1 Maximum Flow

1.1 Definitions

Tarjan: Data Structures and Network Algorithms Ford and Fulkerson, Flows in Networks, 1962 (paper 1956) Ahuja, Magnanti, Orlin Network Flows. Problem: do min-cost. Problem: in a graph, find a flow that is feasible and has maximum value. Directed graph, edge capacities u(e) or u(v, w). Why not c? reserved for costs, later. source s, sink t

Goal: assign a *flow* value to each edge: (net flow form)

- skew symmetry: f(v, w) = -f(w, v)
- conservation: $\sum_{w} f(v, w) = 0$ unless v = s, t
- capacity: $f(e) \le u(e)$ (flow is feasible/legal)

Alternative formulation: (gross flow form)

- conservation: $\sum_{w} f(v, w) = 0$ unless v = s, t
- capacity: $0 \le f(e) \le u(e)$ (flow is feasible/legal)

Equivalence: second formulation has "gross flow" g, first has "net flow" f(v, w) = g(v, w) - g(w, v). To go other way, sign of f defines "direction" of flow in g. We'll focus on net flow model for now.

Flow value $|f| = \sum_{w} f(s, w)$ (in net model).

Water hose intuition. Also routing commodities, messages under bandwidth constraints, etc. Often "per unit time" flows/capacities.

Maximum flow problem: find flow of maximum value.

Path decomposition (another picture):

- claim: any s-t flow can be decomposed into paths with quantities
- proof: induction on number of edges with nonzero flow
- if s has out flow, find an s-t path (why can we? conservation) and kill
- if some vertex has outflow, find a cycle and kill
- corollary: flow into t equals flow out of s (global conservation)

Cuts:

- partition of vertices into 2 groups
- s-t-cut if one has s, other t
- represent as (S, \overline{S}) or just S

- f(S) =net flow leaving S
- lemma: for any s-t cut, f(S) = |f| (all cuts carry same flow)
 - suppose move v from \overline{S} to S.
 - adds f(v, S) to crossing flow value (no longer subtracting from net)
 - also add $f(v, \overline{S})$ (adding to net flow value)
 - total change: $f(v, S \cup \overline{S}) = 0$

Flows versus cuts:

- Deduce: $|f| \le u(S) = \sum_{e \in S \times \overline{S}} u(e).$
- in other words, max-flow \leq minimum s=t cut value.
- soon, we'll see equal (duality)
- first, need more machinery.

Residual network.

- Given: flow f in graph G
- define G_f to have capacities $u'_e = u_e f_e$
- if f feasible, all capacities positive
- Since f_e can be negative, some residual capacities **grow**
- Suppose f' is a feasible flow in G_f
- then f + f' is feasible flow in G of value f + f'
 - flow
 - feasible
- Suppose f' is feasible flow in G
- then f' f is feasible flow in G_f (value |f'| |f|)
- corollary: max-flows in G correspond to max-flows in G_f
- Many algorithms for max-flow:
 - find some flow f
 - recurse on G_f

How can we know a flow is maximum?

- check if residual network has 0 max-flow
- augmenting path: s-t path of positive capacity in G_f

• if one exists, not max-flow

Max-flow Min-cut

- Equivalent statements:
 - -f is max-flow
 - no augmenting path in G_f

-|f| = u(S) for some S

Proof:

- if augmenting path, can increase f
- let S be vertices reachable from S in G_f . All outgoing edges have f(e) = u(e)
- since $|f| \leq u(S)$, equality implies maximum

1.2 Problem Variants

Multiple sinks Edge lower bounds Bipartite matching. Vertex capacities (HW).

1.3 Applications

Matrix Rounding (flow with lower bounds). $P||r_j, pmtn||C_{max}$

1.4 Algorithms

1.4.1 Augmenting paths

Can always find one. If capacities integral, always find an integer.

- So terminate
- Running time O(mf)
- Lemma: flow integrality
- Polynomial for unit-capacity graphs
- Also terminate for rationals (but not polynomial)
- might not terminate for reals.

1.4.2 max capacity augmenting path

How find one? Running time:

- recall flow decomposition bound
- Get 1/m of flow each time.
- So $m \log f$ flows suffice for integer f

weakly vs. strongly polynomial bounds Works for rational capacities too.

1.5 If little time, scaling

Algorithm: repeatedly

- Shift in one bit at a time
- Run plain augmenting paths

Analysis:

- before bit shift, some cut has capacity 0
- after bit shift, cut has capacity at most m
- so *m* augmentations suffice.

Running time: $O(m \log U)$. Discuss relation to max capacity algorithm.

1.5.1 Shortest Augmenting Path

Strongly polynomial.

Lemma: (s.i) and (i, t) distance nondecreasing.

- Among i that got closer to s
- Consider closest to s (after change)
- i has parent j on new shortest path
- j didn't get closer to s
- so (j, i) path got shorter
- so didn't used to have residual (j, i) edge
- so flow went from i to j
- so j was farther than i from s

- now they swapped places
- but *j* didn't get closer!
- so i must be farher—contra.

Lemma: at most mn/2 augmentations.

- Consider edge (i, j) saturated by augmenting path
- Before used again, must push flow on (j, i)
- In first aug, i was closer than j to s
- In next, j was closer than i
- Since no distances go down, must have increased distance of *i*.
- only happens n times per edge

Running time: $O(m^2n)$.

1.5.2 Blocking Flows

definition of blocking flow dinic's alg (alg + proof that s-t dist increases) at the very end I stated BF time bounds and said they'd see more next time