

Design and Analysis of Algorithms Linear Programming

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- An Example
- Standard Form
- Geometry
- Linear Algebra
- Simplex Algorithm

Linear Programming

Linear programming. Optimize a linear function subject to linear inequalities.

$$\max \sum_{j=1}^{n} c_j x_j$$

$$s. t. \sum_{j=1}^{n} a_{ij} x_j \ge b_i \quad 1 \le i \le m$$

$$x_j \ge 0 \quad 1 \le j \le n$$

Ranked among most important scientific advances of 20th century.

Linear Programming

Linear programming. Optimize a linear function subject to linear inequalities.

Generalizes: AX=B, 2-person zero-sum games, shortest path, max flow, assignment problem, ...



Brewery Problem

Small brewery produces ale and beer.

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

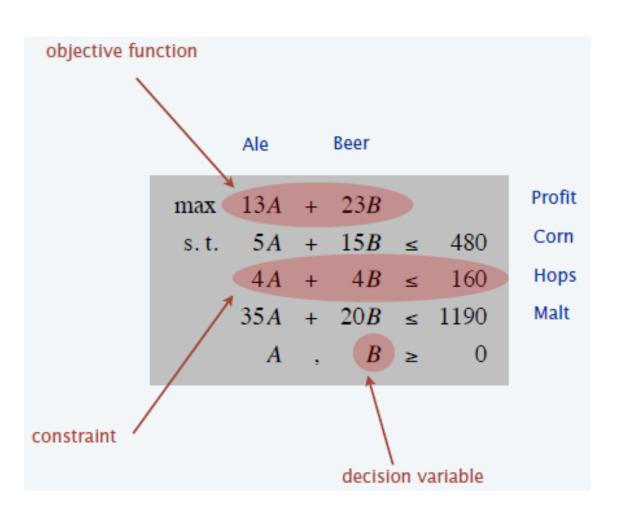
Beverage	Corn (pounds)	Hops (ounces)	Malt (pounds)	Profit (\$)
Ale (barrel)	5	4	35	13
Beer (barrel)	15	4	20	23
constraint	480	160	1190	

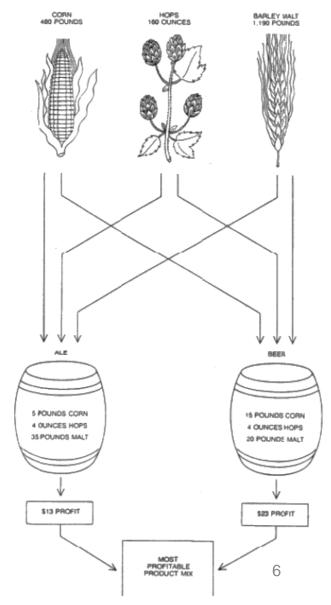
How can brewer maximize profit?

- 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

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Brewery Problem







Standard Form

"Standard form" of a linear program.

- Input: real numbers a_{ij} , c_j , b_i .
- Output: real numbers x_i .
- n = # decision variables, m = # constraints.
- Maximize linear objective function subject to linear equalities.

$$\max \sum_{j=1}^{n} c_j x_j$$

$$s. t. \sum_{j=1}^{n} a_{ij} x_j = b_i \quad 1 \le i \le m$$

$$x_j \ge 0 \quad 1 \le j \le n$$



Brewery Problem: Converting to Standard Form

Original input.

```
\max \quad 13A + 23B
s. t. 5A + 15B \le 480
4A + 4B \le 160
35A + 20B \le 1190
A , B \ge 0
```

Standard form?

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



Basic and Non-basic Variables

Basic variables are selected arbitrarily with the restriction that there will be as many basic variables as the equations. The remaining variables are non-basic variables.

$$x_1 + 2x_2 + s_1 = 32$$
$$3x_1 + 4x_2 + s_2 = 84$$

This system has two equations, we can select any two of the four variables as basic variables. The remaining two variables are then non-basic variables. A solution found by setting the two non-basic variables equal to 0 and solving for the two basic variables is a basic solution. If a basic solution has no negative values, it is a basic feasible solution.

Equivalent Forms

Easy to convert variants to standard form.

$$\max c^T x$$
s. t. $Ax = b$

$$x \ge 0$$

Less than to equality.

$$x + 2y - 3z \le 17$$

Greater than to equality.

$$x + 2y - 3z \ge 17$$

Min to max.

$$\min x + 2y - 3z$$

Unrestricted to nonnegative.

x unrestricted

Equivalent Forms

Easy to convert variants to standard form.

$$\max c^T x$$
s. t. $Ax = b$

$$x \ge 0$$

Less than to equality.

$$x + 2y - 3z \le 17 \rightarrow x + 2y - 3z + s = 17, s \ge 0$$

Greater than to equality.

$$x + 2y - 3z \ge 17 \rightarrow x + 2y - 3z - s = 17, s \ge 0$$

Min to max.

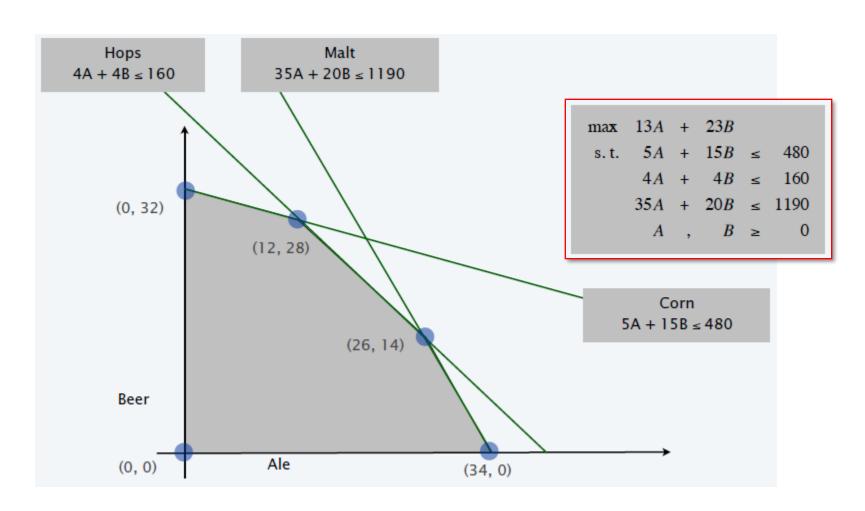
$$\min x + 2y - 3z \rightarrow \max -x - 2y + 3z$$

Unrestricted to nonnegative.

$$x$$
 unrestricted $\rightarrow x = x^+ - x^-, x^+ \ge 0, x^- \ge 0$



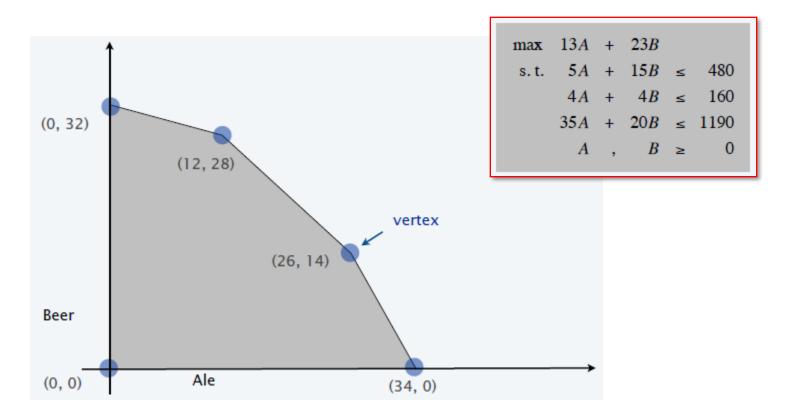
Brewery Problem: Feasible Region





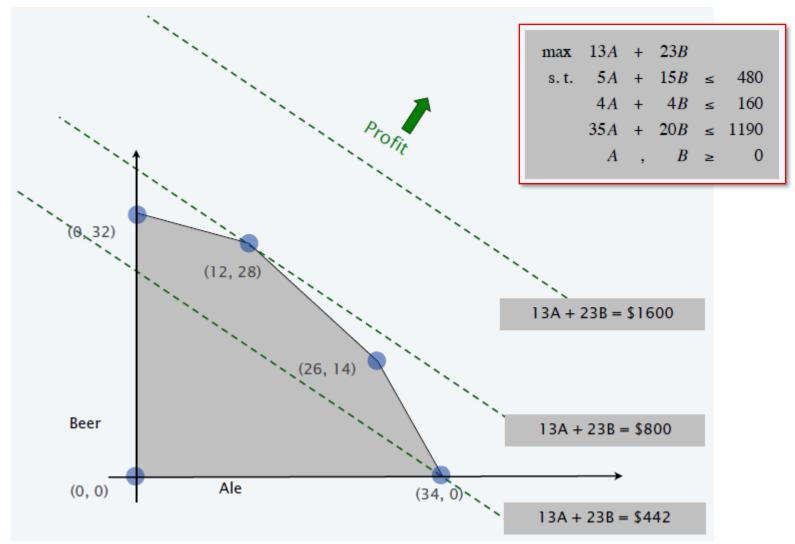
Brewery Problem: Geometry

Brewery problem observation. Regardless of objective function coefficients, an optimal solution occurs at a vertex.





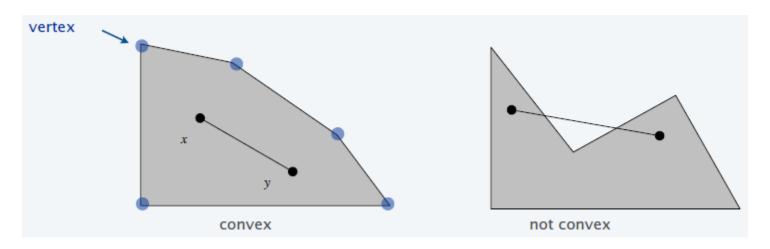
Brewery Problem: Objective Function



Convexity

Convex set. If two points x and y are in the set, then so is $\lambda x + (1 - \lambda)y$ for $0 \le \lambda \le 1$.

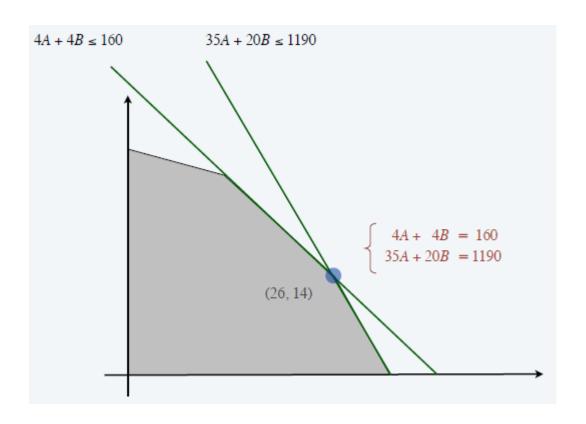
Vertex. A point x in the set that can't be written as a strict convex combination of two distinct points in the set.



Observation. LP feasible region is a convex set.

Vertex

Intuition. A vertex in \mathbb{R}^m is uniquely specified by m linearly independent equations.



Vertex

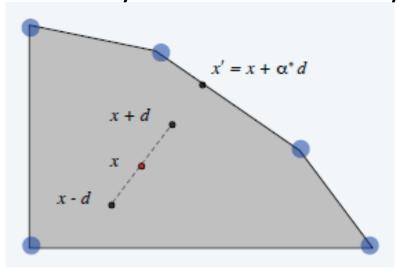
Theorem. If there exists an optimal solution to (P), then there exists one that is a vertex.

$$\max c^T x$$

$$s.t. Ax = b$$

$$x \ge 0$$

Intuition. If the optimum is not a vertex, move in a non-decreasing direction until you reach a boundary.





Theorem. If there exists an optimal solution to (P), then there exists one that is a vertex.

Pf.

Since there exists an optimal solution, there exists an optimal solution x with a minimal number of non-zero components.

Suppose x is not a vertex, so that

$$x = \lambda u + (1 - \lambda)v,$$

for some $u \neq v, \lambda \in (0,1)$.

Vertex

Theorem. If there exists an optimal solution to (P), then there exists one that is a vertex.

Since x is optimal, $c^Tu \le c^Tx$ and $c^Tv \le c^Tx$. But also $c^Tx = \lambda c^Tu + (1-\lambda)c^Tv$ so in fact $c^Tu = c^Tv = c^Tx$. Now consider the line defined by

$$x(\epsilon) = x + \epsilon(u - v)$$

Then

- Ax = Au = Av = b so $Ax(\epsilon) = b$ for all ϵ ,
- $c^T x(\epsilon) = c^T x$ for all ϵ ,
- If $x_i = 0$ then $u_i = v_i = 0$, which implies $x(\epsilon)_i = 0$ for all ϵ ,
- If $x_i > 0$ then $x(0)_i > 0$, and $x(\epsilon)_i$ is continuous in ϵ .

Vertex Vertex

Theorem. If there exists an optimal solution to (P), then there exists one that is a vertex.

So we can increase ϵ from zero, in a positive or a negative direction as appropriate, until at least one extra component of $x(\epsilon)$ becomes zero.

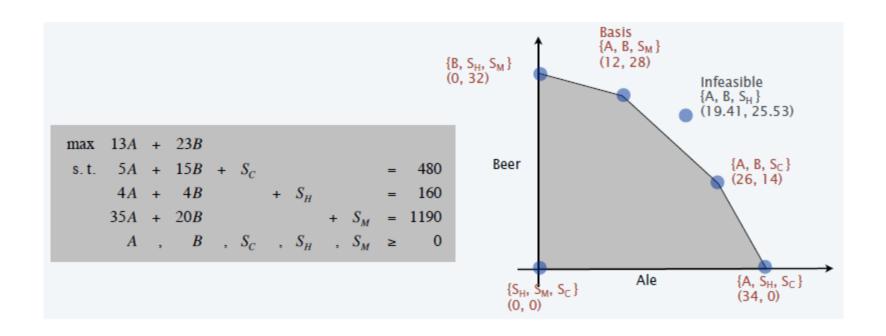
This gives an optimal solution $x(\epsilon)$ with fewer non-zero components than x.

So x must be a vertex.



Basic Feasible Solution: Example

Basic feasible solutions.

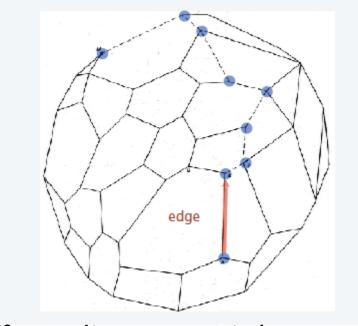




Simplex Algorithm: Intuition

Simplex algorithm. Move from BFS (Basic Feasible Solution) to adjacent BFS, without decreasing objective function (replace one

basic variable with another).



Greedy property. BFS optimal iff no adjacent BFS is better.



Simplex Algorithm: Initialization

max 2	Z su	bject t	0									
13 <i>A</i>	+	23 <i>B</i>						_	\boldsymbol{Z}	=	0	$Basis = \{S_C, S_E\}$
5 <i>A</i>	+	15 <i>B</i>	+	S_C						=	480	A = B = 0 $Z = 0$
4A	+	4 <i>B</i>			+	S_H				=	160	$S_C = 480$
35 <i>A</i>	+	20 <i>B</i>					+	S_M		=	1190	$S_H = 160$
\boldsymbol{A}	,	B	,	S_C	,	S_H	,	S_{M}		≥	0	$S_M = 1190$



Simplex Algorithm: Pivot 1

Basis = $\{S_C, S_H, S_M\}$ A = B = 0 Z = 0 $S_C = 480$ $S_H = 160$ $S_M = 1190$

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

bject to								
-	$\frac{23}{15} S_C$				-	\boldsymbol{Z}	=	-736
B +	$\frac{1}{15} S_C$						=	32
-	$\frac{4}{15} S_C$	+	S_H				=	32
-	$\frac{4}{3}$ S_C			+	S_{M}		=	550
B ,	S_C	,	S_H	,	S_M		≥	0
	B + P	$-\frac{4}{3}S_C$	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$

Basis = $\{B, S_H, S_M\}$ $A = S_C = 0$ Z = 736 B = 32 $S_H = 32$ $S_M = 550$



Simplex Algorithm: Pivot 1

- Q. Why pivot on column 2 (or 1)?
- A. Each unit increase in B increases objective value by \$23.
- Q. Why pivot on row 2.
- A. Preserves feasibility by ensuring RHS (Right Hand Side) ≥ 0 . (min ratio rule: min{480/15, 160/4, 1190/20})



Simplex Algorithm: Pivot 2

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

 $A = S_C = 0$
 $Z = 736$
 $B = 32$
 $S_H = 32$
 $S_M = 550$

Substitute: $A = 3/8 (32 + 4/15 S_C - S_H)$

max Z st	ıbjecı	to								
		-	S_C	-	$2 S_H$		-	\boldsymbol{Z}	=	-800
	В	+	$\frac{1}{10} S_C$	+	$\frac{1}{8}$ S_H				=	28
\boldsymbol{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	, <i>B</i>	,	S_C	,	S_H	,	S_M		≥	0

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



Simplex Algorithm: Optimality

- Q. When to stop pivoting?
- A. When all coefficients in top row are non-positive.
- Q. Why is the resulting solution optimal?
- A. Any feasible solution satisfies systems of equations in tableau.
- In particular: $Z = 800 S_C 2S_H$, $S_C \ge 0$, $S_H \ge 0$.
- Thus, optimal objective value $Z^* \leq 800$.
- Current BFS has value 800 -> optimal.

max Z subjec	t to									
	_	S_{C}	-	$2 S_H$		_	\boldsymbol{Z}	=	-800	Basis = $\{A, B, S\}$
В	+	$\frac{1}{10} S_C$	+	$\frac{1}{8}$ S_H				=	28	$S_C = S_H = 0$ $Z = 800$
\boldsymbol{A}	-	$\frac{1}{10} S_C$	+	$\frac{3}{8}$ S_H				=	12	B = 28
	-	$\frac{25}{6} S_C$	-	$\frac{85}{8} S_{H}$	+	S_M		=	110	$A = 12$ $S_M = 110$
A , B	,	S_C	,	S_H	,	S_{M}		≥	0	202

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Variant Tableau

The constraints are a linear system including m equations and n variables. m of the variables can be evaluated in terms of the other n-m variables

$$x_1 = b_1 - a_{1,m+1}x_{m+1} - \dots - a_{1,n}x_n$$

$$x_2 = b_2 - a_{2,m+1}x_{m+1} - \dots - a_{2,n}x_n$$

.

$$x_m = b_m - a_{m,m+1}x_{m+1} - \dots - a_{m,n}x_n$$

Objective function $z = \sum_{j=1}^{n} c_j x_j$

$$= \sum_{i=1}^{m} c_i b_i + \sum_{j=m+1}^{n} (c_j - \sum_{i=1}^{m} c_i a_{ij}) x_j.$$

Let $z^0 = \sum_{i=1}^m c_i b_i$, $\sigma_j = c_j - \sum_{i=1}^m c_i a_{ij}$, and we have

$$z = z^0 + \sum_{j=m+1}^{n} \sigma_{j} x_j$$
 indicator



Variant Tableau

	Сj	C 1	$C_2 \ldots C_m C_{m+1} \ldots C_n$	7	
Св	Хв	\mathcal{X}_1	\mathcal{X}_2 \mathcal{X}_m \mathcal{X}_{m+1} \mathcal{X}_n	b	θ
c_1	x_1	1	$0 \dots 0 a'_{1,m+1} \dots a'_{1n}$ $1 \dots 0 a'_{2,m+1} \dots a'_{2n}$ $\dots \dots$ $0 \dots 1 a'_{m,m+1} \dots a'_{mn}$	b_1'	
c_2	x_2	0	$1 \dots 0 a'_{2,m+1} \dots a'_{2n}$	b_2'	
•••					
C_{m}	X_m	0	$0 \dots 1 a'_{m,m+1} \dots a'_{mn}$	$b'_{\scriptscriptstyle m}$	
		l	$0 \ldots 0 c_{m+1} - \sum_{i=1}^{m} c_i a'_{i,m+1}$		

Variant Tableau

To solve a linear programming problem, use the following steps:

- 1. Convert each inequality in the set of constraints to an equation by adding slack variables.
- 2. Create the initial simplex tableau.
- 3. Select the pivot column (The column with the "most positive value" element in the last row).
- 4. Select the pivot row (The row with the smallest non-negative result when the last element in the row is divided by the corresponding in the pivot column).
- 5. Use elementary row operations calculate new values for the pivot row so that the pivot is 1.
- 6. Use elementary row operations to make all numbers in the pivot column equal to 0 except for the pivot.
- 7. If all entries in the bottom row are non-positive, this the final tableau. If not, go back to Step 3.



$$\max z = 2x_1 + 3x_2$$

$$s.t. \begin{cases} 2x_1 + x_2 \le 4 \\ x_1 + 2x_2 \le 5 \\ x_1, x_2 \ge 0 \end{cases}$$

$$\max z = 2x_1 + 3x_2$$

$$s.t.\begin{cases} 2x_1 + x_2 + x_3 = 4\\ x_1 + 2x_2 + x_4 = 5\\ x_1, x_2, x_3, x_4 \ge 0 \end{cases}$$



Pivot column. The column of the tableau representing the variable to be entered into the solution mix.

Pivot row. The row of the tableau representing the variable to be replaced in the solution mix.

Basic variable. Variables in the solution mix.

lni	itial tab	leau		Pivot column							
		C _j	2	3	0	0					
	Св	Хв	\mathbf{x}_1	X_2	\mathbf{x}_3	\mathbf{x}_4	b	θ	Min ratio rule		
	0	x ₃	2	1	1	0	4	4/1	/ 10.10		
	0	X_4	1	2	0	1	5	5/2			
Pivot row		$\sigma_{_{j}}$	2	3	0	0			32		



(C _j	2	3	0	0		
Св	Хв	\mathbf{x}_1	(x_2)	X_3	X ₄	b	θ
0	x ₃	2	1	1	0	4	4/1
0	X_4	1	2	0	1	5	5/2
	$\sigma_{_j}$	2	3	0	0		

- Since the entry 3 is the most positive entry in the last row of the tableau, the second column in the tableau is the pivot column.
- Divide each positive number of the pivot column into the corresponding entry in the column of constants. The ratio 5/2 is less then the ratio 4/1, so row 2 is the pivot row.

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(C _j	2	3	0	0		
Св	Хв	\mathbf{x}_1	\mathbf{x}_2	\mathbf{x}_3	\mathbf{x}_4	b	θ
0	x ₃	3/2	0	1	-1/2	3/2	1
3	\mathbf{x}_2	1/2	1	0	1/2	5/2	5
	$\sigma_{_{j}}$	1/2	0	0	-3/2		

- Since the entry 1/2 is the most positive entry in the last row of the tableau, the first column in the tableau is the pivot column.
- Divide each positive number of the pivot column into the corresponding entry in the column of constants. The ratio 3/2 is less then the ratio 5/2, so row 1 is the pivot row.



(C _j	2	3	0	0		
Св	Хв	\mathbf{x}_1	\mathbf{x}_2	\mathbf{x}_3	x ₄	b	θ
2	\mathbf{x}_1	1	0	2/3	-1/3	1	
3	x ₂	0	1	-1/3	2/3	2	
	$\sigma_{_{j}}$	0	0	-1/3	-4/3		

 The last row of the tableau contains no positive numbers, so an optimal solution has been reached.



Initial simplex tableau

$$c_B^T x_B + c_N^T x_N = Z$$

$$A_B x_B + A_N x_N = b$$

$$x_B , x_N \ge 0$$

Simplex tableau corresponding to basis B.

$$(c_N^T - c_B^T A_B^{-1} A_N) x_N = Z - c_B^T A_B^{-1} b \qquad \text{subtract } c_{\mathcal{B}}^T A_{\mathcal{B}}^{-1} \text{ times constraints}$$

$$I x_B + A_B^{-1} A_N x_N = A_B^{-1} b \qquad \text{multiply by } A_{\mathcal{B}}^{-1}$$

$$x_B \quad , \qquad x_N \geq 0$$

$$\begin{array}{ccc} x_B = A_B^{-1}b \geq 0 & & & & & \\ x_N = 0 & & & & & & \\ \end{array}$$

basic feasible solution

optimal basis



Standard form:

$$\max Z = C^T X$$
s. t. $AX = b$

$$X \ge 0$$

Let
$$A = [A_B, A_N]$$
, $X = \begin{bmatrix} X_B \\ X_N \end{bmatrix}$, $C = \begin{bmatrix} C_B \\ C_N \end{bmatrix}$, we have
$$A_B X_B + A_N X_N = b$$
$$\to X_B = A_B^{-1} b - A_B^{-1} A_N X_N$$

For the basis B,

$$Z = C^{T}X = [C_{B}^{T}, C_{N}^{T}] \begin{bmatrix} X_{B} \\ X_{N} \end{bmatrix} = C_{B}^{T}X_{B} + C_{N}^{T}X_{N}$$
$$= C_{B}^{T}(A_{B}^{-1}b - A_{B}^{-1}A_{N}X_{N}) + C_{N}^{T}X_{N}$$
$$= C_{B}^{T}A_{B}^{-1}b + (C_{N}^{T} - C_{B}^{T}A_{B}^{-1}A_{N})X_{N}$$

Matrix Form: Variant Tableau

	C_B^T	C_N^T	
	X_B^T	X_N^T	
$C_B X_B$	I	$A_B^{-1}A_N$	$A_B^{-1}b$
Indicator	0	$C_N^T - C_B^T A_B^{-1} A_N$	