

# Lecture 13: Linear Programming I

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# Outline

- Definition of linear programming and examples
- A linear program to solve max flow and min-cost max flow
- A linear program to solve minimax-optimal strategies in games
- Algorithms for linear programming

# Example

- There are 168 hours in a week. Want to allocate our time between
  - studying (S)
  - going to parties (P)
  - everything else (E)
- To survive:  $E \geq 56$
- For sanity:  $P + E \geq 70$
- To pass courses:  $S \geq 60$
- If party a lot, need to study or eat more:  $2S + E - 3P \geq 150$
- Is there a *feasible* solution? Yes,  $S = 80$ ,  $P = 20$ ,  $E = 68$
- Happiness is  $2P + E$ . Find a feasible solution maximizing this *objective function*

# Linear Program

- This is called a *linear program (LP)*
- All constraints are linear in our variables
- Objective function is linear
- Don't allow  $S \cdot E \geq 100$ , that's a polynomial program. Much harder.

# Formal Definition

- Given:
  - n variables  $x_1, \dots, x_n$
  - m linear inequalities in these variables
    - E.g.,  $3x_1 + 4x_2 \leq 6, 0 \leq x_1, x_1 \leq 3$
- Goal:
  - Find values for the  $x_i$ 's that satisfy constraints and maximize objective
  - In the feasibility problem just satisfy the constraints
  - What would happen if we allowed strict inequalities  $x_1 < 3$ ?
  - $\max x_1$

# Time Allocation Problem

- Variables:  $S, P, E$
- Objective: Maximize  $2P + E$  subject to
- Constraints:  $S + P + E = 168$

$$E \geq 56$$

$$S \geq 60$$

$$2S + E - 3P \geq 150$$

$$P + E \geq 70$$

$$P \geq 0$$

# Operations Research Problem

	labor	materials	pollution
plant 1	2	3	15
plant 2	3	4	10
plant 3	4	5	9
plant 4	5	6	7

What are the variables?

$x_1, x_2, x_3, x_4$  denote the number of cars at plant  $i$

What's our objective?

maximize  $x_1 + x_2 + x_3 + x_4$

- Required to make at least 400 cars at plant 3
- Have 3300 hours of labor and 4000 units of material
- At most 12000 units of pollution
- Maximize number of cars made

	labor	materials	pollution
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Make at least 400 cars at plant 3  
 3300 hours of labor and 4000 units of material  
 At most 12000 units of pollution  
 Maximize number of cars made

**Note:** linear programming does not give an integral solution (NP-hard)

Constraints:  $x_i \geq 0$  for all  $i$

$$x_3 \geq 400$$

$$2x_1 + 3x_2 + 4x_3 + 5x_4 \leq 3300$$

$$3x_1 + 4x_2 + 5x_3 + 6x_4 \leq 4000$$

$$15x_1 + 10x_2 + 9x_3 + 7x_4 \leq 12000$$



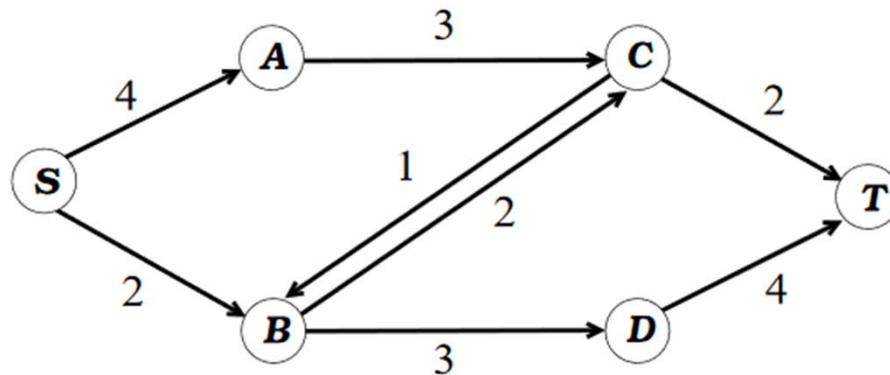
# Modeling Network Flow

**Variables:**  $f_{uv}$  for each edge  $(u,v)$ , representing positive flow

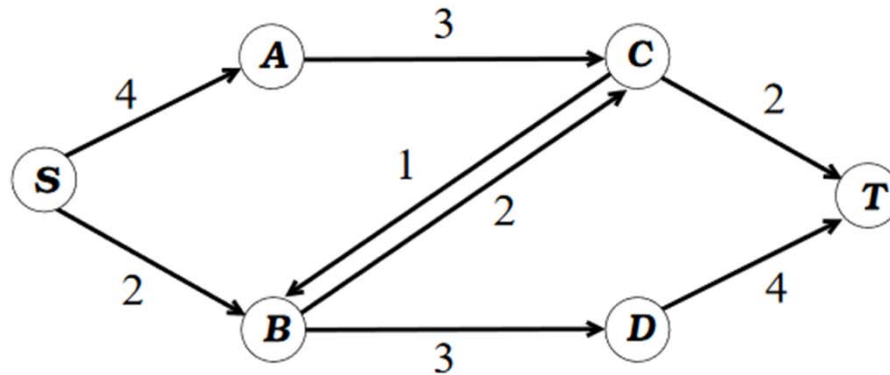
**Objective:** maximize  $\sum_u f_{ut}$

**Constraints:** For all edges  $(u,v)$   $0 \leq f_{uv} \leq c(u,v)$  (capacity constraints)

For all  $v \notin \{s, t\}$ ,  $\sum_u f_{uv} = \sum_u f_{vu}$  (flow conservation)



# Modeling Network Flow



In this case, our LP is: maximize  $f_{ct} + f_{dt}$  subject to the constraints:

$$0 \leq f_{sa} \leq 4, 0 \leq f_{ac} \leq 3, \text{ etc.}$$

$$f_{sa} = f_{ac}, f_{sb} + f_{cb} = f_{bc} + f_{bd}, f_{ac} + f_{bc} = f_{cb} + f_{ct}, f_{bd} = f_{dt}.$$

# Min Cost Max Flow

- Edge  $(u,v)$  has a capacity  $c(u,v)$  and a cost  $w(u,v)$
- Find a max s-t flow of least total cost, where the cost of flow  $f$  is

$$\sum_{(u,v) \in E} w(u,v) f_{uv}$$

- *How to solve this?*
- **Solution 1:** Solve for a maximum flow  $f$ 
  - Add a constraint that flow must equal the flow of  $f$
  - Minimize  $\sum_{(u,v) \in E} w(u,v) f_{uv}$  also subject to original constraints
- **Solution 2:** Add an edge  $(t,s)$  of infinite capacity and very negative cost
  - Minimizing cost automatically maximizes flow

# Zero Sum Games

Row payoffs:

20	-10	5
5	10	-10
-5	0	10

- Given a zero-sum game with  $n$  rows and  $n$  columns, compute a minimax optimal strategy for row player
- *What are the variables?*
  - Probabilities  $p_1, \dots, p_n$  on our actions
  - Linear constraints:  $\sum_{i=1, \dots, n} p_i = 1$  and  $p_i \geq 0$  for all  $i$
  - Maximize the minimum expected payoff, over all column pure strategies
- *How to maximize a minimum with a linear program?*
- Create new “dummy variable”  $v$  to represent minimum

# Zero Sum Games

- $R_{i,j}$  represents payoff to row player with row player action  $i$  and column player action  $j$
- Variables:  $p_1, \dots, p_n$  and  $v$
- Objective: maximize  $v$
- Constraints:
  - $p_i \geq 0$  for all  $i$ , and  $\sum_i p_i = 1$
  - For all columns  $j$ ,  $\sum_i p_i R_{ij} \geq v$

# Linear Programs in Standard Form?

- Many different ways to write the same LP
- Use vector notation, so  $c^T x = \sum_{i=1, \dots, d} c_i x_i$  if there are  $d$  variables
- Any LP can be written in the following form:
- Max  $c^T x$

$$\begin{aligned} \text{Subject to } Ax &\leq b \\ x &\geq 0 \end{aligned}$$

How to handle equality constraints  $d^T x = e$ ?

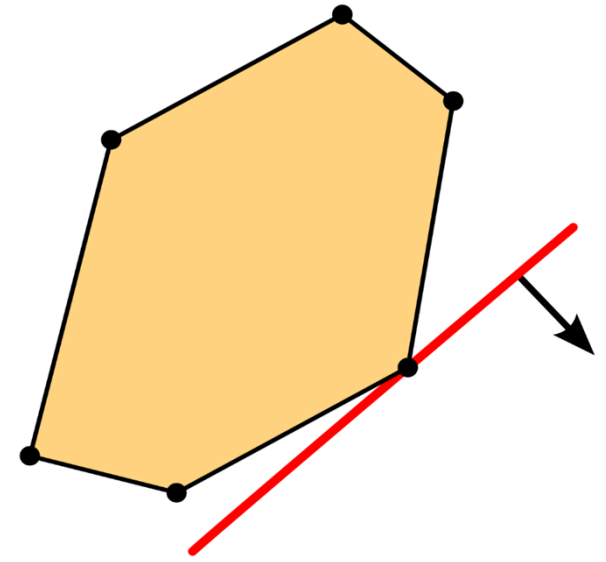
How to convert  $\min c^T x$  to a maximization?

How to handle an unconstrained variable  $x_i$  which could be positive or negative?

Substitute  $x_i = y_i - z_i$ ,  $y_i \geq 0$ ,  $z_i \geq 0$ , everywhere

# Facts about Linear Programs

- Consider the LP
- Max  $c^T x$   
Subject to  $Ax \leq b$   
 $x \geq 0$
- Think of maximizing  $c^T x$  over the set  $Ax \leq b, x \geq 0$
- What does the set  $Ax \leq b, x \geq 0$  look like?
  - Each row is a *halfspace*, cutting  $\mathbb{R}^d$  into two pieces by a hyperplane
  - The intersection of halfspaces could be empty
    - Then the LP is *infeasible*
  - Could be unbounded
  - Could be bounded and then we call it the *feasible region*
- Maximizing  $c^T x$  moves the hyperplane with normal vector  $c$  until it is tangent to the feasible region



# Convexity Properties

- Feasible region  $Ax \leq b, x \geq 0$  is convex
  - If  $p$  and  $q$  are in the feasible region, then so is the line segment joining  $p$  and  $q$ . **Why?**
- Proof by pictures, e.g., convex polygon in two dimensions
- Formally, since  $Ap \leq b$  and  $Aq \leq b$ , for any  $\lambda \in [0,1]$ ,
  - $\lambda Ap \leq \lambda b$  and  $(1 - \lambda)Aq \leq (1 - \lambda)b$
  - So  $A(\lambda p + (1 - \lambda)q) \leq b$
  - Also  $\lambda p \geq 0$  and  $(1 - \lambda)q \geq 0$  since  $p \geq 0$  and  $q \geq 0$
- More generally, intersections of convex sets are convex
- Max  $c^T x$  occurs at a vertex. **Can we just enumerate all vertices?**



# Algorithms for Linear Programming

- Simplex Algorithm
  - Practical, but exponential time in the worst-case
- Ellipsoid Algorithm
  - First polynomial time algorithm, but slow in practice
- Karmarkar's Algorithm (interior point)
  - Polynomial time algorithm and competitive in practice
- Software: LINDO, CPLEX, Solver (in Excel)

# Time Allocation Problem

- Variables: S, P, E
- Objective: Maximize  $2P + E$  subject to
- Constraints:  $S + P + E = 168$

$$E \geq 56$$

$$S \geq 60$$

$$2S + E - 3P \geq 150$$

$$P + E \geq 70$$

$$P \geq 0$$

Substitute  $S = 168 - P - E$ , so two variables P and E, want to maximize  $2P + E$ .

# Intuition for Linear Programming

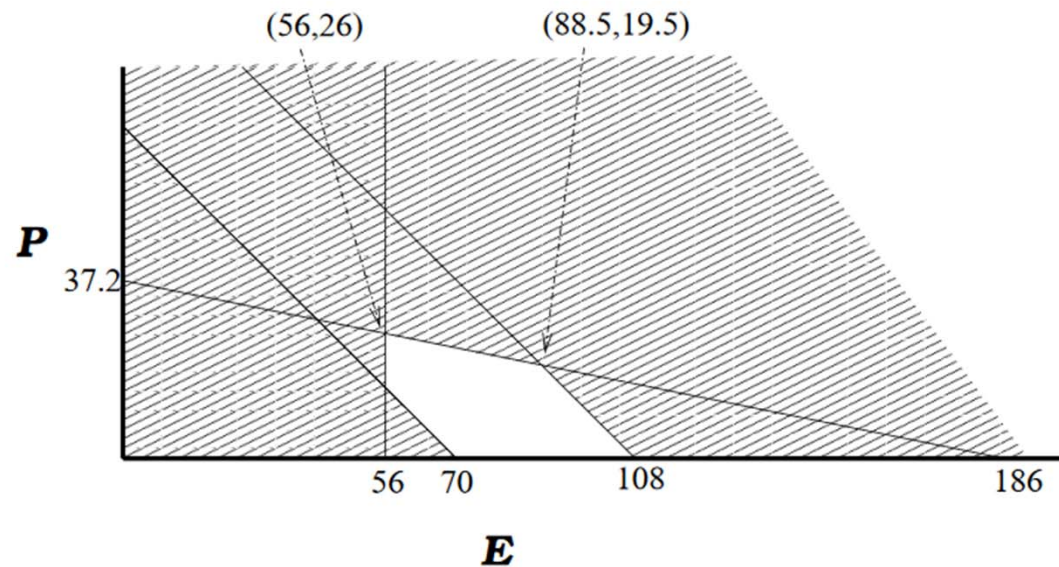
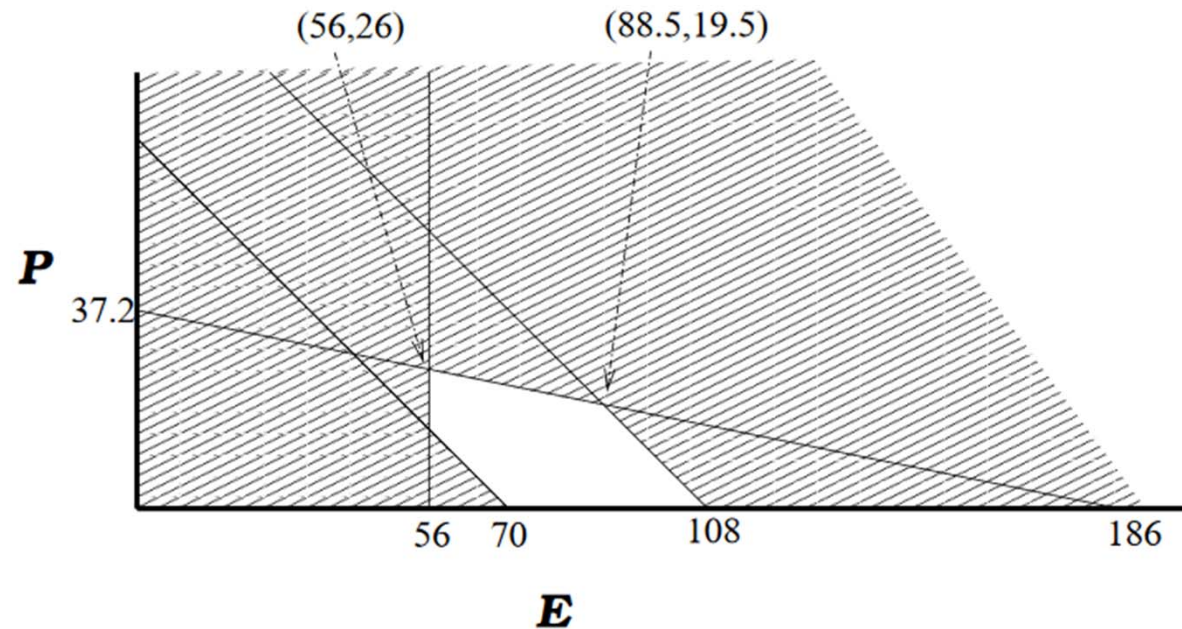


Figure 13.1: Feasible region for our time-planning problem. The constraints are:  $E \geq 56$ ;  $P + E \geq 70$ ;  $P \geq 0$ ;  $S \geq 60$  which means  $168 - P - E \geq 60$  or  $P + E \leq 108$ ; and finally  $2S - 3P + E \geq 150$  which means  $2(168 - P - E) - 3P + E \geq 150$  or  $5P + E \leq 186$ .

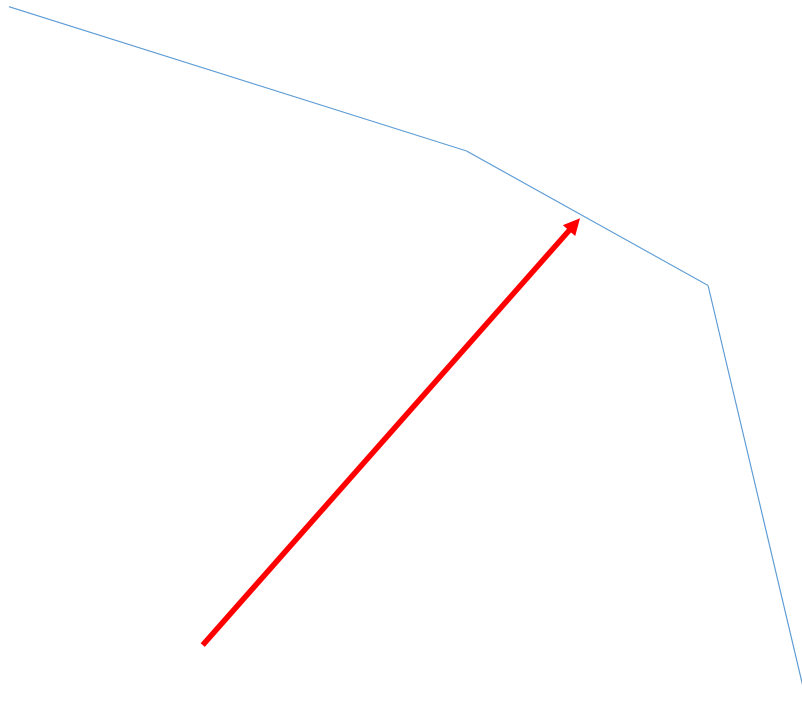
Maximizing  $P$  occurs at  $(56, 26)$ . Maximizing  $2P + E$  occurs at  $(88.5, 19.5)$

# Simplex Algorithm



- Start at vertex of the feasible region (polyhedron in high dimensions)
- Look at cost of objective function at each neighbor
- Move to neighbor of minimum cost
- Always make progress, but could take exponential time (in high dimensions)

# Simplex Algorithm



Get stuck in local maximum?

No, since feasible set is convex

# Other Annoyances I

- How to start at a vertex of the feasible region?
- $Ax \leq b$   
 $x \geq 0$
- What if it's not even feasible?
- Introduce "slack" variable  $s$ . Consider:
- $\min s$   
subject to  $Ax \leq b + s \cdot 1^m$   
 $x \geq 0, s \geq 0, s \leq \max_i -b_i$
- Feasible. Can run simplex starting at  $x = 0^n$  and  $s = \max_i -b_i$
- If original LP is feasible, minimum achieved when  $s = 0$ , and  $x$  that is output is a vertex in the feasible region of original LP

## Other Annoyances II

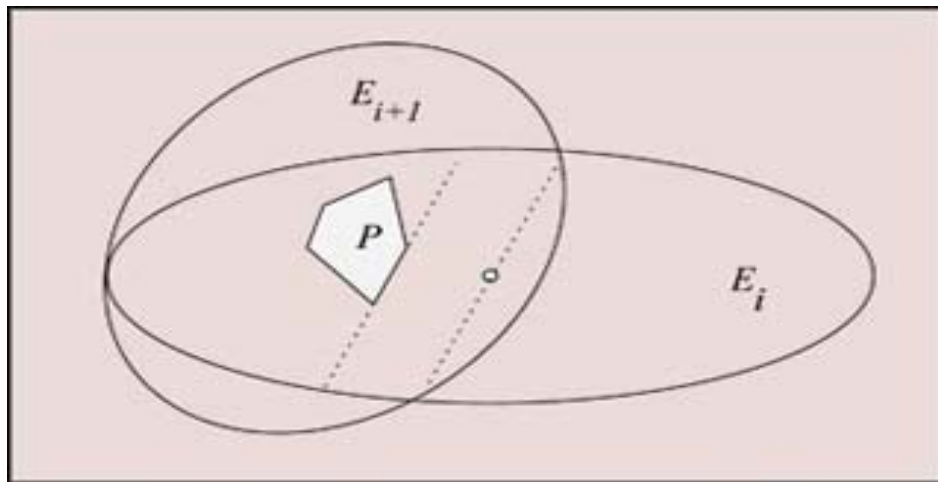
- What if the feasible region is unbounded?
  - Ok, as long as objective function is bounded
- What if objective function is unbounded?
  - Output  $\infty$ , how to detect this?
- Many ways
  - see one based on duality in a few lectures
  - include constraints  $x_i \leq M$  for all  $i$ , for a very large value  $M$
  - can efficiently find  $M$  to ensure if solution is finite, still find the optimum

# Ellipsoid Algorithm

Solves feasibility problem

Replace objective function with constraint, do binary search

Replace “minimize  $x_1 + x_2$ ” with  $x_1 + x_2 \leq \lambda$



Can handle exponential number of constraints if there's a separation oracle



# Karmarkar's Algorithm

- Works with feasible points but doesn't go corner to corner
- Moves in interior of the feasible region – “interior point method”

