2



Lecture 16: max flow Readings: CLRS 26.2

Trevor Brown https://student.cs.uwaterloo.ca/~cs341 trevor.brown@uwaterloo.ca

FLOWS AND PATHS

FLOWS AND PATHS

Edge-disjoint paths problem

- Input: digraph G = (V, E) and two vertices $s, t \in V$
- Output: A maximal number of edge-disjoint paths in G
- Paths P_1 and P_2 are **edge-disjoint** if they do not share any edges





3

s-t FLOWS

1

Let G = (V, E) be a digraph where each edge $e \in E$ has a **capacity** c(e) > 0



s-t FLOWS

- Let G = (V, E) be a digraph
- where each edge $e \in E$ has a **capacity** c(e) > 0An *s*-*t* flow assigns a number f(e) to each edge satisfying:
- Capacity constraints $0 \le f(e) \le c(e)$ for each edge Conservation of flow $f^{in}(v) = f^{out}(v)$ for $v \notin \{s,t\}$ Source $f^{in}(s) = 0$ Sink $f^{out}(t) = 0$ This is the value $f^{in}(s) = 1$ $f^{in}(s) = 1$ f^{in}

DEFINING $f^{in}(u)$ AND $f^{out}(u)$



.

MAX s-t FLOW

- digraph G = (V, E) with capacities c(e) for $e \in E$, Input: and two vertices s, t
- Output: a flow from s to t with maximum value i.e., that maximizes $f^{out}(s)$
- Motivation
 - Liquid flowing through pipes
 - Current through electrical networks Internet/telephony traffic routing
 - Also useful for seemingly unrelated problems (next time)

CONNECTION BETWEEN FLOWS AND PATHS

- In this example, max flow is 3
- Note max flow is limited by the sum of capacities out of s
 - ... and **into** t
- Flows vs paths
- a flow can always be decomposed into "capacity-disjoint" paths



10

LEMMA 1: DECOMPOSITION OF s-t FLOW INTO CAPACITY-DISJOINT s-t PATHS

- Let f be an s-t flow where f(e) is an integer for each $e \in E$, $f^{in}(s) = 0$ and value(f) = k
- Then there are s-t paths P_1, P_2, \ldots, P_k such that each edge e appears in f(e) of these paths
- Proof sketch by induction
- Base case: when k=1 there is only one path



hypothesis applies

11

INDUCTIVE STEP

Suppose lemma holds for k - 1, show it holds for k (where $k \ge 2$) Consider the edges with non-zero flow



There must exist some s-t path P in these edges (why?) Decrease the flow of each edge in P by 1

INDUCTIVE STEP

- Suppose lemma holds for k 1, show it holds for k (where $k \ge 2$)
- Consider the edges with non-zero flow



Decrease the flow of each edge in P by 1



14

But max flow is 3...

16

EXAMPLE APPLICATION OF LEMMA 1

- Given a flow of value k where $f(e) \in \{0,1\}$ for all $e \in E$ The lemma says the flow f can be
- decomposed into k edge-disjoint paths So if our goal is to find k edge-disjoint paths
- we can just focus in finding such a flow instead (so we don't need to worry about which edges belong to which paths during the algorithm)
- Can extract paths from such a flow by repeatedly doing: BFS on the non-zero flow edges, identifying an s-t path, and decrementing the flows along that path

13

15

17

HOME EXERCISE

Find a decomposition of the following flow into capacity-disjoint paths



UPPER BOUNDING THE MAX FLOW FOR G

- What is a good upper bound on the value of a flow? $\sum_{z(e)=17}$ And how do we know a flow is maximal? $\sqrt{\begin{bmatrix} \Sigma z(e)=17\\ m_{flow} \end{bmatrix}}$
- Trivial upper bound
- Sum of capacities of all edges
 Slightly better
- , soligini, solioi



- $\min\left(c^{out}(s), c^{in}(t)\right)$ where $c^{out}(s) = \sum_{e \text{ out } f s} c(e)$ and $c^{in}(t) = \sum_{e \text{ into } t} c(e)$
- Tightly bounds max flow in this case...

BUT WHAT ABOUT THIS CASE?



FLOWS AND CUTS

Estimate using min $(c^{out}(s), c^{in}(t)) = 30$ But real answer is 1...

DEFINITIONS: AN s-t CUT AND ITS CAPACITY

An *s*-*t* cut is a partition (S,V\S) where $s \in S$ and $t \in V \setminus S$ i.e., the partition separates *s* and *t*



UPPER BOUNDING EDGE-DISJOINT PATHS BY S-T CUT

- For the edge-disjoint paths problem, where c(e) = 1 for all e, cut capacity is just the **number of edges crossing the cut**
- If an *s*-*t* cut *S* has at most *k* edges crossing the cut, then are at most *k* edge-disjoint *s*-*t* paths, since each *s*-*t* path has an edge crossing the cut

GENERALIZING TO MAX FLOW

Lemma 2: if an *s*-*t* cut *S* has capacity k, the value of every flow must be $\leq k$

Proof sketch: for contra assume a flow with value k' > k



20

22

- By earlier lemma, a flow with value k'can be decomposed into k' capacity-disjoint paths each w/flow 1 Each such path crosses the cut,
- and consumes one unit of the cut's capacity (up to k' in total) But the cut's capacity is only k,
- so the paths are not capacity-disjoint! Contradiction.

COROLLARY: MAX FLOW \leq MIN *s*-*t* CUT

- Recall lemma 2: if an *s*-*t* cut *S* has capacity *k*, the value of every flow must be $\leq k$
- This holds for any s-t cut
- Including the *s*-*t* cut *S* with the minimum capacity

So, max s-t flow \leq min capacity over all possible s-t cuts

In fact, it turns out max flow is exactly the min cut capacity
 So we can solve max flow by finding a min cut...

21

19

MIN s-t CUT PROBLEM

- Input: digraph G = (V, E) with capacities c(e) > 0 for $e \in E$, and two vertices s, t
- Output: an *s*-*t* cut *S* with minimal capacity $c^{out}(S)$
- This is a natural and useful problem on its own, and we will see some other interesting applications soon...

MAX-FLOW MIN-CUT THEOREM

- **Theorem 3:** every max *s*-*t* flow has value equal to the capacity of a min *s*-*t* cut
- One of the most beautiful and important results in combinatorial optimization and graph theory
- Diverse applications in CS and math
- We give an **algorithmic proof** of this theorem
- (showing that one algorithm solves both max-flow and min-cut at the same time)

FORD-FULKERSON METHOD Algorithm development (mixed together with proof of max-flow min-cut theorem)

24

NAÏVE ALGORITHM ATTEMPT

- For simplicity, try edge-disjoint path problem first (unit capacities) Greedy idea: find a shortest s-t path (to use few edges),
- then repeat on the remaining edges



- Difficult for greedy is to **decide** on a path **permanently**
- Unclear how to find a path that belongs in the optimal solution

25

FORD-FULKERSON METHOD

Same Ford as in Bellman-Ford :)

21

- Ford-Fulkerson is a more general "local search" algorithm which can undo previous decisions to improve the flow
- Greedy flow can be improved by "pushing back" some flow using an augmenting path through a residual graph



FORD-FULKERSON METHOD

Same Ford as in Bellman-Ford :)

- Ford-Fulkerson is a more general "local search" algorithm which can undo previous decisions to improve the flow
- Greedy flow can be improved by "pushing back" some flow using an augmenting path through a residual graph



RESIDUAL GRAPH

- A residual graph R_f is defined for a given flow f and graph G
- R_f has the same vertices as G
- For each edge e = uv in G, If f(e) < c(e), then R_f contains a **forward** edge (u, v)
 - with the remaining capacity c(e) f(e)
 - If f(e) > 0, then R_f contains a **backwards** edge (v, u) with **capacity** f(e) representing flow that could be "pushed back"

 $^{0/1}$ \cap G $O_{\overline{0/2}}$ $O_{\overline{0/1}}$ eedy flow f al g nh for ti Backwards edge: can undo flow Forward edge remaining capacity

ANOTHER EXAMPLE RESIDUAL GRAPH

- Recall: for each edge e = uv in G,
- If f(e) < c(e), then R_f contains a **forward** edge (u, v) with the **remaining capacity** c(e) f(e)
 - If f(e) > 0, then R_f contains a **backwards** edge (v, u) with **capacity** f(e) representing flow that could be "pushed back"

 $O_{\overline{0/1}}$ +O 0/1 flow f with value 2

