

Chapter 19

More Network Flow Applications

CS 473: Fundamental Algorithms, Spring 2013

April 4, 2013

19.1 Baseball Pennant Race

19.1.0.1 Pennant Race



19.1.0.2 Pennant Race: Example

Can Boston win the pennant?

No, because Boston can win at most 91 games.

19.1.0.3 Another Example

Can Boston win the pennant?

Not clear unless we know what the remaining games are!

Example 19.1.1.

Team	Won	Left
New York	92	2
Baltimore	91	3
Toronto	91	3
Boston	89	2

Example 19.1.2.

Team	Won	Left
New York	92	2
Baltimore	91	3
Toronto	91	3
Boston	90	2

19.1.0.4 Refining the Example

Example 19.1.3.

Team	Won	Left	NY	Bal	Tor	Bos
New York	92	2	—	1	1	0
Baltimore	91	3	1	—	1	1
Toronto	91	3	1	1	—	1
Boston	90	2	0	1	1	—

Can Boston win the pennant? Suppose Boston does

- (A) *Boston wins both its games to get 92 wins*
- (B) *New York must lose both games; now both Baltimore and Toronto have at least 92*
- (C) *Winner of Baltimore-Toronto game has 93 wins!*

19.1.0.5 Abstracting the Problem

Given

- (A) A set of teams S
- (B) For each $x \in S$, the current number of wins w_x
- (C) For any $x, y \in S$, the number of remaining games g_{xy} between x and y
- (D) A team z

Can z win the pennant?

19.1.0.6 Towards a Reduction

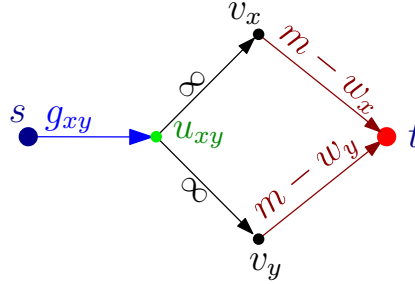
\bar{z} can win the pennant if

- (A) \bar{z} wins at least m games
 - (A) to maximize \bar{z} 's chances we make \bar{z} win all its remaining games and hence $m = w_{\bar{z}} + \sum_{x \in S} g_{x\bar{z}}$
- (B) no other team wins more than m games
 - (A) for each $x, y \in S$ the g_{xy} games between them have to be *assigned* to either x or y .
 - (B) each team $x \neq \bar{z}$ can win at most $m - w_x - g_{x\bar{z}}$ remaining games

Is there an assignment of remaining games to teams such that no team $x \neq \bar{z}$ wins more than $m - w_x$ games?

19.1.0.7 Flow Network: The basic gadget

- (A) s : source
- (B) t : sink
- (C) x, y : two teams
- (D) g_{xy} : number of games remaining between x and y .
- (E) w_x : number of points x has.
- (F) m : maximum number of points x can win before team of interest is eliminated.

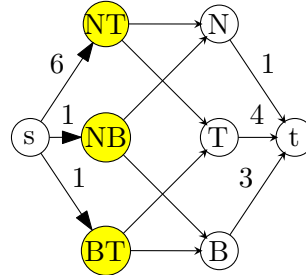


19.1.1 Flow Network: An Example

19.1.1.1 Can Boston win?

Team	Won	Left	NY	Bal	Tor	Bos
New York	90	11	—	1	6	4
Baltimore	88	6	1	—	1	4
Toronto	87	11	6	1	—	4
Boston	79	12	4	4	4	—

- (A) $m = 79 + 12 = 91$: Boston can get at most 91 points.



19.1.1.2 Constructing Flow Network

Notations

- (A) S : set of teams,
- (B) w_x wins for each team, and
- (C) g_{xy} games left between x and y .
- (D) m be the maximum number of wins for \bar{z} ,
- (E) and $S' = S \setminus \{\bar{z}\}$.

Reduction Construct the flow network G as follows

- (A) One vertex v_x for each team $x \in S'$, one vertex u_{xy} for each pair of teams x and y in S'
- (B) A new source vertex s and sink t
- (C) Edges (u_{xy}, v_x) and (u_{xy}, v_y) of capacity ∞
- (D) Edges (s, u_{xy}) of capacity g_{xy}
- (E) Edges (v_x, t) of capacity equal $m - w_x$

19.1.1.3 Correctness of reduction

Theorem 19.1.4. G' has a maximum flow of value $g^* = \sum_{x,y \in S'} g_{xy}$ if and only if \bar{z} can win the most number of games (including possibly tie with other teams).

19.1.1.4 Proof of Correctness

Proof: Existence of g^* flow $\Rightarrow \bar{z}$ wins pennant

- (A) An integral flow saturating edges out of s , ensures that each remaining game between x and y is added to win total of either x or y
- (B) Capacity on (v_x, t) edges ensures that no team wins more than m games

Conversely, \bar{z} wins pennant \Rightarrow flow of value g^*

- (A) Scenario determines flow on edges; if x wins k of the games against y , then flow on (u_{xy}, v_x) edge is k and on (u_{xy}, v_y) edge is $g_{xy} - k$

■

19.1.1.5 Proof that \bar{z} cannot win the pennant

- (A) Suppose \bar{z} cannot win the pennant since $g^* < g$. How do we *prove* to some one *compactly* that \bar{z} cannot win the pennant?
- (B) Show them the min-cut in the reduction flow network!
- (C) See text book for a natural interpretation of the min-cut as a certificate.

19.2 An Application of Min-Cut to Project Scheduling

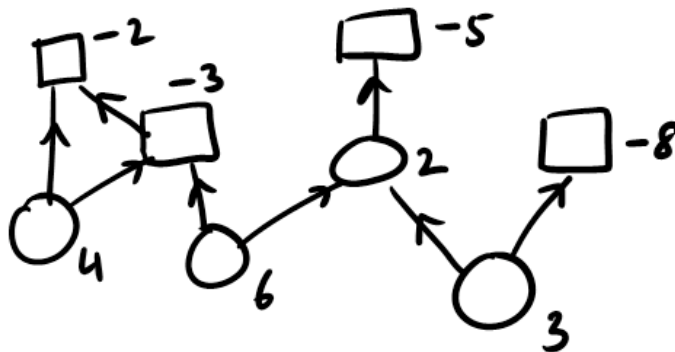
19.2.0.6 Project Scheduling

Problem:

- (A) n projects/tasks $1, 2, \dots, n$
- (B) *dependencies* between projects: i depends on j implies i cannot be done unless j is done. dependency graph is *acyclic*
- (C) each project i has a cost/profit p_i
 - (A) $p_i < 0$ implies i requires a cost of $-p_i$ units
 - (B) $p_i > 0$ implies that i generates p_i profit

Goal: Find projects to do so as to *maximize* profit.

19.2.0.7 Example



19.2.0.8 Notation

For a set A of projects:

- (A) A is a *valid* solution if A is *dependency closed*, that is for every $i \in A$, all projects that i depends on are also in A .
- (B) $profit(A) = \sum_{i \in A} p_i$. Can be negative or positive.

Goal: find valid A to maximize $profit(A)$.

19.2.0.9 Idea: Reduction to Minimum-Cut

Finding a set of projects is partitioning the projects into two sets: those that are done and those that are not done.

Can we express this is a minimum cut problem?

Several issues:

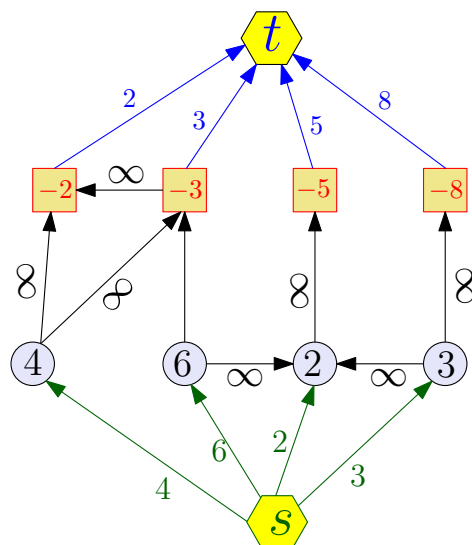
- (A) We are interested in maximizing profit but we can solve minimum cuts.
- (B) We need to convert negative profits into positive capacities.
- (C) Need to ensure that chosen projects is a valid set.
- (D) The cut value captures the profit of the chosen set of projects.

19.2.0.10 Reduction to Minimum-Cut

Note: We are reducing a *maximization* problem to a *minimization* problem.

- (A) projects represented as nodes in a graph
- (B) if i depends on j then (i, j) is an edge
- (C) add source s and sink t
- (D) for each i with $p_i > 0$ add edge (s, i) with capacity p_i
- (E) for each i with $p_i < 0$ add edge (i, t) with capacity $-p_i$
- (F) for each dependency edge (i, j) put capacity ∞ (more on this later)

19.2.0.11 Reduction: Flow Network Example



19.2.0.12 Reduction contd

Algorithm:

- (A) form graph as in previous slide
- (B) compute s - t minimum cut (A, B)
- (C) output the projects in $A - \{s\}$

19.2.0.13 Understanding the Reduction

Let $C = \sum_{i:p_i>0} p_i$: maximum possible profit.

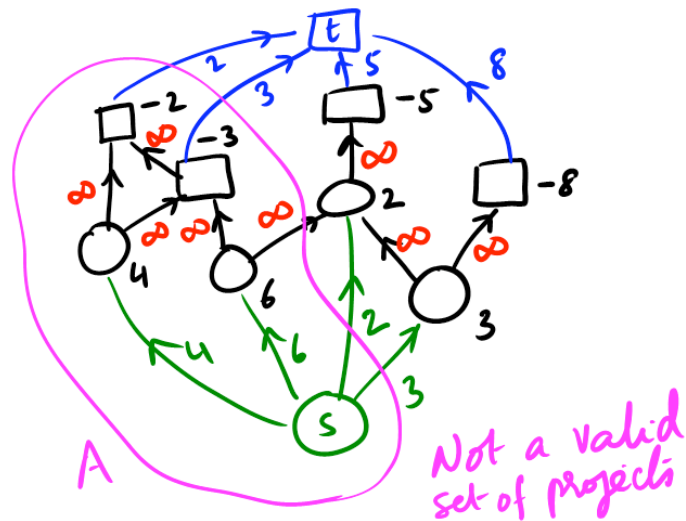
Observation: The minimum s - t cut value is $\leq C$. Why?

Lemma 19.2.1. Suppose (A, B) is an s - t cut of finite capacity (no ∞) edges. Then projects in $A - \{s\}$ are a valid solution.

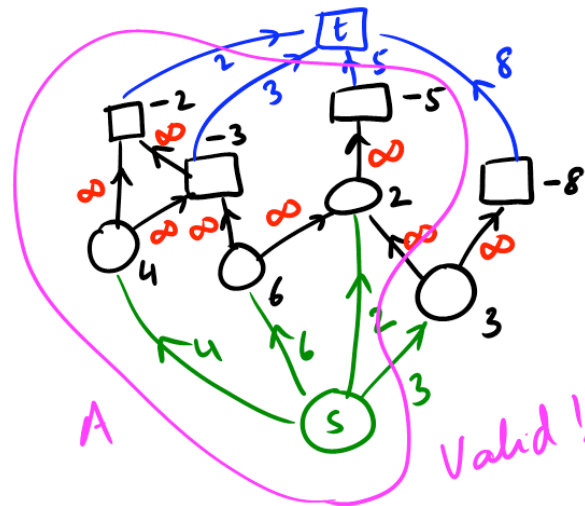
Proof: If $A - \{s\}$ is not a valid solution then there is a project $i \in A$ and a project $j \notin A$ such that i depends on j

Since (i, j) capacity is ∞ , implies (A, B) capacity is ∞ , contradicting assumption. ■

19.2.0.14 Example



19.2.0.15 Example



19.2.0.16 Correctness of Reduction

Recall that for a set of projects X , $profit(X) = \sum_{i \in X} p_i$.

Lemma 19.2.2. Suppose (A, B) is an s - t cut of finite capacity (no ∞) edges. Then $c(A, B) = C - profit(A - \{s\})$.

Proof: Edges in (A, B) :

(A) (s, i) for $i \in B$ and $p_i > 0$: capacity is p_i

- (B) (i, t) for $i \in A$ and $p_i < 0$: capacity is $-p_i$
- (C) cannot have ∞ edges

■

19.2.0.17 Proof contd

For project set A let

- (A) $cost(A) = \sum_{i \in A: p_i < 0} -p_i$
- (B) $benefit(A) = \sum_{i \in A: p_i > 0} p_i$
- (C) $profit(A) = benefit(A) - cost(A)$.

Proof: Let $A' = A \cup \{s\}$.

$$\begin{aligned}
 c(A', B) &= cost(A) + benefit(B) \\
 &= cost(A) - benefit(A) + benefit(A) + benefit(B) \\
 &= -profit(A) + C \\
 &= C - profit(A)
 \end{aligned}$$

■

19.2.0.18 Correctness of Reduction contd

We have shown that if (A, B) is an s - t cut in G with finite capacity then

- (A) $A - \{s\}$ is a valid set of projects
- (B) $c(A, B) = C - profit(A - \{s\})$

Therefore a *minimum* s - t cut (A^*, B^*) gives a *maximum* profit set of projects $A^* - \{s\}$ since C is fixed.

Question: How can we use ∞ in a real algorithm?

Set capacity of ∞ arcs to $C + 1$ instead. Why does this work?

19.3 Extensions to Maximum-Flow Problem

19.3.0.19 Lower Bounds and Costs

Two generalizations:

- (A) flow satisfies $f(e) \leq c(e)$ for all e . suppose we are given *lower bounds* $\ell(e)$ for each e . can we find a flow such that $\ell(e) \leq f(e) \leq c(e)$ for all e ?
- (B) suppose we are given a cost $w(e)$ for each edge. cost of routing flow $f(e)$ on edge e is $w(e)f(e)$. can we (efficiently) find a flow (of at least some given quantity) at minimum cost?

Many applications.

19.3.0.20 Flows with Lower Bounds

Definition 19.3.1. A flow in a network $G = (V, E)$, is a function $f : E \rightarrow \mathbb{R}^{\geq 0}$ such that

- (A) **Capacity Constraint:** For each edge e , $f(e) \leq c(e)$
- (B) **Lower Bound Constraint:** For each edge e , $f(e) \geq \ell(e)$
- (C) **Conservation Constraint:** For each vertex v

$$\sum_{e \text{ into } v} f(e) = \sum_{e \text{ out of } v} f(e)$$

Question: Given G and $c(e)$ and $\ell(e)$ for each e , is there a flow?
As difficult as finding an s - t maximum-flow without lower bounds!

19.3.0.21 Regular flow via lower bounds

Given usual flow network G with source s and sink t , create lower-bound flow network G' as follows:

- (A) set $\ell(e) = 0$ for each e in G
- (B) add new edge (t, s) with lower bound v and upper bound ∞

Claim 19.3.2. *There exists a flow of value v from s to t in G if and only if there exists a feasible flow with lower bounds in G' .*

Above reduction show that lower bounds on flows are naturally related to **circulations**.
With lower bounds, cannot guarantee acyclic flows from s to t .

19.3.0.22 Flows with Lower Bounds

- (A) Flows with lower bounds can be reduced to standard maximum flow problem. See text book. Reduction goes via circulations.
- (B) If all bounds are integers then there is a flow that is integral. Useful in applications.

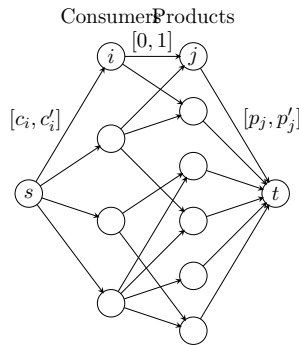
19.3.1 Survey Design

19.3.1.1 Application of Flows with Lower Bounds

- (A) Design survey to find information about n_1 products from n_2 customers.
- (B) Can ask customer questions only about products purchased in the past.
- (C) Customer can only be asked about at most c'_i products and at least c_i products.
- (D) For each product need to ask at east p_i consumers and at most p'_i consumers.

19.3.1.2 Reduction to Circulation

- (A) include edge (i, j) is customer i has bought product j
- (B) Add edge (t, s) with lower bound 0 and upper bound ∞ .
 - (A) Consumer i is asked about product j if the integral flow on edge (i, j) is 1



19.3.1.3 Minimum Cost Flows

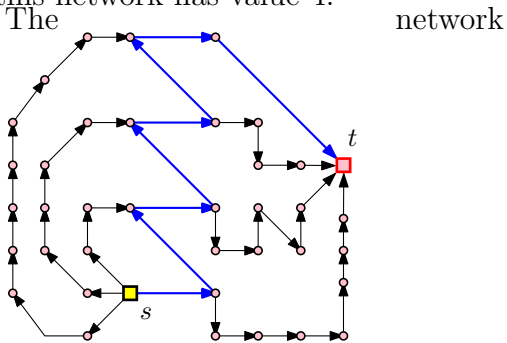
- (A) **Input:** Given a flow network G and also edge costs, $w(e)$ for edge e , and a flow requirement F .
- (B) **Goal;** Find a *minimum cost* flow of value F from s to t
 Given flow $f : E \rightarrow R^+$, cost of flow = $\sum_{e \in E} w(e)f(e)$.

19.3.1.4 Minimum Cost Flow: Facts

- (A) problem can be solved efficiently in polynomial time
- (A) $O(nm \log C \log(nW))$ time algorithm where C is maximum edge capacity and W is maximum edge cost
- (B) $O(m \log n(m + n \log n))$ time strongly polynomial time algorithm
- (B) for integer capacities there is always an optimum solutions in which flow is integral

19.3.1.5 How much damage can a single path cause?

Consider the following network. All the edges have capacity 1. Clearly the maximum flow in this network has value 4.



Why removing the shortest path might ruin everything

- (A) However... The shortest path between s and t is the blue path.
- (B) And if we remove the shortest path, s and t become disconnected, and the maximum flow drop to 0.