

# CS3230 Finals Notes

## Chapter 7 - Network Flow

Define a s-t cut to be a partition  $(A, B)$  with the source in  $A$  and the sink in  $B$ . The capacity of the cut will be the **sum of the capacities of all the edges out of  $A$** .

Flow value lemma: If  $(A, B)$  is an s-t cut, then the net flow of the cut is equal to the amount leaving  $s$ , which is equal to the sum of the flow out of  $A$  - sum of the flow into  $A$ .

Weak duality: The value of the flow across a s-t cut is at most equal to the capacity of the cut.

If  $f$  is any flow and  $(A, B)$  is any cut, if  $v(f) = \text{cap}(A, B)$  then  $f$  is a max flow and  $(A, B)$  is a min cut.

FordFulkerson Algorithm: Add a flow, then reverse the edge, until it is not possible to add any flow into the graph anymore. It will output the maxflow. FF is not a polynomial time algo since it can potentially depend on the size of the input.

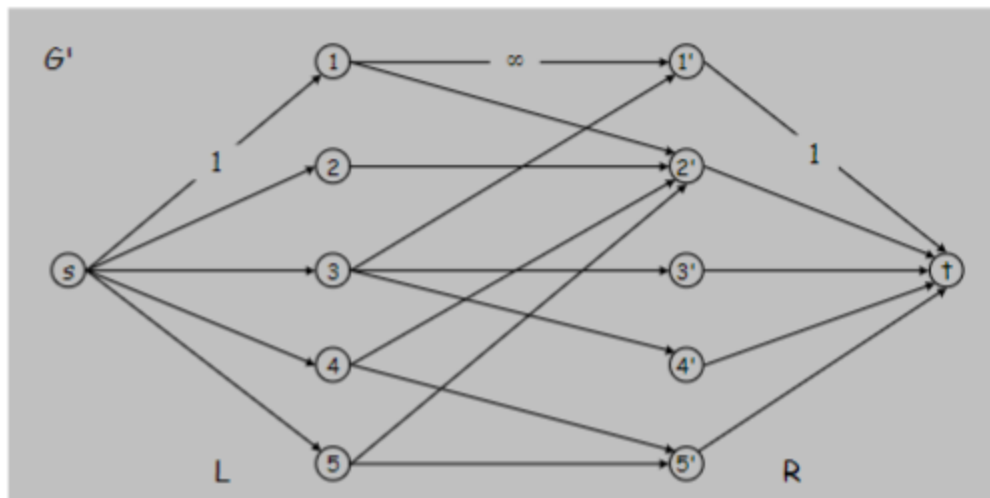
The algorithm terminates at at most  $nC$  iterations, if each augmentation increases the maxflow value by at least 1.  $O(mn)$  time.

Saturated edge: when the flow through the edge is equivalent to its capacity, and you cannot push anymore flow.

Modelling bipartite (max cardinality) matching as a maxflow problem

### Max flow formulation.

- Create digraph  $G' = (L \cup R \cup \{s, t\}, E')$ .
- Direct all edges from L to R, and assign infinite (or unit) capacity.
- Add source  $s$ , and unit capacity edges from  $s$  to each node in L.
- Add sink  $t$ , and unit capacity edges from each node in R to  $t$ .



Max cardinality matching in  $G$  = value of max flow in  $G'$

(since each node in  $L$  and  $R$  participate in at most one edge in  $M$ , the set of edges with flow  $f = 1$  after running the Ford Fulkerson algo)

If a perfect matching exists, then the size of set  $R$ ,  $|R|$  is larger than or equal to  $|L|$ , the size of the set  $L$ . By marriage theorem, iff for any subset  $S$  of  $L$ , the size of  $N(S)$  is larger or equal to  $S$ .

### Maxflow Applications

We can model problems, by adding an edge between  $L$  and  $R$  nodes if the path between them in the problem satisfies a certain requirement ( $L_{\text{path}} < N$ ).

### Edge Disjoint Path

If we set the capacity of every edge to unit capacity, then the max disjoint path will be equivalent to the max flow.


## Chapter 8 - NP Completeness

An algorithm is said to run in polynomial time if the runtime is polynomial in **length of input**. But it depends on what  $n$  means in the context of the problem!

i.e. if we want to check if an element 'a' is the largest element in a  $n$  sized array of  $n$  elements. then length of input is  **$n$ , number of elements**.

$P$  refers to the set of all decision problems with a polynomial runtime.

### Polynomial time reductions

- ▶ Let  $A$  and  $B$  be decision problems 
- ▶ Definition: We say that  $A \leq_p B$  to mean that there exists a polynomial time algorithm which does the following:
  - ▶ Input:  $a$ , an instance of  $A$
  - ▶ Output:  $b$ , an instance of  $B$
  - ▶ The correct answer to  $a$  and  $b$  are the same (i.e. both yes or both no)
- ▶ Theorem: if  $B \in P$ , and  $A \leq_p B$ , then  $A \in P$ 
  - ▶ Given an instance of  $a \in A$ , we can generate  $b \in B$ , solve  $b$  and then use that as our answer to  $a$

If  $A$  reduces to  $B$ , that means if  $B$  is solvable in polynomial time, then  $A$  is also solvable in polynomial time (since we can just get the answer in  $B$  and it will be the

answer for A)

Contrapositive: if A reduces to B, if A is NP hard it simply means that B is NP hard as well (since if there was a solution to B there would have been a solution to A).

**BUT if A reduces to B and A is in p, does not mean that B is in p!! It could be possible that B is not solvable in polynomial time but A is.**

**np refers to the class of decision problems that can be verified in polynomial time with an appropriate certificate.**

Examples of these problems: Does there exist a ....

if yes, then show the example (explicitly) but if no then it is impossible to fool the algorithm

Not an example: Find a example of .... , because there are potentially infinite configurations where you can check and it is not possible to verify all of them

a problem b is np complete if

- for all A in NP, A reduces to B
- B is in np

to show that a problem b is np complete, we find a np complete problem a and show that a reduces to b

## Chapter 11 - Approximation Algorithms

A p-approximation algorithm is an algorithm that:

- Is guaranteed to run in poly time
- Guaranteed to solve an arbitrary instance of the problem
- Guaranteed to find a solution within ratio **p** of true optimum

## **Chapter 12 - Randomized Algorithms**

Expectation = Summation of the (probability that  $X = j$ ) \*  $j$ , for all possible values of  $X$

Just see notes lol