

LP and FFT

EECS 477

Lecture 19, 11/19/2002

A Linear Programming Example

- Politician trying to win a election
- 3 Types of areas: urban, suburban, rural
- Primary issues:
build roads, gun control, farm subsidies
- **To win, need:**
50,000 urban votes
100,000 suburban votes
25,000 rural votes.

Money Spent on Campaign Ads

Policy	Urban	Suburban	Rural
Build Roads	-2	5	3
Gun Control	8	2	-5
Farm Subsidies	0	0	10
Votes Req.	50	100	25

Shown: 1000s of votes won by spending \$1000 an ads

■ **Goal:** spend min \$\$ and win the elections



Formalization

■ Introduce variables:

x_1 = \$\$ spent on building roads

x_2 = \$\$ spent on gun control

x_3 = \$\$ spent on farm subsidies

■ Express constraints through equations:

$$-2 x_1 + 8 x_2 + 0 x_3 \geq 50$$

$$5 x_1 + 2 x_2 + 0 x_3 \geq 100$$

$$3 x_1 - 5 x_2 + 10 x_3 \geq 25$$

Finally...

- Minimize the function:

$$x_1 + x_2 + x_3$$

- Subject to constraints

$$-2x_1 + 8x_2 + 0x_3 \geq 50$$

$$5x_1 + 2x_2 + 0x_3 \geq 100$$

$$3x_1 - 5x_2 + 10x_3 \geq 25$$

$$\text{and } x_1, x_2, x_3 \geq 0$$

- (Any) solution of this linear program is an optimal strategy for the politician

Formal Definitions

- **Linear Function:**

$$f(x_1, x_2, \dots, x_n) = a_1x_1 + a_2x_2 + \dots + a_nx_n$$

where $x_1 \dots x_n$ are variables and $a_1 \dots a_n$ are constants.

- **Linear Equality:**

$$f(x_1, \dots, x_n) = b \text{ where } b \text{ is a real number.}$$

- **Linear Inequalities:**

$$f(x_1, \dots, x_n) \leq b \text{ and}$$

$$f(x_1, \dots, x_n) \geq b$$

More Definitions...

- **Linear Programming Problem:**
 - Minimizing or maximizing linear function
 - E.g., $x_1 + x_2 + x_3$ (Objective Function)
 - Subject to a finite set of linear constraints ($=, \geq, \leq$)
- **Three cases**
 - Max $f(\dots)$
 - Min $f(\dots)$: can turn into Max $-f(\dots)$
 - (Pure) constraint satisfaction (no objective function)
 - E.g., $f(x) = 0$
- **Standard Form:** max subject to linear inequalities
- **Slack Form:** max subject to linear equalities

More Definitions (2)

- **Feasible Solution**
 - Values $x_1, x_2 \dots x_n$ that satisfy all constraints
- **Feasible Region** (in the n -dim space)
 - Graph all the constraint equations
 - Their intersection is the feasible region
- **Objective Value:** value of the objective function at a particular point in the feasible region.

Standard Form

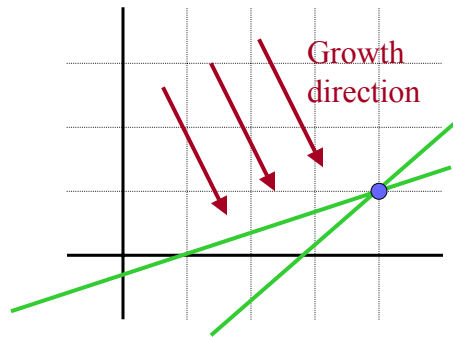
- n real numbers $c_1 \dots c_n$;
 m real numbers b_1, b_2, \dots, b_m ; and mn real numbers a_{ij} for $i = 1 \dots m$ and $j = 1 \dots n$.
- Find $x_1 \dots x_n$ that
maximize $\sum c_j x_j$, $j = 1 \dots n$ (objective function)
subject to
$$\sum a_{ij} x_j \leq b_i \text{ for } i = 1 \dots m \text{ and } j = 1 \dots n \text{ (Eq 1)}$$
$$x_j \geq 0 \text{ for } j = 1 \dots n \text{ (Eq 2)}$$
- Eq 1 and Eq 2 are **constraints**
- Eq 2 : **non negativity** constraints

Summary of Standard Form

- a_{ij} = elements of $m \times n$ matrix A
- b_i = m -dim vector b
- c_j = n -dim vector c
- x_j = n -dim vector x
- **Linear Program** is a tuple (A, b, c) , interpreted as
 - Max $c^T x$
 - Subject to $Ax \leq b$
 $x \geq 0$

Example

- $\max (x_1 - 2 x_2)$
- $x_1 \geq 0, x_2 \geq 0$
- $x_1 - 3 x_2 \leq 1$
- $x_1 - x_2 \leq 3$



- Draw a picture
- Find the solution via simplex method

Scheduling and LP

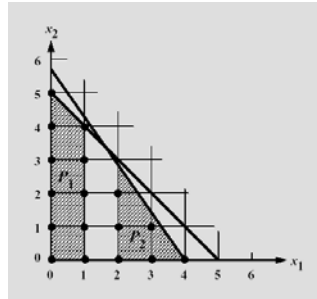
- $m=1..M$ machines
- $j=1..J$ jobs
- $p[m,j]$ – processing time
 $x[m,j] = 0$ or 1 -- assignment
 $\sum_{m=1..M} x[m,j] = 1, \quad j=1..J$
 $\sum_{j=1..J} x[m,j] p[m,j] \leq t, \quad m=1..M$
 $\min t$
LP Relaxation $x[m,j] \geq 0$

Integer programming

■ Maximize $17x+12y$

■ Subject to

- $10x+7y \leq 40$
- $x+y \leq 5$
- $x, y \geq 0$, integers



– from Vanderbei '2001

- Optimal solution $(x, y) = (1.67, 3.33)$
- Rounding $(2, 3)$ infeasible, closest feasible $(1, 3)$
- Can use Branch and Bound for exact solution

DFT

■ $w := \exp(-2\pi i/N)$

- N-th roots of unity $(w^k)^N = 1$
 - Complex number $\exp(iu) = \cos(u) + i \sin(u)$
 - Taking squares get $n/2$ roots of unity
 - Fourier transform: what applications
- $$\hat{a}[k] = \sum_{j=0..N-1} a[j] w^{kj}, \text{ for all } k=0..N-1$$

■ Simple algorithm

N times N-term sums = $\Theta(N^2)$

FFT

- Use the fact that

$$\sum_{j=0..N-1} a[j] x^j = \sum_{j=0..N/2-1} a[2*j] x^{2j} + x \sum_{j=0..N/2-1} a[2*j+1] x^{2j}$$

- Recursive procedure

```
fft(a[0..n-1]) {
    ye = fft((a[0],a[2],...,a[n-2]));
    yo = fft((a[1],a[3],...,a[n-1]));
    for(k=0; k<n/2; ++k) {
        y[k] = ye[k] + w^k yo[k];
        y[k+n/2] = ye[k] - w^k yo[k];
    }
    return y;
}
```

FFT

- $T(N) = 2 T(N/2) + \Theta(N)$
 - Case ? of the Master Theorem

- Fast algorithm

$$T(N) = \Theta(N \log N)$$

[0,1,2,3,4,5,6,7,8]							
[0,2,4,6]				[1,3,5,7]			
[0,4]		[2,6]		[1,5]		[3,7]	
[0]	[4]	[2]	[6]	[1]	[5]	[3]	[7]