CS7800: Advanced Algorithms

Class 9: Linear Programming 1

- · Basic Concepts
- · Simplex Method

Jonathan Ullman October 3, 2025

Exam Info

Topics

- · Greedy
 - "Greedy Stays Ahead"
 - "Exchange Argument"
 - Minimum spanning tree
- · Dynamic programming
 - Weighted interval scheduling
 - Segmented least squares
 - Knapsack Bellman Ford

Strong Duality Thm

- · Flows
 - Definitions/concepts
 - Ford-Fulkerson alg
 - Max Flow / Mm Cut Theorem
 - Foster algorithms
 - · 90 mmutes
 - · One 8.5" ×11" sheet of notes

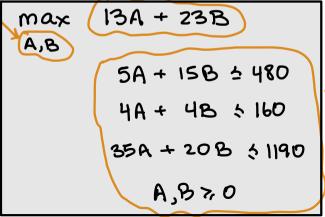
Our Favorite Linear Program

How to maximize the brewery's profits?

Beverage	Corn (pounds)	Hops (ounces)	Malt (pounds)	Profit (\$)
Ale (barrel)	5	4	35	13
Beer (barrel)	15	4	20	23
constraint	480	160	1190	

34 ale, 0 beer ⇒ \$442 0 ale, 32 beer ⇒ \$736 7.5 ale, 29.5 beer ⇒ \$776 12 ale, 28 beer ⇒ \$800

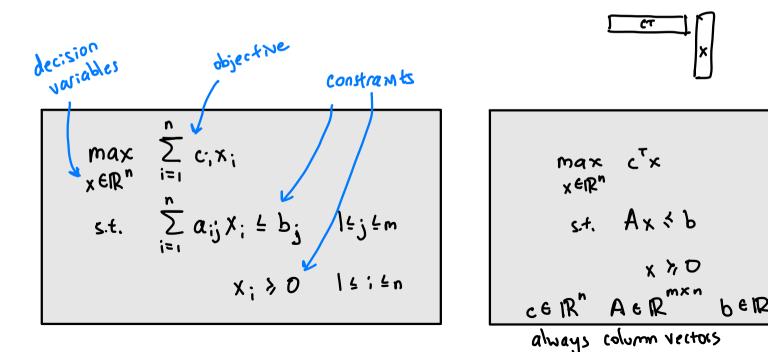
linear objective function



linear mequality

Linear Programming

Optimize a linear objective subject to linear constraints



Standard Form LPs

max
$$\sum_{i=1}^{n} e_i x_i$$

 $x \in \mathbb{R}^n$ $i=1$
 $s.t.$ $\sum_{i=1}^{n} A_{ii}$
 $x_i > 0$ $1 \le i \le n$

Also called slack form for reasons that may be clear soon

Transformation Rules

• Equality to mequality $a^{T}x=b \Rightarrow a^{T}x + b$

• Inequality to equality $a^{T}x \Leftrightarrow b \implies a^{T}x + s = b$

* Minimize to maximize

 $\min_{\mathbf{x}} \mathbf{c}^{\mathsf{T}} \mathbf{x} \implies \max_{\mathbf{x}} \mathbf{c}^{\mathsf{T}} \mathbf{x}$

· Unconstrained to nonnegative

$$x_{i} \in \mathbb{R} \implies x_{i} = x_{i}^{+} - x_{i}^{-}$$

$$x_{i}^{+} x_{i}^{-} > 0$$

Our Favorite LP (now in standard form)

How to maximize the brewery's profits?

Beverage	Corn (pounds)	Hops (ounces)	Malt (pounds)	Profit (\$)
Ale (barrel)	5	4	35	13
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constraint	480	160	1190	

standard form ("slack form")

max
$$13A + 23B$$

A,B,Sc,SH,SM
s.t. $5A + 15B + Sc = 480$
 $4A + 4B + SH = 160$
 $35A + 20B + SM = 1190$
 $A,B,Sc,SH,SM > 0$

Examples of Linear Programs

```
Max Flow
          ∑ fle)
e aut of s
max
{f(e)}eEE
     \sum f(e) - \sum f(e) = 0 (consevotion)
                  e mes
          fle) & cle) (capacity constraint)
          fle) > 0 (non-negativity)
```

```
Mn Cost Plow
min Z $(e).fle)
flex eE
     I f(e) > V (denand)
    eartofs
   5 f(e) - I f(e) = 0 (conservation)
 e out of v e mb v
      fle) & cle) (capacity constraint)
      fle) > 0 (non-negativity)
```

Examples of Linear Programs

Max Bipartite Matching

max Z x(e)

x(e) eeE

x(e) \(\sum_{e \text{inc:dest}} \)

X(e) E {0, 1} Integer Linear Programming
INTRACTABLE

What can ve do for these problems?

01 V

- 1) Show that optimal solutions are integral
- @ Rounding to an integer solution (approximation algorithms)

Algebra

```
max 13A + 23B

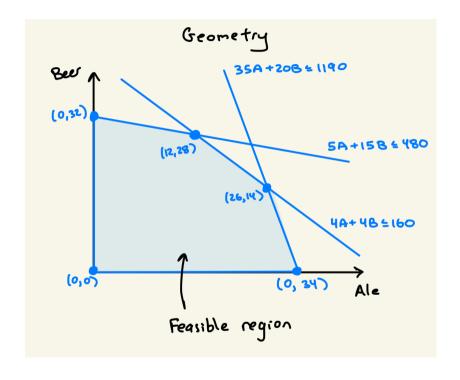
A,B

5.1. 5A+ 15B \( \) 480

4A + 4B \( \) 160

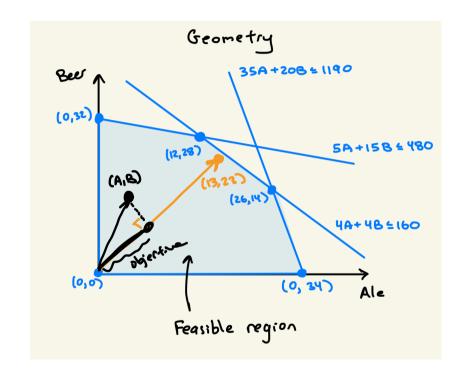
35A + 20B \( \) 1190

A,B\( \) 0
```



Algebra

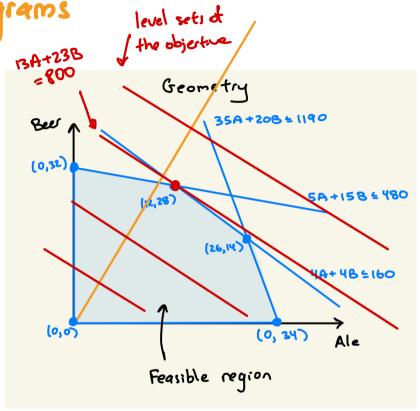
Goal: Find feasible point with the longest projection onto the line defined by the objective



Algebra

max 13A + 23B A,B S.t. 5A+ 15B \(\) 480 4A + 4B \(\) 160 35A + 20B \(\) 1190 A,B \(\) 0

Goal: Find feasible point with the longest projection onto the line defined by the objective



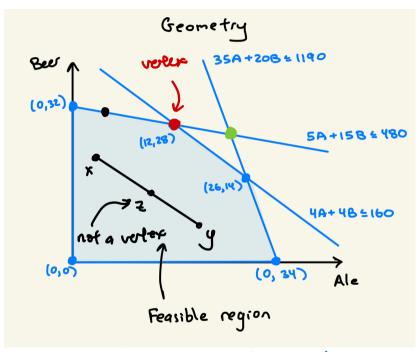
Convexity

A set of points P is convex if for every pair of points x,yEP and every & \(\in (0,i) \) ax + (1-a) y \(\in P \)

A point ZEP is a vertex of P

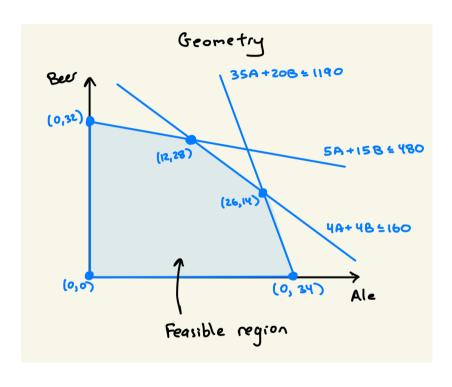
if z cannot be written as an

average of two distinct points x, y EP



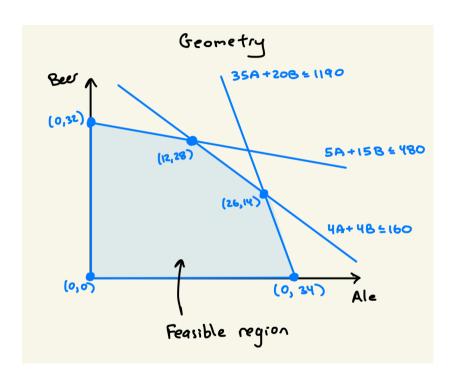
Fact: Vertices are formed by the intersection of n independent constraints

Theorem: If the LP has an optimal solution, then it has an optimal solution at a vertex of the feasible region.



Basic Feasible Solutions (Geometry)

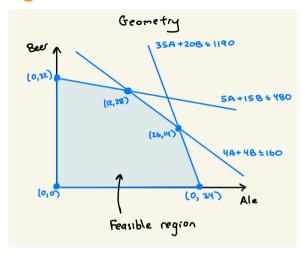
Basic Feasible Solutions:



Basic Feasible Solutions (Algebra)

Slack form LP

constraint matrix



The Simplex Algorithm (30,000' View)

Given an LP in standard form

max c^Tx x Ax=b x>0

Simplex algorithm

- Start with a BFS xo
corresponding to constraint set So - Repeat until optimality: How? - Find on adjacent BFS X; corresponding to contrast cTx; > cT x;-1

Thm: Only terminates at an optimal solution

Simplex in Practice

Theory: Might need exponentially many proofs to termmate

Practice: Can solve LPs with millions of variables/constraints (usually = 2(n+m) proofs)

Many Issues to Resolve:

- 1) What if the UP in infearible / unbounded?
- 2 How to choose a good pivot Ne?
- 3 How to avoid cycling?
- 4 HOW to maintain sparsity?
- (5) How to be numerically stable?
- 1) How to preprocess the LP to be smaller?