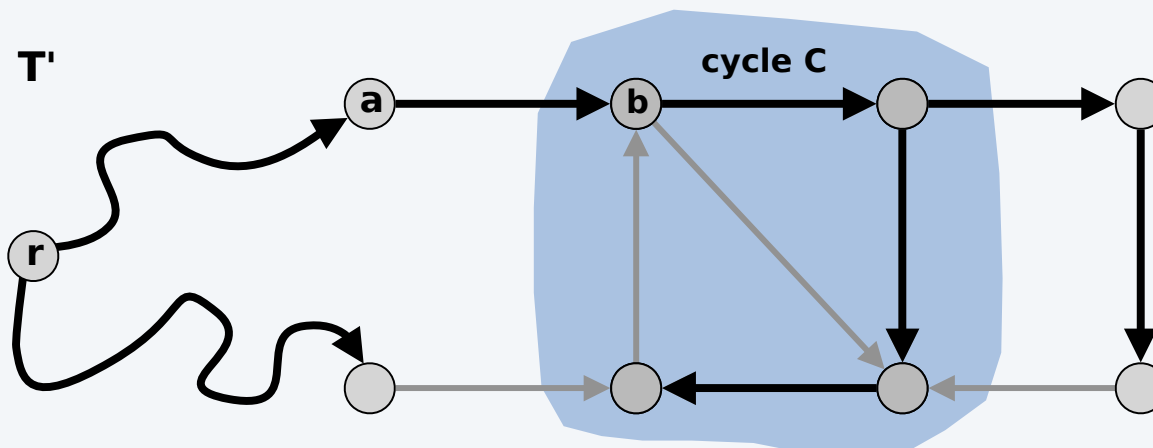
The cost of T' is at most that of T since we add only 0-cost edges.

T' has exactly one edge entering each node.

T' has no directed cycles.

T is an arborescence rooted at r

(T had no cycles before; no cycles with (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

If the edges of E^* form an arborescence, then min-cost arborescence.

Otherwise, we use reduced costs, which is equivalent.

After contracting a 0-cost cycle obtain a smaller graph the algorithm finds a min-cost arborescence' (by induction).

Key lemma: there exists a min-cost arborescence that corresponds to T . ■

Theorem. The greedy algorithm can be implemented in $O(m)$ time.

Pf.

At most n contractions (since each reduces the number of nodes).

Finding and contracting the cycle takes $O(m)$ time.

Transforming \bar{g} into T takes $O(m)$ time. ■

Min-cost arborescence

Theorem. [Gabow-Galil-Spencer-Tarjan 1985] There exists an $O(m \log n)$ 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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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 $O(m \log \beta(m, n))$ time, improved from $O(m \beta(m, n))$ time, where $\beta(m, n) = \min \{i \mid \log^{(i)} n \leq m/n\}$. Our minimum spanning tree algorithm for directed graphs runs in $O(n \log n + m)$ time, improved from $O(n \log n + m \log \log \log_{(m/n+2)} n)$. Both algorithms can be extended to allow a degree constraint at one vertex.

1. Introduction

A *heap* (sometimes called a *priority queue*) is an abstract data structure consisting of a collection of *items*, each with a real-valued *key*, on which at least the following operations are possible: