

CH-231-A

Algorithms and Data Structures

ADS

Lecture 41

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Spring 2020

General: Linear Programming¹

- ▶ Important tool for optimal allocation of scarce resources, among a number of competing activities.
- ▶ Powerful and general problem-solving method.
- ▶ Applications:
 - ▶ Computer science: Compiler register allocation, data mining.
 - ▶ Electrical engineering: VLSI design, optimal clocking.
 - ▶ Economics: Equilibrium theory, two-person zero-sum games.
 - ▶ Environment: Water quality management.
 - ▶ Logistics: Supply-chain management, Berlin airlift.
 - ▶ Manufacturing: Production line balancing, cutting stock.
 - ▶ Telecommunication: Network design, Internet routing.

¹Source of slides: Kevin Wayne: Algorithms and Data Structures, Spring 2004, Princeton University

Brewery Problem: A Toy LP Example

Small brewery produces ale and beer.

- ▶ Production limited by scarce resources: corn, hops, malt.
- ▶ Recipes for ale and beer require different proportions of resources.

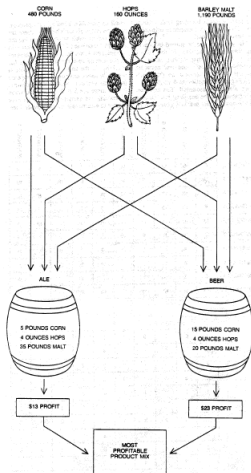
Beverage	Corn (pounds)	Hops (ounces)	Malt (pounds)	Profit (\$)
Ale	5	4	35	13
Beer	15	4	20	23
Quantity	480	160	1190	

How can brewer maximize profits?

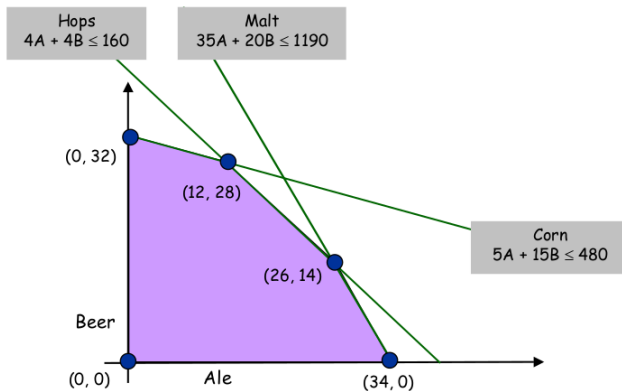
- ▶ Devote all resources to ale: 34 barrels of ale → \$442.
- ▶ Devote all resources to beer: 32 barrels of beer → \$736.
- ▶ 7.5 barrels of ale, 29.5 barrels of beer → \$776.
- ▶ 12 barrels of ale, 28 barrels of beer → \$800.

Brewery Problem

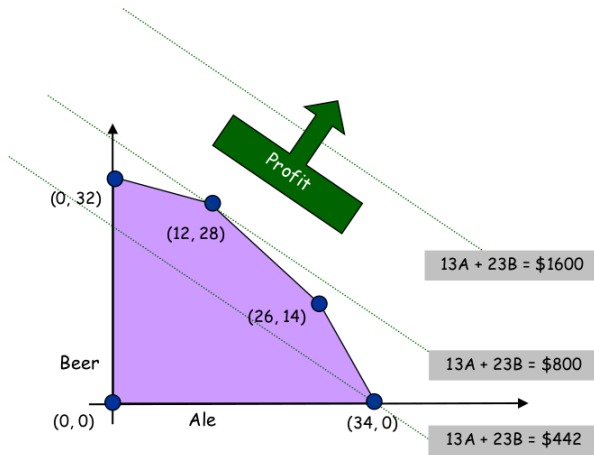
	Ale	Beer	
max	$13A + 23B$		Profit
s. t.	$5A + 15B \leq 480$		Corn
	$4A + 4B \leq 160$		Hops
	$35A + 20B \leq 1190$		Malt
	$A, B \geq 0$		



Brewery Problem: Feasible Region

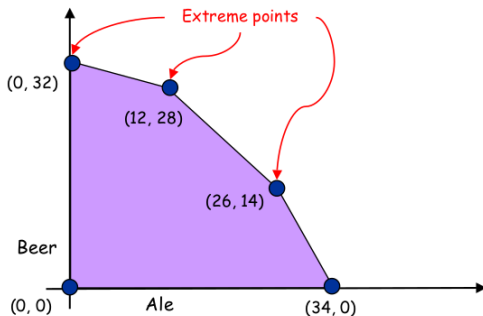


Brewery Problem: Objective Function



Brewery Problem: Geometry

Observation: Regardless of objective function coefficients, an optimal solution occurs at an extreme point.



Linear Programming: Standard Form

Standard form:

- ▶ Input: real numbers c_j , b_i , a_{ij} .
- ▶ Output: real numbers x_j .
- ▶ $n = \#$ nonnegative variables, $m = \#$ constraints.
- ▶ Maximize linear objective function subject to linear inequalities.

$$\begin{aligned} \text{(P)} \quad & \max \sum_{j=1}^n c_j x_j \\ \text{s.t.} \quad & \sum_{j=1}^n a_{ij} x_j = b_i \quad 1 \leq i \leq m \\ & x_j \geq 0 \quad 1 \leq j \leq n \end{aligned}$$

$$\begin{aligned} \text{(P)} \quad & \max c^T x \\ \text{s.t.} \quad & Ax = b \\ & x \geq 0 \end{aligned}$$

Linear: No x^2 , xy , $\arccos(x)$, etc.

Programming: Planning (term predates computer programming).

Brewery Problem: Converting to Standard Form

Original input:

$$\begin{array}{ll}\max & 13A + 23B \\ \text{s. t.} & 5A + 15B \leq 480 \\ & 4A + 4B \leq 160 \\ & 35A + 20B \leq 1190 \\ & A, B \geq 0\end{array}$$

Standard form:

- ▶ Add **slack** variable for each inequality.
- ▶ Now a 5-dimensional problem.

$$\begin{array}{llllll}\max & 13A + 23B & & & & \\ \text{s. t.} & 5A + 15B + S_C & & & & = 480 \\ & 4A + 4B & + S_H & & & = 160 \\ & 35A + 20B & & + S_M & & = 1190 \\ & A, B, S_C, S_H, S_M & \geq & 0 & & \end{array}$$

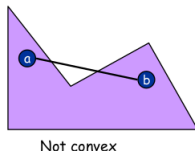
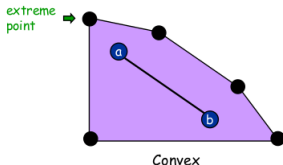
Geometry (1)

Geometry:

- ▶ Inequalities : halfplanes (2D), hyperplanes.
- ▶ Bounded feasible region: convex polygon (2D), (convex) polytope.

Convex: if a and b are feasible solutions, then so is $(a + b)/2$.

Extreme point: feasible solution x that cannot be written as $(a + b)/2$ for any two distinct feasible solutions a and b .



Geometry (2)

Extreme point property: If there exists an optimal solution to (P), then there exists one that is an extreme point.

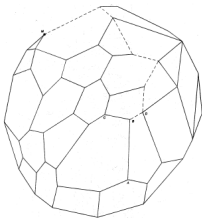
- Only need to consider finitely many possible solutions.

Challenge: Number of extreme points can be exponential.

- Consider n -dimensional hypercube.

Greedy: Local optima are global optima.

- Extreme point is optimal if no neighboring extreme point is better.



Simplex Algorithm

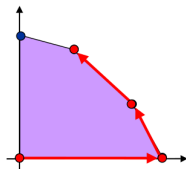
Simplex algorithm: (George Dantzig, 1947)

- ▶ Developed after WWII in response to logistical problems.
- ▶ Used for 1948 Berlin airlift.

Generic algorithm:

- ▶ Start at some extreme point.
- ▶ Pivot from one extreme point to a neighboring one. (never decrease objective function)
- ▶ Repeat until optimal.

How to implement? Linear algebra.



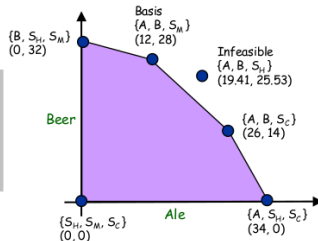
Simplex Algorithm: Basis

Basis: Subset of m of the n variables.

Basic feasible solution (BFS): Set $n - m$ nonbasic variables to 0, solve for remaining m variables.

- Solve m equations in m unknowns.
- If unique and feasible solution \rightarrow BFS.
- BFS corresponds to extreme point.
- Simplex only considers BFS.

$$\begin{array}{llllll} \max & 13A & + & 23B & & \\ \text{s. t.} & 5A & + & 15B & + & S_C & = & 480 \\ & 4A & + & 4B & & + & S_H & = & 160 \\ & 35A & + & 20B & & & + & S_M & = & 1190 \\ & A & , & B & , & S_C & , & S_H & , & S_M & \geq & 0 \end{array}$$



Simplex Algorithm: Pivot 1 (1)

$$\begin{array}{rcll}
 \text{max } Z \text{ subject to} & & & \\
 13A + 23B & & - Z = & 0 \\
 \hline
 5A + 15B + S_C & & = & 480 \\
 4A + 4B + S_H & & = & 160 \\
 35A + 20B + S_M & & = & 1190 \\
 A, B, S_C, S_H, S_M & \geq & & 0
 \end{array}$$

$$\text{Basis} = \{S_C, S_H, S_M\}$$

$$A = B = 0$$

$$Z = 0$$

$$S_C = 480$$

$$S_H = 160$$

$$S_M = 1190$$

Substitute: $B = 1/15 (480 - 5A - S_C)$

$$\begin{array}{rcll}
 \text{max } Z \text{ subject to} & & & \\
 \frac{16}{3}A - \frac{23}{15}S_C & & - Z = & -736 \\
 \hline
 \frac{1}{3}A + B + \frac{1}{15}S_C & & = & 32 \\
 \frac{8}{3}A - \frac{4}{15}S_C + S_H & & = & 32 \\
 \frac{85}{3}A - \frac{4}{3}S_C + S_M & & = & 550 \\
 A, B, S_C, S_H, S_M & \geq & & 0
 \end{array}$$

$$\text{Basis} = \{B, S_H, S_M\}$$

$$A = S_C = 0$$

$$Z = 736$$

$$B = 32$$

$$S_H = 32$$

$$S_M = 550$$

Simplex Algorithm: Pivot 1 (2)

$$\begin{array}{rcll}
 \max Z \text{ subject to} & & & \\
 13A + 23B & & - Z = & 0 \\
 \hline
 5A + 15B + S_C & & = & 480 \\
 4A + 4B + S_H & & = & 160 \\
 35A + 20B + S_M & & = & 1190 \\
 A, B, S_C, S_H, S_M & & \geq & 0
 \end{array}$$

Basis = $\{S_C, S_H, S_M\}$

$A = B = 0$

$Z = 0$

$S_C = 480$

$S_H = 160$

$S_M = 1190$

Why pivot on column 2?

- ▶ Each unit increase in B increases objective value by \$23.
- ▶ Pivoting on column 1 also OK.

Why pivot on row 2?

- ▶ Preserves feasibility by ensuring $RHS \geq 0$.
- ▶ Minimum ratio rule: $\min\{480/15, 160/4, 1190/20\}$.

Simplex Algorithm: Pivot 2

$$\begin{array}{rcll}
 \text{max } Z \text{ subject to} & & & \\
 \frac{16}{3} A & - & \frac{23}{15} S_C & - Z = -736 \\
 \hline
 \frac{1}{3} A + B + \frac{1}{15} S_C & & & = 32 \\
 \frac{8}{3} A & - & \frac{4}{15} S_C + S_H & = 32 \\
 \frac{85}{3} A & - & \frac{4}{3} S_C + S_M & = 550 \\
 A, B, S_C, S_H, S_M & \geq & 0 &
 \end{array}$$

$$\text{Basis} = \{B, S_H, S_M\}$$

$$A = S_C = 0$$

$$Z = 736$$

$$B = 32$$

$$S_H = 32$$

$$S_M = 550$$

$$\text{Substitute: } A = \frac{3}{8} (32 + \frac{4}{15} S_C - S_H)$$

$$\begin{array}{rcll}
 \text{max } Z \text{ subject to} & & & \\
 & - & S_C - 2 S_H & - Z = -800 \\
 \hline
 & B + \frac{1}{10} S_C + \frac{1}{8} S_H & & = 28 \\
 A & - \frac{1}{10} S_C + \frac{3}{8} S_H & & = 12 \\
 & - \frac{25}{6} S_C - \frac{85}{8} S_H + S_M & & = 110 \\
 A, B, S_C, S_H, S_M & \geq & 0 &
 \end{array}$$

$$\text{Basis} = \{A, B, S_M\}$$

$$S_C = S_H = 0$$

$$Z = 800$$

$$B = 28$$

$$A = 12$$

$$S_M = 110$$

Simplex Algorithm: Optimality

When to stop pivoting?

- ▶ If all coefficients in top row are non-positive.

Why is resulting solution optimal?

- ▶ Any feasible solution satisfies system of equations in tableaux.
- in particular: $Z = 800 - S_C - 2S_H$
- ▶ Thus, optimal objective value $Z^* \leq 800$ since $S_C, S_H \geq 0$.
- ▶ Current BFS has value 800 \rightarrow optimal.

max Z subject to						
	-	S_C	- 2	S_H	-	$Z = -800$
	B	+	$\frac{1}{10} S_C$	+	$\frac{1}{8} S_H$	= 28
A	-	$\frac{1}{10} S_C$	+	$\frac{3}{8} S_H$	=	12
	-	$\frac{25}{6} S_C$	-	$\frac{85}{8} S_H$	+	$S_M = 110$
A, B, S_C, S_H, S_M					\geq	0

Basis = $\{A, B, S_M\}$
 $S_C = S_H = 0$
 $Z = 800$
 $B = 28$
 $A = 12$
 $S_M = 110$

Simplex Algorithm: Issues

Remarkable property: In practice, simplex algorithm typically terminates in at most $2(m + n)$ pivots.

- ▶ No polynomial pivot rule known.
- ▶ Most pivot rules known to be exponential in worst-case.

Issues: Which neighboring extreme point?

Degeneracy: New basis, same extreme point.

- ▶ "Stalling" is common in practice.

Cycling: Get stuck by cycling through different bases that all correspond to same extreme point.

- ▶ Does not occur in the wild.
- ▶ Bland's least index rule \rightarrow finite # of pivots.