Module 2: Linear programs (Possible outcomes)





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



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 $x_1 = 1,$ Optimal $x_2 = 0$ Solution



Remark

Sometimes the answer is not so straightforward!!!

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

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Definition

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Feasible solution

$$x_1 = 1$$

 $x_2 = 3$

Problem is feasible

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Definition

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NOT feasible solution But problem is feasible.

 $x_1 = 3$

 $x_2 = 0$

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max	x_1
s.t.	
	$x_1 \leq 1$
	$x_2 \ge 1$

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

- 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.

Remark

An optimization problem can have several optimal solutions.

Does the following linear program have an optimal solution?

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

Does the following linear program have an optimal solution?



Infeasible problem, so no optimal solution

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Does every feasible optimization problem have an optimal solution?

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Does every feasible optimization problem have an optimal solution? NO

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

Does the following linear program have an optimal solution?



Infeasible problem, so no optimal solution

Question

Does every feasible optimization problem have an optimal solution? NO

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

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

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

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We have seen three possible outcomes for an optimization problem:

- It has an optimal solution
- It is infeasible

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Can anything else happen?

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We have seen three possible outcomes for an optimization problem:

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Question

Can anything else happen? YES

 $\begin{array}{ccc} \max & x\\ \mathsf{s.t.} & \\ & x < 1 \end{array}$



• Feasible: set x = 0.



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

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Proof

Suppose for a contradiction x is optimal solution.

 $\begin{array}{c|c} \max & x\\ \mathsf{s.t.}\\ & x < 1 \end{array}$

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Proof

Suppose for a contradiction \boldsymbol{x} is optimal solution. Let

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

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

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Proof

Suppose for a contradiction \boldsymbol{x} is optimal solution. Let

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

Then x' < 1 feasible.

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

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Proof

Suppose for a contradiction x is optimal solution. Let

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

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

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Any other example without strict inequalities?

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

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- Not unbounded: 1 is an upper bound.
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Any other example without strict inequalities? YES

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\min & \frac{1}{x} \\
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Exercise

Check this optimization problem has no optimal solution.



Not a linear program Strict inequality



Not a linear program Objective function non-linear



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Remark

Linear programs are nicer than general optimization problems.

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

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

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

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
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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
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We will prove it later in the course.







LP has an optimal solution

Return an optimal solution $\bar{\boldsymbol{x}}$



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Say the LP is infeasible



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LP is unbounded.



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LP is unbounded.

Say the LP is unbounded

Remark

Algorithms should justify their answers !!!



LP has an optimal solution

Return an optimal solution $\bar{x} + \text{proof}$ 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 !!!

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 - give a proof the answer is correct.