

# Solving Linear Programming Problems

The Simplex Method

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# Outline

Key Takeaways from Last Lecture

LP Standard Form

A Naive Algorithm

Basic and Nonbasic Variables

Prototype example

The states of solutions

Simplex Algorithm

Simplex: Special cases

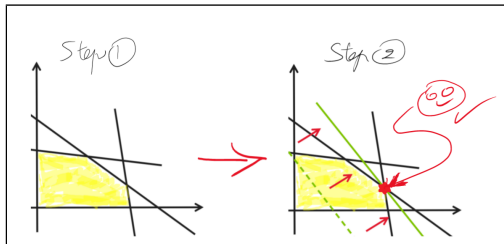
Conclusion

# Key Takeaways from Last Lecture

- If an optimal solution exists for the LP, then at least **one of the corner points is optimal.**

# Key Takeaways from Last Lecture

- If an optimal solution exists for the LP, then at least **one of the corner points is optimal**.
- The feasible region for any LP is a **convex set**.



# LP Standard Form

**Maximize**  $z = c_1x_1 + c_2x_2 + \cdots + c_nx_n$

**Subject to**  $a_{11}x_1 + a_{21}x_2 + \cdots + a_{n1}x_n = b_1,$

$$a_{12}x_1 + a_{22}x_2 + \cdots + a_{n2}x_n = b_2,$$

$$\vdots$$

$$a_{1m}x_1 + a_{2m}x_2 + \cdots + a_{nm}x_n = b_m,$$

$$x_i \geq 0 \quad (1 \leq i \leq n)$$

But ...

Why this LP Standard Form ?!

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$$x_i \geq 0 \quad (1 \leq i \leq n)$$

Because ...

To solve a system of equations.

# How to convert to the LP Standard Form

- Introduce non-negative slack and excess variable variables,  $S_j \geq 0$ , to transform the inequality constraints into equality constraints.
  - $\sum_{i=1}^n a_{ij}x_i \leq b_j$ , add a slack variable  $S_j \geq 0$  such that  $(\sum_{i=1}^n a_{ij}x_i) + S_j = b_j$ .
  - $\sum_{i=1}^n a_{ij}x_i \geq b_j$ , add a surplus variable  $S_j \geq 0$  such that  $(\sum_{i=1}^n a_{ij}x_i) - S_j = b_j$ .

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- When the Slack/Surplus/Original variable is null, it means that a constraint is satisfied with equality ( Binding constraint).
- Introduce two non-negative artificial variables,  $S_i^+ \geq 0$  and  $S_i^- \geq 0$ , to transform unrestricted variable  $x_i$  (no sign restriction).
  - $x_i = S_i^+ - S_i^-$

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- When the Slack/Surplus/Original variable is null, it means that a constraint is satisfied with equality ( Binding constraint).
- Introduce two non-negative artificial variables,  $S_i^+ \geq 0$  and  $S_i^- \geq 0$ , to transform unrestricted variable  $x_i$  (no sign restriction).
  - $x_i = S_i^+ - S_i^-$
- All the original variables  $x_i$  and the added variables  $S_i$  are positive.

# LP Standard Form

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$$x_i \geq 0 \quad (1 \leq i \leq n)$$

But ...

More variables than equations ( $m \ll n$ ).

# LP Standard Form

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$$a_{1m}x_1 + a_{2m}x_2 + \cdots + a_{nm}x_n = b_m,$$

$$x_i \geq 0 \quad (1 \leq i \leq n)$$

But ...

Take  $m$  variables and solve the system.

# Solving LP problem: A Naive Algorithm

- Generate all feasible corner points
  - Determine all the intersection points between constraints.
  - Test whether it is feasible.
- Use the objective function to determine which corner point is the optimal solution.

But ...

What is the **number of intersection points** for a LP problem with  $m$  constraints and  $n$  variables?

# Solving LP problem: A Naive Algorithm

- Generate all feasible corner points
  - Determine all the intersection points between constraints.
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But ...

$$\binom{n}{m} = \frac{n!}{m!(n-m)!}$$

# Solving LP problem: A Naive Algorithm

- Generate all feasible corner points
  - Determine all the intersection points between constraints.
  - Test whether it is feasible.
- Use the objective function to determine which corner point is the optimal solution.

But ...

$$\binom{30}{10} = 30045015 !!!$$



# Basic and Nonbasic Variables

- A **Basic Solution** is obtained by setting  $(n - m)$  variables to 0 and solving the remaining system of  $m$  variables.
- The remaining variables are the **Basic variables**, and the removed ones are the **Non-basic variables**.
- The non basic variables set to zero represent **the fully satisfied constraints**.
- A **Basic Feasible Solution (BFS)** is a basic solution that satisfies all of the constraints.

## Key Idea

A corner point in the feasible region of an LP is a Basic Feasible Solution (BFS).

**Basic Variables**

**Nonbasic Variables**

$$\begin{array}{l} x_1 = b_1 + \sum_{i=m+1}^n a_{1i}x_i \\ \vdots \\ x_m = b_m + \sum_{i=m+1}^n a_{mi}x_i \end{array}$$

## Basic Variables

## Nonbasic Variables

$$\begin{array}{l} x_1 = b_1 + \sum_{i=m+1}^n a_{1i}x_i \\ \vdots \\ x_m = b_m + \sum_{i=m+1}^n a_{mi}x_i \end{array}$$

Assign to 0

## Basic Variables

## Nonbasic Variables

$$x_1 = b_1 + \sum_{i=m+1}^n a_{1i}x_i$$

$\vdots$

$$x_m = b_m + \sum_{i=m+1}^n a_{mi}x_i$$

Assign to b's

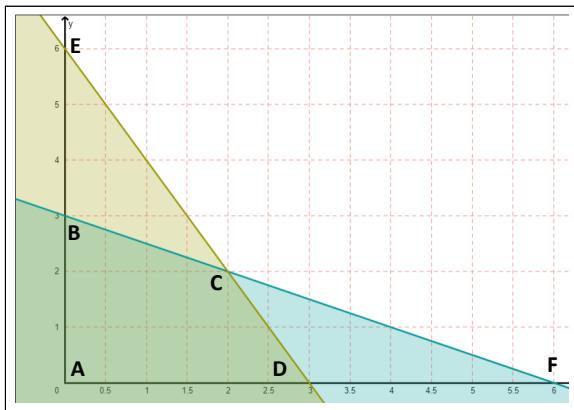
Assign to 0

## Example: Basic Feasible solutions (BFS's)

$$\begin{array}{ll}\text{Maximize} & x + y \\ \text{subject to} & x + 2y \leq 6 \\ & 2x + y \leq 6\end{array}$$

# Example: Basic Feasible Solutions (BFS's)

**Maximize**  $z = x + y$   
**subject to**  $x + 2y \leq 6$   
 $2x + y \leq 6$



## Example: Basic Feasible Solutions (BFS's)

- Variables  $\{x, y, S_1, S_2\}$

**Maximize**  $z = x + y$

**subject to**  $x + 2y + S_1 = 6$

$$2x + y + S_2 = 6$$

$$x, y, S_1, S_2 \geq 0$$

## Example: Basic Feasible Solutions (BFS's)

**Maximize**  $z = x + y$   
**subject to**  $x + 2y + S_1 = 6$   
 $2x + y + S_2 = 6$   
 $x, y, S_1, S_2 \geq 0$

- Variables  $\{x, y, S_1, S_2\}$
- Number of variables  $n = 4$
- Number of constraints  $m = 2$

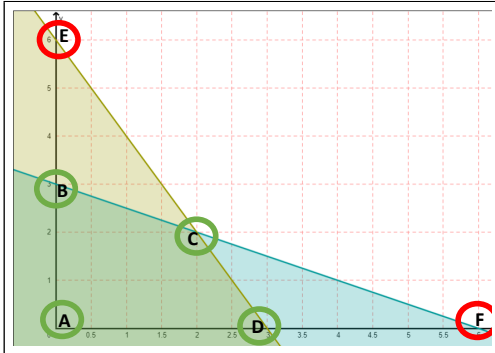


## Example: Basic Feasible Solutions (BFS's)

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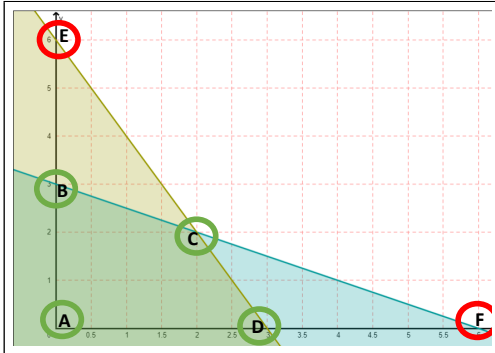
- Variables  $\{x, y, s_1, s_2\}$
- Number of variables  $n = 4$
- Number of constraints  $m = 2$
- Number of Basic solutions  $\binom{n}{m}$
- $\binom{4}{2} = 6$  Basic Solutions

# Example: Basic Feasible Solutions (BFS's)



- Basic solutions
  - A, B, C, D, E, F
- Feasible Basic Solutions
  - A, B, C, D

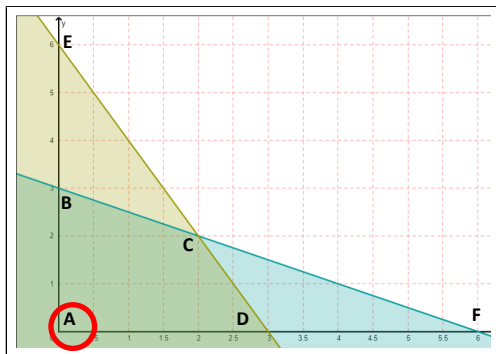
# Example: Basic Feasible Solutions (BFS's)



- Basic solutions
  - A, B, C, D, E, F
- Feasible Basic Solutions
  - A, B, C, D
- **How to find these FBS's ?!**

## Example: BFS $(s_1, s_2)$ - A -

**Maximize**  $z = x + y$   
**subject to**  $s_1 = 6 - x - 2y$   
 $s_2 = 6 - 2x - y$   
 $x, y, s_1, s_2 \geq 0$



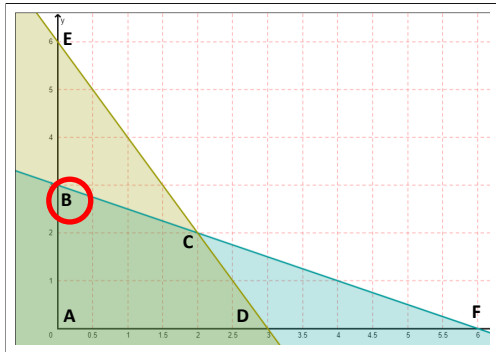
## Example: BFS $(y, s_2)$ - B -

**Maximize**  $z = 3 + \frac{1}{2}x - \frac{1}{2}s_1$

**subject to**  $y = 3 - \frac{1}{2}x - \frac{1}{2}s_1$

$s_2 = 3 - \frac{3}{2}x + \frac{1}{2}s_1$

$x, y, s_1, s_2 \geq 0$



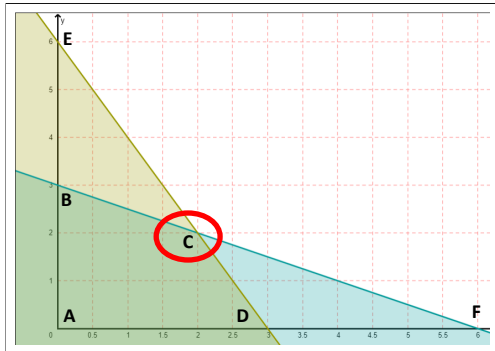
## Example: BFS $(x, y)$ - C -

**Maximize**  $z = 4 - \frac{1}{3}S_1 - \frac{1}{3}S_2$

**subject to**  $x = 2 + \frac{1}{3}S_1 - \frac{2}{3}S_2$

$y = 2 - \frac{2}{3}S_1 + \frac{1}{3}S_2$

$x, y, S_1, S_2 \geq 0$



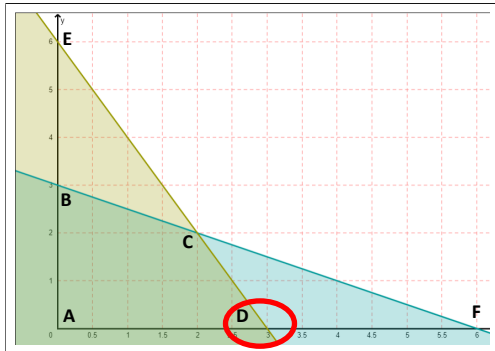
## Example: BFS $(x, s_1)$ - D -

**Maximize**  $z = 3 + \frac{1}{2}y - \frac{1}{2}s_2$

**subject to**  $x = 3 - \frac{1}{2}y - \frac{1}{2}s_2$

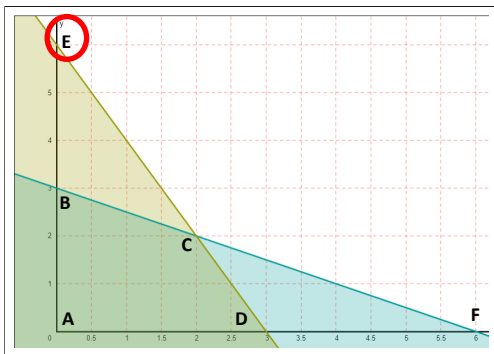
$s_1 = 3 - \frac{3}{2}y + \frac{1}{2}s_2$

$x, y, s_1, s_2 \geq 0$



## Example: Infeasible solution $(y, s_1)$ - E -

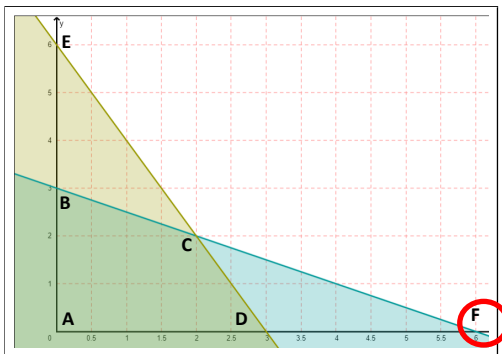
**Maximize**  $z = 6 - x - s_2$   
**subject to**  $y = 6 - 2x - s_2$   
 $s_1 = -6 + 3x + 2s_2$   
 $x, y, s_1, s_2 \geq 0$





## Example: Infeasible solution $(x, s_2)$ - F -

**Maximize**  $z = 6 - y - s_1$   
**subject to**  $x = 6 - 2y - s_1$   
 $s_2 = -6 - 3y + 2s_1$   
 $x, y, s_1, s_2 \geq 0$



## Example: Summary

| Nonbasic (zero) variables | Basic variables | Basic solution | Corner | Feasible ? | Objective          |
|---------------------------|-----------------|----------------|--------|------------|--------------------|
| $x, y$                    | $S_1, S_2$      | (0,0)          | A      | Yes        | 0                  |
| $x, S_1$                  | $y, S_2$        | (0,3)          | B      | Yes        | 3                  |
| $S_1, S_2$                | $x, y$          | (2,2)          | C      | Yes        | <b>4 (Optimum)</b> |
| $y, S_2$                  | $x, S_1$        | (3,0)          | D      | Yes        | 3                  |
| $x, S_2$                  | $y, S_1$        | (0,6)          | E      | No         | $\emptyset$        |
| $y, S_1$                  | $x, S_2$        | (6,0)          | F      | No         | $\emptyset$        |

## State 1: Basic Feasible Solution

$$\begin{array}{ll}\text{Maximize} & z = 3 + \frac{1}{2}x - \frac{1}{2}s_1 \\ \text{subject to} & y = 3 - \frac{1}{2}x - \frac{1}{2}s_1 \\ & s_2 = 3 - \frac{3}{2}x + \frac{1}{2}s_1 \\ & x, y, s_1, s_2 \geq 0\end{array}$$

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## State 2: Infeasible Solution

$$\begin{array}{ll}\text{Maximize} & z = 6 - x - s_2 \\ \text{subject to} & y = 6 - 2x - s_2 \\ & s_1 = -6 + 3x + 2s_2 \\ & x, y, s_1, s_2 \geq 0\end{array}$$

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## State 3: Optimal Basic Feasible Solution

$$\begin{array}{ll}\text{Maximize} & z = 4 - \frac{1}{3}S_1 - \frac{1}{3}S_2 \\ \text{subject to} & x = 2 + \frac{1}{3}S_1 - \frac{2}{3}S_2 \\ & y = 2 - \frac{2}{3}S_1 + \frac{1}{3}S_2 \\ & x, y, S_1, S_2 \geq 0\end{array}$$

## State 3: Optimal Basic Feasible Solution

$$\begin{array}{ll}\text{Maximize} & z = 4 - \frac{1}{3}S_1 - \frac{1}{3}S_2 \\ \text{subject to} & x = 2 + \frac{1}{3}S_1 - \frac{2}{3}S_2 \\ & y = 2 - \frac{2}{3}S_1 + \frac{1}{3}S_2 \\ & x, y, S_1, S_2 \geq 0\end{array}$$



# Simplex Algorithm

$$\text{Maximize } z = b_0 + \sum_{i=m+1}^n c_i x_i$$

$$\text{subject to } x_1 = b_1 + \sum_{i=m+1}^n a_{1i} x_i$$

$$\vdots$$

$$x_m = b_m + \sum_{i=m+1}^n a_{mi} x_i$$

- Convert to the standard form
- While (The BFS is not optimal)
  - Move to a new Basic Feasible Solution
- Output the optimal solution.

# Simplex Algorithm

$$\text{Maximize } z = b_0 + \sum_{i=m+1}^n c_i x_i$$

$$\text{subject to } x_1 = b_1 + \sum_{i=m+1}^n a_{1i} x_i$$

$$\vdots$$

$$x_m = b_m + \sum_{i=m+1}^n a_{mi} x_i$$

- While ( $\exists c_i > 0$ )
  - Select a non-basic variable with a positive coefficient: **Entering variable**
  - Introduce this variable in the basis by removing a basic variable: **Leaving variable**
  - Perform Gaussian elimination
- Output the optimal solution.

# Simplex Algorithm

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- 1: Initialize a feasible basis and corresponding basic feasible solution.
  - 2: **while** the objective function has positive coefficients **do**
  - 3:     Choose a non-basic variable with a positive coefficient as the entering variable.
  - 4:     Choose a basic variable to leave the basis using the minimum ratio test.
  - 5:     Update the basis by replacing the leaving variable with the entering variable.
  - 6:     Recalculate the basic feasible solution.
  - 7: **end while**
  - 8: Output the optimal solution.
-

# The minimum ratio test

$$\begin{array}{ll}\text{Maximize} & z = x + y \\ \text{subject to} & S_1 = 6 - x - 2y \\ & S_2 = 6 - 2x - y \\ & x, y, S_1, S_2 \geq 0\end{array}$$

How to choose the leaving variable?

- We must maintain feasibility.

# The minimum ratio test

$$\begin{array}{ll}\text{Maximize} & z = x + y \\ \text{subject to} & S_1 = 6 - x - 2y \\ & S_2 = 6 - 2x - y \\ & x, y, S_1, S_2 \geq 0\end{array}$$

Chose the leaving variable based on the ration  $\frac{b_i}{-a_i}$  ( $a_i \leq 0$ )

- $x$  is the Entering variable ( $y$  also can be selected).
- $S_2$  is the Leaving variable.
  - $\text{Ratio}(S_1) = \frac{6}{1} = 6$ .
  - $\text{Ratio}(S_2) = \frac{6}{2} = 3$ .
- We only consider constraints with Negative coefficients ( $a_i \leq 0$ ) for the entering variable, so the ratio is necessarily positive.

# The minimum ratio test

$$\begin{array}{ll}\text{Maximize} & z = x + y \\ \text{subject to} & S_1 = 6 - x - 2y \\ & S_2 = 6 - 2x - y \\ & x, y, S_1, S_2 \geq 0\end{array}$$

Old BFS is

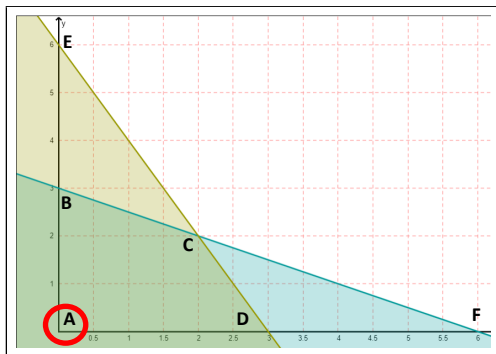
- Basic Variables =  $\{S_1, S_2\}$
- Non Basic variables =  $\{x, y\}$

New BFS is

- Basic Variables =  $\{x, S_1\}$
- Non Basic variables =  $\{y, S_2\}$

## Example: BFS $(s_1, s_2)$ - A -

**Maximize**  $z = x + y$   
**subject to**  $s_1 = 6 - x - 2y$   
 $s_2 = 6 - 2x - y$   
 $x, y, s_1, s_2 \geq 0$



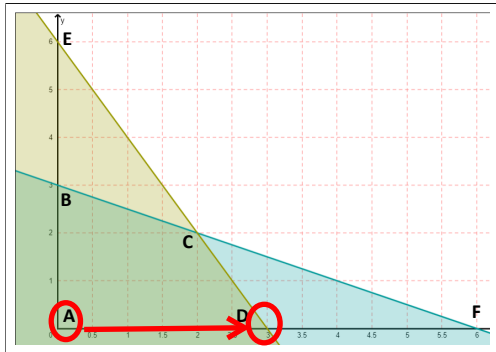
## Example: BFS $(x, s_1)$ - A $\rightarrow$ D -

**Maximize**  $z = 3 + \frac{1}{2}y - \frac{1}{2}s_2$

**subject to**  $x = 3 - \frac{1}{2}y - \frac{1}{2}s_2$

$s_1 = 3 - \frac{3}{2}y + \frac{1}{2}s_2$

$x, y, s_1, s_2 \geq 0$





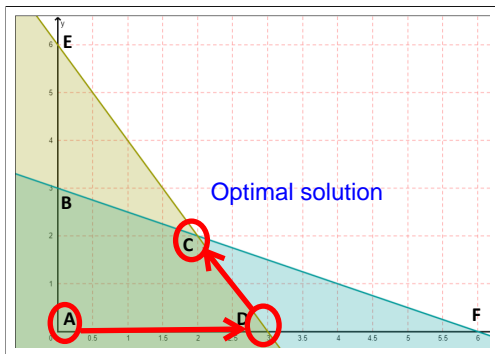
## Example: BFS $(x, y)$ - $A \rightarrow D \rightarrow C$ -

**Maximize**  $z = 4 - \frac{1}{3}S_1 - \frac{1}{3}S_2$

**subject to**  $x = 2 + \frac{1}{3}S_1 - \frac{2}{3}S_2$

$y = 2 - \frac{2}{3}S_1 + \frac{1}{3}S_2$

$x, y, S_1, S_2 \geq 0$



# Simplex Tableau Method

**Maximize**  $z = -(-x - y)$   
**subject to**  $S_1 + x + 2y = 6$   
 $S_2 + 2x + y = 6$   
 $x, y, S_1, S_2 \geq 0$

| Base  | $z$ | $x$ | $y$ | $S_1$    | $S_2$    | $b$ |
|-------|-----|-----|-----|----------|----------|-----|
|       | 1   | -1  | -1  | 0        | 0        | 0   |
| $S_1$ | 0   | 1   | 2   | <b>1</b> | 0        | 6   |
| $S_2$ | 0   | 2   | 1   | 0        | <b>1</b> | 6   |

- The simplex should be updated to consider the sign modifications (optimality test, how to choose leaving/entering variables, minimum ratio test).

# Simplex Tableau Method

| Base | z | x  | y  | S1 | S2 | b | Ratio             |
|------|---|----|----|----|----|---|-------------------|
|      | 1 | -1 | -1 | 0  | 0  | 0 |                   |
| S1   | 0 | 1  | 2  | 1  | 0  | 6 | $\frac{6}{1} = 6$ |
| S2   | 0 | 2  | 1  | 0  | 1  | 6 | $\frac{6}{3} = 2$ |

- **Entering variable:**  $x$

- **Leaving variable:**  $S_2$



Pivot

# Simplex Tableau Method

| Base  | z | x        | y              | S1       | S2             | b | Ratio                      |
|-------|---|----------|----------------|----------|----------------|---|----------------------------|
|       | 1 | 0        | $-\frac{1}{2}$ | 0        | $\frac{1}{2}$  | 3 |                            |
| $S_1$ | 0 | 0        | $\frac{3}{2}$  | <b>1</b> | $-\frac{1}{2}$ | 3 | $3 \times \frac{2}{3} = 2$ |
| x     | 0 | <b>1</b> | $\frac{1}{2}$  | 0        | $\frac{1}{2}$  | 3 | $3 \times \frac{2}{1} = 6$ |

- **Entering variable:** y
- **Leaving variable:**  $S_1$



Pivot

# Simplex Tableau Method

| Base | z | x | y | S1             | S2             | b | Ratio |
|------|---|---|---|----------------|----------------|---|-------|
|      | 1 | 0 | 0 | $\frac{1}{3}$  | $\frac{1}{3}$  | 4 |       |
| y    | 0 | 0 | 1 | $\frac{2}{3}$  | $-\frac{1}{3}$ | 2 |       |
| x    | 0 | 1 | 0 | $-\frac{1}{3}$ | $\frac{2}{3}$  | 2 |       |

Optimality test  
satisfied

- All the objective coefficients are positive.
- **Optimal Basic Feasible Solution detected**

# Simplex: Special cases

- **Unboundedness:** occurs when the objective function can be increased indefinitely without violating any of the constraints.
- **Infeasibility:** occurs when there is no feasible solution that satisfies all of the constraints.
- **Degeneracy:** occurs when one or more basic variables become zero during the iteration process.

# Conclusion

- Simplex method is an efficient algorithm for finding optimal solutions to LP problems by navigating through the corner points of the feasible region.
- It iteratively moves from one Basic Feasible Solution (BFS) to a better neighborhood BFS until the optimal BFS is reached.
- By detecting the optimal BFS, the simplex method provides the optimal values of the decision variables and the objective function.