

Linear Programming Continued!

Algorithms..

Designed..

..multiplication, sorts, depth first search, breadth first search, greedy, dynamic programs.

Used previous algorithms as subroutines.

E.g. sort for greedy mst.

Another method.

- Make a graph.

- Run algorithm.

- Pull out answer.

Examples:

Shortest path.

Dynamic programming and DAG!

Peas and carrots. Express as a linear program!

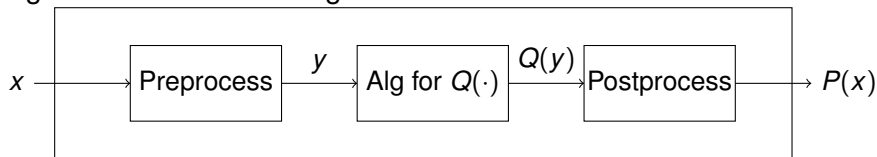
Production. Express as a linear program!

Different from subroutine call.

Pre/post process is “easy”.

Reduction.

Algorithm for Pfrom algorithm for Q .



Peas and Carrots, Production. Instances.

General problem reduction to linear program.

New problem: Maximum Flow.

Max-Flow Problem.

1. Capacity Constraints: $0 \leq f_e \leq c_e$.

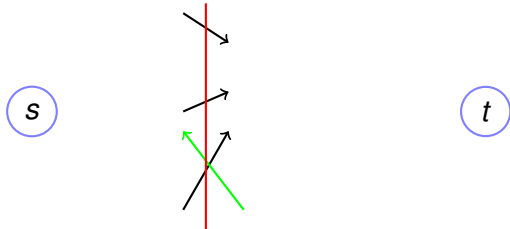
2. Conservation Constraints:

“flow into v ” = “flow out of v ” (if not s or t .)

Algorithm adds flow, say f , to path from s to t .

Optimality: upper bound.

s - t Cut: $V = S \cup T$ and $s \in S$ and $t \in T$.



Lemma: Capacity of any $s - t$ cut is an upper bound on the flow.

$C(S, T)$ - sum of capacities of all arcs from S to T

$$C(S, T) = \sum_{e=(u,v): u \in S, v \in T} c_e$$

For valid flow:

Flow out of (S) = Flow out of s .

Flow into (T) = Flow into t .

For any valid flow, $f: E \rightarrow \mathbb{Z}_+$, the flow out of S (into T)

$$\sum_{e \in S \times T} f_e - \sum_{e \in T \times S} f_e \leq \sum_{e \in S \times T} c_e - \sum_{e \in T \times S} 0 = C(S, T).$$

→ The value of any valid flow is at most $C(S, T)$!

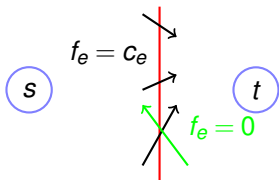


Optimality: max flow = min cut.

At termination of augmenting path algorithm.

No path with residual capacity!

Depth first search only starting at s does not reach t .



S be reachable nodes.

No arc with positive residual capacity leaving S

\implies All arcs leaving S are full.

\implies No arcs into S have flow.

Total flow leaving S is $C(S, T)$.

Valid flow \implies all that flow from source.

Value of flow equals value of $C(S, T)$. and Optimal is $\leq C(S, T)$.

\rightarrow Flow is maximum!!

Cut is minimum $s - t$ cut too!

“any flow” \leq “any cut” and this flow = this cut.

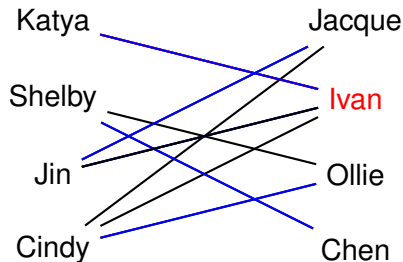
\rightarrow Maximum flow and minimum $s - t$ cut!

Celebrated max flow -minimum cut theorem.

Theorem: In any flow network, the maximum s - t flow is equal to the minimum cut.

Bipartite Matching

Given a bipartite graph: $B = (L, R, E)$ where $E \subseteq L \times R$.



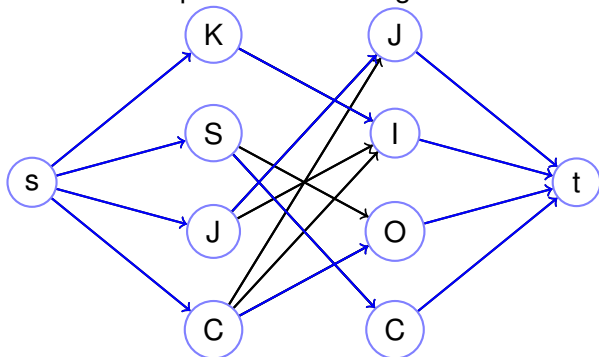
Find largest subset of edges (“matches”) which are one to one.

Bipartite Matching

Algorithm by “Reduction.”:

From matching problem produce flow problem.

From flow solution produce matching solution.



Max flow = Max Matching Size.

Flow is not integer necessarily....

Augmenting path algorithm gives integer flow.

Duality.

$$\max x_1 + 8x_2$$

$$x_1 \leq 4$$

$$x_2 \leq 3$$

$$x_1 + 2x_2 \leq 7$$

$$x_1, x_2 \geq 0$$

One Solution: $x_1 = 1, x_2 = 3$. Value is 25.

Best possible?

For any solution.

$$x_1 \leq 4 \text{ and } x_2 \leq 3 \dots$$

$$\dots \text{so } x_1 + 8x_2 \leq 4 + 8(3) = 28.$$

Added equation 1 and 8 times equation 2

yields bound on objective..

Better solution?

Better upper bound?

Duality.

$$\max x_1 + 8x_2$$

$$x_1 \leq 4$$

$$x_2 \leq 3$$

$$x_1 + 2x_2 \leq 7$$

$$x_1, x_2 \geq 0$$

Solution value: 25.

Add equation 1 and 8 times equation 2 gives..

$$x_1 + 8x_2 \leq 4 + 24 = 28.$$

Better way to add equations to get bound on function?

Sure: 6 times equation 2 and 1 times equation 3.

$$x_1 + 8x_2 \leq 6(3) + 7 = 25.$$

Thus, the value is at most 25.

The upper bound is same as solution!

Proof of optimality!

Duality:example

Idea: Add up positive linear combination of inequalities to “get” upper bound on optimization function.

Will this always work?

How to find best upper bound?

Duality: computing upper bound.

Best Upper Bound.

Multiplier	Inequality
y_1	$x_1 \leq 4$
y_2	$x_2 \leq 3$
y_3	$x_1 + 2x_2 \leq 7$

Adding equations thusly...

The left hand side should "dominate" optimization function:

If $y_1, y_2, y_3 \geq 0$

and $y_1 + y_3 \geq 1$ and $y_2 + 2y_3 \geq 8$ then..

$$x_1 + 8x_2 \leq 4y_1 + 3y_2 + 7y_3$$

Find best y_i 's to minimize upper bound?

The dual, the dual, the dual.

Find best y_i 's to minimize upper bound?

Again: If you find $y_1, y_2, y_3 \geq 0$
and $y_1 + y_3 \geq 1$ and $y_2 + 2y_3 \geq 8$ then..
 $x_1 + 8x_2 \leq 4y_1 + 3y_2 + 7y_3$

$$\min 4y_1 + 3y_2 + 7y_3$$

$$y_1 + y_3 \geq 1$$

$$y_2 + 2y_3 \geq 8$$

$$y_1, y_2, y_3 \geq 0$$

A linear program.

The **Dual** linear program.

Primal: $(x_1, x_2) = (1, 3)$; Dual: $(y_1, y_2, y_3) = (0, 6, 1)$.

Value of both is 25!

Primal is optimal ... and dual is optimal!

The dual.

In general.

Primal LP

$$\max c \cdot x$$

$$Ax \leq b$$

$$x \geq 0$$

Dual LP

$$\min y^T b$$

$$y^T A \geq c$$

$$y \geq 0$$

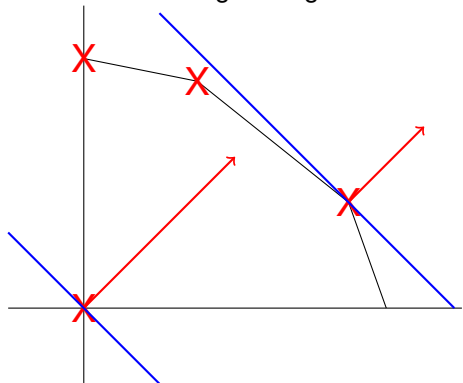
Theorem: If a linear program has a bounded value, then its dual is bounded and has the same value.

Simplex Algorithm.

Start at a vertex.

Move to better neighboring vertex.

Until no better neighboring vertex.



$$\begin{aligned} \max(x_1 + x_2) \\ 7x_1 + 5x_2 &\leq 20 \\ 4x_1 + 5x_2 &\leq 21 \\ 2x_1 + 10x_2 &\leq 33 \\ x_1 \geq 0, x_2 &\geq 0 \end{aligned}$$

Why optimal? Draw line corresponding to $cx = \text{current value}$.
Entire feasible region on “wrong” side.

Example: review.

$$\max x_1 + 8x_2$$

$$x_1 \leq 4$$

$$x_2 \leq 3$$

$$x_1 + 2x_2 \leq 7$$

$$y_1, y_2, y_3 \geq 0$$

$$\min 4y_1 + 3y_2 + 7y_3$$

$$y_1 + y_3 \geq 1$$

$$y_2 + 2y_3 \geq 8$$

$$x_1, x_2 \geq 0$$

“Matrix form”

$$\max [1, 8] \cdot [x_1, x_2]$$

$$\begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 2 \end{pmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \leq \begin{bmatrix} 4 \\ 3 \\ 7 \end{bmatrix}$$

$$[x_1, x_2] \geq 0$$

$$\min [4, 3, 7] \cdot [y_1, y_2, y_3]$$

$$[y_1, y_2, y_3] \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 2 \end{pmatrix} \geq \begin{bmatrix} 1 \\ 8 \end{bmatrix}$$

$$[y_1, y_2, y_3] \geq 0$$

Matrix equations.

$$\max[1, 8] \cdot [x_1, x_2]$$

$$\begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 2 \end{pmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \leq \begin{bmatrix} 4 \\ 3 \\ 7 \end{bmatrix}$$

$$[x_1, x_2] \geq 0$$

$$\min[4, 3, 7] \cdot [y_1, y_2, y_3]$$

$$[y_1, y_2, y_3] \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 2 \end{pmatrix} \geq \begin{bmatrix} 1 \\ 8 \end{bmatrix}$$

$$[y_1, y_2, y_3] \geq 0$$

We can rewrite the above in matrix form.

$$A = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 2 \end{pmatrix}$$

$$c = [1, 8] \quad b = [4, 3, 7]$$

The primal is $Ax \leq b, \max c \cdot x, x \geq 0$.

The dual is $y^T A \geq c, \min b \cdot y, y \geq 0$.

Generality of Linear Programming.

Linear program solves many problems.

How applicable is it?

Circuit Evaluation.

Circuit Evaluation:

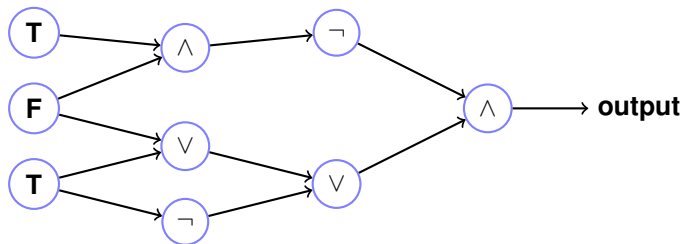
Given: DAG of boolean gates:

two input **AND/OR**.

One input **NOT**.

TRUE/FALSE inputs.

Problem: What is the output of a specified Output Gate?



What is the value of the output?

Translation to linear program.

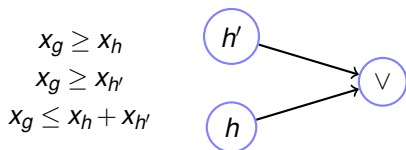
Variable for gate g : x_g .

Constraints:

$$0 \leq x_g \leq 1$$

Gate g is true gate: $x_g = 1$.

Gate g is false gate: $x_g = 0$.



For \wedge gate:

$$x_g \leq x_h, \quad x_g \leq x_{h'} \quad x_g \geq x_h + x_{h'} - 1$$

For \neg gate: $x_g = 1 - x_h$.

x_o is 1 if and only if the circuit evaluates to true.

What does this mean?

The circuit value problem is completely general!

A computer program can be unfolded into a circuit.

Each level is the circuit for a computer.

The number of levels is the number of steps.

⇒ circuit value problems model computation.

⇒ linear programs can model any polynomial time problem!

Warning: existence proof, not generally efficient.