

# Lecture 18: Linear Programming and Optimization

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Sam McCauley

November 14, 2025

Williams College

# Admin

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- Assignment Resubmissions open! Will give more detail next slide
- Colloquium after class today: “Programming for Everyone: Developers, Scientists, and Creators of Tomorrow”
  - Faculty candidate! Come and (optionally) give us anonymous feedback
- Questions?

# Assignment Resubmission

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- Can bring your grade up 10 points out of 100
- Reminder: cap at 95 unless extra credit
- Submit your new version under (e.g.) “Assignment 1 Resubmission” on gradescope
- Please *say where you fixed it* by typing RESUBMIT at the beginning of any problem/code you want me to look at
- For code, just need to git push the new version (and write RESUBMIT on the relevant parts of the pdf)
- Reminder that honor code/etc. is still in effect; no LLMs allowed

# **Linear Programming and Optimization**

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# What is an algorithmic problem?

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- Constraints
- Objective
- What if we had a *single tool* that could solve *any* problem with *certain kinds* of constraints and objectives?

## Next section of the course

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- Framework to phrase large varieties of algorithmic problems



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- Allow practical solutions for a wide variety of otherwise-intractable problems

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- Framework to phrase large varieties of algorithmic problems
- Allow practical solutions for a wide variety of otherwise-intractable problems
- “Optimization” problems that come up frequently in practice
- This topic is much older and much much broader than anything else we’ve covered
- Focus for this class: using linear programming and integer linear programming (and their solvers) to obtain optimal solutions to difficult problems. (Won’t be focusing on structure, mathematical properties.)

## Context

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“*I have a strong interest in the question of where mathematical ideas come from, and a strong conviction that they always result from a fairly systematic process—and that the opposite impression, that some ideas are incredible bolts from the blue that require “genius” or “sudden inspiration” to find, is an illusion.*”

Timothy Gowers


# History

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- Starts with a legend



# (You may have seen heard this story before)



AI Mode

All

**Images**

Videos


News

Forums


Short videos

More ▾


Tools ▾




George b




George bernard dantzig




Solved



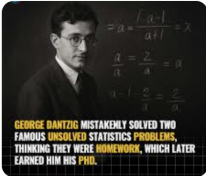
Unsolved math problem



Linear programming




Simplex meth



**GEORGE DANTZIG MISTAKENLY SOLVED TWO FAMOUS UNSOLVED STATISTICS PROBLEMS, THINKING THEY WERE HOMEWORK, WHICH LATER EARNED HIM HIS PHD.**

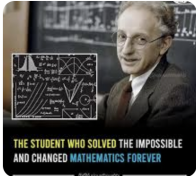
Facebook

In 1939, George Dantzig unk...



TikTok


George Dantzig: T...



**THE STUDENT WHO SOLVED THE IMPOSSIBLE AND CHANGED MATHEMATICS FOREVER**

Facebook


In 1939, George Dantzig u...



**HE SOLVED A MATH MYSTERY BY ACCIDENT**

TikTok


George Dantzig: T...



**After arriving late to class, George Dantzig copied down two problems from the blackboard, thinking they were homework. He solved them over the next few days, not realizing they were actually famous unsolved problems in statistics. Weeks later, his professor revealed the true nature of the**


Facebook

In 1939, George Dantzi...




Berkeley Engi


George Dantzi



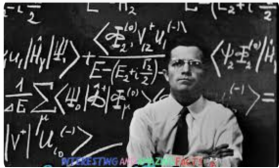
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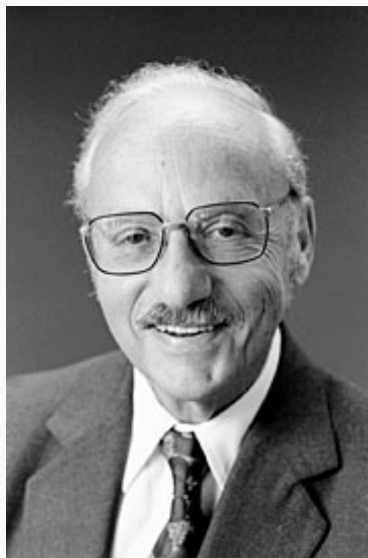
# History

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- Starts with a legend

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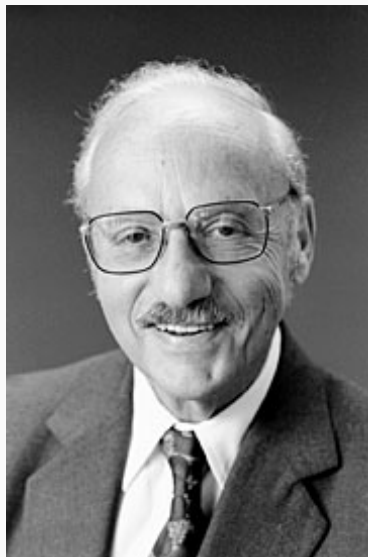
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- Father of Linear Programming

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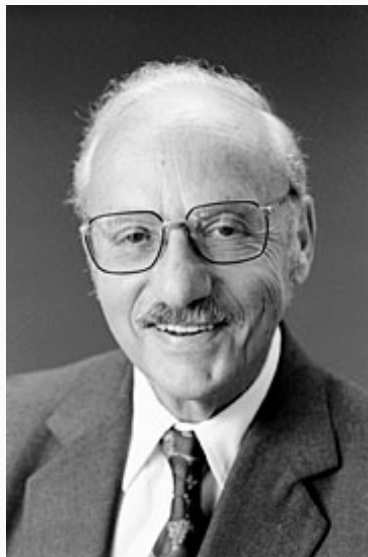
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- Father of Linear Programming
- Worked for military during World War 2

# George Dantzig

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- Father of Linear Programming
- Worked for military during World War 2
- Invented the simplex algorithm

# Where we're headed

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- A “big hammer” to solve an extremely wide variety of algorithmic problems

# Where we're headed

College Class Timetable

Monday	Monday	Tuesday	Tuesday	Tuesday	Cheestry	Sriay	Friuary	Friday	Friday
									Arts
		Mathematics Phorens					8:00	Mathematics Engl. Literatury	
	Trts					Arts		19	
8:00 Am			18						
12:00 Am		English Literature		Hilos	History		Physics		Arts
14:00 Am									
10:00 Am	106	Physical		104			125	Physical	104
12:00 Am		Chemistry					Education		
11:00 Am	300				18	300	Arts		
10:30 Am		Physical Science							
11:00 Am	250			260			230		6:00
11:00 Am						Computer Science			
18:00 Am	2:00	Arts				Arts			
18:00 Am									
18:00 Am									

- A “big hammer” to solve an extremely wide variety of algorithmic problems

# Where we're headed


- There are five houses.
- The white-haired girl lives in the red house.
- The black-haired girl worships the Fire god.
- The person fighting with a bow lives in the green house.
- The redhead fights with scimitars.
- The green house is immediately to the right of the ivory house.
- The person of house Bear worships the Moon god.
- The person of house Dragon lives in the yellow house.
- Daggers are used by the person living the middle house.
- The blonde lives in the first house.
- The person of house Badger lives in the house next to the house of the person who worships the Rising Sun god.
- The person of House dragon lives next to the house of the person who worships the Star god.
- The person of house Lion fights with an axe.
- The brunette belongs to house Stag.
- The blonde lives next to the blue house.

2			5			8			10			12			14		
3			6			9			11			13			15		
4			7														

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- a linear **objective function**, and
- a set of linear **constraints**.
- (We'll discuss what we mean by linear in a moment.)

Goal: achieve the best possible objective function value while satisfying the constraints

# Why linear programming

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- Black-box tools to solve important optimization problems that would be otherwise intractable

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  - Essentially gives a free method to solve continuous optimization problems—as well as some others



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  - More powerful (in a sense) than dynamic programming
  - Strictly generalizes network flows
  - Essentially gives a free method to solve continuous optimization problems—as well as some others
- 2004 survey: 85% of fortune 500 companies report using linear programming

# What do I mean by “linear”?

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- Let's say our variables are  $x_1, \dots, x_n$ .

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- Example 2:  $4x_1x_2 + x_1 = 3$  is not
- Example 3:  $|\sqrt{x_3} - x_7| \geq 5$  is not

# Linear Programming

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A linear program consists of:

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- **Goal:** achieve the best possible objective function value while satisfying *all of* the constraints
- Note that variables need not be integer or positive

# Example of a Linear Program

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Objective:

$$\max \quad 3x_1 + 4x_2$$

Subject to:

$$2x_1 + x_2 \leq 120$$

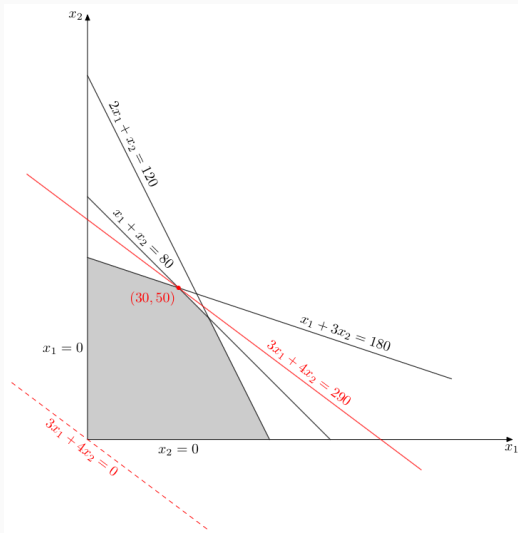
$$x_1 + 3x_2 \leq 180$$

$$x_1 + x_2 \leq 80$$

$$x_1 \geq 0$$

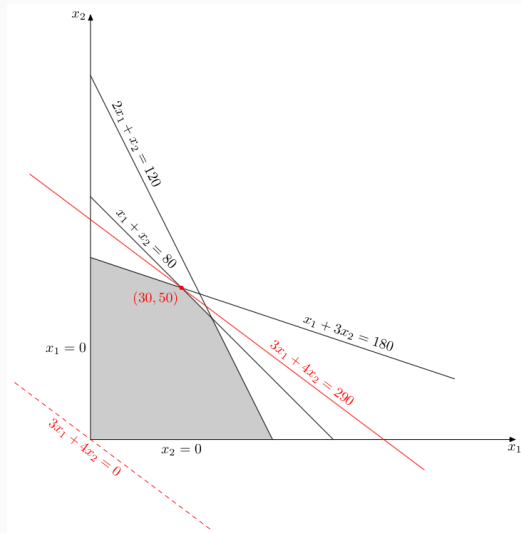
$$x_2 \geq 0$$

# Feasibility



- An LP is *feasible* if there exists an assignment of variables that satisfies the constraints

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- An LP is *feasible* if there exists an assignment of variables that satisfies the constraints
- **Nontrivial result:** feasibility is not trivial to determine. In the worst case, it is *as difficult* as solving the entire LP.

# Matrix Representation

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Objective:

$$[3 \ 4]$$

Subject to:

$$\begin{bmatrix} 2 & 1 \\ 0 & 3 \\ 0 & 1 \\ -1 & 0 \\ 0 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \leq \begin{bmatrix} 120 \\ 180 \\ 80 \\ 0 \\ 0 \end{bmatrix}$$

- Can represent with a matrix and vector
- Useful! Can use linear algebra techniques and optimizations for these problems.
- I don't plan to use this representation again in this class

# Visual representation

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- We can plot these inequalities
- Works best for instances with 2 or 3 variables
- We'll use extensively as it gives good intuition

# Plotting an LP

Objective:

$$\max \quad 3x_1 + 4x_2$$

Subject to:

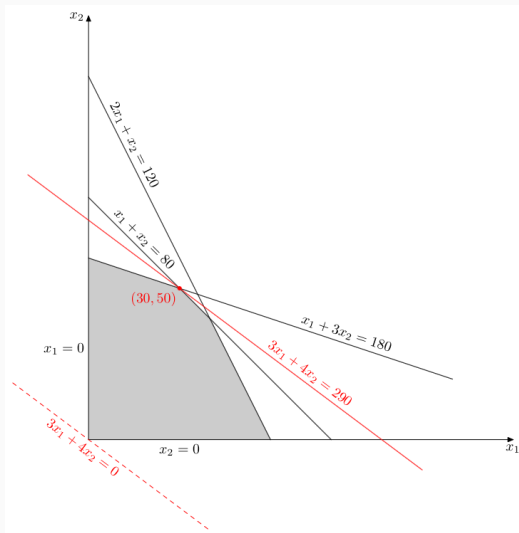
$$2x_1 + x_2 \leq 120$$

$$x_1 + 3x_2 \leq 180$$

$$x_1 + x_2 \leq 80$$

$$x_1 \geq 0$$

$$x_2 \geq 0$$





# Why are we looking at this?

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  - I will! It's called CPLEX. (It's free and open source so I'm not really "giving" it.)

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  - I will! It's called CPLEX. (It's free and open source so I'm not really "giving" it.)
- Guarantees correct, optimal solutions!
- Frequently very fast in practice (though these tools are often exponential in the worst case)
- **Our goal:** use this tool to solve computational problems

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- Many problems can be phrased as a linear program
  - We'll start with some slightly contrived problems to build intuition, and eventually get to problems with more real-world importance.

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- For today: take as a given that efficient solving is possible. How can we use linear programming to solve these problems?



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  - We'll start with some slightly contrived problems to build intuition, and eventually get to problems with more real-world importance.
- Linear programs can be solved efficiently
- For today: take as a given that efficient solving is possible. How can we use linear programming to solve these problems?
- Essentially a reduction: similar to using Network Flow to solve problems

# **Solving Problems with Linear Programming**

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# Optimization Problems

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## Example 1: Diet

- You need to eat 46 grams of protein and 130 grams of carbs every day



# Optimization Problems

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  - 100g Peanuts: 25.8g of protein, 16.1g carbs, \$1.61

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What is the cheapest way you can hit your diet goals?

**In Pairs:** try to phrase this problem as a linear program. What variables do we want? What is the objective function? What are the constraints?



# Diet Problem

---

How can we phrase this as a linear program? (I'll write the problem on the board; you should think about how you would do it.)

- Let  $p$  be the amount of peanuts,  $r$  be the amount of rice, and  $c$  be the amount of chicken you buy.

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- Then what is our *objective function*?

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- **Answer:** The price is  $1.61p + .79r + .7c$

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- Do we want to maximize or minimize this?

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- **Answer:** The price is  $1.61p + .79r + .7c$
- Do we want to maximize or minimize this?
- $\min 1.61p + .79r + .7c$

# Diet Problem Constraints

---

$$\min 1.61p + .79r + .7c$$

Reminder:

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# Diet Problem Constraints

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$$\min 1.61p + .79r + .7c$$

- Protein:  $25.8p + 2.5r + 13.5c \geq 46$

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- Anything else?

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- Carbs:  $16.1p + 28.7r \geq 130$
- Anything else?
- $p \geq 0, r \geq 0, c \geq 0$

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# Diet Problem Solution

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- Protein:  $25.8p + 2.5r + 13.5c \geq 46$
- Carbs:  $16.1p + 28.7r \geq 130$
- $p \geq 0, r \geq 0, c \geq 0$

Solution:  $p = 0, r = 2.9216..., c = 2.86636...$

So we want to buy about 293g of rice, and 287g of chicken, for total cost \$4.32

# Diet Problem Solution

---



A mathematically optimal meal

$$\min 1.61p + .79r + .7c$$

- Protein:  $25.8p + 2.5r + 13.5c \geq 46$
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  - **First:** how would we write these constraints if we don't require that they are *linear*?

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  - $|r - c| \leq .5, |c - p| \leq .5, |r - p| \leq .5$



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$$\begin{array}{lll} r - c \leq .5 & c - p \leq .5 & r - p \leq .5 \\ c - r \leq .5 & p - c \leq .5 & p - r \leq .5 \end{array}$$

## Example 3: Facility Location (Harder Problem)

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- Given coordinates for  $n$  roommates  $(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$

## Example 3: Facility Location (Harder Problem)

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- Given coordinates for  $n$  roommates  $(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$
- Goal: find location for a router that minimizes the average distance to each roommate

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- Cannot have distance more than 10 from any roommate

## Example 3: Facility Location

---

Objective:

Constraints:

- Given roommates at  $(3, 4)$  and  $(13, 5)$
- Goal: find location for a router that minimizes the average distance to each roommate
- Distance from  $(x, y)$  to  $(x_i, y_i)$  is  $|x - x_i| + |y - y_i|$
- Cannot have distance  $> 10$  from any roommate



## Example 3: Facility Location

---

Objective:

Constraints:

$$(x - 3) + (y - 4) \leq 10$$

$$(-x + 3) + (y - 4) \leq 10$$

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Can't make  
objective  
function.

Idea: add  
new variables!

- Given roommates at  $(3, 4)$  and  $(13, 5)$
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- Distance from  $(x, y)$  to  $(x_i, y_i)$  is  $|x - x_i| + |y - y_i|$
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## Example 3: Facility Location

---

Objective:  $\min d_1 + d_2$

Constraints:

- Given roommates at  $(3, 4)$  and  $(13, 5)$
- Goal: find location for a router that minimizes the average distance to each roommate
- Distance from  $(x, y)$  to  $(x_i, y_i)$  is  $|x - x_i| + |y - y_i|$
- Cannot have distance  $> 10$  from any roommate

## Example 3: Facility Location

---

Objective:  $\min d_1 + d_2$

Constraints:

$$(x - 3) + (y - 4) \leq d_1$$

$$(-x + 3) + (y - 4) \leq d_1$$

$$(-x + 3) + (-y + 4) \leq d_1$$

$$(x - 3) + (-y + 4) \leq d_1$$

$$(x - 13) + (y - 5) \leq d_2$$

$$(-x + 13) + (y - 5) \leq d_2$$

$$(-x + 13) + (-y + 5) \leq d_2$$

$$(x - 13) + (-y + 5) \leq d_2$$

$$d_1 \leq 10$$

$$d_2 \leq 10$$

- Given roommates at  $(3, 4)$  and  $(13, 5)$
- Goal: find location for a router that minimizes the average distance to each roommate
- Distance from  $(x, y)$  to  $(x_i, y_i)$  is  $|x - x_i| + |y - y_i|$
- Cannot have distance  $> 10$  from any roommate

## Example 3: Facility Location (Equations Simplified)

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Objective:  $\min d_1 + d_2$

Constraints:

$$\begin{aligned}x + y - d_1 &\leq 7 \\ -x + y - d_1 &\leq 1 \\ -x - y - d_1 &\leq -7 \\ x - y - d_1 &\leq -1 \\ x + y - d_2 &\leq 18 \\ -x + y - d_2 &\leq -8 \\ -x - y - d_2 &\leq -18 \\ x - y - d_2 &\leq 8\end{aligned}$$

- Given roommates at  $(3, 4)$  and  $(13, 5)$
- Goal: find location for a router that minimizes the average distance to each roommate
- Distance from  $(x, y)$  to  $(x_i, y_i)$  is  $|x - x_i| + |y - y_i|$
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# Proving Correctness

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- How can we show that the above LP works?

# Proving Correctness

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- Idea: an LP is feasible if and only if it corresponds to a correct router placement



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- Idea: an LP is feasible if and only if it corresponds to a correct router placement
- 1st: if there exists a feasible LP solution has values  $d_1, d_2, x, y$  then there exists a router placement at  $(x, y)$  with distance *at most*  $d_1$  and  $d_2$  from roommates 1 and 2, with  $d_1 \leq 10$  and  $d_2 \leq 10$

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- 2nd: any placement of a router at location  $(x, y)$ , with distance  $d_1 \leq 10$  and  $d_2 \leq 10$  from the first and second roommate respectively corresponds to a feasible LP solution with variables  $d_1, d_2, x, y$

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- 2nd: any placement of a router at location  $(x, y)$ , with distance  $d_1 \leq 10$  and  $d_2 \leq 10$  from the first and second roommate respectively corresponds to a feasible LP solution with variables  $d_1, d_2, x, y$
- If we can prove these claims then solving this LP solves the router placement problem: we get the min total distance placement

# Proving Correctness

## Lemma

*If there exists a feasible LP solution with variables  $d_1, d_2, x, y$  then a router at  $(x, y)$  has distance at most  $d_1$  and  $d_2$  from roommates 1 and 2, with  $d_1 \leq 10$  and  $d_2 \leq 10$*

*Proof:* Router at  $(x, y)$  has distance  $\hat{d}_1 = |x - 3| + |y - 4|$  from roommate 1. Because the LP soln is feasible, we have:

$$\begin{array}{ll} (x - 3) + (y - 4) \leq d_1 & (-x + 3) + (y - 4) \leq d_1 \\ (-x + 3) + (-y + 4) \leq d_1 & (x - 3) + (-y + 4) \leq d_1 \end{array}$$

Since  $\hat{d}_1$  is equal to the left side of one of these equations,  $\hat{d}_1 \leq d_1$ . Furthermore, since the LP solution is feasible,  $d_1 \leq 10$ , so  $\hat{d}_1 \leq 10$ .

Same argument works for roommate 2

# Proving Correctness

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## Lemma

*Any placement of a router at location  $(x, y)$ , with distance  $d_1 \leq 10$  and  $d_2 \leq 10$  from the first and second roommate respectively corresponds to a feasible LP solution with variables  $d_1, d_2, x, y$*

*Proof summary:* We have  $d_1, d_2 \leq 10$  by definition. We need to show the roommate constraints are satisfied. Let's focus on  $d_1$ . We have  $d_1 = |x - 3| + |y - 4|$ .

For any  $x, y$  we have:

$$x - 3 \leq |x - 3|$$

$$-x + 3 \leq |x - 3|$$

$$y - 4 \leq |y - 4|$$

$$-y + 4 \leq |y - 4|$$

Substituting, all equations for  $d_1$  are satisfied.

# Proving Correctness

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- Therefore, the best LP solution gives the best router placement!

# Proving Correctness

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- Therefore, the best LP solution gives the best router placement!
- So we can solve this problem by solving an LP

# Router Example Discussion

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- Can we add new roommates?



# Router Example Discussion

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- New constraints? (E.g. can't have the router in a certain portion of the house, or can't be too close to one of the roommates)

# Router Example Discussion

---

- Can we add new roommates?
  - Yes!
- New constraints? (E.g. can't have the router in a certain portion of the house, or can't be too close to one of the roommates)
  - Yes—if they're linear