

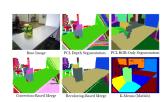
Max-Flow and Min-Cut

Two important algorithmic problems, which yield a beautiful duality

Myriad of non-trivial applications, it plays an important role in the optimization of many problems:

Network connectivity, airline schedule (extended to all means of transportation), image segmentation, bipartite matching, distributed computing, data mining,







Flow Networks

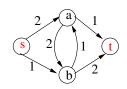
Network diagraph G = (V, E) s.t. it has

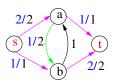
- ▶ source vertex $s \in V$
- ▶ sink vertex $t \in V$
- ▶ edge capacities $c: E \to \mathbb{R}^+ \cup \{0\}$

Flow $f: V \times V \to \mathbb{R}^+ \cup \{0\}$ s.t. Kirchoff's laws:

- $\forall (u,v) \in E, \ 0 \leq f(u,v) \leq c(u,v),$
- ► (Flow conservation) $\forall v \in V \{s, t\}$, $\sum_{u \in V} f(u, v) = \sum_{z \in V} f(v, z)$
- ► The value of a flow

$$|f| = \sum_{v \in V} f(s, v) = f(s, V) = f(V, t).$$

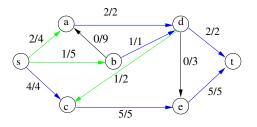




Value |f|=3

The Maximum flow problem

INPUT: Given a flow network (G = (V, E), s, t, c) QUESTION: Find a flow of maximum value on G.



The value of the max-flow is 7 = 4 + 1 + 2 = 5 + 2.

Notice: Although the flow exiting s is not maximum, the flow going into t is maximum (= max. capacity).

Therefore the total flow is maximum.

The s-t cut

Given (G = (V, E), s, t, c) a s - t cut is a partition of $V = S \cup T$ $(S \cap T = \emptyset)$, with $s \in S$ and $t \in T$.

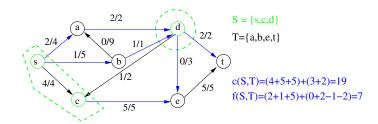
The flow across the cut:

$$f(S) = \sum_{u \in S} \sum_{v \in T} f(u, v) - \sum_{v \in S} \sum_{u \in T} f(v, u).$$

The capacity of the cut: $c(S) = \sum_{u \in S} \sum_{v \in T} c(u, v)$

capacity of cut (S, T) = sum of weights leaving S.

Notice because of the capacity constrain: $f(S) \le c(S)$



The s-t cut

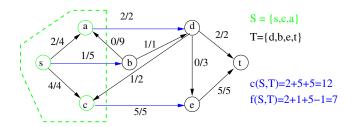
Given (G = (V, E), s, t, c) a s - t cut is a partition of S, T of V (i.e. $V = S \cup T$ and $S \cap T = \emptyset$), with $s \in S$ and $t \in T$.

The flow across the cut:

$$f(S) = \sum_{u \in S} \sum_{v \in T} f(u, v) - \sum_{v \in T} \sum_{u \in S} f(v, u).$$

The capacity of the cut: $c(S) = \sum_{u \in S} \sum_{v \in T} c(u, v)$

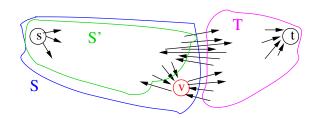
Notice because of the capacity constrain: $f(S) \le c(S)$



Notation

Given $v \in G$ and cut (S, T) and a $v \in S$, let $S' = S - \{v\}$. Then

- ▶ Denote f(S', T) flow between S' and T (without going by v). i.e. $f(S', T) = \sum_{u \in S'} \sum_{w \in T} f(u, w) - \sum_{w \in T} \sum_{u \in S'} f(w, u)$ with $(u, w) \in E$ and $(u, w) \in E$,
- ▶ denote f(v, T) flow $v \to T$ i.e. $f(v, T) = \sum_{u \in T} f(v, u)$,
- ▶ denote f(T, v) flow $T \to v$ i.e. $f(T, v) = \sum_{u \in T} f(u, v)$,
- ▶ denote f(S', v) flow $S' \to v$ i.e. $f(S', v) = \sum_{u \in S'} f(u, v)$,
- ▶ denote f(v, S') flow $v \to S'$ i.e. $f(v, S') = \sum_{u \in S'} f(v, u)$,



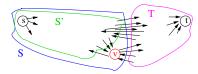
Any s - t cut has the same flow

Theorem

Given (G, s, t, c) the flow through any s - t cut (S, T) is f(S) = |f|.

Proof (Induction on |S|)

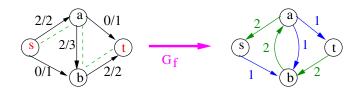
- If $S = \{s\}$ then f(S) = |f|.
- Assume it is true for $S' = S \{v\}$, i.e. f(S') = |f|. Notice f(S') = f(S', T) + f(S', v) - f(v, S'). Moreover from the flow conservation, f(S', v) + f(T, v) = f(v, S') + f(v, T) $\Rightarrow \underbrace{f(v, T) - f(T, v) = f(S', v) - f(v, S')}$
- ► Then f(S) = f(S', T) + f(v, T) f(T, v), using (*) f(S) = f(S') = |f|



Residual network

Given a network (G = (V, E), s, t, c) together with a flow f on it, the residual network, ($G_f = (V, E_f), c_f$) is the network with the same vertex set and edge set:

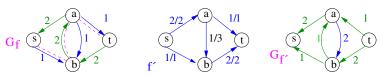
- if c(u,v) f(u,v) > 0 then $(u,v) \in E_f$ and $c_f(u,v) = c(u,v) f(u,v) > 0$ (forward edges), and
- ▶ if f(u, v) > 0 then $(v, u) \in E_f$ and $c_f(v, u) = f(u, v)$ (backward edges). i.e. there are f(u, v) units of flow we can undo, by pushing flow backward. Notice, if c(u, v) = f(u, v) then there is only a backward edge.
- ightharpoonup the c_f are denoted residual capacity.



Residual network: Augmenting paths

Given G = (V, E) and a flow f on G, an augmenting path P is any simple path in G_f (using forward and backward edges, but $P : s \leadsto t$).

Given $f: s \leadsto t$ in G and P in G_f define the bottleneck (P, f) to be the minimum residual capacity of any edge in P, with respect to f.



P: dotted line

Residual network: Augmenting paths

```
Given G = (V, E) and a flow f on G, an augmenting path P is
any simple path in G_f.
Given f s \to t in G and P in G_f define the bottleneck (P, f) to be
the minimum residual capacity of any edge in P.
  Augment(P, f)
  b=bottleneck (P, f)
  for each (u, v) \in P do
    if (u, v) is forward edge in G then
       Increase f(u, v) in G by b
    else
       Decrease f(u, v) in G by b
    end if
  end for
  return f
```

Residual network: Augmenting paths

Lemma

Consider f' = Augment(P, f), then f' is a flow in G.

Proof: We have to prove that (1) $\forall e \in E$, $0 \le f(e) \le c(e)$ and that $\forall v$ flow to v = flow out of v.

- ▶ Capacity law Forward edges $(u, v) \in P$ we increase f(u, v) by b, as $b \le c(u, v) f(u, v)$ then $f'(u, v) = f(u, v) + b \le c(u, v)$. Backward edges $(u, v) \in P$ we decrease f(v, u) by b, as $b \le f(v, u), f'(v, u) = f(u, v) b \ge 0$.
- ▶ Conservation law, $\forall v \in P$ given edges e_1, e_2 in P and incident to v, it is easy to check the 4 cases based whether e_1, e_2 are forward or backward edges.

Max-Flow Min-Cut theorem

Theorem: For any (G, s, t, c) the value of the max flow f^* is equal to the capacity of the min (S, T)-cut (over all s - t cuts in G)

$$f^* = \max\{|f|\} = \min_{\forall (S,T)} \{c(S,T)\}.$$

Ford and Fulkerson (1954); Peter Elias, Amiel Feinstein and Claude Shannon (1956) (in framework of information-theory).

Proof:

- ▶ For any s t cut (S, T) in $G \Rightarrow f^*(S) \leq c(S, T)$.
- ▶ If f^* in G is a max flow then G_{f^*} has no augmenting path $s \rightsquigarrow t$ so it is disconnected.

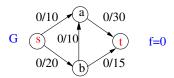
Let
$$S_s = \{v \in V | \exists s \leadsto v \text{ in } G_{f^*} \}$$
, then $(S_s, V - \{S_s\})$ is a $s - t$ cut in $G_{f^*} \Rightarrow \forall v \in S_s, u \in V - \{S_s\}, (v, u)$ is not a residual edges, so in $G(f^*(v, u)) = c(v, u)$, i.e. $c(S_s, V - \{S_s\}) = f^*(S_s, V - \{S_s\})$ in $G(S_s, V - \{S_s\})$ is a min-cut in $G(S_s, V - \{S_s\})$ in $G(S_s, V - \{S_s\})$ is a min-cut in $G(S_s, V - \{S_s\})$ in $G(S_s, V - \{S_s\})$ is a min-cut in $G(S_s, V - \{S_s\})$ in $G(S_s, V - \{S_s\})$ in $G(S_s, V - \{S_s\})$ is a min-cut in $G(S_s, V - \{S_s\})$ in $G(S_s, V - \{S_$

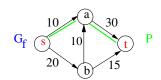
L.R. Ford, D.R. Fulkerson: Maximal flow through a network. Canadian J. of Math. 1956.





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Ford-Fulkerson(G, s, t, c)
for all (u, v) \in E let f(u, v) = 0
G_f = G
while there is an s - t path in G_f do
find a simple path P in G_f (use DFS)
f' = \operatorname{Augment}(f, P)
Update f to f'
Update G_f to G_{f'}
end while
return f
```



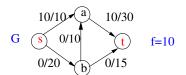


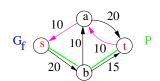
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Ford-Fulkerson(G, s, t, c)for all $(u, v) \in E$ let f(u, v) = 0 $G_f = G$ while there is an s - t path in G_f do find a simple path P in G_f (use DFS) $f' = \operatorname{Augment}(f, P)$ Update f to f'Update G_f to $G_{f'}$ end while return f



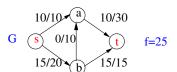


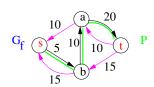
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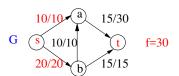


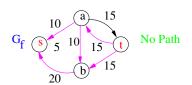
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Analysis of Ford Fulkerson

We are considering networks that initial flow and capacities are integers,

Lemma (Integrality invariant)

At every iteration of the Ford-Fulkerson algorithm, the flow values f(e) and the residual capacities in G_f are integers.

Proof: (induction)

- ▶ The statement is true before the **while** loop.
- ▶ Inductive Hypothesis: The statement is true after j iterations.
- ▶ iteration j+1: As all residual capacities in G_f are integers, then bottleneck $(P,f) \in \mathbb{Z}$, for the augmenting path found in iteration j+1. Thus the flow f' will have integer values \Rightarrow so will the capacities in the new residual graph. \Box

Corollary: Integrality theorem

Theorem (Integrality theorem)

There exists a max-flow f^* for which every flow value f^* is an integer.

Proof:

Since the algorithm terminates, the theorem follows from the integrality invariant lemma.

Analysis of Ford Fulkerson

Lemma

If f is a flow in G and f' is the flow after an augmentation, then |f| < |f'|.

Proof: Let P be the augmenting path in G_f . The first edge $e \in P$ leaves s, and as G has no incoming edges to s, e is a forward edge. Moreover P is simple \Rightarrow never returns to s. Therefore, the value of the flow increases in edge e.

Correctness of Ford-Fulkerson

Consequence of the Max-flow min-cut theorem.

Theorem

The flow returned by Ford-Fulkerson f^* is the max-flow.

Proof:

- ▶ For any flow f and s t cut (S, T) we have $|f| \le c(S, T)$.
- ▶ The flow f^* is such that $|f^*| = c(S^*, T^*)$, for some s t cut $(S^*, T^*) \Rightarrow f^*$ is the max-flow.

Therefore, for any (G, s, t, c) the value of the max s - t flow is equal to the capacity of the minimum s - t cut.

Analysis of Ford Fulkerson: Running time

Lemma

Let C be the min cut capacity (=max. flow value), Ford-Fulkerson terminates after finding at most C augmenting paths.

Proof: The value of the flow increases by ≥ 1 after each augmentation.

- ▶ The number of iterations is $\leq C$. At each iteration:
- ▶ We have to modify G_f , with $E(G_f) \le 2m$, to time O(m).
- ▶ Using DFS, the time to find an augmenting path P is O(n+m)
- ▶ Total running time is O(C(n+m)) = O(Cm)
- ▶ Is that polynomic?

П

Running time of Ford-Fulkerson

The number of iterations of Ford-Fulkerson could be $\Omega(C)$ As it is described Ford-Fulkerson can alternate C times between the blue and red paths if the figure.



C=1000000000 2000 million iteractions in a G with 4 vertices!!

Recall a pseudopolynomial algorithm is an algorithm that is polynomial in the unary encoding of the input.

Is there a polynomial time algorithm for the max-flow problem?

Dinic and Edmonds-Karp algorithm

J.Edmonds, R. Karp: *Theoretical improvements in algorithmic efficiency for network flow problems*. Journal ACM 1972.

Yefim Dinic: Algorithm for solution of a problem of maximum flow in a network with power estimation. Doklady Ak.N. 1970

Choosing a good augmenting path can lead to a faster algorithm. Use BFS to find shorter augmenting paths in G_f .







Using BFS on G_f we can find the shortest augmenting path P in O(m+n), independently of max capacity C.

Edmonds-Karp algorithm

Greedy type algorithm: Using BFS, hoose the augmenting path G_f with the smallest number of edges. Uses BFS to find the augmenting path at each G_f with fewer number of edges.

```
Edmonds-Karp(G, s, t, c)

For all e = (u, v) \in E let f(u, v) = 0

G_0 = G

while there is an s \rightsquigarrow t path in G_f

do

P = \mathsf{BFS}(G_f, s, t)

f' = \mathsf{Augment}(f, P)

Update G_f = G_{f'} and f = f'

end while

return f
```



The BFS in EK will choose: → or →

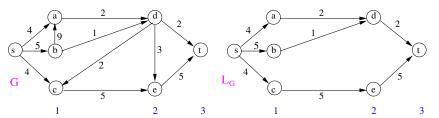
Level graph

Given G = (V, E), s, define $L_G = (V, E_G)$ to be its the level graph by:

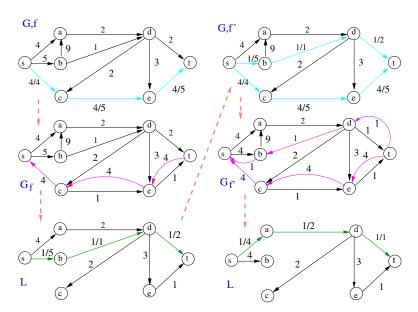
- ▶ $\ell(v)$ = number of edges in shortest path $s \rightsquigarrow v$ in G,
- ▶ $L_G = (V, E_G)$ is the subgraph of G that contains only edges $(v, w) \in E$ s.t. $\ell(w) = \ell(v) + 1$.

Notice:

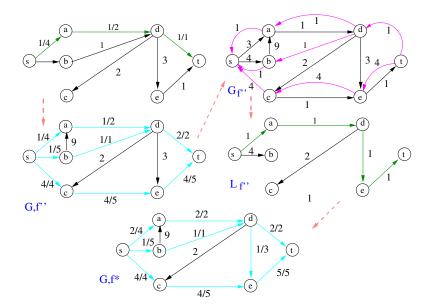
- ▶ Using BFS we can compute L_G in O(n+m)
- ▶ Important property: P is a shortest $s \rightsquigarrow t$ in G iff P is an $s \rightsquigarrow t$ path in L_G .



The working of the EK algorithm



The working of the EK algorithm



EK algorithm: Properties

Lemma

Throughout the algorithm, the length of the shortest path never decreases.

Proof:

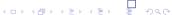
- ▶ Let *f* and *f'* be the flow before and after a shortest path augmentation
- ▶ let L and L' be the levels graphs of G_f and $G_{f'}$.
- ▶ Only back edges added to $G_{f'}$.

Lemma

After at most m shortest path augmentations, the length of P is monotonically increasing.

Proof:

- ▶ The bottleneck edge is deleted from *L* after each augmentation.
- No new edge is added to L until length of shortest path strictly increases



Complexity of Edmonds-Karp algorithm

Using the the previous lemmas, we prove

Theorem

The EK algorithms runs in $O(m^2n)$ steps. Therefore it is a polynomial time algorithm.

Proof:

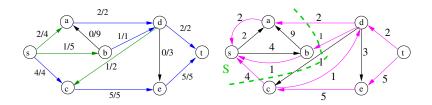
- ▶ Need time O(m+n) to find the augmenting path using BFS.
- ▶ Need O(m) augmentations for paths of length k.
- ▶ Every augmentation path is simple $\Rightarrow 1 \le k \le n \Rightarrow O(nm)$ augmentations

Finding a min-cut

Given (G, s, t, c) to find a min-cut:

- 1. Compute the max-flow f^* in G.
- 2. Obtain G_{f^*} .
- 3. Find the set $S = \{v \in V | s \rightsquigarrow v\}$ in G_{f^*} .
- 4. Output the cut $(S, V \{S\}) = \{(v, u) | v \in S \text{ and } u \in V \{S\}\} \text{ in } G.$

The running time is the same than the algorithm to find the max-flow.



The max-flow problems: History

- ▶ Ford-Fulkerson (1956) O(mC), where C is max capacity.
- ▶ Dinic (1970) (blocking flow) $O(n^2m)$
- ▶ Edmond-Karp (1972) (shortest augmenting path) $O(nm^2)$
- ▶ Karzanov (1974), $O(n^2m)$ Goldberg-Tarjant (1986) (push re-label preflow + dynamic trees) $O(nm \lg(n^2/m))$ (for this time uit uses parallel implementation)
- ► King-Rao-Tarjan (1998) $O(nm \log_{m/n \lg n} n)$.
- ▶ J. Orlin (2013) O(nm) (clever follow up to KRT-98)

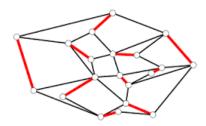
So: Maximum flows can be computed in O(nm) time!

MAXIMUM MATCHING problem

Given an undirected graph G = (V, E) a subset of edges $M \subseteq E$ is a matching if each node appears at most in one edge (a node may not appear at all).

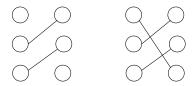
A perfect matching in G is a matching M such that |M|=|V|/2

The $MAXIMUM\ MATCHING\ problem$: Given as input a graph G, find a matching with maximum cardinality.

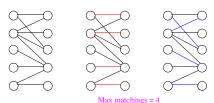


Maximum matching in graphs bipartite

A graph G=(V,E) is said to be bipartite if V can be partite in L and R, $L \cup R = V$, $L \cap R = \emptyset$, such that every $e \in E$ connects L with R.



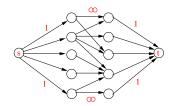
The MAXIMUM MATCHING BIPARTITE GRAPH problem: Given as input a bipartite graph $G = (L \cup R, E)$ with 2n vertices, find a maximum matching.



MAXIMUM MATCHING: flow formulation

Given a bipartite graph $G = (L \cup R, E)$ construct $\hat{G} = (\hat{V}, \hat{E})$:

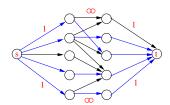
- ▶ Add vertices s and t: $\hat{V} = L \cup R \cup \{s, t\}$.
- Add directed edges s → L with capacity 1. Add directed edges R → t with capacity 1.
- ▶ Direct the edges E from L to R, and give them capacity ∞ .
- $\hat{E} = \{s \to L\} \cup E \cup \{R \to t\}.$



MAXIMUM MATCHING: flow formulation

Given a bipartite graph $G = (L \cup R, E)$ construct $\hat{G} = (\hat{V}, \hat{E})$:

- ▶ Add vertices s and t: $\hat{V} = L \cup R \cup \{s, t\}$.
- Add directed edges s → L with capacity 1. Add directed edges R → t with capacity 1.
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Maximum matching algorithm: Analysis

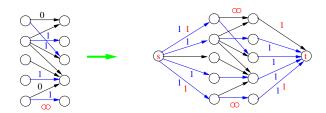
Theorem

Max flow in \hat{G} =Max bipartite matching in G.

Proof ≤

Given a matching M in G with k-edges, consider the flow F that sends 1 unit along each one of the k paths.

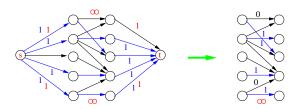
f is a flow and has value k.



Maximum matching algorithm: Analysis

Max flow ≤Max bipartite matching

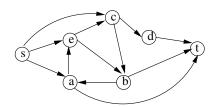
- ▶ If there is a flow f in \hat{G} , |f| = k, as capacities are $\mathbb{Z}^* \Rightarrow$ an integral flow exists is.
- ▶ Consider the cut $C = (\{s\} \cup L, R \cup \{t\})$ in \hat{G} .
- ▶ Let F be the set of edges in C with flow=1, then |F| = k.
- ▶ Each node in L is in at most one $e \in F$ and every node in R is in at most one head of an $e \in F$
- ▶ Therefore, exists a bipartite matching F in G with $|F| \le |f|$ □



DISJOINT PATH problem

Given a digraph (G = (V, E), s, t), a set of paths is edge-disjoint if their edges are disjoint (although them may go through some of the same vertices)

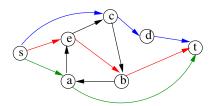
DISJOINT PATH problem: Given as input G, s, t, find the max number of edge disjoint paths $s \rightsquigarrow t$



DISJOINT PATH problem

Given a digraph (G = (V, E), s, t), a set of paths is edge-disjoint if their edges are disjoint (although them may go through some of the same vertices)

The disjoint path problem given G, s, t find the max number of edge disjoint paths $s \rightsquigarrow t$

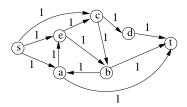


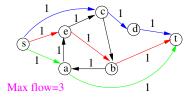
DISJOINT PATH: Max flow formulation

Assign unit capacity to every edge

Theorem

The max number of edge disjoint paths $s \rightsquigarrow t$ is equal to the max flow value





DISJOINT PATH: Proof of the Theorem

= 1.

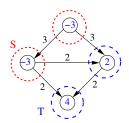
Number of disjoints paths \leq max flow If we have k edge-disjoints paths $s \rightsquigarrow t$ in G then making f(e) = 1 for each e in a path, we get a flow = k Number of disjoints paths \geq max flow If max flow $|f^*| = k \Rightarrow \exists \ 0\text{-}1 \text{ flow } f^* \text{ with value } k \Rightarrow \exists k \text{ edges } (s,v) \text{ s.t. } f(s,v) = 1, \text{ by flow conservation we can extend to } k \text{ paths } s \rightsquigarrow t, \text{ where each edge is a path carries flow}$

If we have an undirected graph, with two distinguised nodes u, v, how would you apply the max flow formulation to solve the problem of finding the max number of disjoint paths between u and t?

Circulation with demands

Given a graph G = (V, E) with capacities c in the edges, such that each $v \in V$ is associate with a demand d(v), where

- ▶ If $d(v) > 0 \Rightarrow v$ is a sink, v can receive d(v) units of flow more than it sends.
- ▶ If $d(v) < 0 \Rightarrow v$ is a source, v can send d(v) units of flow more than it receives.
- ▶ If d(v) = 0 then v is neither a source or a sink.
- ► Define *S* to be the set of sources and *T* the set of sinks.

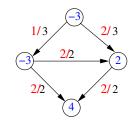


Circulation with demands problem

Given G = (V, E) with $c \ge 0$ and $\{d(v)\}_{v \in V}$, define a circulation as a function $f : E \to \mathbb{R}^+$ s.t.

- 1. capacity: For each $e \in E$, $0 \le f(e) \le c(e)$,
- 2. conservation: For each $v \in V$,

$$\sum_{(u,v)\in E} f(u,v) - \sum_{(v,z)\in E} f(v,z) = d(v).$$



Circulation with demands feasibility problem: Given G = (V, E) with $c \ge 0$ and $\{d(v)\}_{v \in V}$, does it exists a feasible circulation? Feasible circulation: a function f on G with $c \ge 0$ and $\{d(v)\}_{v \in V}$, such that it satisfies (1) and (2)?

Circulation with demands problem

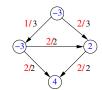
Notice that if f is a feasible circulation, then

$$\sum_{v \in V} d(v) = \sum_{v \in V} \left(\underbrace{\sum_{(u,v) \in E} f(u,v)}_{\text{edges to } v} - \underbrace{\sum_{(v,z) \in E} f(v,z)}_{\text{edges out of } v} \right).$$

Notice $\sum_{v \in V} d(v) = 0$, so we have,

So If there is a feasible circulation with demands $\{d(v)\}_{v\in V}$, then $\sum_{v\in V}d(v)=0$.

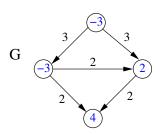
Therefore as $S = \{v \in V | d(v) > 0\}$ and $T = \{v \in V | d(v) < 0\}$, we can define $D = -\sum_{v \in S} d(v) = \sum_{v \in T} d(v)$.

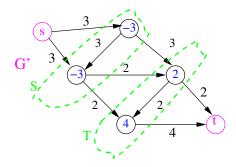


Circulation with demands: Max-flow formulation

Extend G = (V, E) to G' = (V', E') by

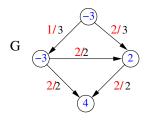
- Add new source s and sink t.
- ▶ For each $v \in S$ (d(v) < 0) add (s, v) with capacity -d(v).
- ▶ For each $v \in T$ (d(v) > 0) add (v, s) with capacity d(v).

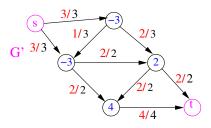




Analysis

- 1.- Every flow $f: s \leadsto t$ in G' must be $|f| \le D$ The capacity $c(\{s\}, V) = D \Rightarrow$ by max-flow min-cut Thm. any max-flow f in G', $|f| \le D$.
- 2.- If there is a feasible circulation f with $\{d(v)\}_{v \in V}$ in G, then we have a max-flow $f: s \leadsto t$ in G with |f| = D $\forall (s,v) \in E', \ f'(s,v) = -d(v)$ and $\forall (u,t) \in E', \ f'(u,t) = d(v)$.

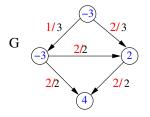


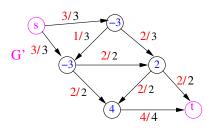


Analysis

- **3.-** If there is a flow $f': s \rightsquigarrow t$ in G' with |f| = D:
 - 1. $\forall (s, v) \in E'$ and $\forall (u, t) \in E'$ must be saturated \Rightarrow if we delete these edges in G' we obtain a circulation f in G.

2.
$$f$$
 satisfies $d(v) = \underbrace{\sum_{(u,v)\in E} f(u,v)}_{\text{edges to } v} - \underbrace{\sum_{(v,z)\in E} f(v,z)}_{\text{edges out of } v}$.





Main results

Theorem (Circulation integrality theorem)

If all capacities and demands are integers, and there exists a circulation, then there exists an integer valued circulation.

Sketch Proof Max-flow formulation + integrality theorem for max-flow

From the previous discussion, we can conclude:

Theorem (Necessary and sufficient condition)

There is a feasible circulation with $\{d(v)\}_{v \in V}$ in G iff the max-flow in G' has value D.

Circulations with demands and lower bounds: Max-flow formulation

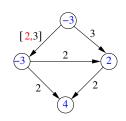
Generalization of the previous problem: besides satisfy demands at nodes, we want to force the flow to use certain edges.

Introduce a new constrain $\ell(e)$ on each $e \in E$, indicating the min-value the flow must be on e.

Given G = (V, E) with c(e), $c(e) \ge \ell(e) \ge 0$, for each $e \in E$ and $\{d(v)\}_{v \in V}$, define a circulation as a function $f : E \to \mathbb{R}^+$ s.t.

- 1. capacity: For each $e \in E$, $\ell(e) \le f(e) \le c(e)$,
- 2. conservation: For each $v \in V$,

$$\sum_{(u,v)\in E} f(u,v) - \sum_{(v,z)\in E} f(v,z) = d(v).$$

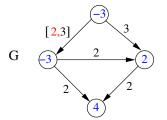


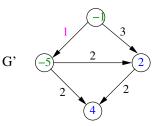
Circulation problems with lower bounds: Given $(G, c, \ell, \{d(v)\})$, does there exists a feasible circulation?

Circulations with demands and lower bounds: Max-flow formulation

Let $G=((V,E),c,\ell,d(\cdot))$ be a graph, construct G'=((V,E),c',d'), where for each $e=(u,v)\in E$, with $\ell(e)>0$:

- $c'(e) = c(e) \ell(e)$ (sent $\ell(e)$ units along e).
- ▶ Update the demands on both ends of e: $(d'(u) = d(u) + \ell(e))$ and $d'(v) = d(v) \ell(e)$





Main result

Theorem

There exists a circulation in G iff there exists a circulation in G'. Moreover, if all demands, capacities and lower bounds in G are integers, then there is a circulation in G that is integer-valued.

Sketch Proof Need to prove f(e) is a circulation in G iff $f'(e) = f(e) - \ell(e)$ is a circulation in G'.

The integer-valued circulation part is a consequence of the integer-value circulation Theorem for f' in G'.



Applications: Generic reduction to Max Flow

Consider a generic problem \mathcal{GP} where we have as input s finite sets X_1, \ldots, X_d , each representing a different resources.

Our goal is to chose the "largest" number of d-tuples, each d-tuple containing exactly one element from each X_i , subject to the constrains:

- ▶ For each $i \in [d]$, each $x \in X_i$ can appears in at most c(x) selected tuples.
- ▶ For each $i \in [d]$, any two $x \in X_i$ and $y \in X_{i+1}$ can appear in at most c(x, y) selected tuples.
- ▶ The values for c(x) and c(x, y) are either in \mathbb{Z}^+ or ∞ .

Notice that only pairs of objects between adjacent X_i and X_{i+1} are constrained.

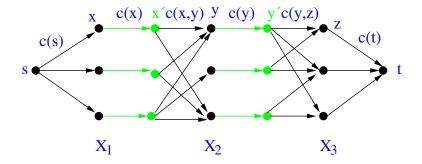
Applications: Generic reduction to Max-Flow

Make the reduction from \mathcal{GP} to the following input s-t network \vec{G} :

- ▶ G contains a vertex for each element x in each X_i , and a copy x' for each element $x \in X_i$ for $1 \le i < d$.
- We add vertex s and vertex t.
- Add an edge $s \to x$ for each $x \in X_1$ and add an edge $y \to t$ for every $y \in X_d$. Give capacities c(s,x) = c(x) and c(y,t) = c(y).
- ▶ Add an edge $x' \to y$ for every pair $x \in X_i$ and $y \in X_{i+1}$. Give a capacity c(x, y). Omit the edges with capacity 0.
- ▶ For every $x \in X_i$ for $1 \le i < d$, add an edge $x \to x'$ with c(x, x') = c(x).

Every path $s \rightsquigarrow t$ in \vec{G} is a feasible d-tuple, conversely every d-tuple that satisfies the constrains is a path $s \rightsquigarrow t$.

Flow Network: The reduction



So to solve \mathcal{GP} we construct from the input a s-t network, and find the maximum flow f^* .

FINAL'S SCHEDULING

We have as input:

- ▶ n classes, each one with a final. Each exam must be given in one room. Each class c_i has E[i] students.
- r rooms. Each r_j has a capacity S[j],
- ightharpoonup au time slots. For each room and time slot we only can give one final.
- **p** professors to watch exams. Each exam needs one professor in each class and time. Each professor has its own restrictions of availability and no professor supposed to oversee more than 6 finals. For each p_ℓ and τ_k define a Boolean variable $A[k,\ell] = T$ if p_ℓ is available at τ_k .

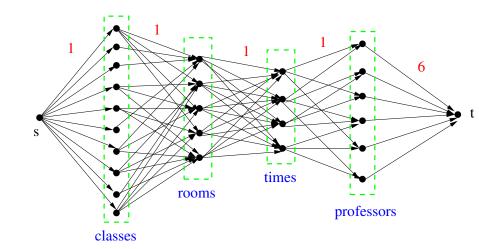
Design an efficient algorithm that correctly schedules a room, a time slot and a professor to every final, or report that not such schedule is possible.

Construction of the network

Construct the network \vec{G} with vertices $\{s, t, \{c_i\}, \{r_j\}, \{r_k\}, \{p_\ell\}\}$. Edges:

- $s \rightarrow c_i, c(s, c_i) = 1$ (each class has one final)
- ▶ $c_i \rightarrow r_j$, **if** $E[i] \leq S[j]$. Then, $c(c_i, r_j) = \infty$
- $\forall j, k, r_j \rightarrow \tau_k, c(r_j, \tau_k) = 1$ (one final per room).
- ▶ $\tau_k \to p_\ell$ if $A[k, \ell] = T$, $c(\tau_k, p_\ell) = 1$ (p can watch one final if p is available at τ_k).
- ▶ $p_\ell \rightarrow t$, $c(p_\ell,t) = 6$ (each p can watch ≤ 6 finals)

FINAL'S SCHEDULING: Flow Network



FINAL'S SCHEDULING

Notice the input size to the problem is $N = n + r + \tau + p + 2$. and size of the network is O(N) vertices and $O(N^2)$ edges, why?

Every path $s \rightsquigarrow t$ is a valid assignment of room-time-professor a single final, and every valid assignment room-time-professor a class final is represented by a path $s \rightsquigarrow t$.

To maximize the number of finals to be given, we compute the max-flow f^* from s to t.

If $|f^*| = n$ then we can schedule all finals, otherwise we can not schedule all the finals.

Complexity: To construct \vec{G} from the input, we need $O(\vec{E})$.

As $|f^*| \le n$, we can use Ford-Fulkerson to compute f^* , so the complexity of solving the problem is $O(n\vec{E})$.



SURVEY DESIGN problem

Problem: Design a survey among customers of products (KT-7.8)

- ► Each customer will receive questions about some products.
- ▶ Each customer i can only be asked about a number of products between c_i (l.b.) and c'_i ([c_i , c'_i]) which he has purchased.
- For each product j we want to collect date for a minimum of p_j distinct customers and a maximum of p_j' ($[p_j, p_j']$)



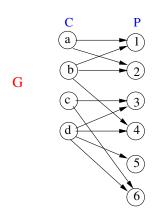
SURVEY DESIGN problem

Measuring customer satisfaction. Consider n customers and m products.

Formally we want to model the problem as:

- ▶ A bipartite graph $G = (C \cup P, E)$, where $C = \{i\}$ is the set of customers and $P = \{i\}$ is the set of products.
- ▶ There is an $(i,j) \in E$ is i has purchased product j.
- ▶ For each $i \in \{1, ..., n\}$, we we have bounds $[c_i, c'_i]$ on the number of products i can be asked about.
- ▶ For each $j \in \{1, ..., n\}$, we we have bounds $([p_j, p'_j])$ on the number of customers that can be asked about it.

SURVEY DESIGN: Bipartite graph G



Customers $C=\{a,b,c,d\}$

Products $P=\{1,2,3,4,5,6\}$

Customer	Buys
a	1,2
b	1,2,4
С	3,6
d	3,4,5,6

a:[1,2]1: [1,2]

b:[1,3] 2: [1,2] c:[1,2]3: [1,2]

d:[2,4]4: [1,2]

5: [0,1]

6: [1,2]

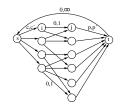
SURVEY DESIGN: Max flow formulation

We construct G' from G, by adding: Edges: $s \to \{C\}$, $\{P\} \to t$, and (t,s).

Capacities:
$$c(t, s)$$

l.b. $= \sum_i c_i$, cap. $= \sum_i c_i'$
 $c(i,j) = 1$,
 $c(s,i) = [c_i, c_i']$,

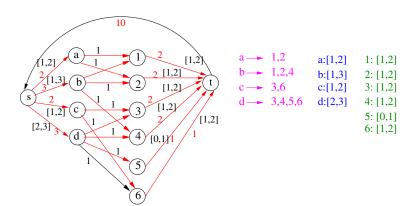
$$c(j,t)=[p_j,p'_j].$$



Notice if *f* is the flow:

- ▶ $f(i,j) = 1 \Rightarrow$ customer i is asked about product j,
- ▶ f(s,i) # products to ask customer i for opinion,
- f(j,t) = # customers to be asked to review product j,
- f(t,s) is the number of questions asked.

Max flow formulation: Example



Main result

Theorem G' has a feasible circulation iff there is a feasible way to design the survey.

Proof if there is a feasible way to design the survey:

- ▶ if *i* is asked about *j* then f(i,j) = 1,
- f(s, i) = number questions asked to i,
- f(j, t) = number of customers who were asked about j,
- f(t,s) = total number of questions.
- easy to verify that f is feasible in G'

If there is an integral, feasible circulation in G':

- if f(i,j) = 1 then i will be asked about j,
- ▶ the constrains (c_i, c'_i, p_j, p'_i) will be satisfied.

Conclusions

Max-Flow/ Min-Cut problem is an intuitively easy problem with lots of applications.

We just presented a few ones.

An alternative point of view can be obtained from duality in Linear Programming

The material in this talk has been basically obtained from two textbooks:

- Chapter 7 of Kleinberg, Tardos: Algorithm Design.
- ► The Dinik-Edmond-Karp algorithm are basically inspired from an old set of lectures notes by Larry Harper in 1975, which I don't think there are in internet.