#### CMPS 3130/6130 Computational Geometry Spring 2015



#### Linear Programming and Halfplane Intersection Carola Wenk

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# Word Problem

A company produces tables and chairs. The profit for a chair is \$2, and for a table \$4. Machine group *A* needs 4 hours to produce a chair, and 6 hours to produce a table. Machine group *B* needs 2 hours to produce a chair, and 6 hours to produce a table. Per day there are at most 120 working hours for group *A* and at most 72 hours for group *B*.

#### How can the company maximize profit?

#### Variables:

 $c_A = \#$  chairs produced on machine group *A*  $c_B = \#$  chairs produced on machine group *B*  $t_A = \#$  tables produced on machine group *A*  $t_B = \#$  tables produced on machine group *B* 

#### **Objective function (profit):**

Maximize  $2(c_A + c_B) + \overline{4}(t_A + t_B)$ 

#### **Constraints:**

 $4c_A + 6t_A \le 120$  $2c_B + 6t_B \le 72$ 

# **Linear Programming**

Variables:  $x_1, \ldots, x_d$ 

**Objective function:** Maximize  $f_{\vec{c}}(\vec{x}) = c_1 x_1 + \ldots + c_d x_d$ 

#### **Constraints:**

 $h_1: \quad a_{11}x_1 + \ldots + a_{1d}x_d \le b_1$ 

$$h_2: \quad a_{21}x_1 + \ldots + a_{2d}x_d \le b_2$$

$$h_{\mathrm{n}}: \qquad a_{\mathrm{n}1} x_1 + \ldots + a_{\mathrm{n}d} x_d \le b_{\mathrm{n}}$$

Linear program in *d* variables with *n* constraints

- Each constraint  $h_i$  is a half-space in  $\mathbb{R}^d$
- $\bigcap_{i=1}^{n} h_i$  is the feasible region of the linear program
- Maximizing  $f_{\vec{c}}(\vec{x})$  corresponds to finding a point  $\vec{x}$  that is extreme feasible regree in direction  $\vec{c}$ .

h1

# **Sub-Problem: Halfspace Intersection** (in R<sup>2</sup>: Halfplane Intersection)

**Given:** A set  $H = \{h_1, h_2, ..., h_n\}$  of halfplanes  $h_i: a_i x + b_i y \le c_i$ with constants  $a_i, b_i, c_i$ ; for i=1,...,n. **Find:**  $\bigcap_{i=1}^n h_i$ , i.e., the feasible region of all points  $(x,y) \in \mathbb{R}^2$ satisfying all *n* constraints at the same time. This is a convex polygonal region bounded by at most *n* edges.



# **D&C Halfplane Intersection**

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Algorithm Intersect_Halfplanes(H):

Input: A set H of n halfplanes in \mathbb{R}^2

Output: The convex polygonal region \mathbb{C} = \bigcap_{h \in H} h

if |H|=1 then

\mathbb{C} = \mathbb{h}, where H = \{h\}

else

split H into two sets H_1 and H_2 of size n/2 each

C_1 = \text{Intersect}\_\text{Halfplanes}(H_1)

C_2 = \text{Intersect}\_\text{Halfplanes}(H_2)

C = \text{Intersect}\_\text{Convex}\_\text{Regions}(C_1, C_2)

return C
```

- Use a plane-sweep to develop an O(*n*)-time algorithm for Intersect\_Convex\_Regions
- $T(n) = 2T(n/2) + n \implies T(n) \in O(n \log n)$

# **Incremental Linear Programming**

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- 2D linear program (LP)
- Assume the LP is bounded (otherwise add constraints)
- Assume there is one unique solution (if any); • take the lexicographically smallest solution
- **Incremental approach:** Add one halfplane after the other.  $H_i = \{h_1, \dots, h_i\}$   $C_i = h_1 \cap \dots \cap h_i$   $C = C_n = \bigcap_{h \in H} h$ Let  $v_i$  = unique optimal vertex for feasible region  $C_i$ , for  $i \ge 2$ .

Then  $C_1 \supseteq C_2 \supseteq \dots \supseteq C_n = C$ , and hence if  $C_i = \emptyset$  for some *i* then  $C_j = \emptyset$  for all  $j \ge i$ .

#### **Incremental Linear Programming**

#### Lemma: Let $2 \le i \le n$ . (i) If $v_{i-1} \in h_i$ then $v_i = v_{i-1}$ (ii) If $v_{i-1} \notin h_i$ then $C_i = \emptyset$ or $v_i \in l_i$ = the line bounding $h_i$

Handling case (ii) involves solving a 1-dimensional LP on  $l_i$ :

• The feasible region is just an interval, that can be computed in linear time [rightmost left-bounded halfplane, leftmost right-bounded halfplane]



•  $\Rightarrow$  We can compute a new  $v_i$ , or decide that the LP is infeasible, in O(*i*) time

# 2D\_Bounded\_LP



## **Randomized Incremental LP**

Depending on the insertion order of the halfplanes the runtime varies between O(n) and  $O(n^2)$ .  $\Rightarrow$  Randomize the input order of the halfplanes.

**Theorem:** 2D\_Randomized\_Bounded\_LP runs in O(n) expected time and O(n) deterministic space.

**Proof:** Define a random variable  $X_i = \begin{cases} 1, v_{i-1} \notin h_i \\ 0, else \end{cases}$ The total time spent to resolve case (ii), over all  $h_1, \dots, h_n$  is  $\sum_{i=1}^n O(i)X_i$ 

### **Randomized Incremental LP**

We now need to bound the expected value  $E(\sum_{i=1}^{n} O(i)X_i) = \sum_{i=1}^{n} O(i)E(X_i)$ and we know that  $E(X_i) = P(X_i) = P(v_{i-1} \notin h_i)$ . Apply backwards analysis to bound  $E(X_i)$ :

- Fix  $H_i = \{h_1, \dots, h_i\}$  which determines  $C_i$ .
- Analyze what happened in last step when  $h_i$  was added.
- P(had to compute new optimal vertex when adding  $h_i$ )
  - = P(optimal vertex changes when we remove a halfplane from  $C_i$ )



 $\Rightarrow$  Total expected runtime is  $\sum_{i=1}^{n} O(i) \frac{2}{i} = O(n)$