

2011 - אלגוריתמים דינמיים

10:5

תוכנית דינמית של אלגוריתם דינמי

- Consider a matrix  $A \in M_{m \times n}$  and vectors  $\begin{cases} x \in \mathbb{R}^n \\ b \in \mathbb{R}^m \end{cases}$
- System of equations  $Ax = b$

solution: an (affine) subspace of  $\mathbb{R}^n$   
 $z + \{x : Ax = b\}$  where  $Az = b$   
algorithm: Gauss elimination

What about inequalities?

$$\begin{cases} Ax = b \\ x \geq 0 \end{cases}$$

$$\begin{cases} Ax \leq b \end{cases}$$

$$\begin{cases} x_1 + 2x_2 + 3x_3 \leq 4 \\ x_1 + x_2 + x_3 = 1 \\ x_1 \geq 0 \\ x_2 \geq 0 \\ x_3 \geq 0 \end{cases}$$

Question #1 Find an efficient algorithm for the following problem:

**Input:** a set  $\Pi$  of constraints  
(in)equalities over variables  $x_1, \dots, x_n \in \mathbb{R}$

[ Each constraint is of the form:

$$a_{i1}x_1 + a_{i2}x_2 + \dots + a_{in}x_n \begin{cases} \leq \\ \geq \end{cases} b_i ]$$

**Output:** "not feasible" or  $\vec{x} \in \mathbb{R}^n$  that satisfies all the constraints

Question #2: Find an efficient algorithm for the following problem:

**Input:**  $m \times n$  matrix  $A \in M_{m \times n}$   
vector  $b \in \mathbb{R}^m$   
vector  $c \in \mathbb{R}^n$

**Goal:** find  $x \in \mathbb{R}^n$  that minimizes  $c^t x \hat{=} \sum_j c_j x_j$   
subject to  $A \cdot x = b, x \geq 0$ .

**Output:** "not feasible" or  $x^* = \operatorname{argmin} \{c^t x / Ax = b, x \geq 0\}$ .

**Example:**  $\min x_1 + x_2$

s.t.  $x_1 + 10x_2 \geq 21, x_1 \geq 0, x_2 \geq 0$ .

Both questions are about **linear programming**!

- Q1: Is a polyhedron empty?
- Q2: Linear programming in standard form.

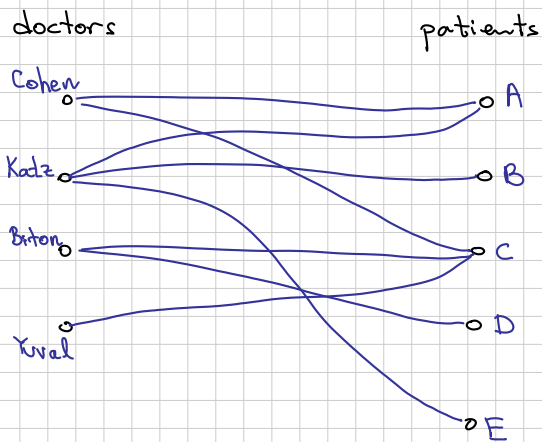
Lots of: structure  
& algorithms...

## Combinatorial Optimization Problems

find "best" solution for a combinatorial problem (as opposed to a "numerical problem").

Examples: Set Cover  
Vertex Cover  
Minimum weight matching  
Maximum flow

### Set Cover - Example



**Rules:**

Every doctor can serve a subset of the patients.

**Goal:**

Choose fewest # doctors, so that each patient is served.

Definition: Set-Cover

**Input:** a set of elements  $[1..n]$  (patients)  
a family of subsets:  $S_1, \dots, S_m \subseteq [1..n]$  (doctors)

$\{S_{i_1}, S_{i_2}, \dots, S_{i_k}\}$  is a **Set Cover** if  $S_{i_1} \cup S_{i_2} \cup \dots \cup S_{i_k} = [1..n]$  (every patient served by a doctor)

**Goal:** find a set cover of minimum size.

## History: Set Cover

- 1) Lots of applications & special cases.
- 2) Among the first problems proven to be NP-Complete [Karp]
- 3) Greedy algorithm gives  $H_n \approx \ln n$  approximation.
- 4) NP-Complete to find "better" approximations [Feige, RS]

What is the greedy algorithm for Set Cover?

What does linear programming have to say?

Set Cover: the greedy algorithm

Input:  $S_1, \dots, S_m \subseteq [1..n]$

goal: find smallest set  $\{S_{i_1}, \dots, S_{i_k}\}$  s.t.  $S_{i_1} \cup \dots \cup S_{i_k} = [1..n]$

method: (Greedy)

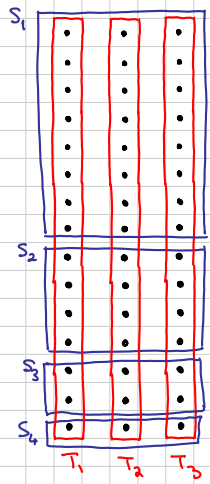
$C \leftarrow \emptyset$

while  $C \neq [1..n]$  do

$S \leftarrow \operatorname{argmax} \{ |S_i \setminus C| : 1 \leq i \leq m \}$   
 $C \leftarrow C \cup S$  (add  $S$  to the cover)

Question: How many sets in the cover (compared to an optimal solution)?

## Greedy - Example



Greedy picks:

$\{S_1, S_2, S_3, S_4\}$

But OPT picks

$\text{opt} = \{T_1, T_2, T_3\}$

so  $|\text{Greedy}| = 4$

&  $|\text{OPT}| = 3.$

Question 1: Give a sequence of examples where  $\text{Greedy}/\text{OPT} \geq \ln n$  (easier:  $\geq \Omega(\ln n)$ )

Set Cover: the greedy algorithm (COST)

Theorem [Johnson]

greedy returns an  $(H_n)$ -approx.

$$H_n = 1 + \frac{1}{2} + \dots + \frac{1}{n} \leq 1 + \ln n$$

$$(|\text{Greedy}| \leq (1 + \ln n) \cdot \text{OPT})$$

proof - not hard, we will prove using LP & duality...

## An Integer Programming formulation for Set Cover

For each set  $S$ , define a variable  $x_S \in \{0, 1\}$

$$\begin{aligned} \min \quad & \sum_S x_S \\ \text{s.t.} \quad & \sum_{\{S: e \in S\}} x_S \geq 1 \quad \forall \text{element } e \in [1..n] \end{aligned}$$

Now relax integrality constraint  $x_S \in \{0, 1\}$   
and use fractional constraint  $0 \leq x_S \leq 1$ :

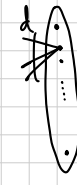
$$\begin{aligned} \min \quad & \sum_S x_S \\ \text{s.t.} \quad & \sum_{\{S: e \in S\}} x_S \geq 1 \quad \forall \text{element } e \\ & 0 \leq x_S \leq 1 \quad \forall \text{set } S \end{aligned} \quad \left. \begin{array}{l} \text{fractional} \\ \text{relaxation} \\ \text{for Set-Cover} \end{array} \right\}$$

## Example of fractional Set cover

$n$  sets



$n$  elements



$\leftarrow$   $d$ -regular bipartite graph (superposition of  $d$  matchings)

$$\begin{aligned} \min \quad & \sum_S x_S \\ \text{s.t.} \quad & \sum_{\{S: e \in S\}} x_S \geq 1 \quad \forall \text{element } e \\ & 0 \leq x_S \leq 1 \quad \forall \text{set } S \end{aligned}$$

Suppose  $\forall \text{set } S: |S| = d$   
 $\forall \text{element } e: |\{S: e \in S\}| = d$

$\Rightarrow x_S = \frac{1}{d}$  is a feasible solution.

Question #2: prove that  $\sum x_S = \frac{n}{d}$  is optimal

What can we do with (optimal) solutions  $\{x_S^*\}$  of LP?

$$\begin{aligned} \min \quad & \sum_S x_S \\ \text{s.t.} \quad & \sum_{\{S: e \in S\}} x_S \geq 1 \quad \forall \text{element } e \\ & 0 \leq x_S \leq 1 \quad \forall \text{set } S \end{aligned}$$

1) Lower bound:  $\sum_S x_S^* \triangleq \text{OPT}_f \leq \text{OPT}$

2) If (almost) integral then  $\{x_S^*\}$  is (almost) a set cover.

3) Round  $\{x_S^*\}$  ...

First, let's use it to analyze Greedy.

$C \leftarrow \emptyset$  Greedy - with charging

while  $C \neq [1..n]$  do

$$\begin{aligned} & S \leftarrow \text{argmax} \{ |S_i \setminus C| : 1 \leq i \leq m \} \\ & \forall e \in S \setminus C: p(e) \leftarrow \frac{1}{|S \setminus C|} \quad (\text{charge covered elements}) \\ & C \leftarrow C \cup S \quad (\text{add } S \text{ to the cover}) \end{aligned}$$

In each round  $\sum_{e \in S \setminus C} p(e) = 1$  pays for  $S$ .

$\Rightarrow$  In the end:  $\sum_e p(e) = |\text{cover}|$

Why does this help?

### Primal LP (covering)

$$\begin{aligned} \text{Min } & \sum_S x_S \\ \text{s.t. } & \sum_{S: e \in S} x_S \geq 1 \quad \forall \text{ element } e \\ & 0 \leq x_S \leq 1 \quad \forall \text{ set } S \end{aligned}$$

### Dual LP (packing)

$$\begin{aligned} \text{max } & \sum_e y_e \\ \text{s.t. } & \sum_{e \in S} y_e \leq 1 \quad \forall S \\ & 0 \leq y_e \end{aligned}$$

Claim (weak duality):  $\sum_S x_S \geq \sum_e y_e$

proof:  $\sum_S 1 \cdot x_S \geq \sum_S \left( \sum_{e \in S} y_e \right) x_S = \sum_e \left( \sum_{S: e \in S} x_S \right) y_e \geq \sum_e y_e$

Hope: If  $\{p_e\}$  is a dual solution, then

$$|\text{Greedy}| = \sum_e p_e \leq \text{OPT}_f \leq \text{OPT} \dots$$

Question #3: show that  $\{p_e\}_e$  is NOT a dual solution.

Lovasz:  $y_e = \frac{p_e}{1 + \ln n}$  is a dual solution!

$$\begin{aligned} \text{and then } |\text{Greedy}| &= \sum_e p_e = (1 + \ln n) \sum_e y_e \\ &\leq (1 + \ln n) \cdot \text{OPT}_f \leq (1 + \ln n) \cdot \text{OPT} \end{aligned}$$

$\Rightarrow$  Greedy is an  $O(\lg n)$ -approximation.

So, now left to prove feasibility of  $\{y_e\}$ .

Consider any set  $S'$

$$S' = \{e_1, e_2, \dots, e_l\}$$

let  $T(e) \triangleq$  iteration

in which  $e$  is covered

while  $G \neq [1..n]$  do  
 $S \leftarrow \text{argmax} \{ |S_i \setminus C| : 1 \leq i \leq m \}$   
 $\forall e \in S \setminus C : p(e) \leftarrow \frac{1}{|S \setminus C|}$  (change covered elements)  
 $C \leftarrow C \cup S$  (add  $S$  to the cover)

assume:  $T(e_1) \leq T(e_2) \leq \dots \leq T(e_l)$

$$\Rightarrow p(e_1) \leq p(e_2) \leq \dots \leq p(e_l)$$

Consider iteration in which  $e_i$  is covered.

$$\{e_1, \dots, e_l\} \subseteq S' \setminus C \quad \& \quad |S \setminus C| \geq |S' \setminus C|$$

$$\Rightarrow p(e_i) = \frac{1}{|S \setminus C|} \leq \frac{1}{|S' \setminus C|} = \frac{1}{l - i + 1}$$

$$\Rightarrow p(e_1) \leq \frac{1}{l}, p(e_2) \leq \frac{1}{l-1}, \dots, p(e_l) \leq 1$$

$$\Rightarrow \forall S : \sum_{e \in S} p(e) \leq \sum_{i=1}^n \frac{1}{i} \triangleq H_n \leq 1 + \ln n$$

$$\Rightarrow y_e = \frac{p(e)}{1 + \ln n} \text{ is dual feasible. } \square$$

Question #4: prove that  $\ln(n+1) \leq H_n \leq 1 + \ln n$

## Recap

- 1) Set Cover Problem
- 2) Greedy Algorithm
- 3) Integer Programming Formulation.
- 4) Linear Programming Formulation.
- 5) Primal & Dual LP's.
- 6) Weak Duality
- 7) Analysis of approximation ratio of Greedy.

Hint for Q1: Find an example in which  
 $OPT \geq \Omega(\lg n) \cdot OPT_f$ .

Hint: Elements  $U = \{0,1\}^k - 0^k$  (all nonzero strings)

Sets  $\{S_i\}_{i \in U}$

where  $S_i = \{u \in U \mid \sum_{j=1}^k u_j \cdot i_j \text{ is odd}\}$

show:  $\forall i: |S_i| = \frac{2^k}{2}$

$\forall u: |\{S_i: u \in S_i\}| = \frac{2^k}{2}$

use prob. (2) to prove  $OPT_f \leq 2$ .

prove  $OPT \geq k$  (system of  $\leq k$  equations in  $GF(2)$  has nonzero solution)

## MORE QUESTIONS

5) Suppose that  $\forall \text{element } e: |\{S: e \in S\}| \leq d$ .

Find an approximation algorithm with

$$|ALG| \leq d \cdot OPT.$$

6) Improve the analysis of the approximation ratio of the Greedy algorithm if

$\forall \text{Set } S: |S| \leq k$ . (say,  $k=3$ ).

\* what if  $k=2$  ???

7) Weak duality

<u>Primal</u>	<u>Dual</u>
$\min c^t x$	$\max y^t b$
s.t. $Ax \geq b$	s.t. $y^t A \leq c^t$
$x \geq 0$	$y \geq 0$

prove that: if  $x$  is a primal solution &  $y$  is a dual solution, then  $c^t x \geq y^t b$ .

## Encore: Rounding a fractional set-cover

Suppose  $\{x_s\}$  is an optimal fractional set-cover,  
I.E.,  $\forall e: \sum_{e \in S} x_s \geq 1$  and  $\sum_s x_s = \text{OPT}_f$ .

How can we obtain an integral set-cover  
 $\hat{x}_s \in \{0, 1\}$  from  $x_s$ ?

2 Goals:

**feasibility:**  $\forall e \exists S: e \in S \ \& \ \hat{x}_s = 1$

**cost:**  $\frac{\sum_s \hat{x}_s}{\sum_s x_s}$  is small ( $\sum_s \hat{x}_s \leq O(\lg n) \cdot \text{OPT}_f$ )

## Randomized Rounding

For each set  $S$ , randomly flip a "coin"  $\hat{x}_s$   
such that  $\Pr[\hat{x}_s = 1] = x_s$ .

Now:  $\forall$  element  $e$ :

$$\begin{aligned}\Pr[e \text{ is covered}] &= \Pr[\exists S: e \in S \ \& \ \hat{x}_s = 1] \\ &= 1 - \Pr[\forall S: e \in S \Rightarrow \hat{x}_s = 0]\end{aligned}$$

$$\begin{aligned}&= 1 - \prod_{S: e \in S} (1 - x_s) \\ &\stackrel{e^{-x} \geq 1-x}{\geq} 1 - e^{-\sum_{e \in S} x_s} \geq 1 - e^{-1} \approx 0.632\end{aligned}$$

$\Rightarrow$  Each element is covered with constant prob.

Repeat this experiment  $\ln(4n)$  times  
( $S$  is in the cover if  $\hat{x}_s = 1$  in one of experiments)

$$\Pr(e \text{ is not covered}) \leq \left(\frac{1}{e}\right)^{\ln(4n)} = \frac{1}{4n}$$

$$\Rightarrow \Pr(\exists e \text{ not covered}) \leq \sum_e \Pr(e \text{ not covered}) = \frac{1}{4}$$

$\Rightarrow$  with prob  $\geq \frac{3}{4}$  every element is covered  
if we repeat  $\ln(4n)$  times.

Amplify prob. of success by repeating  $k \cdot \ln(4n)$   
times  $\Rightarrow \Pr(\text{Success}) \geq 1 - \left(\frac{1}{4}\right)^k$ .

What about cost?

$$E\left[\sum_s \hat{x}_s\right] = \sum_s E[\hat{x}_s] = \sum_s x_s$$

We repeat  $\ln(4n)$  times  $\Rightarrow E(\text{cover}) = \ln(4n) \cdot \text{OPT}_f$ .

Markov Inequality:

$$\Pr(\text{cover} \geq 4 \cdot \ln(4n) \cdot \text{OPT}_f) \leq \frac{1}{4}.$$