

COL106: Data Structures and Algorithms

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Network Flow

Algorithm

Ford-Fulkerson

- Start with a flow f such that $f(e) = 0$
- While there is an $s - t$ path P in G_f
 - Augment flow along an $s - t$ path and let f' be resulting flow
 - Update f to f' and G_f to $G_{f'}$
- return(f)

- What is the running time of the above algorithm? $O(m \cdot C)$
 - Claim 2: $v(f') > v(f)$.
 - Claim 3: The while loop runs for $C = \sum_{e \text{ out of } s} c(e)$ iterations.
 - Claim 4: Finding augmenting path and augmenting flow along this path takes $O(m)$ time.

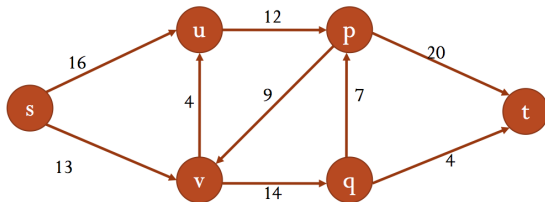
Network Flow

Maximum flow

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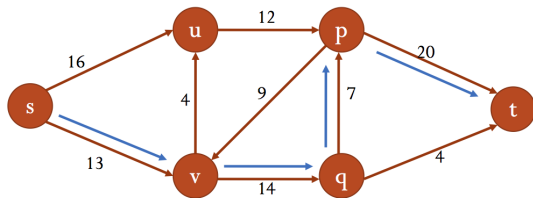
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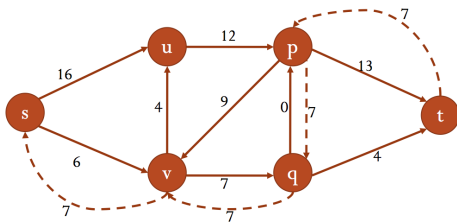


Figure: Graph G_f , where $f(s, u) = 0, f(s, v) = 7, f(v, u) = 0, f(v, q) = 7, f(u, p) = 0, f(p, v) = 0, f(p, t) = 7, f(q, p) = 7, f(q, t) = 0$

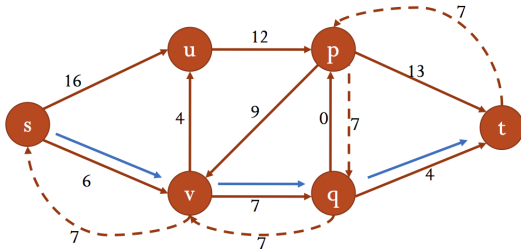
Network Flow

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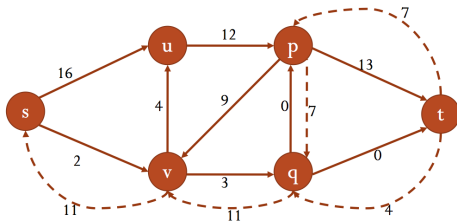


Figure: Graph G_f , where $f(s, u) = 0, f(s, v) = 11, f(v, u) = 0, f(v, q) = 11, f(u, p) = 0, f(p, v) = 0, f(p, t) = 7, f(q, p) = 7, f(q, t) = 4$

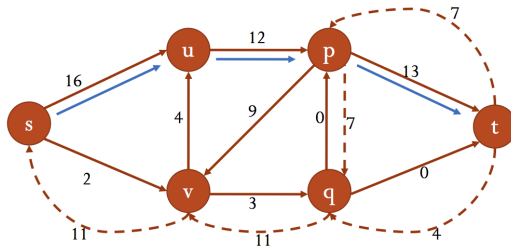
Network Flow

Maximum flow

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Network Flow

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Algorithm

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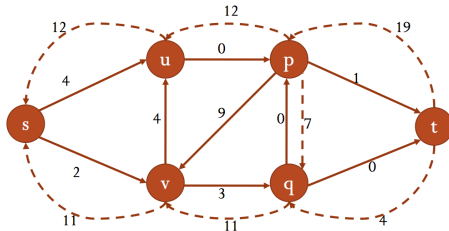


Figure: Graph G_f , where $f(s, u) = 12, f(s, v) = 11, f(v, u) = 0, f(v, q) = 11, f(u, p) = 12, f(p, v) = 0, f(p, t) = 19, f(q, p) = 7, f(q, t) = 4$

Network Flow

Maximum flow

Algorithm

Ford-Fulkerson

- Start with a flow f such that $f(e) = 0$
- While there is an $s - t$ path P in G_f
 - Augment flow along an $s - t$ path and let f' be resulting flow
 - Update f to f' and G_f to $G_{f'}$
- return(f)

- How do we prove that the flow returned by the Ford-Fulkerson algorithm is the maximum flow?

Network Flow

Maximum flow

- Theorem 1: Let f be the flow returned by the Ford-Fulkerson algorithm. Then f maximizes $v(f) = \sum_{e \text{ out of } s} f(e)$.

Definition (f^{in} and f^{out})

Let S be a subset of vertices and f be a flow. Then

$$f^{in}(S) = \sum_{e \text{ into } S} f(e) \quad \text{and} \quad f^{out}(S) = \sum_{e \text{ out of } S} f(e)$$

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Definition ($s - t$ cut)

A partition of vertices (A, B) is called an $s - t$ cut iff A contains s and B contains t .

Definition (Capacity of $s - t$ cut)

The capacity of an $s - t$ cut (A, B) is defined as

$$C(A, B) = \sum_{e \text{ out of } A} c(e).$$

Network Flow

Maximum flow

- Theorem 1: Let f be the flow returned by the Ford-Fulkerson algorithm. Then f maximizes $v(f) = \sum_{e \text{ out of } s} f(e)$.

Proof

- Claim 1.1: For any $s - t$ cut (A, B) and any $s - t$ flow f ,
 $v(f) = f^{out}(A) - f^{in}(A)$.

Network Flow

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Proof

- Claim 1.1: For any $s - t$ cut (A, B) and any $s - t$ flow f , $v(f) = f^{out}(A) - f^{in}(A)$.

Proof of claim 1.1.

$v(f) = f^{out}(\{s\}) - f^{in}(\{s\})$ and for all other nodes $v \in A$, $f^{out}(\{v\}) - f^{in}(\{v\}) = 0$. So,

$$v(f) = \sum_{v \in A} (f^{out}(\{v\}) - f^{in}(\{v\})) = f^{out}(A) - f^{in}(A).$$



Network Flow

Maximum flow

- Theorem 1: Let f be the flow returned by the Ford-Fulkerson algorithm. Then f maximizes $v(f) = \sum_{e \text{ out of } s} f(e)$.

Proof

- Claim 1.1: For any s - t cut (A, B) and any s - t flow f , $v(f) = f^{out}(A) - f^{in}(A)$.
- Claim 1.2: Let f be any s - t flow and (A, B) be any s - t cut. Then $v(f) \leq C(A, B)$.

Network Flow

Maximum flow

- Theorem 1: Let f be the flow returned by the Ford-Fulkerson algorithm. Then f maximizes $v(f) = \sum_{e \text{ out of } s} f(e)$.

Proof

- Claim 1.1: For any s - t cut (A, B) and any s - t flow f , $v(f) = f^{out}(A) - f^{in}(A)$.
- Claim 1.2: Let f be any s - t flow and (A, B) be any s - t cut. Then $v(f) \leq C(A, B)$.

Proof of claim 1.2.

$$v(f) = f^{out}(A) - f^{in}(A) \leq f^{out}(A) \leq C(A, B). \quad \square$$

Network Flow

Maximum flow

- Theorem 1: Let f be the flow returned by the Ford-Fulkerson algorithm. Then f maximizes $v(f) = \sum_{e \text{ out of } s} f(e)$.

Proof

- Claim 1.1: For any s - t cut (A, B) and any s - t flow f ,
 $v(f) = f^{out}(A) - f^{in}(A)$.
- Claim 1.2: Let f be any s - t flow and (A, B) be any s - t cut. Then
 $v(f) \leq C(A, B)$.
- Claim 1.3: Let f be an s - t flow such that there is no s - t path in G_f . Then there is an s - t cut (A^*, B^*) such that
 $v(f) = C(A^*, B^*)$. Furthermore, f is a flow with maximum value and (A^*, B^*) is an s - t cut with minimum capacity.



Network Flow

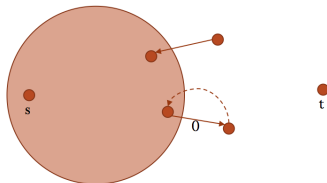
Maximum flow

- Claim 1.3: Let f be an s - t flow such that there is no s - t path in G_f . Then there is an s - t cut (A^*, B^*) such that $v(f) = C(A^*, B^*)$. Furthermore, f is a flow with maximum value and (A^*, B^*) is an s - t cut with minimum capacity.

Proof of claim 1.3

- Let A^* be all vertices reachable from s in the graph G_f (see figure below). Then we have:

$$\begin{aligned}v(f) &= f^{out}(A^*) - f^{in}(A^*) \\ &= f^{out}(A^*) - 0 \\ &= C(A^*, B^*)\end{aligned}$$



A^* (all vertices reachable from s in G_f)

Network Flow

Maximum flow

Theorem (Max-flow-min-cut theorem)

In every flow network, the maximum value of s - t flow is equal to the minimum capacity of s - t cut.

Network Flow

Maximum flow

- Summary:
 - Ford-Fulkerson Algorithm:
 - Given network with integer capacities, find a source-to-sink path and push as much flow along the path as possible.
 - Update the residual capacity of edges in the residual graph.
 - Repeat.
 - Proof of correctness:
 - The algorithm terminates (since the capacities are integers).
 - Max-flow-min-cut theorem: In every flow network, the maximum value of s - t flow is equal to the minimum capacity of s - t cut.

Applications of Network Flow

Network Flow

Bipartite Matching

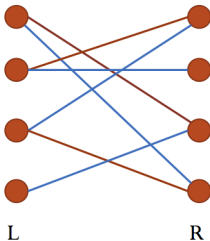
Definition (Matching in bipartite graphs)

A subset M of edges such that each node appears in at most one edge in M .

Problem

Given a bipartite graph $G = (L, R, E)$, design an algorithm to give a maximum matching in the graph.

- Example:



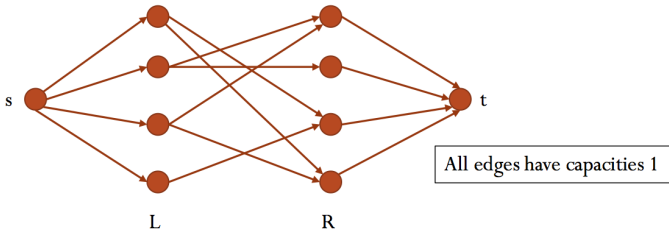
Network Flow

Bipartite Matching

Problem

Given a bipartite graph $G = (L, R, E)$, design an algorithm to give a maximum matching in the graph.

- Consider the network graph below constructed from the bipartite graph.



- Claim 1: Suppose there is an integer flow of value k in the network graph. Then the bipartite graph has a matching of size k .

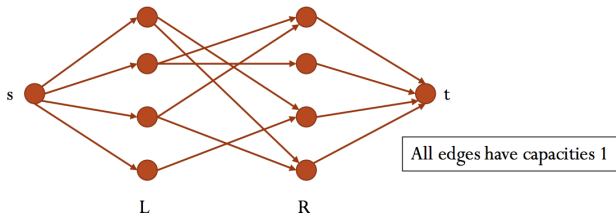
Network Flow

Bipartite Matching

Problem

Given a bipartite graph $G = (L, R, E)$, design an algorithm to give a maximum matching in the graph.

- Consider the network graph below constructed from the bipartite graph.



- Claim 1: Suppose there is an integer flow of value k in the network graph. Then the bipartite graph has a matching of size k .
- Claim 2: Suppose the bipartite graph has a matching of size k . Then there is an integer flow of value k in the network graph.

- Suppose there are four teams in IPL with their current number of wins:
 - Daredevils: 10
 - Sunrisers: 10
 - Lions: 10
 - Supergiants: 8
- There are 7 more games to be played. These are as follows:
 - Supergiants plays all other 3 teams.
 - Daredevils Vs Sunrisers, Sunrisers Vs Lions, Daredevils Vs Lions, Sunrisers Vs Daredevils

Network Flow

Team Elimination

- Suppose there are four teams in IPL with their current number of wins:
 - Daredevils: 10
 - Sunrisers: 10
 - Lions: 10
 - Supergiants: 8
- There are 7 more games to be played. These are as follows:
 - Supergiants plays all other 3 teams.
 - Daredevils Vs Sunrisers, Sunrisers Vs Lions, Daredevils Vs Lions, Sunrisers Vs Daredevils
- A team is said to be eliminated if it cannot end with maximum number of wins.
- Can we say that Supergiants have been eliminated give the current scenario?

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 - Lions: 9
 - Supergiants: 8
- There are 7 more games to be played. These are as follows:
 - Supergiants plays all other 3 teams.
 - 4 games between Daredevils and Sunrisers.
- Can we say that Supergiants have been eliminated give the current scenario?

Problem

There are n teams. Each team i has a current number of wins denoted by $w(i)$. There are $G(i, j)$ games yet to be played between team i and j . Design an algorithm to determine whether a given team x has been eliminated.

End