COS 226 Lecture 22: Mincost Flow

MAXFLOW: assign flows to edges that

- equalize inflow and outflow at every vertex
- maximize total flow through the network

MINCOST MAXFLOW: find the BEST maxflow

Mincost maxflow is important for two primary reasons

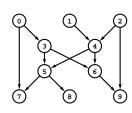
- it is a GENERAL PROBLEM-SOLVING MODEL
 - solves (through reduction) numerous practical problems
- it is TRACTABLE and PRACTICAL
 - we know fast algorithms that solve mincost flow problems
 - basic data structures play a critical role

One step closer to a single ADT for combinatorial problems

Distribution problem

| SUPPLY vertices (produce goods) | supply | channels |
|---|--------------|----------|
| DEMAND vertices (consume goods) | 0: 3 | 0-3: 2 |
| PERMIP VERMEES (CONSume goods) | 1: 4 | 0-7: 1 |
| DISTRIBUTION points (transfer goods) | 2: 6 | 1-4: 5 |
| | distribution | 2-4: 3 |
| | 3 | 2-9: 1 |
| | 4 | 3-5: 3 |
| Feasible flow problem | 5 | 3-6: 4 |
| • | 6 | 4-5: 2 |
| Can we make supply to meet demand? | demand | 4-6: 1 |
| | 7: 7 | 5-7: 6 |
| Distribution problem | 8: 3 | 5-8: 3 |
| Add costs, find the lowest-cost way | 9: 4 | 6-9: 4 |

- Ex: Walmart
- Ex: McDonald's



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THM: Feasible flow reduces to maxflow

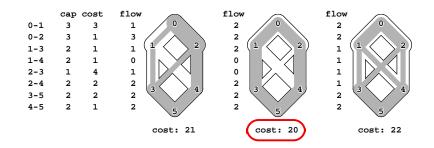
THM: Distribution reduces to mincost maxflow

Proof: Add source to provide supply, sink to take demand

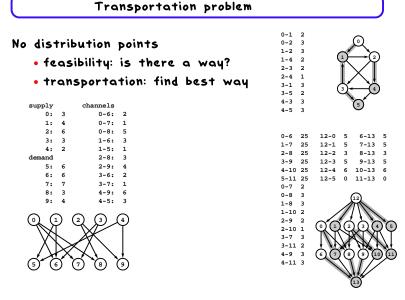
Mincost flow

Add COST to each edge in a flow network FLOW COST: sum of cost*flow over all edges

Maxflows have different costs



MINCOST FLOW: find a minimal-cost maxflow



Seems easier, but that is not the case (!) THM: Maxflow reduces to maxflow for acyclic networks THM: Transportation reduces to mincost maxflow

ST TIOW

Mincost flow reductions

SHORTEST PATHS MAXFLOW DISTRIBUTION and TRANSPORTATION

ASSIGNMENT Minimal weight matching in weighted bipartite graph

MAIL CARRIER Find a cyclic path that includes each edge AT LEAST once

SCHEDULING (example) Given a sport's league schedule, which teams are eliminated?

POINT MATCHING Given two sets of N points, find minimal-distance pairing

ALL of these problems reduce to mincost flow 22.5

Cycle canceling

RESIDUAL NETWORK

for each edge in original network

flow f in edge u-v with capacity c and cost x

define TWO edges in residual network

- FORWARD edge: capacity c-f and cost x in edge u-v
- BACKWARD edge: capacity f and cost -x in edge v-u

THM: A maxflow is mincost iff

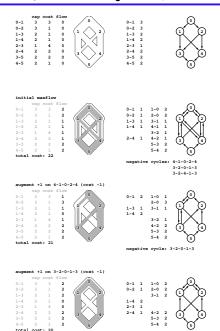
there are NO negative-cost cycles in its residual network

GENERIC method for solving mincost flow problems:

start with ANY maxflow REPEAT until no negative cycles are left • increase the flow along ANY negative cycle

Implementation: use Bellman-Ford to find negative cycles^{22.6}

Cycle canceling example



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Cycle canceling implementation

```
void addflow(link u, int d)
{ u->flow += d; u->dup->flow -=d; }
int GRAPHmincost(Graph G, int s, int t)
{ int d, x, w; link u, st[maxV];
    GRAPHmaxflow(G, s, t);
    while ((x = GRAPHnegcycle(G, st)) != -1)
        {
            u = st[x]; d = Q;
            for (w=u->dup->v; w != x; w=u->dup->v)
            { u = st[w]; d = ( Q > d ? d : Q ); }
            u = st[x]; addflow(u, d);
            for (w=u->dup->v; w != x; w=u->dup->v)
            { u = st[x]; addflow(u, d);
            for (w=u->dup->v; w != x; w=u->dup->v)
            { u = st[w]; addflow(u, d); }
            for (w=u->dup->v; w != x; w=u->dup->v)
            { u = st[w]; addflow(u, d); }
            for (w=u->dup->v; w != x; w=u->dup->v)
            { u = st[w]; addflow(u, d); }
            }
        return GRAPHcost(G);
        }
    }
}
```

Cycle canceling analysis

No need to compute initial maxflow

use dummy edge from sink to source that carries maxflow

THM: Generic cycle canceling alg takes O(VE^2CM) time Proof:

- each edge has at most capacity C and cost M
- total cost could be ECM
- each augment reduces cost by at least i
- Bellman-Ford takes O(VE) time

There exist O(VE^2log^2 V) cycle-canceling implementations

mincost maxflow is therefore TRACTABLE

EXTREMELY pessimistic UPPER bounds

- not useful for predicting performance in practice
- algs that achieve such bounds would be useless
- algs are typically fast on practical problems

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Network simplex concepts (continued)

REDUCED COST (reweighted edge cost)

• c*(u, v) = c(u, v) - (phi(u) - phi(v))

- VALID vertex potentials for a spanning tree
 - all tree edges have reduced cost o

ELIGIBLE EDGE

nontree edge that creates negative cycle with tree edges

THM: A nontree edge is eligible iff it is either

- a full edge with positive reduced cost, or
- an empty edge with negative reduced cost Proof:
 - cycle cost equals cycle reduced cost
 - edge cost is negative of cycle reduced cost (since reduced costs of tree edges are all zero)

THEREFORE, it is easy to identify eligible edges

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Network simplex algorithm

An implementation of the cycle-canceling algorithm

Identify negative cycles quickly by

- maintaining a tree data structure
- reweighting costs at vertices

Edge classification

- EMPTY
- FULL
- PARTIAL

FEASIBLE SPANNING TREE

Any spanning tree that contains all the partial edges

VERTEX POTENTIALS

- a set of vertex weights (vertex-indexed array phi)
 - 22.10

Network simplex algorithm

still a generic algorithm for the mincost flow problem

start with ANY feasible spanning tree REPEAT until no eligible edges are left

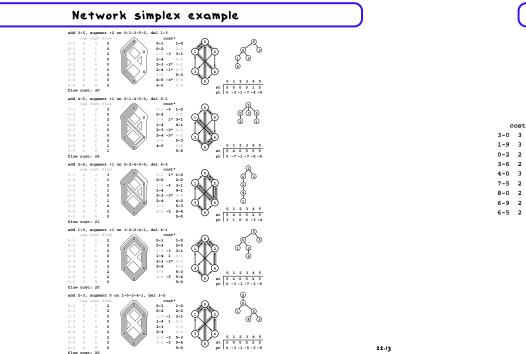
- ensure that vertex potentials are valid
- add to the tree an eligible edge
- increase the flow along the negative cycle formed
- remove from the tree an edge that is filled or emptied

Problem: could have zero flow on cycle

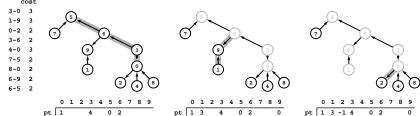
THM: IF the algorithm terminates, it computes a maxflow

Implementation challenges

- cope with zero-flow cycles
- strategy to choose eligible edges
- data structure to represent tree



Computing vertex potentials (example)



Feasible spanning tree data structure

Operations to support

- compute valid vertex potentials
- find cycle created by nontree edge
- replace tree edge by nontree edge

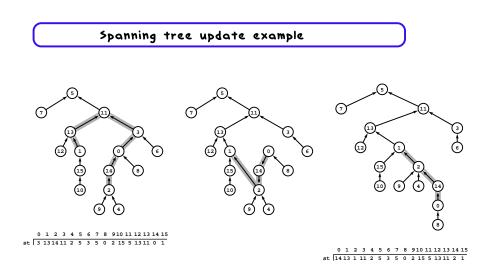
use PARENT-LINK representation!

to compute vertex potentials

- start with root at potential o
- for each vertex

follow parent links to vertex with known potential (recursively) set each vertex potential on path to make reduced edge costs o

- to follow cycle created by nontree edge u-v
 - follow parent links from each to their LCA
- to delete nontree edge that fills or empties
 - REVERSE the parent links from u or v



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Network simplex basic implementation

```
#define R(u) u->cost - phi[u->v]+phi[u->dup->v]
int GRAPHmincost(Graph G, int s, int t)
{ int v; link u, x, st[maxV];
  GRAPHinsertE(G, EDGE(t, s, M, 0, C));
  initialize(G, s, t, st);
  for (valid = 1; valid++; )
  {
     for (v = 0; v < G ->V; v++)
       phi[v] = phiR(st, v);
     for (v = 0, x = G->adj[v]; v < G->V; v++)
       for (u = G->adj[v]; u!=NULL; u = u->next)
            if (R(u) < R(x)) x = u;
     if (R(x) == 0) break;
     update(st, augment(st, x), x);
  }
  return GRAPHcost(G);
}
```

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Network simplex variations

OBJECTIVES

- guarantee terminimation
- reduce number of iterations
- reduce cost per iteration

Eligible edge selection strategies

- random
- find next
- queue of eligible edges
- Lazy vertex potential calculation

Tree representations

triply-linked, threaded

Guided by practical performance, not worst-case bounds

• DATA STRUCTURES are the key to good performance

Different implementations for different reductions??

BOTTOM LINE

• accessible code for powerful problem-solving model ^{22.18}