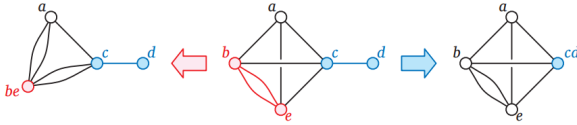


Algorithms Final Cheat Sheet

RandomizedAlgorithms

Max Flows and Min Cuts



Blind Guess

```

1: function GUESSMINCUT( $G$ )
2:   for  $i \leftarrow n, 2$  do
3:     pick a random edge  $e$  in  $G$ 
4:      $G \leftarrow G/e$ 
5:   end for
6:   return the only cut in  $G$ 
7: end function

```

$$P(n) = \frac{2}{n(n-1)}$$

Repeated Guessing

```

1: function KARGERMINCUT( $G$ )
2:    $min_k \leftarrow \infty$ 
3:   for  $i \leftarrow 1, N$  do
4:      $X \leftarrow \text{GUESSMINCUT}(G)$ 
5:     if  $|X| < min_k$  then
6:        $min_k \leftarrow |X|$ 
7:        $minX \leftarrow X$ 
8:     end if
9:   end for
10:  return  $minX$ 
11: end function

```

Set $N = c \binom{n}{2} \ln n$ for some constant c . $P(n) \geq 1 - \frac{1}{n^c}$.

KARGERMINCUT computes the min cut of any n -node graph with high probability in $O(n^4 \log n)$ time.

Not-So-Blindly Guessing

```

1: function CONTRACT( $G, m$ )
2:   for  $i \leftarrow n, m$  do
3:     pick a random edge  $e$  in  $G$ 
4:      $G \leftarrow G/e$ 
5:   end for
6: end function
7: function BETTERGUESS( $G$ )
8:   if  $G$  has more than 8 vertices then
9:      $G_1 \leftarrow \text{CONTRACT}(G, n/\sqrt{2} + 1)$ 

```

```

10:     $G_2 \leftarrow \text{CONTRACT}(G, n/\sqrt{2} + 1)$ 
11:     $X_1 \leftarrow \text{BETTERGUESS}(G_1)$ 
12:     $X_2 \leftarrow \text{BETTERGUESS}(G_2)$ 
13:    return  $\min(X_1, X_2)$ 
14:  else
15:    use brute force
16:  end if
17: end function

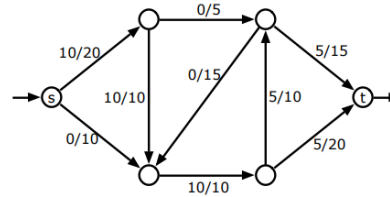
```

$P(n) \geq 1/\log n$. The running time is $O(n^2 \log n)$.

Flows

A *flow* is a function f that satisfies the *conservation constraint* at every vertex v : the total flow into v is equal to the total flow out of v .

A flow f is *feasible* if $f(e) \leq c(e)$ for each edge e . A flow *saturates* edge e if $f(e) = c(e)$, and *avoids* edge e if $f(e) = 0$.

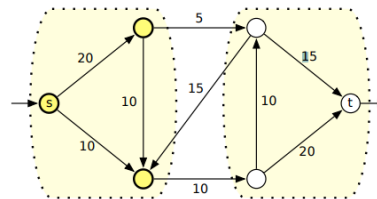


An (s, t) -flow with value 10. Each edge is labeled with its flow/capacity.

Cuts

A *cut* is a partition of the vertices into disjoint subsets S and T - meaning $S \cup T = V$ and $S \cap T = \emptyset$ - where $s \in S$ and $t \in T$.

If we have a capacity function c , the *capacity* of a cut is the sum of the capacities of the edges that start in S and end in T . The definition is asymmetric; edges that start in T and end in S are unimportant. The *min-cut problem* is to compute a cut whose capacity is as large as possible.

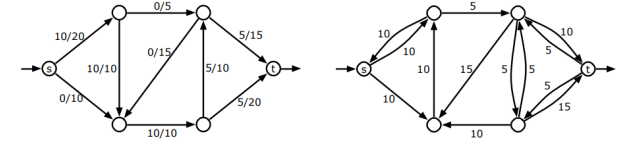


An (s, t) -cut with capacity 15. Each edge is labeled with its capacity.

Theorem 1 (Maxflow Mincut Theorem) In any flow network, the value of the maximum flow is equal to the capacity of the minimum cut.

Residual Capacity

$$c_f(u \rightarrow v) = \begin{cases} c(u \rightarrow v) - f(u \rightarrow v) & \text{if } u \rightarrow v \in E \\ f(v \rightarrow u) & \text{if } v \rightarrow u \in E \\ 0 & \text{otherwise} \end{cases}$$



A flow f in a weighted graph G and the corresponding residual graph G_f .

Augmenting Paths

Suppose there is a path $s = v_0 \rightarrow v_1 \rightarrow \dots \rightarrow v_r$ in the residual graph G_f . This is an *augmenting path*. Let $F = \min_i c_f(v_i \rightarrow v_{i+1})$ denote the maximum amount of flow that we can push through the augmenting path in G_f . We can augment the flow into a new flow function f' :

$$f'(u \rightarrow v) = \begin{cases} f(u \rightarrow v) + F & \text{if } u \rightarrow v \in s \\ f(u \rightarrow v) - F & \text{if } v \rightarrow u \in s \\ f(u \rightarrow v) & \text{otherwise} \end{cases}$$

Ford-Fulkerson

Starting with the zero flow, repeatedly augment the flow along *any* path from s to t in the residual graph, until there is no such path.

Further Work

The fastest known maximum flow algorithm, announced by James Orlin in 2012, runs in $O(VE)$ time.

Flow/Cut Applications

Edge-Disjoint Paths

A set of paths in G is *edge-disjoint* if each edge in G appears in at most one of the paths; several edge-disjoint paths may pass through the same vertex, however. Assign each edge capacity 1. The number of edge-disjoint paths is exactly equal to the value of the flow. Using Orlin's algorithm is overkill; the maximum flow has value at most $V - 1$, so Ford-Fulkerson's original augmenting path algorithm also runs in $O(|f^*| E) = O(VE)$ time.

Vertex Capacities and Vertex-Disjoint Paths

If we require the total flow into (and out of) any vertex v other than s and t is at most some value $c(v)$, we transform the input into a new graph. We replace each vertex v with two vertices v_{in} and v_{out} , connected by an edge $v_{in} \rightarrow v_{out}$ with capacity $c(v)$, and then replace every directed edge $u \rightarrow v$ with the edge $u_{out} \rightarrow v_{in}$ (keeping the same capacity).

Computing the maximum number of *vertex-disjoint* paths from s to t in any directed graph simply involves giving every vertex capacity 1, and computing a maximum flow. SAT and CNFSAT

NP-Hardness

Definition 2 P is the set of decision problems that can be solved in polynomial time. Intuitively, P is the set of problems that can be solved quickly.

Definition 3 NP is the set of decision problems with the following property: if the answer is Yes, then there is a proof of this fact that can be checked in polynomial time. Intuitively, NP is the set of decision problems where we can verify a Yes answer quickly if we have the solution in front of us.

Definition 4 co-NP is essentially the opposite of NP. If the answer to a problem in co-NP is No, then there is a proof of this fact that can be checked in polynomial time.

Every decision problem in P is also in NP and also in co-NP.

Definition 5 A problem Π is NP-hard if a polynomial-time algorithm for Π would imply a polynomial-time algorithm for every problem in NP.

Definition 6 A problem Π is NP-complete if it is both NP-hard and an element of NP.

Theorem 7 (Cook-Levin Theorem) Circuit satisfiability is NP-complete.

To prove that problem A is NP-hard, reduce a known NP-hard problem to A .

Definition 8 A many-one reduction from one language $L' \subseteq \Sigma^*$ is a function $f : \Sigma^* \rightarrow \Sigma^*$ such that $x \in L'$ iff $f(x) \in L$. A language L is NP-hard iff, for any language $L' \in NP$, there is a many-one reduction from L' to L that can be computed in polynomial time.

NP-Hard Problems

- SAT
- 3SAT
- Maximum Independent Set: find the size of the largest subset of the vertices of a graph with no edges between them
- Clique: Compute the number of nodes in its largest complete subgraph
- Vertex Cover: Smallest set of vertices that touch every edge in the graph
- Graph Coloring: Find the smallest possible number of colors in a legal coloring such that every edge has two different colors at its endpoints
- Hamiltonian Cycle: find a cycle that visits each vertex in a graph exactly once
- Subset Sum: Given a set X of positive integers and an integer t , determine whether X has a subset whose elements sum to t
- Planar Circuit SAT: Given a boolean circuit that can be embedded in the plane so that no two wires

cross, is there an input that makes the circuit output True

- Not All Equal 3SAT: Given a 3CNF formula, is there an assignment of values to the variables so that every clause contains at least one True literal and at least one False literal?
- Exact 3-Dimensional Matching: Given a set S and a collection of three-element subsets of S , called *triples*, is there a sub-collection of disjoint triples that exactly cover S ?
- Partition: Given a set S of n integers, are there subsets A and B such that $A \cup B = S$, $A \cap B = \emptyset$, and $\sum_{a \in A} a = \sum_{b \in B} b$?
- 3Partition: Given a set S of $3n$ integers, can it be partitioned into n disjoint three-element subsets, such that every subset has exactly the same sum?
- Set Cover: Given a collection of sets $S = \{S_1, S_2, \dots, S_m\}$, find the smallest sub-collection of S_i 's that contains all the elements of $\bigcup_i S_i$
- Hitting Set: Given a collection of sets $S = \{S_1, S_2, \dots, S_m\}$, find the minimum number of elements of $\bigcup_i S_i$ that hit every set in S
- Hamiltonian Path: Given a graph G , is there a path in G that visits every vertex exactly once?
- Longest Path: Given a non-negatively weighted graph G and two vertices u and v , what is the longest simple path from u to v in the graph? A path is *simple* if it visits each vertex at most once.
- Steiner Tree: Given a weighted, undirected graph G with some of the vertices marked, what is the minimum-weight subtree of G that contains every marked vertex?