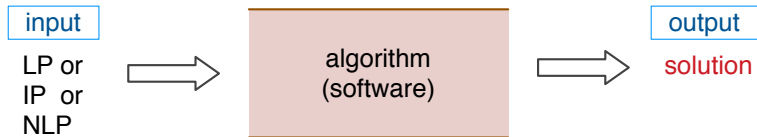


## Module 2: Linear programs (Possible outcomes)

# What does solving an optimization problem mean?



# What does solving an optimization problem mean?

input

LP or  
IP or  
NLP



algorithm  
(software)

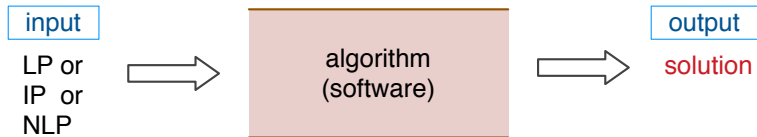


output

solution

$$\begin{array}{ll} \max & 2x_1 - 3x_2 \\ \text{s.t.} & \\ & x_1 + x_2 \leq 1 \\ & x_1, x_2 \geq 0 \end{array}$$

# What does solving an optimization problem mean?



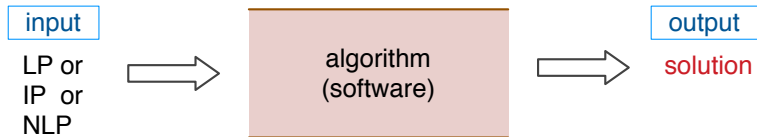
$$\begin{array}{ll} \max & 2x_1 - 3x_2 \\ \text{s.t.} & \\ & x_1 + x_2 \leq 1 \\ & x_1, x_2 \geq 0 \end{array}$$



$$\begin{array}{l} x_1 = 1, \\ x_2 = 0 \end{array}$$

Optimal  
Solution

# What does solving an optimization problem mean?



$$\begin{array}{ll} \max & 2x_1 - 3x_2 \\ \text{s.t.} & \\ & x_1 + x_2 \leq 1 \\ & x_1, x_2 \geq 0 \end{array}$$



$$\begin{array}{l} x_1 = 1, \\ x_2 = 0 \end{array}$$

Optimal  
Solution

## Remark

Sometimes the answer is not so straightforward!!!

## Definition

An assignment of values to each of the variables is a **feasible solution** if all the constraints are satisfied.

## Definition

An assignment of values to each of the variables is a **feasible solution** if all the constraints are satisfied.

## Definition

An **optimization problem** is **feasible** if it has at least one feasible solution. It is **infeasible** otherwise.

## Definition

An assignment of values to each of the variables is a **feasible solution** if all the constraints are satisfied.

## Definition

An **optimization problem** is **feasible** if it has at least one feasible solution. It is **infeasible** otherwise.

$$\max \quad x_1 - x_2$$

s.t.

$$4x_1 + x_2 \geq 2$$

$$2x_1 - 3x_2 \leq 4$$

$$x_1, x_2 \geq 0$$



## Definition

An assignment of values to each of the variables is a **feasible solution** if all the constraints are satisfied.

## Definition

An **optimization problem** is **feasible** if it has at least one feasible solution. It is **infeasible** otherwise.

$$\max \quad x_1 - x_2$$

s.t.

$$4x_1 + x_2 \geq 2$$

$$2x_1 - 3x_2 \leq 4$$

$$x_1, x_2 \geq 0$$

$$x_1 = 1$$

$$x_2 = 3$$

## Definition

An assignment of values to each of the variables is a **feasible solution** if all the constraints are satisfied.

## Definition

An **optimization problem** is **feasible** if it has at least one feasible solution. It is **infeasible** otherwise.

$$\max \quad x_1 - x_2$$

s.t.

$$4x_1 + x_2 \geq 2$$

$$2x_1 - 3x_2 \leq 4$$

$$x_1, x_2 \geq 0$$

$$x_1 = 1$$

$$x_2 = 3$$

Feasible solution



Problem is  
feasible

## Definition

An assignment of values to each of the variables is a **feasible solution** if all the constraints are satisfied.

## Definition

An **optimization problem** is **feasible** if it has at least one feasible solution. It is **infeasible** otherwise.

$$\max \quad x_1 - x_2$$

s.t.

$$4x_1 + x_2 \geq 2$$

$$2x_1 - 3x_2 \leq 4$$

$$x_1, x_2 \geq 0$$

$$x_1 = 3$$

$$x_2 = 0$$

**NOT**  
**feasible solution**

**But problem is**  
**feasible.**

## Definition

- For a **maximization** problem, an **optimal solution** is a feasible solution that **maximizes** the objective function.

## Definition

- For a **maximization** problem, an **optimal solution** is a feasible solution that **maximizes** the objective function.
- For a **minimization** problem, an **optimal solution** is a feasible solution that **minimizes** the objective function.

## Definition

- For a **maximization** problem, an **optimal solution** is a feasible solution that **maximizes** the objective function.
- For a **minimization** problem, an **optimal solution** is a feasible solution that **minimizes** the objective function.

$$\max \quad x_1$$

s.t.

$$x_1 \leq 1$$

$$x_2 \geq 1$$

$x_1 = 1, x_2 = \alpha$  optimal for all  $\alpha \geq 1$ .

## Definition

- For a **maximization** problem, an **optimal solution** is a feasible solution that **maximizes** the objective function.
- For a **minimization** problem, an **optimal solution** is a feasible solution that **minimizes** the objective function.

$$\max \quad x_1$$

s.t.

$$x_1 \leq 1$$

$$x_2 \geq 1$$

$x_1 = 1, x_2 = \alpha$  optimal for all  $\alpha \geq 1$ .

## Remark

An optimization problem can have several optimal solutions.

## Question

Does the following linear program have an optimal solution?

$$\max x_1$$

s.t.

$$x_1 \geq 2$$

$$x_1 \leq 1$$



## Question

Does the following linear program have an optimal solution?

$$\max x_1$$

s.t.

$$x_1 \geq 2$$

$$x_1 \leq 1$$

Infeasible problem, so  
no optimal solution

## Question

Does the following linear program have an optimal solution?

$$\begin{array}{ll} \max & x_1 \\ \text{s.t.} & \\ & x_1 \geq 2 \\ & x_1 \leq 1 \end{array}$$

Infeasible problem, so  
no optimal solution

## Question

Does every **feasible** optimization problem have an optimal solution?

## Question

Does the following linear program have an optimal solution?

$$\begin{array}{ll} \max & x_1 \\ \text{s.t.} & \\ & x_1 \geq 2 \\ & x_1 \leq 1 \end{array}$$

Infeasible problem, so  
no optimal solution

## Question

Does every **feasible** optimization problem have an optimal solution? **NO**

$$\begin{array}{ll} \max & x_1 \\ \text{s.t.} & \\ & x_1 \geq 1 \end{array}$$

## Question

Does the following linear program have an optimal solution?

$$\begin{array}{ll} \max & x_1 \\ \text{s.t.} & \\ & x_1 \geq 2 \\ & x_1 \leq 1 \end{array}$$

Infeasible problem, so  
no optimal solution

## Question

Does every **feasible** optimization problem have an optimal solution? **NO**

$$\begin{array}{ll} \max & x_1 \\ \text{s.t.} & \\ & x_1 \geq 1 \end{array}$$

Feasible ( $x_1 = 1$ ),  
but still no optimal solution!!!

## Definition

- A maximization problem is unbounded if for every value  $M$  there exists a feasible solution with objective value greater than  $M$ .

## Definition

- A **maximization** problem is **unbounded** if for every value  $M$  there exists a feasible solution with objective value **greater** than  $M$ .
- A **minimization** problem is **unbounded** if for every value  $M$  there exists a feasible solution with objective value **smaller** than  $M$ .

## Definition

- A **maximization** problem is **unbounded** if for every value  $M$  there exists a feasible solution with objective value **greater** than  $M$ .
- A **minimization** problem is **unbounded** if for every value  $M$  there exists a feasible solution with objective value **smaller** than  $M$ .



We have seen three possible outcomes for an optimization problem:

## Definition

- A **maximization** problem is **unbounded** if for every value  $M$  there exists a feasible solution with objective value **greater** than  $M$ .
- A **minimization** problem is **unbounded** if for every value  $M$  there exists a feasible solution with objective value **smaller** than  $M$ .



We have seen three possible outcomes for an optimization problem:

- It has an optimal solution



## Definition

- A **maximization** problem is **unbounded** if for every value  $M$  there exists a feasible solution with objective value **greater** than  $M$ .
- A **minimization** problem is **unbounded** if for every value  $M$  there exists a feasible solution with objective value **smaller** than  $M$ .



We have seen three possible outcomes for an optimization problem:

- It has an optimal solution
- It is infeasible

## Definition

- A **maximization** problem is **unbounded** if for every value  $M$  there exists a feasible solution with objective value **greater** than  $M$ .
- A **minimization** problem is **unbounded** if for every value  $M$  there exists a feasible solution with objective value **smaller** than  $M$ .



We have seen three possible outcomes for an optimization problem:

- It has an optimal solution
- It is infeasible
- It is unbounded

## Definition

- A **maximization** problem is **unbounded** if for every value  $M$  there exists a feasible solution with objective value **greater** than  $M$ .
- A **minimization** problem is **unbounded** if for every value  $M$  there exists a feasible solution with objective value **smaller** than  $M$ .



We have seen three possible outcomes for an optimization problem:

- It has an optimal solution
- It is infeasible
- It is unbounded

## Question

Can anything else happen?

## Definition

- A **maximization** problem is **unbounded** if for every value  $M$  there exists a feasible solution with objective value **greater** than  $M$ .
- A **minimization** problem is **unbounded** if for every value  $M$  there exists a feasible solution with objective value **smaller** than  $M$ .



We have seen three possible outcomes for an optimization problem:

- It has an optimal solution
- It is infeasible
- It is unbounded

## Question

Can anything else happen? **YES**

Consider,

$\max x$

s.t.

$x < 1$

Consider,

$$\max x$$

s.t.

$$x < 1$$

- Feasible: set  $x = 0$ .

Consider,

$$\begin{array}{ll} \max & x \\ \text{s.t.} & \\ & x < 1 \end{array}$$

- Feasible: set  $x = 0$ .
- Not unbounded: 1 is an upper bound.

Consider,

$$\begin{array}{ll} \max & x \\ \text{s.t.} & \\ & x < 1 \end{array}$$

- Feasible: set  $x = 0$ .
- Not unbounded: 1 is an upper bound.
- But no optimal solution!



Consider,

$$\begin{array}{ll} \max & x \\ \text{s.t.} & \\ & x < 1 \end{array}$$

- Feasible: set  $x = 0$ .
- Not unbounded: 1 is an upper bound.
- But no optimal solution!

## Proof

Suppose for a contradiction  $x$  is optimal solution.

Consider,

$$\begin{array}{ll} \max & x \\ \text{s.t.} & \\ & x < 1 \end{array}$$

- Feasible: set  $x = 0$ .
- Not unbounded: 1 is an upper bound.
- But no optimal solution!

## Proof

Suppose for a contradiction  $x$  is optimal solution. Let

$$x' := \frac{x + 1}{2}.$$

Consider,

$$\begin{array}{ll} \max & x \\ \text{s.t.} & \\ & x < 1 \end{array}$$

- Feasible: set  $x = 0$ .
- Not unbounded: 1 is an upper bound.
- But no optimal solution!

## Proof

Suppose for a contradiction  $x$  is optimal solution. Let

$$x' := \frac{x + 1}{2}.$$

Then  $x' < 1$  feasible.

Consider,

$$\begin{array}{ll} \max & x \\ \text{s.t.} & \\ & x < 1 \end{array}$$

- Feasible: set  $x = 0$ .
- Not unbounded: 1 is an upper bound.
- But no optimal solution!

## Proof

Suppose for a contradiction  $x$  is optimal solution. Let

$$x' := \frac{x + 1}{2}.$$

Then  $x' < 1$  feasible. Moreover,  $x' > x$ .

Consider,

$$\begin{array}{ll} \max & x \\ \text{s.t.} & \\ & x < 1 \end{array}$$

- Feasible: set  $x = 0$ .
- Not unbounded: 1 is an upper bound.
- But no optimal solution!

## Proof

Suppose for a contradiction  $x$  is optimal solution. Let

$$x' := \frac{x + 1}{2}.$$

Then  $x' < 1$  feasible. Moreover,  $x' > x$ .

Thus  $x$  not optimal, contradiction.

Consider,

$$\begin{array}{ll} \max & x \\ \text{s.t.} & \\ & x < 1 \end{array}$$

- Feasible: set  $x = 0$ .
- Not unbounded: 1 is an upper bound.
- But no optimal solution!

## Proof

Suppose for a contradiction  $x$  is optimal solution. Let

$$x' := \frac{x + 1}{2}.$$

Then  $x' < 1$  feasible. Moreover,  $x' > x$ .

Thus  $x$  not optimal, contradiction.

## Question

Any other example without strict inequalities?

Consider,

$$\begin{array}{ll} \max & x \\ \text{s.t.} & \\ & x < 1 \end{array}$$

- Feasible: set  $x = 0$ .
- Not unbounded: 1 is an upper bound.
- But no optimal solution!

## Proof

Suppose for a contradiction  $x$  is optimal solution. Let

$$x' := \frac{x + 1}{2}.$$

Then  $x' < 1$  feasible. Moreover,  $x' > x$ .

Thus  $x$  not optimal, contradiction.

## Question

Any other example without strict inequalities? YES

Consider,

$$\begin{array}{ll} \min & \frac{1}{x} \\ \text{s.t.} & \\ & x \geq 1 \end{array}$$



Consider,

$$\begin{array}{ll} \min & \frac{1}{x} \\ \text{s.t.} & \\ & x \geq 1 \end{array}$$

- Feasible: set  $x = 1$ .

Consider,

$$\begin{array}{ll} \min & \frac{1}{x} \\ \text{s.t.} & \\ & x \geq 1 \end{array}$$

- Feasible: set  $x = 1$ .
- Not unbounded: 0 is a lower bound.

Consider,

$$\begin{array}{ll} \min & \frac{1}{x} \\ \text{s.t.} & \\ & x \geq 1 \end{array}$$

- Feasible: set  $x = 1$ .
- Not unbounded: 0 is a lower bound.
- But no optimal solution!

Consider,

$$\begin{array}{ll} \min & \frac{1}{x} \\ \text{s.t.} & \\ & x \geq 1 \end{array}$$

- Feasible: set  $x = 1$ .
- Not unbounded: 0 is a lower bound.
- But no optimal solution!

## Exercise

Check this optimization problem has no optimal solution.

max  $x$

s.t.

$x < 1$

Not a linear program

Strict inequality

$$\begin{array}{ll} \min & \frac{1}{x} \\ \text{s.t.} & \\ & x \geq 1 \end{array}$$

Not a linear program  
Objective function non-linear

$$\begin{array}{ll} \min & \frac{1}{x} \\ \text{s.t.} & \\ & x \geq 1 \end{array}$$

Not a linear program  
Objective function non-linear

## Remark

Linear programs are nicer than general optimization problems.

$$\begin{array}{ll} \min & \frac{1}{x} \\ \text{s.t.} & \\ & x \geq 1 \end{array}$$

Not a linear program  
Objective function non-linear

## Remark

Linear programs are nicer than general optimization problems.

## Fundamental theorem of linear programming

For any linear program one of the following holds:

- It has an optimal solution
- It is infeasible
- It is unbounded



$$\begin{array}{ll} \min & \frac{1}{x} \\ \text{s.t.} & \\ & x \geq 1 \end{array}$$

Not a linear program  
Objective function non-linear

## Remark

Linear programs are nicer than general optimization problems.

## Fundamental theorem of linear programming

For any linear program **exactly one** of the following holds:

- It has an optimal solution
- It is infeasible
- It is unbounded

$$\begin{array}{ll} \min & \frac{1}{x} \\ \text{s.t.} & \\ & x \geq 1 \end{array}$$

Not a linear program  
Objective function non-linear

## Remark

Linear programs are nicer than general optimization problems.

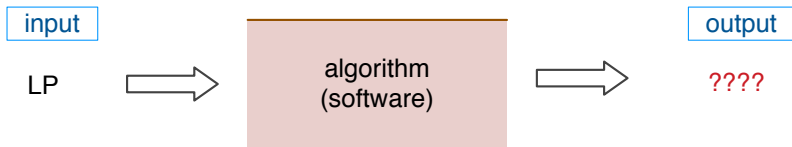
## Fundamental theorem of linear programming

For any linear program **exactly one** of the following holds:

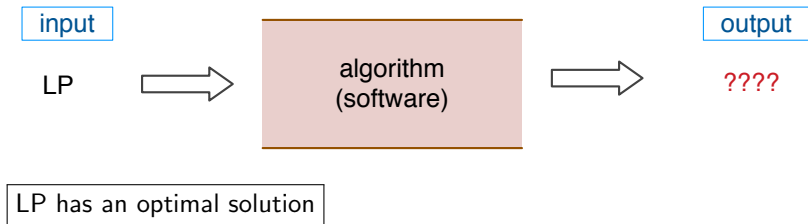
- It has an optimal solution
- It is infeasible
- It is unbounded

We will prove it later in the course.

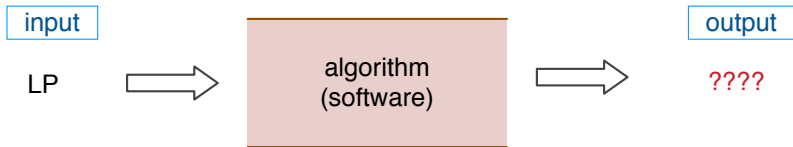
We can now describe what we mean by solving a linear program,



We can now describe what we mean by solving a linear program,



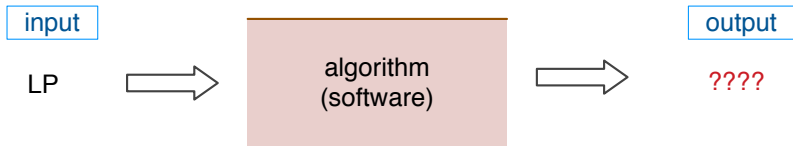
We can now describe what we mean by solving a linear program,



LP has an optimal solution

Return an optimal solution  $\bar{x}$

We can now describe what we mean by solving a linear program,

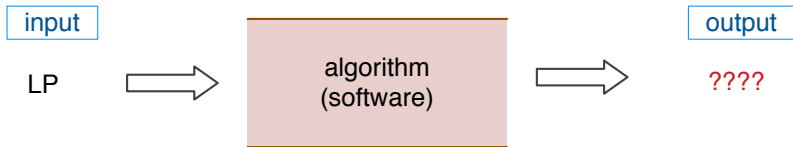


LP has an optimal solution

Return an optimal solution  $\bar{x}$

LP is infeasible.

We can now describe what we mean by solving a linear program,



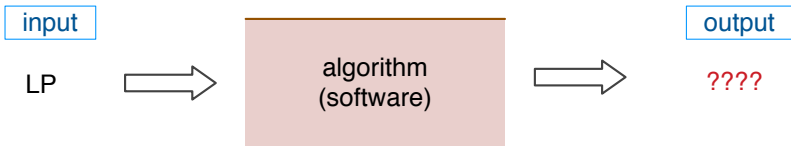
LP has an optimal solution

Return an optimal solution  $\bar{x}$

LP is infeasible.

Say the LP is infeasible

We can now describe what we mean by solving a linear program,



LP has an optimal solution

Return an optimal solution  $\bar{x}$

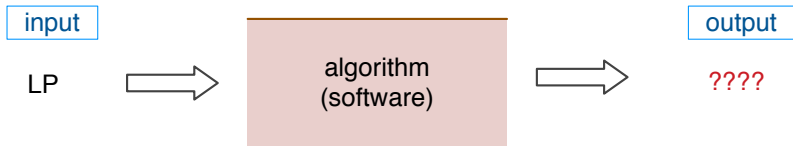
LP is infeasible.

Say the LP is infeasible

LP is unbounded.



We can now describe what we mean by solving a linear program,



LP has an optimal solution

Return an optimal solution  $\bar{x}$

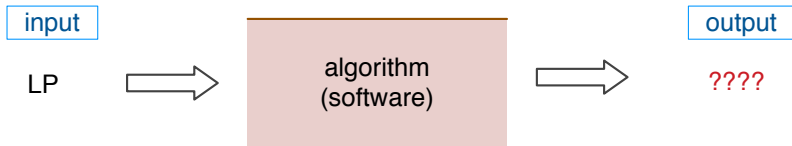
LP is infeasible.

Say the LP is infeasible

LP is unbounded.

Say the LP is unbounded

We can now describe what we mean by solving a linear program,



LP has an optimal solution

Return an optimal solution  $\bar{x}$

LP is infeasible.

Say the LP is infeasible

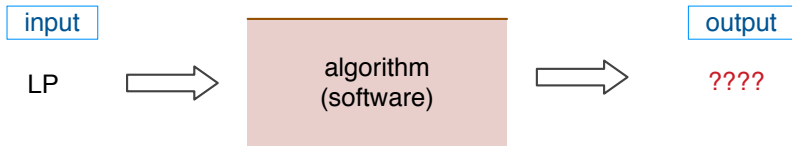
LP is unbounded.

Say the LP is unbounded

## Remark

Algorithms should justify their answers !!!

We can now describe what we mean by solving a linear program,



LP has an optimal solution

Return an optimal solution  $\bar{x}$  + **proof** that  $\bar{x}$  is optimal.

LP is infeasible.

Return a **proof** the LP is infeasible.

LP is unbounded.

Return a **proof** the LP is unbounded.

## Remark

Algorithms always need to justify their answers !!!

## Recap

## Recap

1. Optimization problems can be:

## Recap

1. Optimization problems can be:
  - (A) infeasible,

## Recap

1. Optimization problems can be:
  - (A) infeasible,
  - (B) unbounded, or

## Recap

1. Optimization problems can be:
  - (A) infeasible,
  - (B) unbounded, or
  - (C) have an optimal solution.



## Recap

1. Optimization problems can be:
  - (A) infeasible,
  - (B) unbounded, or
  - (C) have an optimal solution.
2. There are optimization where none of (A), (B), (C) hold,

## Recap

1. Optimization problems can be:
  - (A) infeasible,
  - (B) unbounded, or
  - (C) have an optimal solution.
2. There are optimization where none of (A), (B), (C) hold,
3. For LPs exactly one of (A), (B), (C) holds,

## Recap

1. Optimization problems can be:
  - (A) infeasible,
  - (B) unbounded, or
  - (C) have an optimal solution.
2. There are optimization where none of (A), (B), (C) hold,
3. For LPs exactly one of (A), (B), (C) holds,
4. By solving an LP we mean

## Recap

1. Optimization problems can be:
  - (A) infeasible,
  - (B) unbounded, or
  - (C) have an optimal solution.
2. There are optimization where none of (A), (B), (C) hold,
3. For LPs exactly one of (A), (B), (C) holds,
4. By solving an LP we mean
  - indicating which of (A), (B), (C) holds,

## Recap

1. Optimization problems can be:
  - (A) infeasible,
  - (B) unbounded, or
  - (C) have an optimal solution.
2. There are optimization where none of (A), (B), (C) hold,
3. For LPs exactly one of (A), (B), (C) holds,
4. By solving an LP we mean
  - indicating which of (A), (B), (C) holds,
  - if (C) holds give an optimal solution,

## Recap

1. Optimization problems can be:
  - (A) infeasible,
  - (B) unbounded, or
  - (C) have an optimal solution.
2. There are optimization where none of (A), (B), (C) hold,
3. For LPs exactly one of (A), (B), (C) holds,
4. By solving an LP we mean
  - indicating which of (A), (B), (C) holds,
  - if (C) holds give an optimal solution,
  - **give a proof the answer is correct.**