EECS 336: Introduction to Algorithms Lecture 7 network flow, reduction, bipartite matching

Reading: 7.1,7.5

Last time:

• Interval Pricing

Today:

- Reductions
- Network flow
- Bipartite matching

Reductions

"to solve problem B given solution to problem A, transform instances from problem B into instances of A, solve, transform solution back"

Problem A: Network Flow

"given a network with bandwidth constraints on links, how much data can we send from source to sink"

Def: a flow graph G = (V, E) is a directed graph with:

- c(e) =capacity of edge e
- $s \in V$ is source.
- $t \in V$ is \mathbf{sink} .

Def: a flow f in G is an assignment of flow to edges "f(e)" satisfying:

- capacity: $\forall e, f(e) \leq c(e)$
- conservation: $\forall v \neq s, t,$ $\sum_{e \text{ into } v} f(e) = \sum_{e \text{ out of } v} f(e)$

Def: the **value** of a flow is

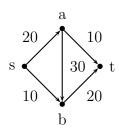
$$|f| = \sum_{e \text{ out of } s} f(e) = \sum_{e \text{ into } t} f(e)$$

Problem: Network Flow

input: flow graph $G, s, t, c(\cdot)$.

output: flow f with maximum value.

Example:



Max flow = 30.

Theorem 1: there is an algorithm to compute the max flow in polynomial time.

Theorem 2: if capacities are integral, then max flow is integral (on each edge).

Problem B: bipartite matching

Def: G = (V, E) is a bipartite if exists partitioning of V into A and B s.t.,

- $u, v \in A \Rightarrow (u, v) \notin E$,
- $u, v \in B \Rightarrow (u, v) \notin E$,

Recall: a **matching** is a set of edges $M \subseteq E$ each node is connected by at most one edge in M

- a **perfect** matching is one where all nodes are connected by exactly one edge.
- a **maximum** matching is one with maximum cardinality.

Problem: bipartite matching

input: bipartite graph G = (A, B, E)

output: a maximum matching M.

Reducing bipartite matching to Runtime max flow

 $T_{\text{matching}}(n, m) = O(n + m) + T_{\text{max flow}}(n, m)$

"use max flow alg to solve bipartite matching."

Steps:

- 1. convert matching instance into flow instance.
- 2. run flow alg flow instance.
- 3. convert flow soln to matching soln with same value.
- 4. prove flow soln optimal iff matching soln optimal.
 - (a) (convert flow soln to matching soln with same value)
 - (b) convert matching soln to flow soln with same value.

Step 1:

- (a) connect s to each $v \in A$ with capacity 1.
- (b) connect t to each $u \in B$ with capacity 1.
- (c) set capacity of each edge $e \in E$ to 1.

Step 2: compute (integral) max flow f

Step 3: matching is $M = \{e \in E \ : \ f(e) = 1\}$

- $\bullet \ |M| = |f|$
- (capacity constraints imply matching)

Step 4: Proof:

- any matching M' can be turned into a flow f' with |f'| = |M'|
 (send form s to each matched edge to t one unit of flow)
- any integral flow f' can be turned into a matching M' with |f'| = |M'| (Step 3)
- \Rightarrow size of output matching = value of max flow = size of max matching.

Reductions

Def: \underline{Y} reduces to \underline{X} in polynomial time (notation: $\underline{Y} \leq_P X$ if any instance of \underline{Y} can be solved in a polynomial number of computational steps and a polynomial number of calls to black-box that solves instances of X.

Note: to prove correctness of general reduction, must show that correctness (e.g., optimality) of algorithm for X implies correctness of algorithm for Y.

Def: one-call reduction maps instance of Y to instance of X, solution of Y to solution of X. (also called a Karp reduction)

Note: a one-call reduction gives two algorithms:

- (a) construction of X^Y instance from Y instance.
- (b) construction of Y solution from X^Y solution (with same value).

Note: the proof of correctness of a one-call reduction gives one algorithm:

(c) construction of X^Y solution from Y solution (with same value).

(Only need to consider X^Y instances not general X instances.)

Theorem: reduction from "(a) and (b)" is correct if (a), (b), and (c) are correct.

Proof:

- for instance y of Y, let instance $x^y of X^Y$ be outcome of (a).
- (b) correct \Rightarrow OPT $(y) \ge$ OPT (x^y) .
- (c) correct $\Rightarrow OPT(x^y) \ge OPT(y)$.
- $\Rightarrow \text{OPT}(y) = \text{OPT}(x^y)$
- \Rightarrow output of reduction has value OPT(y).