

CS 341: ALGORITHMS

Lecture 16: max flow
Readings: CLRS 26.2

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FLOWS AND PATHS

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FLAWS 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

This is a special case of the **maximum flow** problem
... where the union of paths defines a **flow**

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s-t FLOWS

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

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s-t FLOWS

- Let $G = (V, E)$ be a digraph where each edge $e \in E$ has a **capacity** $c(e) > 0$
- An $s-t$ flow assigns a number $f(e)$ to each edge satisfying:
 - Capacity constraints
 - $0 \leq f(e) \leq c(e)$ 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** of the flow

$f^{out}(s) = 3$ $f^{in} = f^{out} = 1$ $f^{in} = f^{out} = 2$ $f^{in}(t) = 3$

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

$f^{in}(u) = \sum_{e \text{ into } u} f(e)$

$f^{out}(u) = \sum_{e \text{ out of } u} f(e)$

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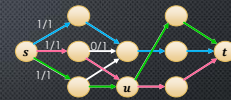
MAX $s-t$ FLOW

- Input: digraph $G = (V, E)$ with capacities $c(e)$ for $e \in E$, 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)

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CONNECTION BETWEEN FLOWS AND PATHS

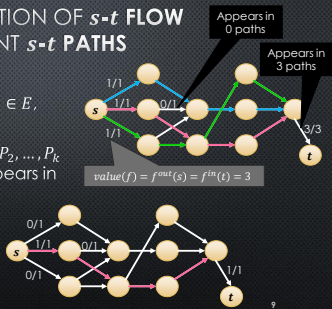
- In this example, max flow is 3
- Note max flow is limited by the sum of capacities **out of** s
 - ... and **into** t
- Flows vs paths
 - a flow can always be decomposed into "capacity-disjoint" paths



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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, \dots, 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



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INDUCTIVE STEP

- Suppose lemma holds for $k - 1$, show it holds for k (where $k \geq 2$)
 - 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

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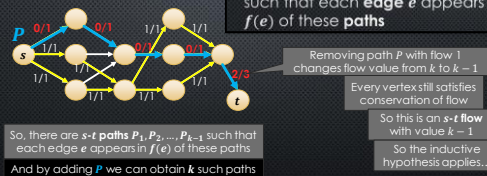
INDUCTIVE STEP

- Suppose lemma holds for $k - 1$, show it holds for k (where $k \geq 2$)
 - Consider the edges with **non-zero flow**
-
- There must exist some $s-t$ path P in these edges (So this is an $s-t$ flow with value $k - 1$)
 - **Decrease** the flow of each edge in P by 1 (So the inductive hypothesis applies...)

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INDUCTIVE STEP

- Lemma**
- 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, \dots, P_k such that each **edge** e appears in $f(e)$ of these **paths**



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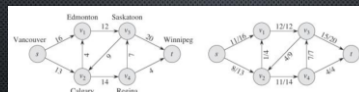
EXAMPLE APPLICATION OF LEMMA 1

- 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

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HOME EXERCISE

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



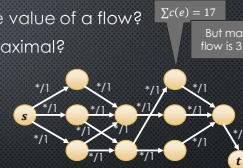
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FLows AND CUTS

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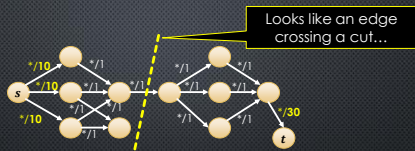
UPPER BOUNDING THE MAX FLOW FOR G

- What is a good upper bound on the value of a flow?
 - And how do we know a flow is maximal?
- Trivial upper bound
 - Sum of capacities of all edges
- Slightly better
 - $\min(c^{out}(s), c^{in}(t))$ where $c^{out}(s) = \sum_{e \text{ out of } s} c(e)$ and $c^{in}(t) = \sum_{e \text{ into } t} c(e)$
 - Tightly bounds max flow in this case...



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BUT WHAT ABOUT THIS CASE?

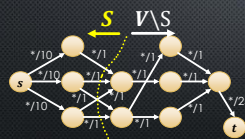


Estimate using $\min(c^{out}(s), c^{in}(t)) = 30$
But real answer is 1...

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DEFINITIONS: AN s - t CUT AND ITS CAPACITY

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



(Recall S does not need to be connected)

Let $\delta^{out}(S)$ be the set of edges directed out from S
 $\delta^{out}(S) = \{(u,v) \in E : u \in S, v \in V \setminus S\}$
 The capacity of the cut is the sum of the capacities of these edges
 $c^{out}(S) = \sum_{e \in \delta^{out}(S)} c(e)$

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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

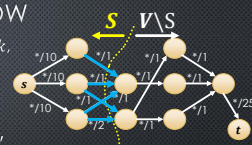


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GENERALIZING TO MAX FLOW

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 $k' > k$
- 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.



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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...

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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 $c^{cut}(S)$
- This is a natural and useful problem on its own, and we will see some other interesting applications soon...

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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)

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FORD-FULKERSON METHOD

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

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NAÏVE ALGORITHM ATTEMPT

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



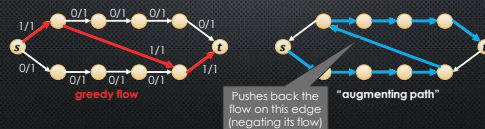
- Difficult for greedy is to **decide** on a path **permanently**
- Unclear how to find a path that **belongs** in the optimal solution

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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**

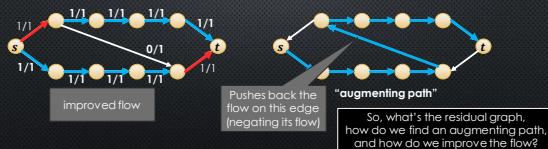


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FORD-FULKERSON METHOD

Same Ford as in Bellman-Ford :)

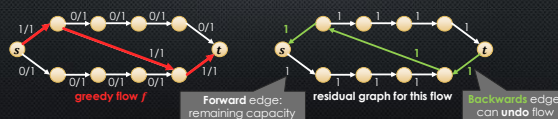
- Ford-Fulkerson is a more general "local search" algorithm which can **undo** previous decisions to improve the flow
- Greedy flow can be improved by "pushing back" some flow using an **augmenting path** through a **residual graph**



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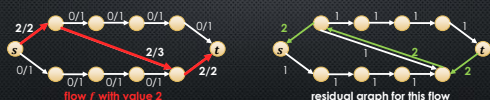
RESIDUAL GRAPH

- 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"



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"



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