Randomised Algorithms

Lecture 6-7: Linear Programming

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Lent 2022



Outline

Introduction

A Simple Example of a Linear Program

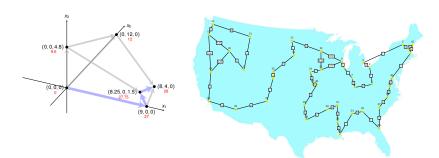
Formulating Problems as Linear Programs

Standard and Slack Forms

Simplex Algorithm

Finding an Initial Solution

Introduction



- linear programming is a powerful tool in optimisation
- inspired more sophisticated techniques such as quadratic optimisation, convex optimisation, integer programming and semi-definite programming
- we will later use the connection between linear and integer programming to tackle several problems (Vertex-Cover, Set-Cover, TSP, satisfiability)

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What are Linear Programs?

Linear Programming (informal definition) —

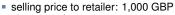
- maximise or minimise an objective, given limited resources (competing constraint)
- constraints are specified as (in)equalities
- objective function and constraints are linear

Laptop

- Laptop
 - selling price to retailer: 1,000 GBP

- Laptop
 - selling price to retailer: 1,000 GBP
 - glass: 4 units

Laptop



glass: 4 unitscopper: 2 units



Laptop



glass: 4 units

copper: 2 units

rare-earth elements: 1 unit





Laptop



glass: 4 units

copper: 2 units

rare-earth elements: 1 unit

Smartphone

Laptop

selling price to retailer: 1,000 GBP

glass: 4 units

copper: 2 units

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rare-earth elements: 1 unit









Smartphone

selling price to retailer: 1,000 GBP

glass: 1 unit

copper: 1 unit

Laptop

selling price to retailer: 1,000 GBP

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selling price to retailer: 1,000 GBP

glass: 1 unit copper: 1 unit

rare-earth elements: 2 units

Laptop

selling price to retailer: 1,000 GBP

glass: 4 units

copper: 2 units

rare-earth elements: 1 unit









selling price to retailer: 1,000 GBP

glass: 1 unit copper: 1 unit

rare-earth elements: 2 units

You have a daily supply of:

Laptop

selling price to retailer: 1,000 GBP

glass: 4 units

copper: 2 units

rare-earth elements: 1 unit









Smartphone

selling price to retailer: 1,000 GBP

glass: 1 unit copper: 1 unit

rare-earth elements: 2 units



You have a daily supply of:

glass: 20 units

Laptop

- selling price to retailer: 1,000 GBP
- glass: 4 units
- copper: 2 units
- rare-earth elements: 1 unit



Smartphone

- selling price to retailer: 1,000 GBP
- glass: 1 unit
- copper: 1 unit
- rare-earth elements: 2 units



You have a daily supply of:

- glass: 20 units
- copper: 10 units

Laptop

- selling price to retailer: 1,000 GBP
- glass: 4 units
- copper: 2 units
- rare-earth elements: 1 unit



Smartphone

- selling price to retailer: 1,000 GBP
- glass: 1 unit
- copper: 1 unitrare-earth elements: 2 units



- glass: 20 unitscopper: 10 units
- rare-earth elements: 14 units









Laptop

- selling price to retailer: 1,000 GBP
- glass: 4 units
- copper: 2 units
- rare-earth elements: 1 unit



Smartphone

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- glass: 20 units
- copper: 10 units
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- (and enough of everything else...)







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- glass: 1 unit copper: 1 unit
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- glass: 20 units
- copper: 10 units
- rare-earth elements: 14 units
- (and enough of everything else...)



How to maximise your daily earnings?

Linear Program for the Production Problem ——

Linear Program for the Production Problem -

The solution of this linear program yields the optimal production schedule.

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Formal Definition of Linear Program -

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• Given a_1, a_2, \ldots, a_n and a set of variables x_1, x_2, \ldots, x_n , a linear function f is defined by

$$f(x_1, x_2, \ldots, x_n) = a_1x_1 + a_2x_2 + \cdots + a_nx_n.$$

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- Linear Equality: $f(x_1, x_2, \dots, x_n) = b$
- Linear Inequality: $f(x_1, x_2, ..., x_n) \ge b$

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Linear Program for the Production Problem -

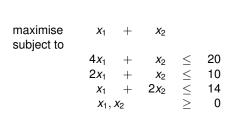
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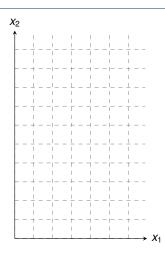
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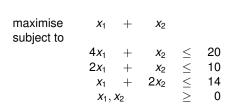
• Given a_1, a_2, \ldots, a_n and a set of variables x_1, x_2, \ldots, x_n , a linear function f is defined by

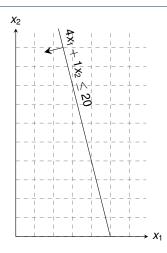
$$f(x_1, x_2, \ldots, x_n) = a_1 x_1 + a_2 x_2 + \cdots + a_n x_n.$$

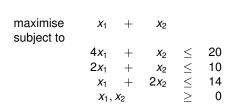
- Linear Equality: $f(x_1, x_2, ..., x_n) = b$ Linear Inequality: $f(x_1, x_2, ..., x_n) \ge b$ Linear Constraints
- Linear-Progamming Problem: either minimise or maximise a linear function subject to a set of linear constraints

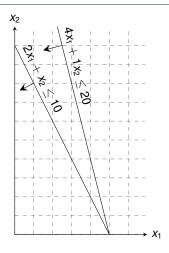


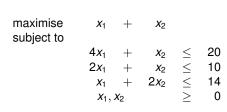


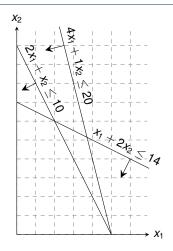


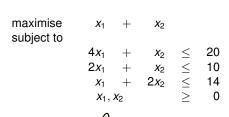


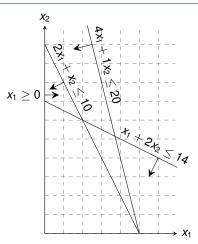


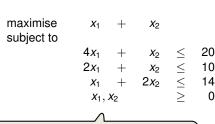


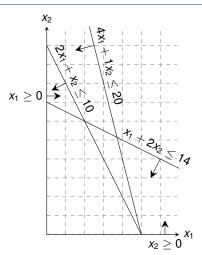


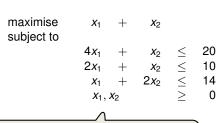




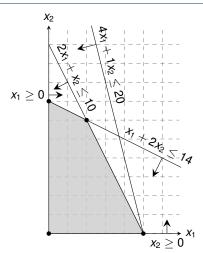


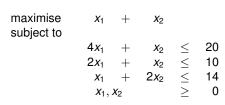


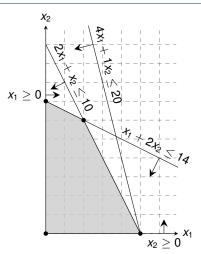


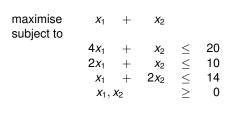


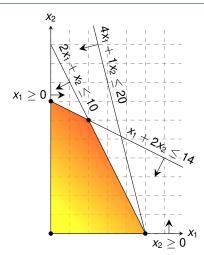
Any setting of x_1 and x_2 satisfying all constraints is a feasible solution

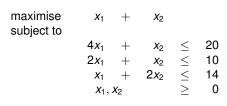


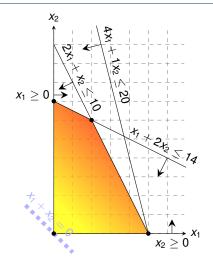


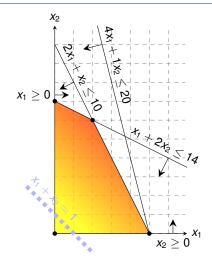


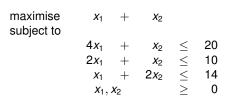


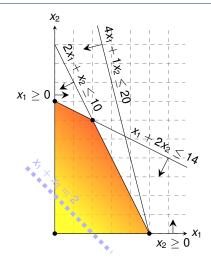


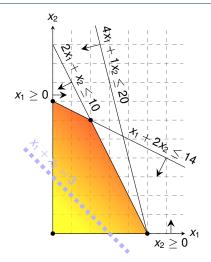


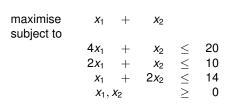


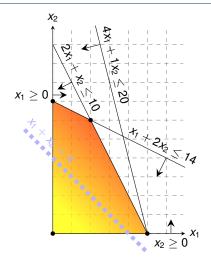


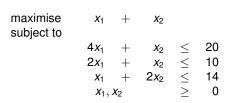


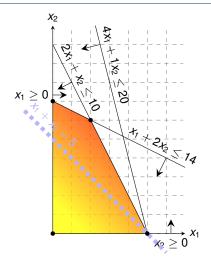


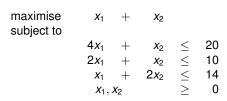


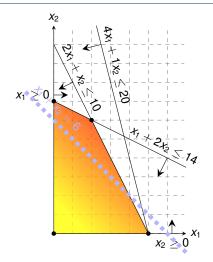


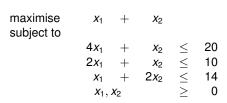


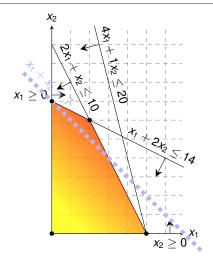


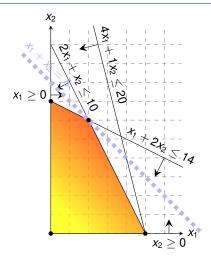


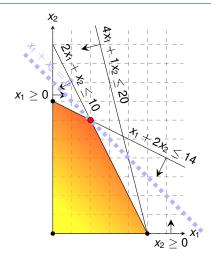


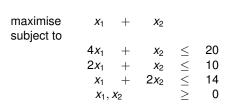












 $x_2 \geq 0$

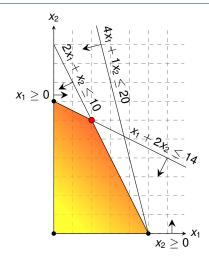
 χ_2

Graphical Procedure: Move the line $x_1 + x_2 = z$ as far up as possible.



Exercise: Which aspect did we ignore in the formulation of the linear program?

Graphical Procedure: Move the line $x_1 + x_2 = z$ as far up as possible.



While the same approach also works for higher-dimensions, we need to take a more systematic and algebraic procedure.

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Formulating Problems as Linear Programs

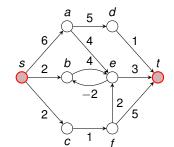
Standard and Slack Forms

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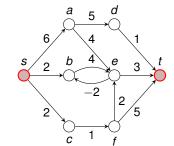
Single-Pair Shortest Path Problem

■ Given: directed graph G = (V, E) with edge weights $w : E \to \mathbb{R}$, pair of vertices $s, t \in V$



Single-Pair Shortest Path Problem

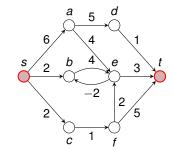
- Given: directed graph G = (V, E) with edge weights $w : E \to \mathbb{R}$, pair of vertices $s, t \in V$
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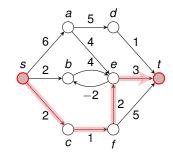
$$p = (v_0 = s, v_1, \dots, v_k = t)$$
 such that $w(p) = \sum_{i=1}^k w(v_{k-1}, v_k)$ is minimised.



Single-Pair Shortest Path Problem

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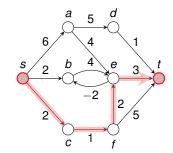
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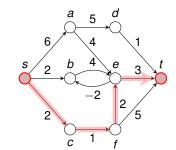
- Shortest Paths as LP -

subject to

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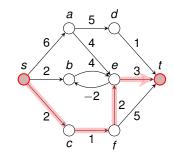
subject to

$$d_v \le d_u + w(u,v)$$
 for each edge $(u,v) \in E$, $d_s = 0$.

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Shortest Paths as I P =

$$d_t$$

$$egin{array}{lcl} \emph{d}_v & \leq & \emph{d}_u & + & \emph{w}(\emph{u},\emph{v}) & \mbox{for each edge } (\emph{u},\emph{v}) \in \emph{E}, \ \emph{d}_s & = & 0. \end{array}$$

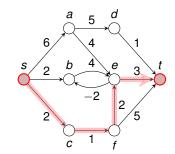
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,

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Shortest Paths as I P =

maximise subject to d_t

this is a maxim-

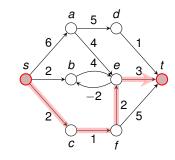
 $\leq d_u + w(u,v)$ for each edge $(u,v) \in E$, = 0.

isation problem!

Single-Pair Shortest Path Problem

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Shortest Paths as LP —

maximise d_t subject to

 d_v

 $d_s \leq d_u$ $d_s = 0.$

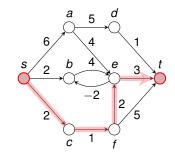
Recall: When Bellman-Ford terminates, all these inequalities are satisfied.

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Single-Pair Shortest Path Problem

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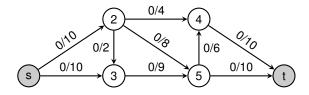
Shortest Paths as LP Recall: When Bellman-Ford terminates, all these inequalities are satisfied. Subject to $d_v \leq d_u + w(u,v) \text{ for each edge } (u,v) \in E,$ this is a maximisation problem! Solution \overline{d} satisfies $\overline{d}_v = \min_{u \in (u,v) \in E} \{\overline{d}_u + w(u,v)\}$

Maximum Flow Problem -

• Given: directed graph G=(V,E) with edge capacities $c:E\to\mathbb{R}^+$ (recall c(u,v)=0 if $(u,v)\not\in E$), pair of vertices $s,t\in V$

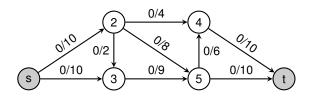
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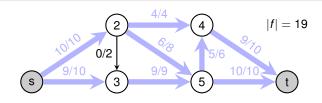
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- Goal: Find a maximum flow $f: V \times V \to \mathbb{R}$ from s to t which satisfies the capacity constraints and flow conservation



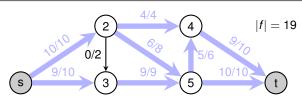
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Maximum Flow as LP

Extension of the Maximum Flow Problem

Minimum-Cost-Flow Problem

Extension of the Maximum Flow Problem

Minimum-Cost-Flow Problem -

• Given: directed graph G = (V, E) with capacities $c : E \to \mathbb{R}^+$, pair of vertices $s, t \in V$, cost function $a : E \to \mathbb{R}^+$, flow demand of d units

Extension of the Maximum Flow Problem

Minimum-Cost-Flow Problem

- Given: directed graph G = (V, E) with capacities $c : E \to \mathbb{R}^+$, pair of vertices $s, t \in V$, cost function $a : E \to \mathbb{R}^+$, flow demand of d units
- Goal: Find a flow $f: V \times V \to \mathbb{R}$ from s to t with |f| = d while minimising the total cost $\sum_{(u,v)\in E} a(u,v)f_{uv}$ incurred by the flow.

Extension of the Maximum Flow Problem

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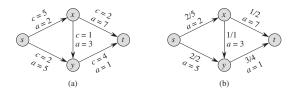


Figure 29.3 (a) An example of a minimum-cost-flow problem. We denote the capacities by c and the costs by a. Vertex s is the source and vertex t is the sink, and we wish to send 4 units of flow from s to t. (b) A solution to the minimum-cost flow problem in which 4 units of flow are sent from s to t. For each edge, the flow and capacity are written as flow/capacity.

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Optimal Solution with total cost:
$$\sum_{(u,v)\in E} a(u,v)f_{uv} = (2\cdot2) + (5\cdot2) + (3\cdot1) + (7\cdot1) + (1\cdot3) = 27$$

(b)

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(a)

Minimum Cost Flow as a LP

Minimum Cost Flow as LP ----

minimise
$$\sum_{(u,v)\in \mathcal{E}} a(u,v) f_{uv}$$
 subject to
$$f_{uv} \leq c(u,v) \quad \text{for } u,v\in V,$$

$$\sum_{v\in V} f_{vu} - \sum_{v\in V} f_{uv} = 0 \quad \text{for } u\in V\setminus \{s,t\},$$

$$\sum_{v\in V} f_{sv} - \sum_{v\in V} f_{vs} = d,$$

$$f_{uv} \geq 0 \quad \text{for } u,v\in V.$$

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Real power of Linear Programming comes from the ability to solve **new problems**!

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Standard and Slack Forms

Simplex Algorithm

Finding an Initial Solution

Standard Form -

maximise
$$\sum_{j=1}^{n} c_{j} x_{j}$$

subject to

$$\sum_{j=1}^{n} a_{ij} x_{j} \le b_{i} \quad \text{for } i = 1, 2, \dots, m$$
$$x_{j} \ge 0 \quad \text{for } j = 1, 2, \dots, n$$

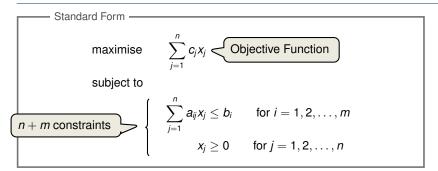
$$x_j \ge 0$$
 for $j = 1, 2, ..., r$

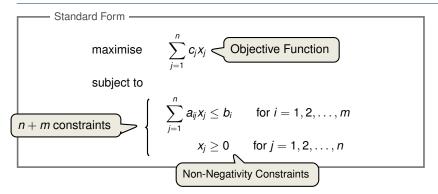
Standard Form -

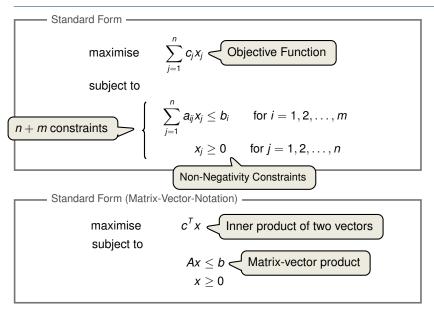
maximise
$$\sum_{j=1}^{n} c_j x_j$$
 Objective Function

subject to

$$\sum_{j=1}^{n} a_{ij} x_{j} \le b_{i} \quad \text{for } i = 1, 2, \dots, m$$
$$x_{j} \ge 0 \quad \text{for } j = 1, 2, \dots, n$$







Converting Linear Programs into Standard Form

Reasons for a LP not being in standard form:

- 1. The objective might be a minimisation rather than maximisation.
- 2. There might be variables without nonnegativity constraints.
- 3. There might be equality constraints.
- 4. There might be inequality constraints (with \geq instead of \leq).

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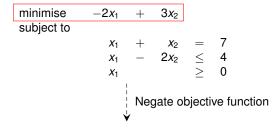
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Equivalence: a correspondence (not necessarily a bijection) between solutions.

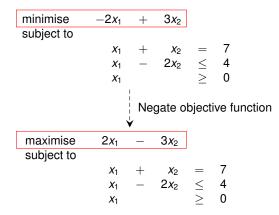
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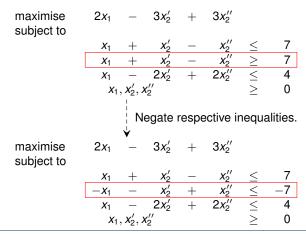
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s measures the slack between the two sides of the inequality.

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$$s > 0$$

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$$s > 0$$

• Denote slack variable of the *i*-th inequality by x_{n+i}

subject to

 $X_1, X_2, X_3, X_4, X_5, X_6$

maximise
$$2x_1 - 3x_2 + 3x_3$$
 subject to
$$x_4 = 7 - x_1 - x_2 + x_3 \\ x_5 = -7 + x_1 + x_2 - x_3 \\ x_6 = 4 - x_1 + 2x_2 - 2x_3 \\ x_1, x_2, x_3, x_4, x_5, x_6 \geq 0$$

maximise subject to

$$2x_1$$
 - $3x_2$ + $3x_3$
 x_4 = 7 - x_1 - x_2 + x_3
 x_5 = -7 + x_1 + x_2 - x_3
 x_6 = 4 - x_1 + $2x_2$ - $2x_3$
 $x_1, x_2, x_3, x_4, x_5, x_6$ \geq 0
Use variable z to denote objective fun

Use variable z to denote objective function and omit the nonnegativity constraints.

Z	=			$2x_{1}$	_	$3x_{2}$	+	3 <i>x</i> ₃
<i>X</i> ₄	=	7	_	<i>X</i> ₁	_	<i>X</i> ₂	+	<i>X</i> 3
<i>X</i> ₅	=	-7	+	<i>X</i> ₁	+	<i>X</i> ₂	_	<i>X</i> ₃
<i>X</i> ₆	=	4	_	<i>X</i> ₁	+	$2x_{2}$	_	$2x_3$

This is called slack form.

$$z = 2x_1 - 3x_2 + 3x_3$$

 $x_4 = 7 - x_1 - x_2 + x_3$
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Slack Form (Formal Definition) ————

Slack form is given by a tuple (N, B, A, b, c, v) so that

$$z = v + \sum_{j \in N} c_j x_j$$

 $x_i = b_i - \sum_{i \in N} a_{ij} x_j$ for $i \in B$,

and all variables are non-negative.

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Variables/Coefficients on the right hand side are indexed by B and N.

$$z = 28 - \frac{x_3}{6} - \frac{x_5}{6} - \frac{2x_6}{3}$$

$$x_1 = 8 + \frac{x_3}{6} + \frac{x_5}{6} - \frac{x_6}{3}$$

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- Conversion ("pivoting") is achieved by switching the roles of one basic and one non-basic variable

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- Each iteration converts one slack form into an equivalent one while the objective value will not decrease In that sense, it is a greedy algorithm.
- Conversion ("pivoting") is achieved by switching the roles of one basic and one non-basic variable

Extended Example: Conversion into Slack Form

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$$z = 3x_1 + x_2 + 2x_3$$

 $x_4 = 30 - x_1 - x_2 - 3x_3$
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Basic solution: $(\overline{x_1}, \overline{x_2}, \dots, \overline{x_6}) = (0, 0, 0, 30, 24, 36)$

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This basic solution is **feasible**
Objective value is 0.

Increasing the value of x_1 would increase the objective value.

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Solving for x₁ yields:

$$x_1 = 9 - \frac{x_2}{4} - \frac{x_3}{2} - \frac{x_6}{4}$$
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$$z = 3x_1 + x_2 + 2x_3$$

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Solving for x₁ yields:

$$x_1 = 9 - \frac{x_2}{4} - \frac{x_3}{2} - \frac{x_6}{4}$$
.

• Substitute this into x_1 in the other three equations

$$z = 27 + \frac{x_2}{4} + \frac{x_3}{2} - \frac{3x_6}{4}$$

$$x_1 = 9 - \frac{x_2}{4} - \frac{x_3}{2} - \frac{x_6}{4}$$

$$x_4 = 21 - \frac{3x_2}{4} - \frac{5x_3}{2} + \frac{x_6}{4}$$

$$x_5 = 6 - \frac{3x_2}{2} - 4x_3 + \frac{x_6}{2}$$

$$z = 27 + \frac{x_2}{4} + \frac{x_3}{2} - \frac{3x_6}{4}$$

$$x_1 = 9 - \frac{x_2}{4} - \frac{x_3}{2} - \frac{x_6}{4}$$

$$x_4 = 21 - \frac{3x_2}{4} - \frac{5x_3}{2} + \frac{x_6}{4}$$

$$x_5 = 6 - \frac{3x_2}{2} - 4x_3 + \frac{x_6}{2}$$

Basic solution: $(\overline{x_1}, \overline{x_2}, \dots, \overline{x_6}) = (9, 0, 0, 21, 6, 0)$ with objective value 27

Increasing the value of x_3 would increase the objective value.

$$z = 27 + \frac{x_2}{4} + \frac{x_3}{2} - \frac{3x_1}{4}$$

$$x_1 = 9 - \frac{x_2}{4} - \frac{x_3}{2} - \frac{x_4}{4}$$

$$x_4 = 21 - \frac{3x_2}{4} - \frac{5x_3}{2} + \frac{x_4}{4}$$

$$x_5 = 6 - \frac{3x_2}{2} - 4x_3 + \frac{x_4}{4}$$

Basic solution: $(\overline{x_1}, \overline{x_2}, \dots, \overline{x_6}) = (9, 0, 0, 21, 6, 0)$ with objective value 27

Increasing the value of x_3 would increase the objective value.

$$z = 27 + \frac{x_2}{4} + \frac{x_3}{2} - \frac{3x_6}{4}$$

$$x_1 = 9 - \frac{x_2}{4} - \frac{x_3}{2} - \frac{x_6}{4}$$

$$x_4 = 21 - \frac{3x_2}{4} - \frac{5x_3}{2} + \frac{x_6}{4}$$

$$x_5 = 6 - \frac{3x_2}{2} - 4x_3 + \frac{x_6}{2}$$

The third constraint is the tightest and limits how much we can increase x_3 .

Increasing the value of x_3 would increase the objective value.

$$z = 27 + \frac{x_2}{4} + \frac{x_3}{2} - \frac{3x_6}{4}$$

$$x_1 = 9 - \frac{x_2}{4} - \frac{x_3}{2} - \frac{x_6}{4}$$

$$x_4 = 21 - \frac{3x_2}{4} - \frac{5x_3}{2} + \frac{x_6}{4}$$

$$x_5 = 6 - \frac{3x_2}{2} - 4x_3 + \frac{x_6}{2}$$

The third constraint is the tightest and limits how much we can increase x_3 .

Switch roles of x_3 and x_5 :

Increasing the value of x_3 would increase the objective value.

$$z = 27 + \frac{x_2}{4} + \frac{x_3}{2} - \frac{3x_6}{4}$$

$$x_1 = 9 - \frac{x_2}{4} - \frac{x_3}{2} - \frac{x_6}{4}$$

$$x_4 = 21 - \frac{3x_2}{4} - \frac{5x_3}{2} + \frac{x_6}{4}$$

$$x_5 = 6 - \frac{3x_2}{2} - 4x_3 + \frac{x_6}{2}$$

The third constraint is the tightest and limits how much we can increase x_3 .

Switch roles of x_3 and x_5 :

Solving for x₃ yields:

$$x_3 = \frac{3}{2} - \frac{3x_2}{8} - \frac{x_5}{4} - \frac{x_6}{8}$$

Increasing the value of x_3 would increase the objective value.

$$z = 27 + \frac{x_2}{4} + \frac{x_3}{2} - \frac{3x_6}{4}$$

$$x_1 = 9 - \frac{x_2}{4} - \frac{x_3}{2} - \frac{x_6}{4}$$

$$x_4 = 21 - \frac{3x_2}{4} - \frac{5x_3}{2} + \frac{x_6}{4}$$

$$x_5 = 6 - \frac{3x_2}{2} - 4x_3 + \frac{x_6}{2}$$

The third constraint is the tightest and limits how much we can increase x_3 .

Switch roles of x_3 and x_5 :

Solving for x₃ yields:

$$x_3 = \frac{3}{2} - \frac{3x_2}{8} - \frac{x_5}{4} - \frac{x_6}{8}.$$

• Substitute this into x_3 in the other three equations

$$z = \frac{111}{4} + \frac{x_2}{16} - \frac{x_5}{8} - \frac{11x_6}{16}$$

$$x_1 = \frac{33}{4} - \frac{x_2}{16} + \frac{x_5}{8} - \frac{5x_6}{16}$$

$$x_3 = \frac{3}{2} - \frac{3x_2}{8} - \frac{x_5}{4} + \frac{x_6}{8}$$

$$x_4 = \frac{69}{4} + \frac{3x_2}{16} + \frac{5x_5}{8} - \frac{x_6}{16}$$

$$z = \frac{111}{4} + \frac{x_2}{16} - \frac{x_5}{8} - \frac{11x_6}{16}$$

$$x_1 = \frac{33}{4} - \frac{x_2}{16} + \frac{x_5}{8} - \frac{5x_6}{16}$$

$$x_3 = \frac{3}{2} - \frac{3x_2}{8} - \frac{x_5}{4} + \frac{x_6}{8}$$

$$x_4 = \frac{69}{4} + \frac{3x_2}{16} + \frac{5x_5}{8} - \frac{x_6}{16}$$

Basic solution: $(\overline{x_1},\overline{x_2},\ldots,\overline{x_6})=(\frac{33}{4},0,\frac{3}{2},\frac{69}{4},0,0)$ with objective value $\frac{111}{4}=27.75$

Increasing the value of x_2 would increase the objective value.

$$z = \frac{111}{4} + \frac{x_2}{16} - \frac{x_5}{8} - \frac{11x_6}{16}$$

$$x_1 = \frac{33}{4} - \frac{x_2}{16} + \frac{x_5}{8} - \frac{5x_6}{16}$$

$$x_3 = \frac{3}{2} - \frac{3x_2}{8} - \frac{x_5}{4} + \frac{x_6}{8}$$

$$x_4 = \frac{69}{4} + \frac{3x_2}{16} + \frac{5x_5}{8} - \frac{x_6}{16}$$

Basic solution: $(\overline{X_1}, \overline{X_2}, \dots, \overline{X_6}) = (\frac{33}{4}, 0, \frac{3}{2}, \frac{69}{4}, 0, 0)$ with objective value $\frac{111}{4} = 27.75$

Increasing the value of x_2 would increase the objective value.

$$z = \frac{111}{4} + \frac{x_2}{16} - \frac{x_5}{8} - \frac{11x_6}{16}$$

$$x_1 = \frac{33}{4} - \frac{x_2}{16} + \frac{x_5}{8} - \frac{5x_6}{16}$$

$$x_3 = \frac{3}{2} - \frac{3x_2}{8} - \frac{x_5}{4} + \frac{x_6}{8}$$

$$x_4 = \frac{69}{4} + \frac{3x_2}{16} + \frac{5x_5}{8} - \frac{x_6}{16}$$

The second constraint is the tightest and limits how much we can increase x_2 .

Increasing the value of x_2 would increase the objective value.

$$z = \frac{111}{4} + \frac{x_2}{16} - \frac{x_5}{8} - \frac{11x_6}{16}$$

$$x_1 = \frac{33}{4} - \frac{x_2}{16} + \frac{x_5}{8} - \frac{5x_6}{16}$$

$$x_3 = \frac{3}{2} - \frac{3x_2}{8} - \frac{x_5}{4} + \frac{x_6}{8}$$

$$x_4 = \frac{69}{4} + \frac{3x_2}{16} + \frac{5x_5}{8} - \frac{x_6}{16}$$

The second constraint is the tightest and limits how much we can increase x_2 .

Switch roles of x_2 and x_3 :

Increasing the value of x_2 would increase the objective value.

$$z = \frac{111}{4} + \frac{x_2}{16} - \frac{x_5}{8} - \frac{11x_6}{16}$$

$$x_1 = \frac{33}{4} - \frac{x_2}{16} + \frac{x_5}{8} - \frac{5x_6}{16}$$

$$x_3 = \frac{3}{2} - \frac{3x_2}{8} - \frac{x_5}{4} + \frac{x_6}{8}$$

$$x_4 = \frac{69}{4} + \frac{3x_2}{16} + \frac{5x_5}{8} - \frac{x_6}{16}$$

The second constraint is the tightest and limits how much we can increase x_2 .

Switch roles of x_2 and x_3 :

Solving for x₂ yields:

$$x_2 = 4 - \frac{8x_3}{3} - \frac{2x_5}{3} + \frac{x_6}{3}$$
.

Increasing the value of x_2 would increase the objective value.

$$z = \frac{111}{4} + \frac{x_2}{16} - \frac{x_5}{8} - \frac{11x_6}{16}$$

$$x_1 = \frac{33}{4} - \frac{x_2}{16} + \frac{x_5}{8} - \frac{5x_6}{16}$$

$$x_3 = \frac{3}{2} - \frac{3x_2}{8} - \frac{x_5}{4} + \frac{x_6}{8}$$

$$x_4 = \frac{69}{4} + \frac{3x_2}{16} + \frac{5x_5}{8} - \frac{x_6}{16}$$

The second constraint is the tightest and limits how much we can increase x_2 .

Switch roles of x_2 and x_3 :

Solving for x₂ yields:

$$x_2 = 4 - \frac{8x_3}{3} - \frac{2x_5}{3} + \frac{x_6}{3}$$
.

• Substitute this into x_2 in the other three equations

$$z = 28 - \frac{x_3}{6} - \frac{x_5}{6} - \frac{2x_6}{3}$$

$$x_1 = 8 + \frac{x_3}{6} + \frac{x_5}{6} - \frac{x_6}{3}$$

$$x_2 = 4 - \frac{8x_3}{3} - \frac{2x_5}{3} + \frac{x_6}{3}$$

$$x_4 = 18 - \frac{x_3}{3} + \frac{x_5}{3}$$

$$z = 28 - \frac{x_3}{6} - \frac{x_5}{6} - \frac{2x_6}{3}$$

$$x_1 = 8 + \frac{x_3}{6} + \frac{x_5}{6} - \frac{x_6}{3}$$

$$x_2 = 4 - \frac{8x_3}{3} - \frac{2x_5}{3} + \frac{x_6}{3}$$

$$x_4 = 18 - \frac{x_3}{2} + \frac{x_5}{2}$$

Basic solution: $(\overline{x_1}, \overline{x_2}, \dots, \overline{x_6}) = (8, 4, 0, 18, 0, 0)$ with objective value 28

All coefficients are negative, and hence this basic solution is optimal!

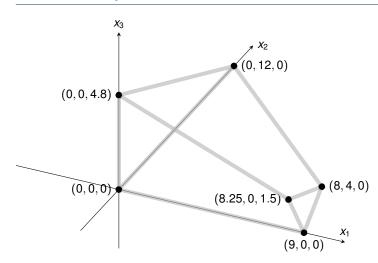
$$z = 28 - \frac{x_3}{6} - \frac{x_5}{6} - \frac{2x_6}{3}$$

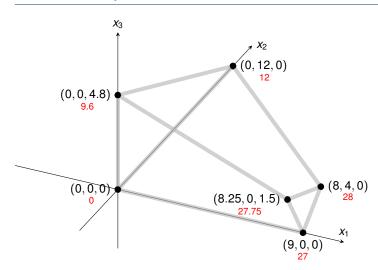
$$x_1 = 8 + \frac{x_3}{6} + \frac{x_5}{6} - \frac{x_6}{3}$$

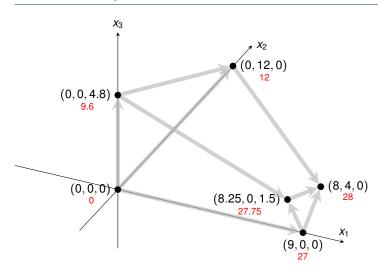
$$x_2 = 4 - \frac{8x_3}{3} - \frac{2x_5}{3} + \frac{x_6}{3}$$

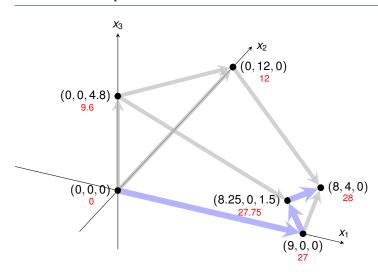
$$x_4 = 18 - \frac{x_3}{2} + \frac{x_6}{2}$$

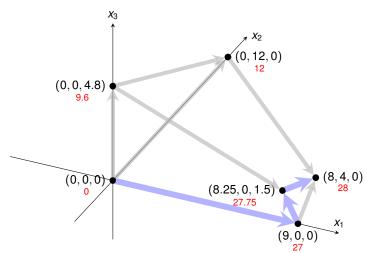
Basic solution: $(\overline{x_1}, \overline{x_2}, \dots, \overline{x_6}) = (8, 4, 0, 18, 0, 0)$ with objective value 28













Exercise: How many basic solutions (including non-feasible ones) are there?

$$z = 3x_1 + x_2 + 2x_3$$

 $x_4 = 30 - x_1 - x_2 - 3x_3$
 $x_5 = 24 - 2x_1 - 2x_2 - 5x_3$
 $x_6 = 36 - 4x_1 - x_2 - 2x_3$

$$z = 3x_1 + x_2 + 2x_3$$

 $x_4 = 30 - x_1 - x_2 - 3x_3$
 $x_5 = 24 - 2x_1 - 2x_2 - 5x_3$
 $x_6 = 36 - 4x_1 - x_2 - 2x_3$

Switch roles of x_1 and x_6 _____

Switch roles of x_1 and x_6 _____

$$z = \frac{111}{4} + \frac{x_2}{16} - \frac{x_5}{8} - \frac{11x_6}{16}$$

$$x_1 = \frac{33}{4} - \frac{x_2}{16} + \frac{x_5}{8} - \frac{5x_6}{16}$$

$$x_3 = \frac{3}{2} - \frac{3x_2}{8} - \frac{x_5}{4} + \frac{x_6}{8}$$

$$x_4 = \frac{69}{4} + \frac{3x_2}{16} + \frac{5x_5}{8} - \frac{x_6}{16}$$

$$z = 3x_1 + x_2 + 2x_3$$

$$x_4 = 30 - x_1 - x_2 - 3x_3$$

$$x_5 = 24 - 2x_1 - 2x_2 - 5x_3$$

$$x_6 = 36 - 4x_1 - x_2 - 2x_3$$

$$x_6 = 36 - 4x_1 - x_2 - 2x_3$$

$$x_6 = 36 - 4x_1 - x_2 - 2x_3$$
Switch roles of x_3 and x_5

$$z = \frac{48}{5} + \frac{11x_1}{5} + \frac{x_2}{5} - \frac{2x_5}{5}$$

$$x_4 = \frac{78}{5} + \frac{x_1}{5} + \frac{x_2}{5} + \frac{3x_5}{5}$$

$$x_3 = \frac{24}{5} - \frac{2x_1}{5} - \frac{2x_2}{5} - \frac{x_5}{5}$$

$$x_6 = \frac{132}{5} - \frac{16x_1}{5} - \frac{x_2}{5} + \frac{2x_3}{5}$$
Switch roles of x_1 and x_6

$$x_6 = \frac{132}{5} - \frac{16x_1}{5} - \frac{x_2}{5} + \frac{2x_3}{5}$$
Switch roles of x_2 and x_3

$$x_6 = \frac{132}{5} - \frac{11x_6}{16}$$

$$x_6 = \frac{132}{5} - \frac{16x_1}{5} - \frac{x_2}{5} + \frac{2x_3}{5}$$

$$x_6 = \frac{132}{5} - \frac{16x_1}{5} - \frac{x_2}{5} + \frac{2x_3}{5}$$
Switch roles of x_2 and x_3

$$x_6 = \frac{13x_2}{5} - \frac{x_5}{16} + \frac{x_5}{8} - \frac{11x_6}{16}$$

$$x_6 = \frac{3x_2}{8} - \frac{x_5}{8} - \frac{5x_5}{16}$$

$$x_6 = \frac{13x_5}{16} - \frac{x_5}{8} - \frac{11x_6}{16}$$

$$x_6 = \frac{3x_2}{8} - \frac{x_5}{8} - \frac{5x_5}{16}$$

$$x_6 = \frac{13x_5}{16} - \frac{x_5}{8} - \frac{11x_6}{16}$$

 $\frac{x_2}{16}$

 X_1 *X*3

<u>69</u>

$$z = 3x_1 + x_2 + 2x_3$$

$$x_4 = 30 - x_1 - x_2 - 3x_3$$

$$x_5 = 24 - 2x_1 - 2x_2 - 5x_3$$

$$x_6 = 36 - 4x_1 - x_2 - 2x_3$$

$$y = 3x_1 + x_2 + 2x_3$$

$$x_6 = 3x_1 + x_2 + 2x_3$$

$$x_6 = 3x_1 + x_2 + 2x_3$$

$$x_6 = 3x_1 - x_2 - 3x_3$$

$$x_6 = 3x_1 - x_2 - 3x_2$$

$$x_6 = 3x_1 - x_2 - 3x_3$$

$$x_6 = 3x_1 - x_2 - 3x_2$$

$$x_6 = 3x_1 - x_2 - 3x_2$$

$$x_1 - x_2 - 2x_2 - 3x_3$$

$$x_2 - x_3 - x_4 - x_4 - x_2 - 2x_3$$

$$x_3 - x_4 - x_4 - x_5 - x_5 - x_5$$

$$x_4 = x_1 - x_2 - x_3 - x_5$$

$$x_4 = x_1 - x_2 - x_3 - x_5$$

$$x_4 = x_1 - x_2 - x_3 - x_5$$

$$x_4 = x_1 - x_2 - x_3$$

$$x_4 = x_1 - x_2 - x_3$$

$$x_4 = x_1 - x_2 - x_3$$

$$x_5 - x_5 - x_5 - x_5$$

$$x_6 = x_1 - x_2 - x_3$$

$$x_1 - x_2 - x_3$$

$$x_2 - x_3 - x_4 - x_5 - x_5$$

$$x_1 = x_1 - x_2 - x_3$$

$$x_2 - x_3 - x_4 - x_5 - x_5$$

$$x_3 - x_4 - x_5 - x_5$$

$$x_4 - x_5 - x_5$$

$$x_4 - x_5 - x_5$$

$$x_4 - x_5 - x_5$$

$$x_5 - x_$$

X1

Xз

<u>69</u>

```
PIVOT(N, B, A, b, c, v, l, e)
      // Compute the coefficients of the equation for new basic variable x_e.
 2 let \widehat{A} be a new m \times n matrix
 \hat{b}_e = b_l/a_{le}
 4 for each i \in N - \{e\}
       \hat{a}_{ei} = a_{li}/a_{le}
 6 \hat{a}_{el} = 1/a_{le}
 7 // Compute the coefficients of the remaining constraints.
 8 for each i \in B - \{l\}
     \hat{b}_i = b_i - a_{ie}\hat{b}_e
10 for each j \in N - \{e\}
              \hat{a}_{ii} = a_{ii} - a_{ie}\hat{a}_{ei}
     \hat{a}_{il} = -a_{ie}\hat{a}_{el}
13 // Compute the objective function.
14 \hat{v} = v + c_{\theta} \hat{b}_{\theta}
15 for each j \in N - \{e\}
16
     \hat{c}_i = c_i - c_e \hat{a}_{ei}
      \hat{c}_l = -c_e \hat{a}_{el}
18 // Compute new sets of basic and nonbasic variables.
19 \hat{N} = N - \{e\} \cup \{l\}
20 \hat{B} = B - \{l\} \cup \{e\}
21 return (\hat{N}, \hat{B}, \hat{A}, \hat{b}, \hat{c}, \hat{v})
```

```
PIVOT(N, B, A, b, c, v, l, e)
      // Compute the coefficients of the equation for new basic variable x_e.
     let \widehat{A} be a new m \times n matrix
 \hat{b}_e = b_l/a_{le}
                                                                                     Rewrite "tight" equation
 4 for each i \in N - \{e\}
                                                                                     for enterring variable x_e.
        \hat{a}_{ei} = a_{li}/a_{le}
 6 \hat{a}_{el} = 1/a_{le}
     // Compute the coefficients of the remaining constraints.
 8 for each i \in B - \{l\}
      \hat{b}_i = b_i - a_{ie}\hat{b}_e
     for each j \in N - \{e\}
                \hat{a}_{ii} = a_{ii} - a_{ie}\hat{a}_{ei}
     \hat{a}_{il} = -a_{ia}\hat{a}_{al}
     // Compute the objective function.
14 \hat{v} = v + c_{\theta} \hat{b}_{\theta}
15 for each j \in N - \{e\}
      \hat{c}_i = c_i - c_e \hat{a}_{ei}
16
     \hat{c}_1 = -c_{\alpha}\hat{a}_{\alpha 1}
18 // Compute new sets of basic and nonbasic variables.
19 \hat{N} = N - \{e\} \cup \{l\}
20 \hat{B} = B - \{l\} \cup \{e\}
21 return (\hat{N}, \hat{B}, \hat{A}, \hat{b}, \hat{c}, \hat{v})
```

```
PIVOT(N, B, A, b, c, v, l, e)
      // Compute the coefficients of the equation for new basic variable x_e.
     let \widehat{A} be a new m \times n matrix
 \hat{b}_e = b_l/a_{le}
                                                                                    Rewrite "tight" equation
 4 for each i \in N - \{e\}
                                                                                   for enterring variable x_e.
        \hat{a}_{ei} = a_{li}/a_{le}
 6 \hat{a}_{el} = 1/a_{le}
     // Compute the coefficients of the remaining constraints.
 8 for each i \in B - \{l\}
      \hat{b}_i = b_i - a_{ie}\hat{b}_e
                                                                                    Substituting x_e into
     for each j \in N - \{e\}
                                                                                      other equations.
                \hat{a}_{ii} = a_{ii} - a_{ie}\hat{a}_{ei}
     \hat{a}_{il} = -a_{ia}\hat{a}_{al}
     // Compute the objective function.
14 \hat{v} = v + c_{\theta} \hat{b}_{\theta}
15 for each i \in N - \{e\}
      \hat{c}_i = c_i - c_e \hat{a}_{ei}
16
     \hat{c}_1 = -c_{\alpha}\hat{a}_{\alpha 1}
    // Compute new sets of basic and nonbasic variables.
19 \hat{N} = N - \{e\} \cup \{l\}
20 \hat{B} = B - \{l\} \cup \{e\}
21 return (\hat{N}, \hat{B}, \hat{A}, \hat{b}, \hat{c}, \hat{v})
```

```
PIVOT(N, B, A, b, c, v, l, e)
      // Compute the coefficients of the equation for new basic variable x_e.
     let \widehat{A} be a new m \times n matrix
 \hat{b}_e = b_I/a_{Ie}
                                                                                  Rewrite "tight" equation
   for each j \in N - \{e\}
                                                                                  for enterring variable x_e.
        \hat{a}_{ei} = a_{li}/a_{le}
 6 \hat{a}_{el} = 1/a_{le}
     // Compute the coefficients of the remaining constraints.
 8 for each i \in B - \{l\}
      \hat{b}_i = b_i - a_{ie}\hat{b}_e
                                                                                   Substituting x_e into
     for each j \in N - \{e\}
                                                                                     other equations.
                \hat{a}_{ii} = a_{ii} - a_{ie}\hat{a}_{ei}
     \hat{a}_{il} = -a_{ia}\hat{a}_{al}
     // Compute the objective function.
14 \hat{v} = v + c_{\theta} \hat{b}_{\theta}
                                                                                   Substituting x_e into
     for each i \in N - \{e\}
                                                                                   objective function.
      \hat{c}_i = c_i - c_e \hat{a}_{ei}
16
     \hat{c}_1 = -c_{\alpha}\hat{a}_{\alpha 1}
     // Compute new sets of basic and nonbasic variables.
19 \hat{N} = N - \{e\} \cup \{l\}
20 \hat{B} = B - \{l\} \cup \{e\}
21 return (\hat{N}, \hat{B}, \hat{A}, \hat{b}, \hat{c}, \hat{v})
```

```
PIVOT(N, B, A, b, c, v, l, e)
      // Compute the coefficients of the equation for new basic variable x_e.
     let \widehat{A} be a new m \times n matrix
 \hat{b}_e = b_l/a_{le}
                                                                                Rewrite "tight" equation
   for each i \in N - \{e\}
                                                                                for enterring variable x_e.
        \hat{a}_{ei} = a_{li}/a_{le}
 6 \hat{a}_{el} = 1/a_{le}
     // Compute the coefficients of the remaining constraints.
 8 for each i \in B - \{l\}
      \hat{b}_i = b_i - a_{ia}\hat{b}_a
                                                                                 Substituting x_e into
     for each j \in N - \{e\}
                                                                                   other equations.
                \hat{a}_{ii} = a_{ii} - a_{ie}\hat{a}_{ei}
     \hat{a}_{il} = -a_{ia}\hat{a}_{al}
     // Compute the objective function.
14 \hat{v} = v + c_{\theta} \hat{b}_{\theta}
                                                                                 Substituting x_e into
     for each i \in N - \{e\}
                                                                                 objective function.
     \hat{c}_i = c_i - c_e \hat{a}_{ei}
16
     \hat{c}_1 = -c_{\alpha}\hat{a}_{\alpha 1}
     // Compute new sets of basic and nonbasic variables.
19 \hat{N} = N - \{e\} \cup \{l\}
                                                                                  Update non-basic
20 \hat{B} = B - \{l\} \cup \{e\}
                                                                                and basic variables
21 return (\hat{N}, \hat{B}, \hat{A}, \hat{b}, \hat{c}, \hat{v})
```

```
PIVOT(N, B, A, b, c, v, l, e)
      // Compute the coefficients of the equation for new basic variable x_e.
     let \widehat{A} be a new m \times n matrix
 \hat{b}_e = b_I/a_{Ie}
                                                                                Rewrite "tight" equation
    for each j \in N - \{e\} Need that a_{le} \neq 0!
          \hat{a}_{ei} = a_{li}/a_{le}
                                                                               for enterring variable x_e.
 6 \hat{a}_{el} = 1/a_{le}
     // Compute the coefficients of the remaining constraints.
 8 for each i \in B - \{l\}
      \hat{b}_i = b_i - a_{ia}\hat{b}_a
                                                                                Substituting x_e into
     for each j \in N - \{e\}
                                                                                  other equations.
               \hat{a}_{ii} = a_{ii} - a_{ie}\hat{a}_{ei}
    \hat{a}_{il} = -a_{ia}\hat{a}_{al}
     // Compute the objective function.
14 \hat{v} = v + c_{\theta} \hat{b}_{\theta}
                                                                                Substituting x_e into
   for each i \in N - \{e\}
                                                                                 objective function.
     \hat{c}_i = c_i - c_e \hat{a}_{ei}
16
     \hat{c}_1 = -c_{\alpha}\hat{a}_{\alpha 1}
     // Compute new sets of basic and nonbasic variables.
19 \hat{N} = N - \{e\} \cup \{l\}
                                                                                 Update non-basic
20 \hat{B} = B - \{l\} \cup \{e\}
                                                                                and basic variables
21 return (\hat{N}, \hat{B}, \hat{A}, \hat{b}, \hat{c}, \hat{v})
```

Lemma 29.1

Consider a call to PIVOT(N, B, A, b, c, v, l, e) in which $a_{le} \neq 0$. Let the values returned from the call be $(\widehat{N}, \widehat{B}, \widehat{A}, \widehat{b}, \widehat{c}, \widehat{v})$, and let \overline{x} denote the basic solution after the call. Then

Lemma 29.1

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- 1. $\overline{x}_j = 0$ for each $j \in \widehat{N}$.
- 2. $\overline{x}_e = b_l/a_{le}$.
- 3. $\overline{x}_i = b_i a_{ie}\widehat{b}_e$ for each $i \in \widehat{B} \setminus \{e\}$.

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Proof:

Lemma 29.1

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Proof:

- 1. holds since the basic solution always sets all non-basic variables to zero.
- 2. When we set each non-basic variable to 0 in a constraint

$$x_i = \widehat{b}_i - \sum_{j \in \widehat{N}} \widehat{a}_{ij} x_j,$$

we have $\overline{x}_i = \hat{b}_i$ for each $i \in \hat{B}$. Hence $\overline{x}_e = \hat{b}_e = b_l/a_{le}$.

3. After substituting into the other constraints, we have

$$\overline{x}_i = \widehat{b}_i = b_i - a_{ie}\widehat{b}_e.$$

Lemma 29.1

Consider a call to PIVOT(N,B,A,b,c,v,l,e) in which $a_{le}\neq 0$. Let the values returned from the call be $(\widehat{N},\widehat{B},\widehat{A},\widehat{b},\widehat{c},\widehat{v})$, and let \overline{x} denote the basic solution after the call. Then

- 1. $\overline{x}_j = 0$ for each $j \in \widehat{N}$.
- 2. $\overline{x}_e = b_l/a_{le}$.
- 3. $\overline{x}_i = b_i a_{ie}\widehat{b}_e$ for each $i \in \widehat{B} \setminus \{e\}$.

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3. After substituting into the other constraints, we have

$$\overline{X}_i = \widehat{b}_i = b_i - a_{ie}\widehat{b}_e.$$

Formalizing the Simplex Algorithm: Questions

Questions:

- How do we determine whether a linear program is feasible?
- What do we do if the linear program is feasible, but the initial basic solution is not feasible?
- How do we determine whether a linear program is unbounded?
- How do we choose the entering and leaving variables?

Formalizing the Simplex Algorithm: Questions

Questions:

- How do we determine whether a linear program is feasible?
- What do we do if the linear program is feasible, but the initial basic solution is not feasible?
- How do we determine whether a linear program is unbounded?
- How do we choose the entering and leaving variables?

Example before was a particularly nice one!

```
SIMPLEX(A, b, c)
     (N, B, A, b, c, v) = \text{INITIALIZE-SIMPLEX}(A, b, c)
     let \Delta be a new vector of length m
     while some index j \in N has c_i > 0
           choose an index e \in N for which c_e > 0
          for each index i \in B
                if a_{ie} > 0
                     \Delta_i = b_i/a_{ie}
                else \Delta_i = \infty
 9
          choose an index l \in B that minimizes \Delta_i
10
          if \Delta_I == \infty
11
                return "unbounded"
12
          else (N, B, A, b, c, v) = PIVOT(N, B, A, b, c, v, l, e)
     for i = 1 to n
14
          if i \in B
               \bar{x}_i = b_i
15
          else \bar{x}_i = 0
16
     return (\bar{x}_1, \bar{x}_2, \dots, \bar{x}_n)
```

```
SIMPLEX(A, b, c)
                                                                            Returns a slack form with a
     (N, B, A, b, c, v) = \text{INITIALIZE-SIMPLEX}(A, b, c)
                                                                        feasible basic solution (if it exists)
     let \Delta be a new vector of length m
     while some index j \in N has c_i > 0
           choose an index e \in N for which c_e > 0
          for each index i \in B
                if a_{ie} > 0
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```

```
SIMPLEX(A, b, c)
                                                                             Returns a slack form with a
     (N, B, A, b, c, v) = \text{INITIALIZE-SIMPLEX}(A, b, c)
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    let \Delta be a new vector of length \underline{m}
    while some index j \in N has c_i > 0
           choose an index e \in N for which c_e > 0
          for each index i \in B
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                     \Delta_i = b_i/a_{ie}
                else \Delta_i = \infty
          choose an index l \in B that minimizes \Delta_i
          if \Delta_I == \infty
10
11
                return "unbounded"
          else (N, B, A, b, c, v) = PIVOT(N, B, A, b, c, v, l, e)
     for i = 1 to n
          if i \in B
14
               \bar{x}_i = b_i
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     return (\bar{x}_1, \bar{x}_2, \dots, \bar{x}_n)
```

```
SIMPLEX(A, b, c)
                                                                          Returns a slack form with a
     (N, B, A, b, c, v) = \text{INITIALIZE-SIMPLEX}(A, b, c)
                                                                     feasible basic solution (if it exists)
    let \Delta be a new vector of length \underline{m}
    while some index j \in N has c_i > 0
                                                                              Main Loop:
          choose an index e \in N for which c_e > 0
          for each index i \in B
               if a_{ie} > 0
                    \Delta_i = b_i/a_{ie}
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          if \Delta_I == \infty
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11
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          else (N, B, A, b, c, v) = PIVOT(N, B, A, b, c, v, l, e)
     for i = 1 to n
          if i \in B
14
               \bar{x}_i = b_i
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          else \bar{x}_i = 0
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```

return $(\bar{x}_1, \bar{x}_2, \dots, \bar{x}_n)$

```
SIMPLEX(A, b, c)
                                                                        Returns a slack form with a
     (N, B, A, b, c, v) = \text{INITIALIZE-SIMPLEX}(A, b, c)
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    let \Delta be a new vector of length \underline{m}
    while some index j \in N has c_i > 0
                                                                            Main Loop:
          choose an index e \in N for which c_e > 0
          for each index i \in B

    terminates if all coefficients in

                                                                                 objective function are negative
               if a_{ia} > 0
                    \Delta_i = b_i/a_{ie}
                                                                              Line 4 picks enterring variable
               else \Delta_i = \infty
                                                                                 x<sub>e</sub> with negative coefficient
          choose an index l \in B that minimizes \Delta_i
                                                                               ■ Lines 6 — 9 pick the tightest
          if \Delta_I == \infty
10
                                                                                 constraint, associated with x1
11
               return "unbounded"
                                                                               Line 11 returns "unbounded" if
12
          else (N, B, A, b, c, v) = PIVOT(N, B, A, b, c, v, l, e)
                                                                                 there are no constraints
     for i = 1 to n
                                                                              Line 12 calls PIVOT, switching
14
          if i \in R
                                                                                 roles of x_i and x_e
              \bar{x}_i = b_i
15
          else \bar{x}_i = 0
16
```

return $(\bar{x}_1, \bar{x}_2, \dots, \bar{x}_n)$

```
SIMPLEX(A, b, c)
                                                                          Returns a slack form with a
     (N, B, A, b, c, v) = \text{INITIALIZE-SIMPLEX}(A, b, c)
                                                                       feasible basic solution (if it exists)
    let \Delta be a new vector of length \underline{m}
    while some index j \in N has c_i > 0
                                                                               Main Loop:
          choose an index e \in N for which c_e > 0
          for each index i \in B

    terminates if all coefficients in

                                                                                    objective function are negative
               if a_{ia} > 0
                    \Delta_i = b_i/a_{ie}

    Line 4 picks enterring variable

               else \Delta_i = \infty
                                                                                    x<sub>e</sub> with negative coefficient
          choose an index l \in B that minimizes \Delta_i
                                                                                 ■ Lines 6 — 9 pick the tightest
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                                                                                    constraint, associated with x1
11
               return "unbounded"
                                                                                 Line 11 returns "unbounded" if
          else (N, B, A, b, c, v) = PIVOT(N, B, A, b, c, v, l, e)
                                                                                    there are no constraints
     for i = 1 to n
                                                                                 Line 12 calls PIVOT, switching
14
          if i \in R
                                                                                    roles of x_i and x_e
               \bar{x}_i = b_i
15
          else \bar{x}_i = 0
16
     return (\bar{x}_1, \bar{x}_2, \dots, \bar{x}_n)
```

Return corresponding solution.

```
SIMPLEX(A, b, c)
                                                                            Returns a slack form with a
     (N, B, A, b, c, v) = \text{INITIALIZE-SIMPLEX}(A, b, c)
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          else (N, B, A, b, c, v) = PIVOT(N, B, A, b, c, v, l, e)
     for i = 1 to n
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          if i \in R
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               \bar{x}_i = b_i
          else \bar{x}_i = 0
16
     return (\bar{x}_1, \bar{x}_2, \dots, \bar{x}_n)
```

- Lemma 29 2

Suppose the call to INITIALIZE-SIMPLEX in line 1 returns a slack form for which the basic solution is feasible. Then if SIMPLEX returns a solution, it is a feasible solution. If SIMPLEX returns "unbounded", the linear program is unbounded.

```
SIMPLEX (A,b,c) Returns a slack form with a feasible basic solution (if it exists)

1 (N,B,A,b,c,\nu) = \text{INITIALIZE-SIMPLEX}(A,b,c) feasible basic solution (if it exists)

2 \underline{\text{let } \Delta} be a new vector of length \underline{m}

3 while some index j \in N has c_j > 0

4 choose an index e \in N for which c_e > 0

5 for each index i \in B

6 if a_{ie} > 0

7 \Delta_i = b_i/a_{ie}

8 \underline{\text{else } \Delta_i = \infty}

9 choose an index l \in B that minimizes \Delta_i

10 if \Delta_l == \infty

11 return "unbounded"
```

Proof is based on the following three-part loop invariant:

Lemma 29 2 =

Suppose the call to INITIALIZE-SIMPLEX in line 1 returns a slack form for which the basic solution is feasible. Then if SIMPLEX returns a solution, it is a feasible solution. If SIMPLEX returns "unbounded", the linear program is unbounded.

```
SIMPLEX (A,b,c)

1 (N,B,A,b,c,\nu) = INITIALIZE-SIMPLEX (A,b,c)

2 \det \Delta be a new vector of length m

3 while some index j \in N has c_j > 0

4 choose an index e \in N for which c_e > 0

5 for each index i \in B

6 if a_{ie} > 0

7 \Delta_i = b_i/a_{ie}

8 else \Delta_i = \infty

9 choose an index l \in B that minimizes \Delta_i

10 if \Delta_l = \infty

11 return "unbounded"
```

Proof is based on the following three-part loop invariant:

- 1. the slack form is always equivalent to the one returned by INITIALIZE-SIMPLEX,
- 2. for each $i \in B$, we have $b_i \ge 0$,
- 3. the basic solution associated with the (current) slack form is feasible.

Lemma 29.2 -

Suppose the call to INITIALIZE-SIMPLEX in line 1 returns a slack form for which the basic solution is feasible. Then if SIMPLEX returns a solution, it is a feasible solution. If SIMPLEX returns "unbounded", the linear program is unbounded.

$$z = x_1 + x_2 + x_3$$

 $x_4 = 8 - x_1 - x_2$
 $x_5 = x_2 - x_3$

$$z = x_1 + x_2 + x_3$$

$$x_4 = 8 - x_1 - x_2$$

$$x_5 = x_2 - x_3$$

$$\begin{vmatrix} \text{Pivot with } x_1 \text{ entering and } x_4 \text{ leaving} \end{vmatrix}$$

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$$x_4 = 8 - x_1 - x_2$$

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$$\begin{vmatrix} \text{Pivot with } x_1 \text{ entering and } x_4 \text{ leaving} \end{vmatrix}$$

$$z = 8 + x_3 - x_4$$

$$x_1 = 8 - x_2 - x_3$$

$$\begin{vmatrix} \text{Pivot with } x_3 \text{ entering and } x_5 \text{ leaving} \end{vmatrix}$$

$$y$$

Degeneracy: One iteration of SIMPLEX leaves the objective value unchanged.

Linear Programming @ Thomas Sauerwald

z

*X*₁

*X*₃

8

8

 X_2

 X_2

 X_2

 X_4

 X_4

 X_5

*X*₅



Exercise: Execute one more step of the Simplex Algorithm on the tableau from the previous slide.

Termination and Running Time

Cycling: SIMPLEX may fail to terminate.

Termination and Running Time

It is theoretically possible, but very rare in practice.

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Anti-Cycling Strategies

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Cycling: SIMPLEX may fail to terminate.

Anti-Cycling Strategies

1. Bland's rule: Choose entering variable with smallest index

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Anti-Cycling Strategies -

- 1. Bland's rule: Choose entering variable with smallest index
- 2. Random rule: Choose entering variable uniformly at random

It is theoretically possible, but very rare in practice.

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- Anti-Cycling Strategies
- 1. Bland's rule: Choose entering variable with smallest index
- 2. Random rule: Choose entering variable uniformly at random
- 3. Perturbation: Perturb the input slightly so that it is impossible to have two solutions with the same objective value

It is theoretically possible, but very rare in practice.

Cycling: SIMPLEX may fail to terminate.

Anti-Cycling Strategies

- 1. Bland's rule: Choose entering variable with smallest index
- 2. Random rule: Choose entering variable uniformly at random
- 3. Perturbation: Perturb the input slightly so that it is impossible to have two solutions with the same objective value

Replace each b_i by $\hat{b}_i = b_i + \epsilon_i$, where $\epsilon_i \gg \epsilon_{i+1}$ are all small.

It is theoretically possible, but very rare in practice.

Cycling: SIMPLEX may fail to terminate.

Anti-Cycling Strategies -

- 1. Bland's rule: Choose entering variable with smallest index
- 2. Random rule: Choose entering variable uniformly at random
- 3. Perturbation: Perturb the input slightly so that it is impossible to have two solutions with the same objective value

Replace each
$$b_i$$
 by $\widehat{b}_i = b_i + \epsilon_i$, where $\epsilon_i \gg \epsilon_{i+1}$ are all small.

- Lemma 29.7

Assuming Initialize-Simplex returns a slack form for which the basic solution is feasible, Simplex either reports that the program is unbounded or returns a feasible solution in at most $\binom{n+m}{m}$ iterations.

It is theoretically possible, but very rare in practice.

Cycling: SIMPLEX may fail to terminate.

Anti-Cycling Strategies

- 1. Bland's rule: Choose entering variable with smallest index
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Replace each
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Lemma 29.7

Assuming Initialize-Simplex returns a slack form for which the basic solution is feasible, Simplex either reports that the program is unbounded or returns a feasible solution in at most $\binom{n+m}{m}$ iterations.

Every set *B* of basic variables uniquely determines a slack form, and there are at most $\binom{n+m}{m}$ unique slack forms.

Outline

Introduction

A Simple Example of a Linear Program

Formulating Problems as Linear Programs

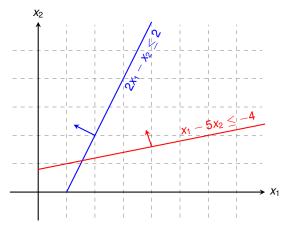
Standard and Slack Forms

Simplex Algorithm

maximise
$$2x_1 - x_2$$
 subject to
$$2x_1 - x_2 \le 2$$
 $x_1 - 5x_2 \le -4$ $x_1, x_2 \ge 0$ Conversion into slack form
$$z = 2x_1 - x_2$$
 $x_3 = 2 - 2x_1 - x_2$ $x_4 = -4 - x_1 + 5x_2$
Basic solution $(x_1, x_2, x_3, x_4) = (0, 0, 2, -4)$ is not feasible!

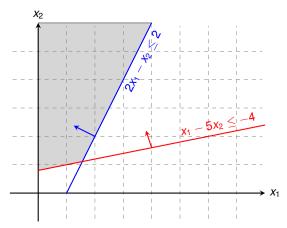
maximise subject to

$$2x_1 - x_2$$



maximise subject to

$$2x_1 - x_2$$



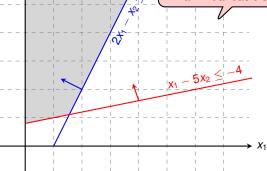
maximise subject to

$$2x_1 - x_2$$



Questions:

- How to determine whether there is any feasible solution?
- If there is one, how to determine an initial basic solution?



maximise subject to

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

$$\begin{array}{ccc} \sum_{j=1}^n a_{ij} x_j & \leq & b_i & \text{ for } i=1,2,\ldots,m, \\ x_j & \geq & 0 & \text{ for } j=1,2,\ldots,n \end{array}$$

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

$$\begin{array}{cccc} \sum_{j=1}^n a_{ij} x_j & \leq & b_i & \text{for } i=1,2,\ldots,m, \\ x_j & \geq & 0 & \text{for } j=1,2,\ldots,n \end{array}$$

$$\ \ \, \downarrow \text{Formulating an Auxiliary Linear Program}$$

maximise
$$\sum_{j=1}^{n} c_j x_j$$
 subject to $\sum_{j=1}^{n} a_{ij} x_j \leq b_i$ for $i=1,2,\ldots,m$, $x_j \geq 0$ for $j=1,2,\ldots,n$ Formulating an Auxiliary Linear Program maximise $-x_0$ subject to $\sum_{j=1}^{n} a_{ij} x_j - x_0 \leq b_i$ for $i=1,2,\ldots,m$, $x_i \geq 0$ for $j=0,1,\ldots,n$

maximise subject to $\sum_{j=1}^{n} c_{j}x_{j}$ subject to $\sum_{j=1}^{n} a_{ij}x_{j} \leq b_{i} \quad \text{for } i=1,2,\ldots,m,$ $x_{j} \geq 0 \quad \text{for } j=1,2,\ldots,n$ Formulating an Auxiliary Linear Program maximise subject to $\sum_{j=1}^{n} a_{ij}x_{j} - x_{0} \leq b_{i} \quad \text{for } i=1,2,\ldots,m,$ $x_{i} > 0 \quad \text{for } j=0,1,\ldots,n$

Lemma 29.11

Let L_{aux} be the auxiliary LP of a linear program L in standard form. Then L is feasible if and only if the optimal objective value of L_{aux} is 0.

maximise subject to
$$\sum_{j=1}^{n} c_{j}x_{j}$$
 subject to
$$\sum_{j=1}^{n} a_{ij}x_{j} \leq b_{i} \quad \text{for } i=1,2,\ldots,m,$$

$$x_{j} \geq 0 \quad \text{for } j=1,2,\ldots,n$$
 Formulating an Auxiliary Linear Program maximise subject to
$$\sum_{j=1}^{n} a_{ij}x_{j} - x_{0} \leq b_{i} \quad \text{for } i=1,2,\ldots,m,$$

$$x_{i} > 0 \quad \text{for } j=0,1,\ldots,n$$

Lemma 29.11

Let L_{aux} be the auxiliary LP of a linear program L in standard form. Then L is feasible if and only if the optimal objective value of L_{aux} is 0.

maximise subject to $\sum_{j=1}^{n} c_{j}x_{j}$ subject to $\sum_{j=1}^{n} a_{ij}x_{j} \leq b_{i} \quad \text{for } i=1,2,\ldots,m,$ $x_{j} \geq 0 \quad \text{for } j=1,2,\ldots,n$ Formulating an Auxiliary Linear Program maximise subject to $\sum_{j=1}^{n} a_{ij}x_{j} - x_{0} \leq b_{i} \quad \text{for } i=1,2,\ldots,m,$ $x_{i} > 0 \quad \text{for } j=0,1,\ldots,n$

- Lemma 29.11

Let L_{aux} be the auxiliary LP of a linear program L in standard form. Then L is feasible if and only if the optimal objective value of L_{aux} is 0.

Proof.

• " \Rightarrow ": Suppose *L* has a feasible solution $\overline{x} = (\overline{x}_1, \overline{x}_2, \dots, \overline{x}_n)$

maximise
$$\sum_{j=1}^{n} c_j x_j$$
 subject to $\sum_{i=1}^{n}$

$$\begin{array}{cccc} \sum_{j=1}^n a_{ij} x_j & \leq & b_i & \text{for } i=1,2,\ldots,m, \\ x_j & \geq & 0 & \text{for } j=1,2,\ldots,n \\ & & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ \end{array}$$

maximise $-x_0$ subject to

$$\begin{array}{cccc} \sum_{j=1}^{n} a_{ij} x_{j} - x_{0} & \leq & b_{i} & \text{for } i = 1, 2, \dots, m, \\ x_{j} & \geq & 0 & \text{for } j = 0, 1, \dots, n \end{array}$$

- Lemma 29.11

Let L_{aux} be the auxiliary LP of a linear program L in standard form. Then L is feasible if and only if the optimal objective value of L_{aux} is 0.

- " \Rightarrow ": Suppose *L* has a feasible solution $\overline{x} = (\overline{x}_1, \overline{x}_2, \dots, \overline{x}_n)$
 - $\overline{x}_0 = 0$ combined with \overline{x} is a feasible solution to L_{aux} with objective value 0.

maximise
$$\sum_{j=1}^{n} c_j x_j$$
 subject to

$$\sum_{j=1}^{n} a_{ij} x_{j} \leq b_{i} \quad \text{for } i = 1, 2, \dots, m,$$

$$x_{j} \geq 0 \quad \text{for } j = 1, 2, \dots, n$$

$$\downarrow \text{Formulating an Auxiliary Linear Program}$$

maximise $-x_0$ subject to

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 Formulating an Auxiliary Linear Program

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 Since x̄₀ ≥ 0 and the objective is to maximise -x₀, this is optimal for L_{aux}
- " \Leftarrow ": Suppose that the optimal objective value of L_{aux} is 0

maximise
$$\sum_{j=1}^{n} c_{j}$$
 subject to

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

$$\sum_{j=1}^{n} a_{ij} x_{j} \leq b_{i} \quad \text{for } i = 1, 2, \dots, m,$$

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$$\downarrow \text{Formulating an Auxiliary Linear Program}$$

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 - Then $\overline{x}_0 = 0$, and the remaining solution values $(\overline{x}_1, \overline{x}_2, \dots, \overline{x}_n)$ satisfy L.

maximise subject to

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

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 Formulating an Auxiliary Linear Program

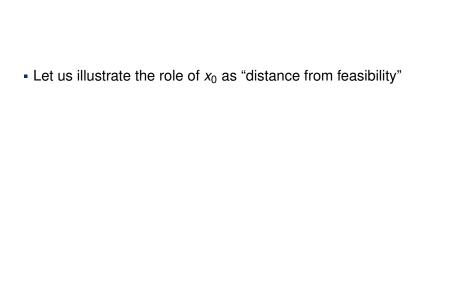
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Lemma 29.11

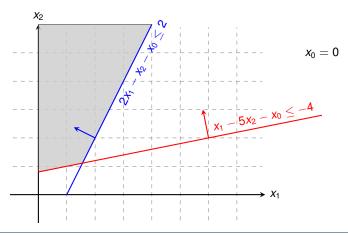
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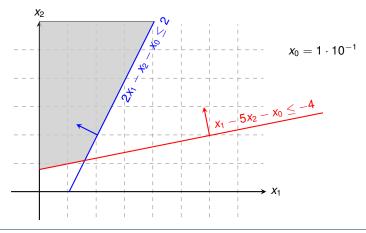


- Let us illustrate the role of x₀ as "distance from feasibility"
- We will also see that increasing x_0 enlarges the feasible region.

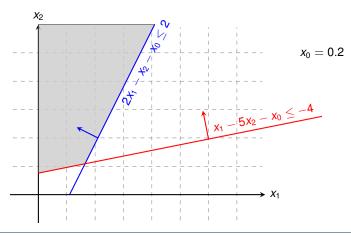
$$-x_0$$



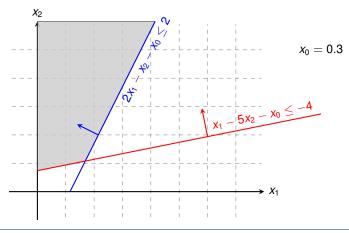
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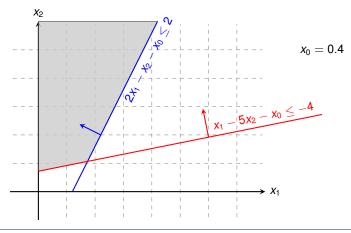
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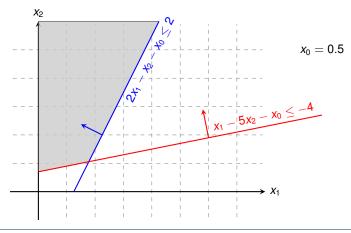
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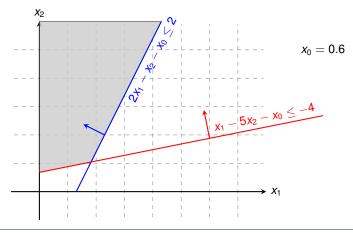
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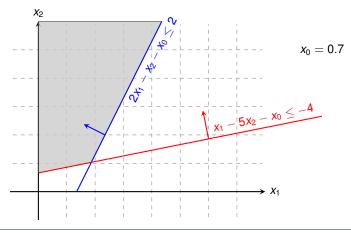
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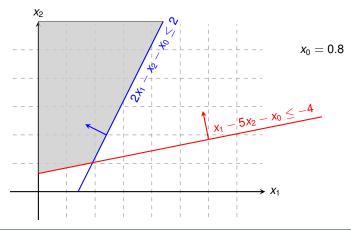
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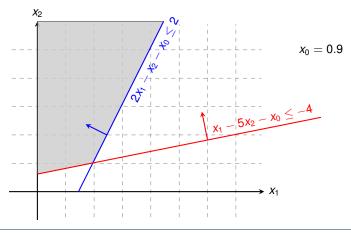
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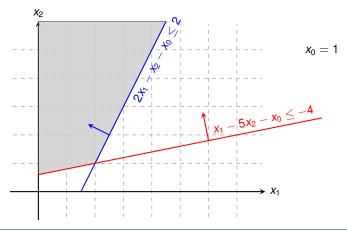
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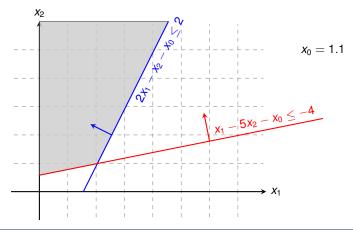
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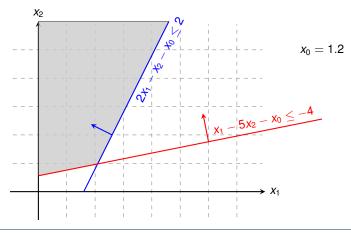
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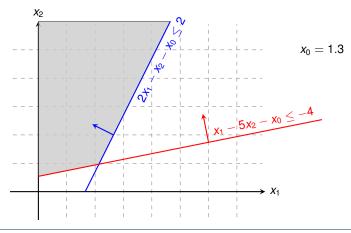
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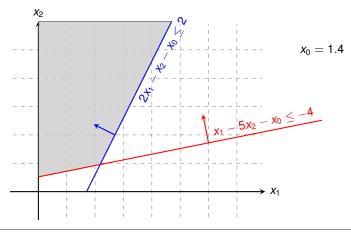
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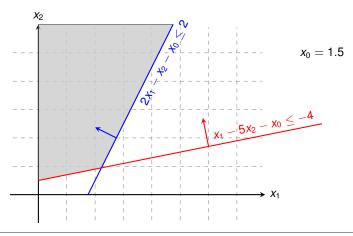
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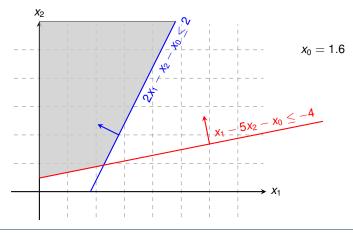
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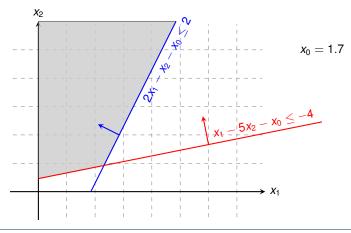
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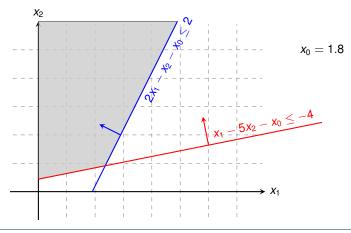
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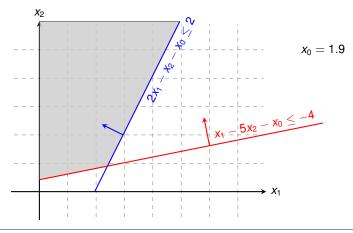
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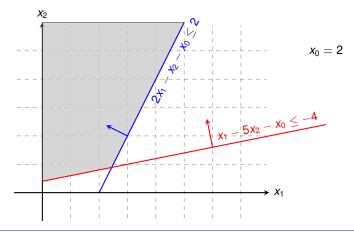
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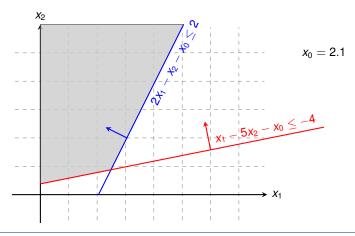
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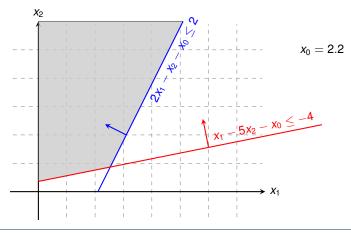
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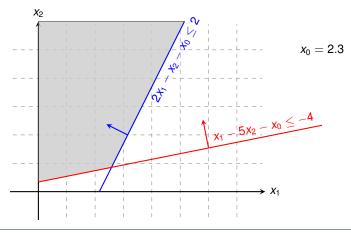
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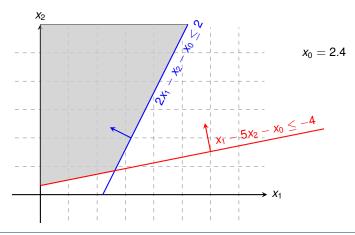
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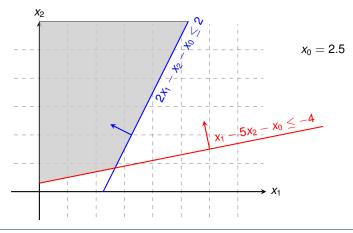
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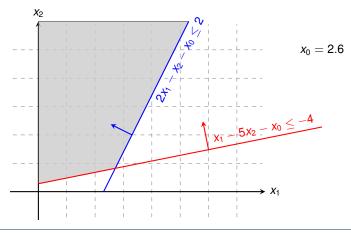
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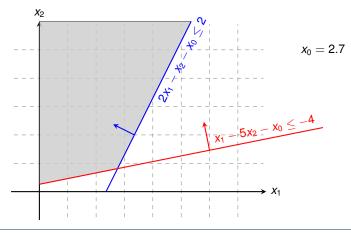
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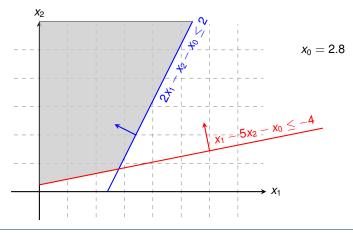
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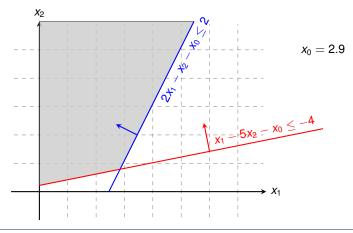
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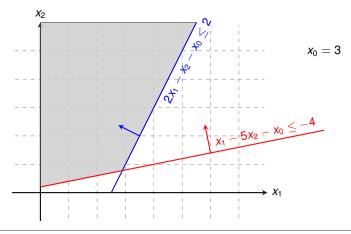
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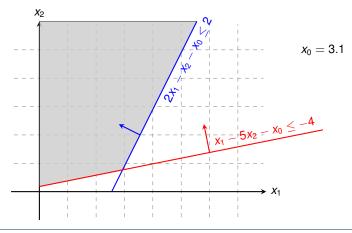
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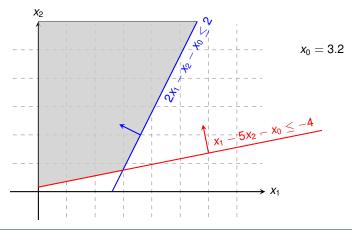
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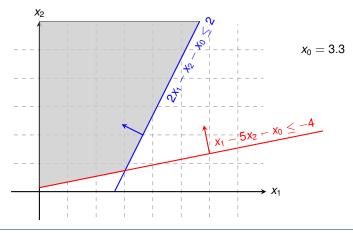
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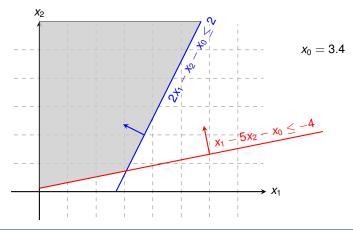
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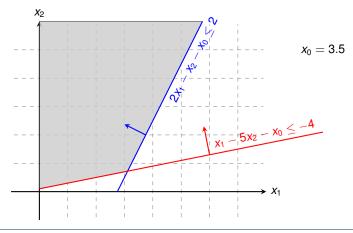
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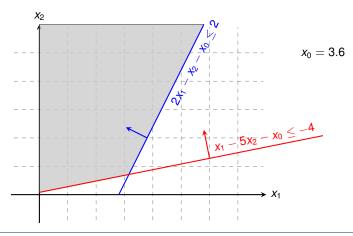
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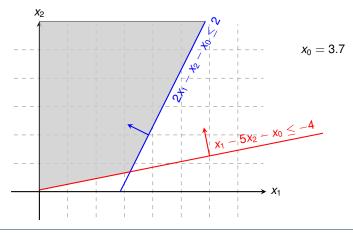
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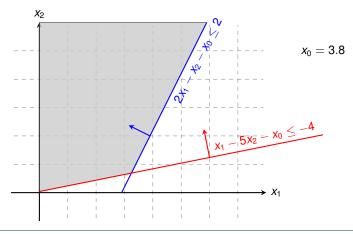
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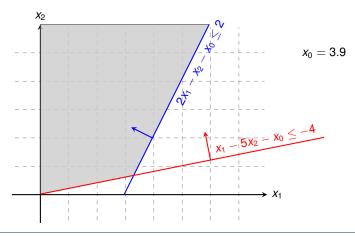
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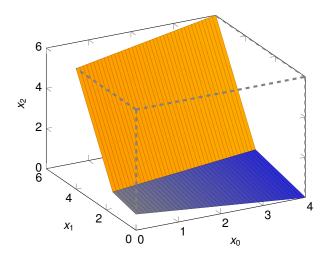
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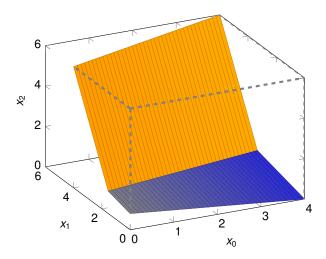


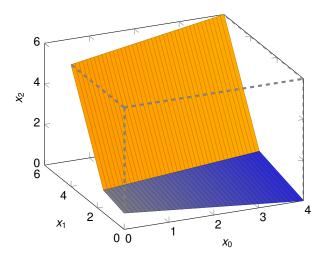
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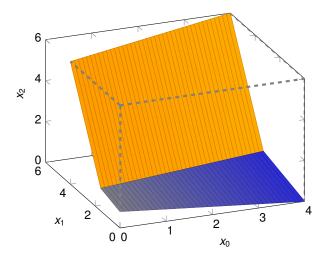


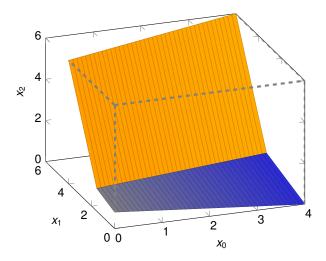
Now the Feasible Region of the Auxiliary LP in 3D

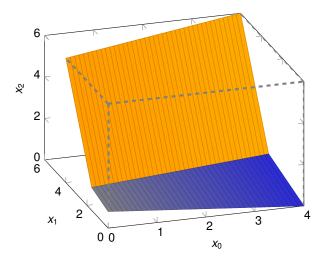


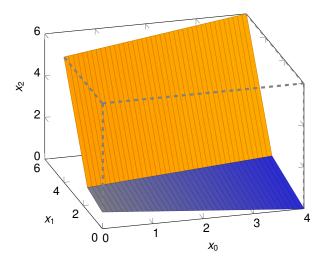


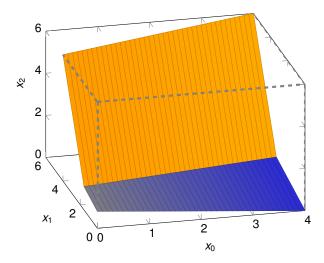


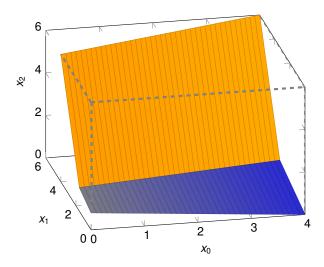


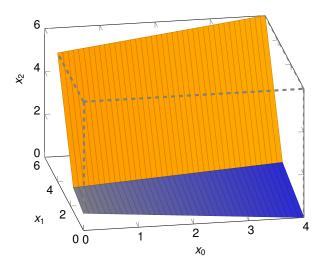


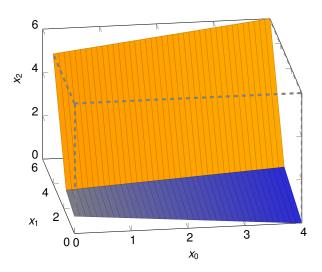


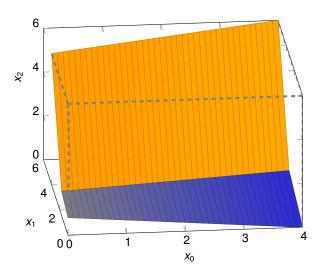


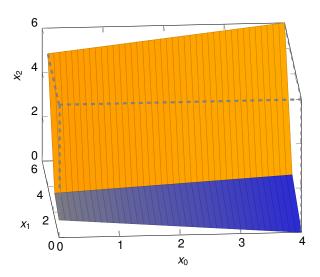


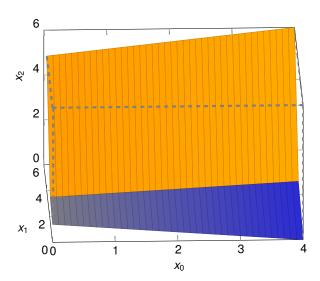


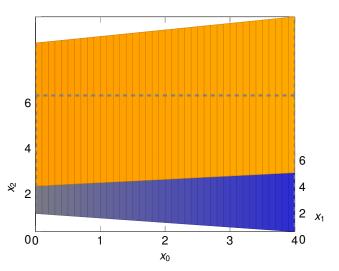


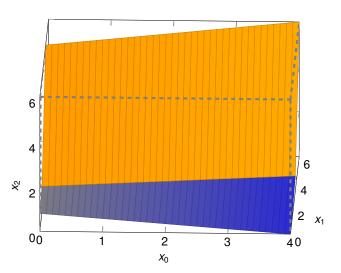


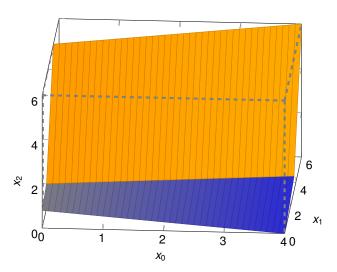


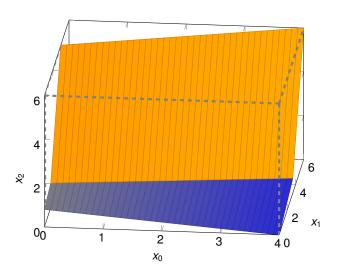


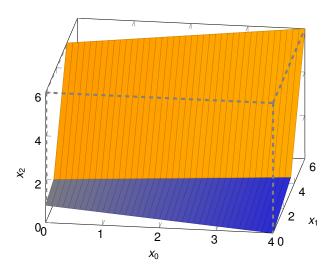


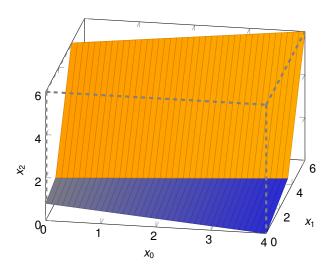


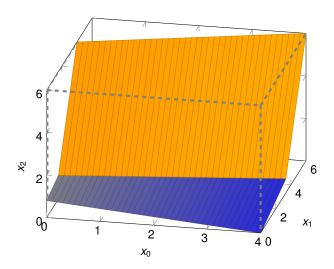


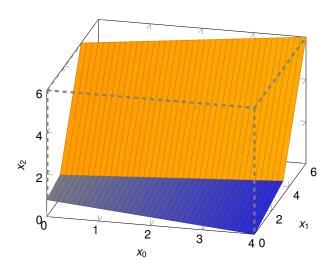


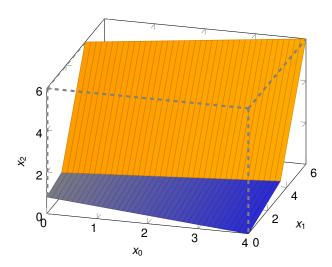


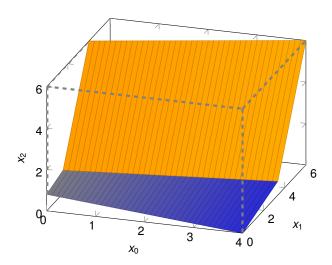


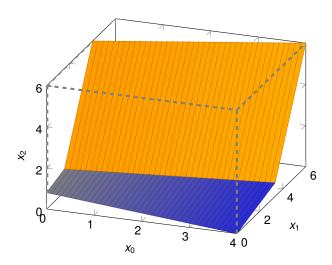


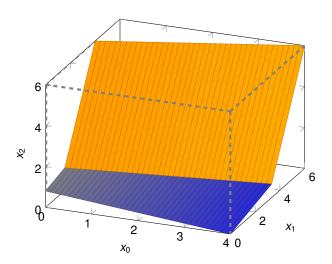


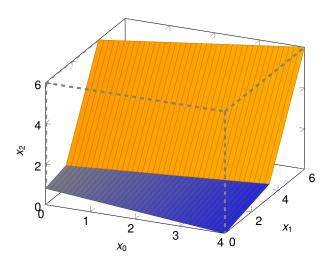


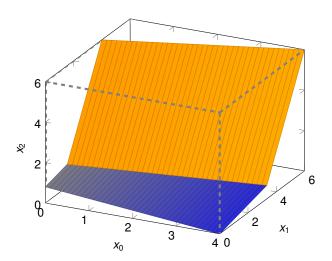


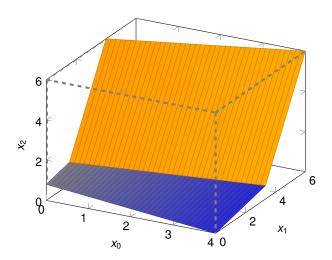


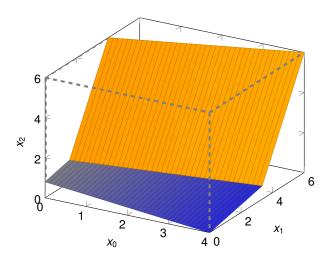


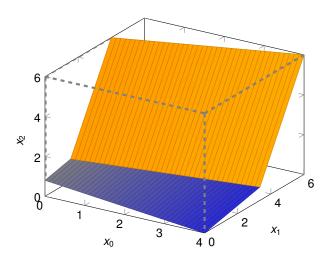


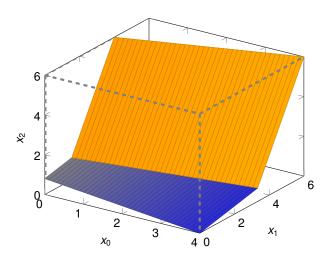


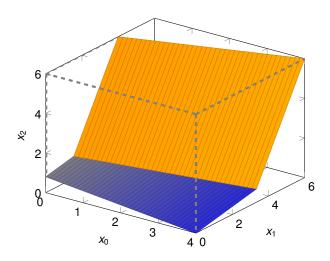


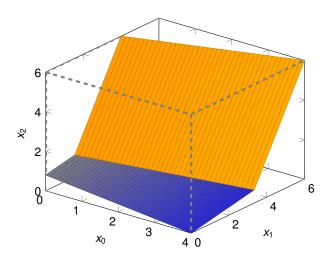


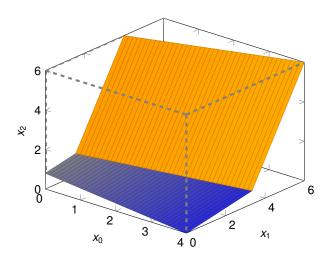


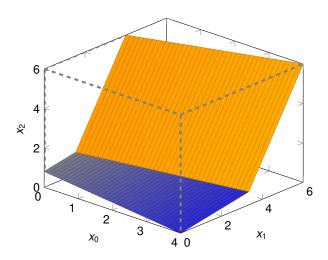


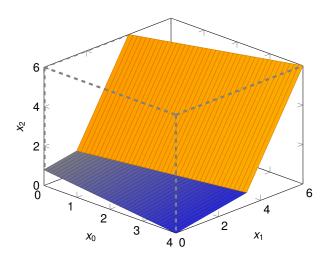


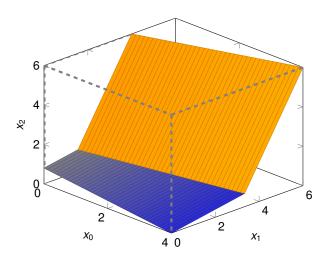


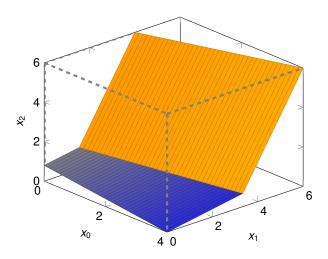


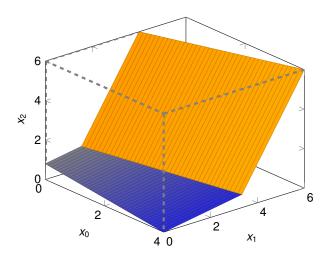


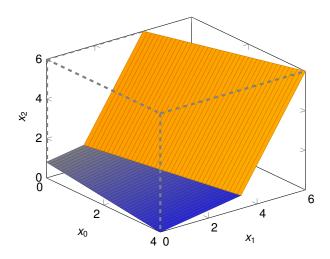


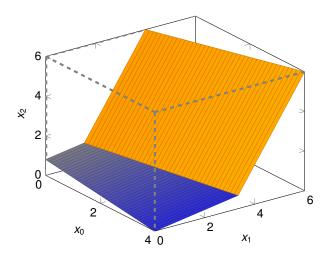


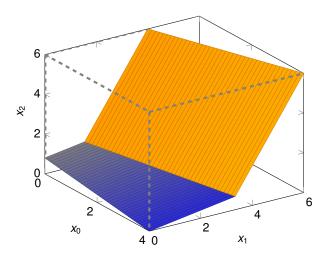


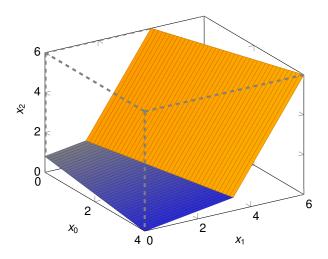


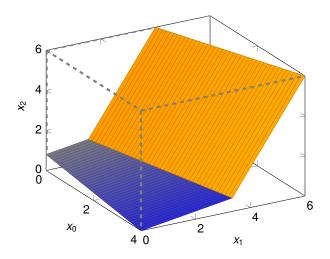


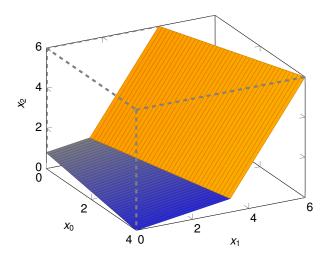


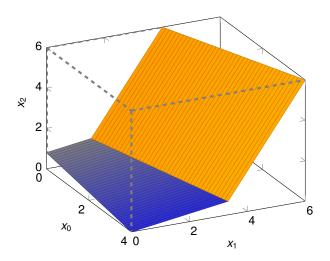


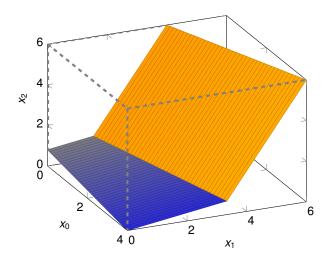


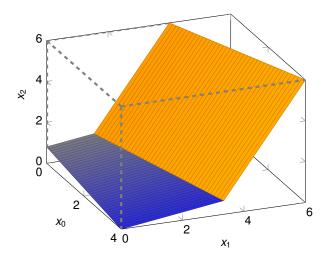


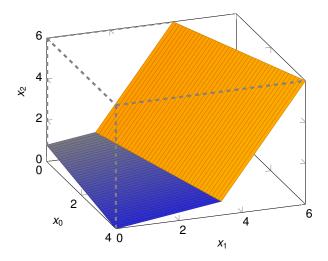


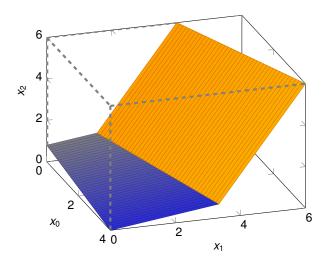


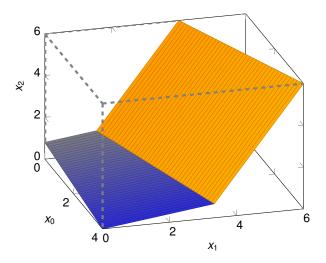


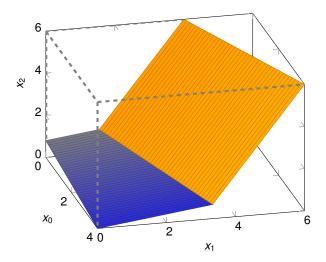


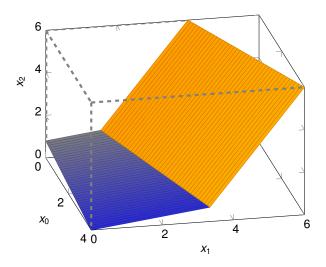


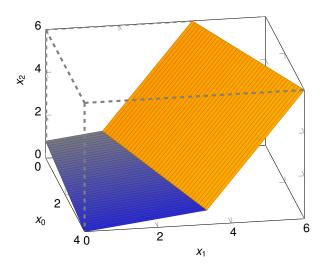


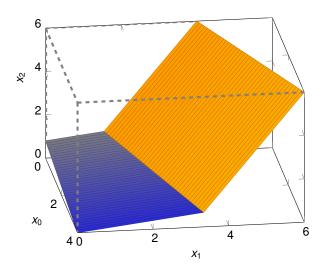


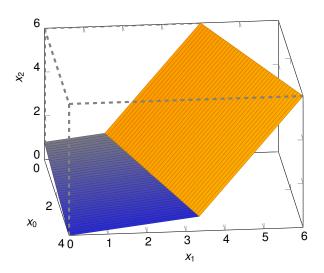


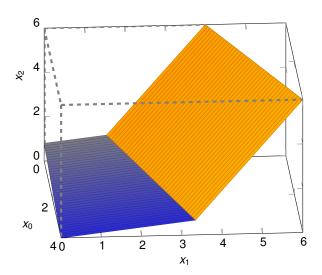


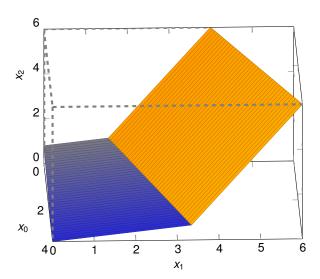


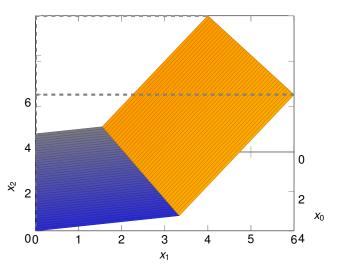


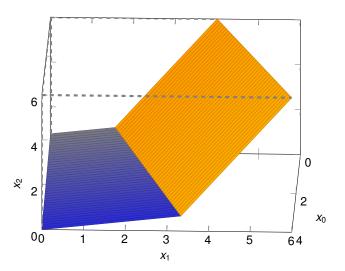


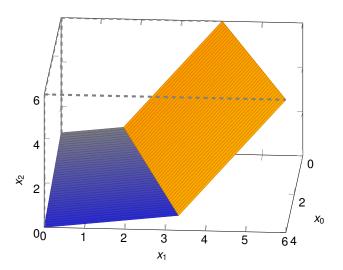


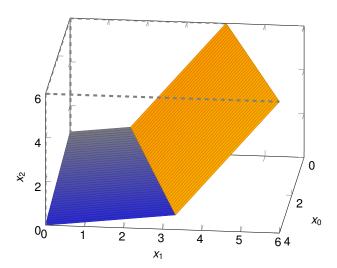


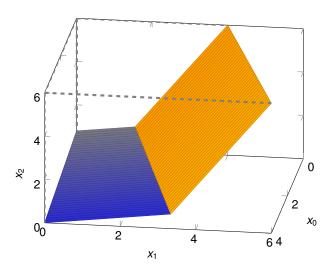


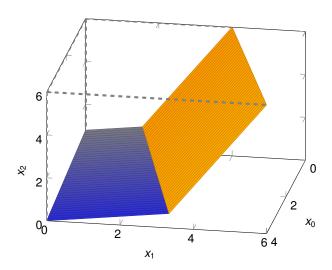


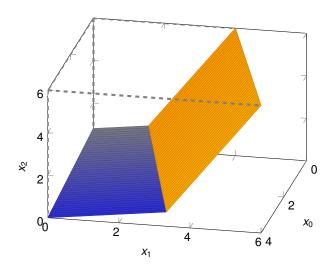


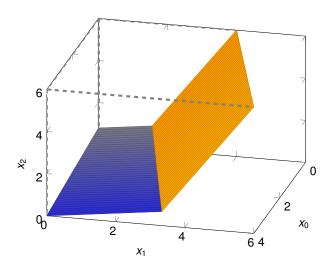


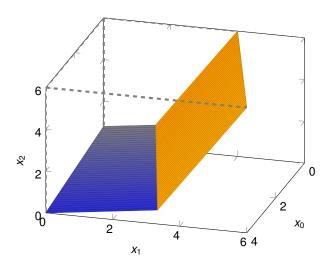


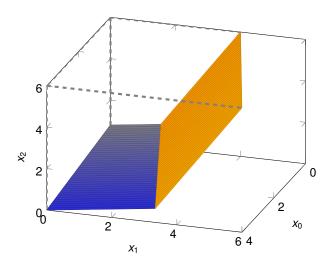


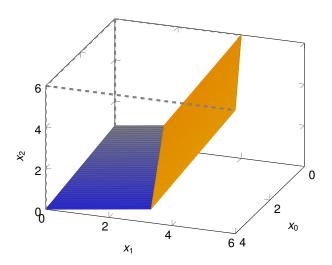


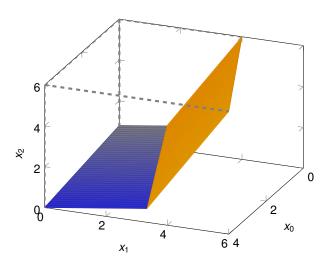


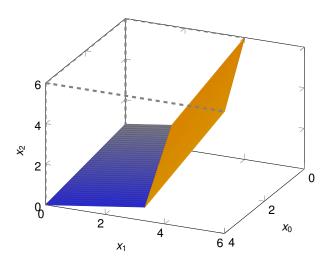


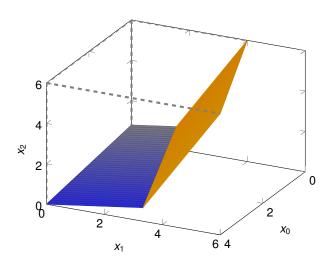


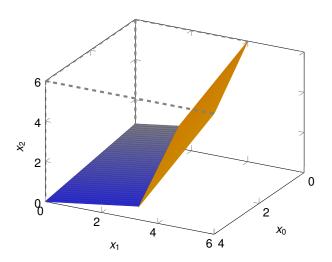


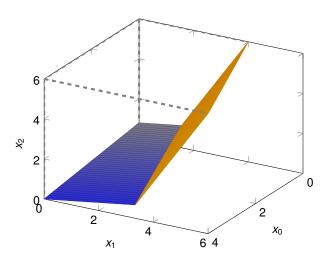


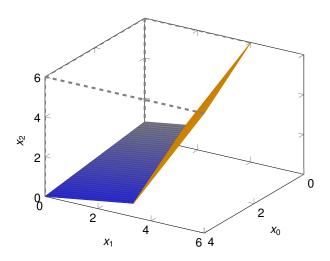


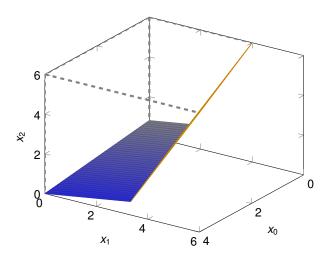


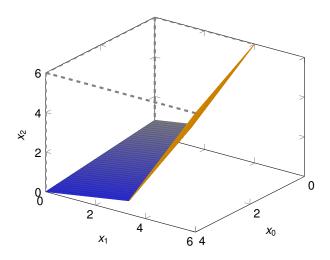


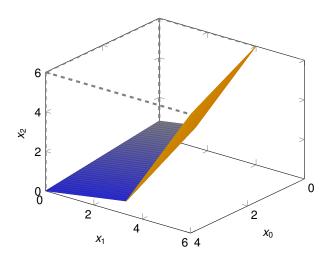


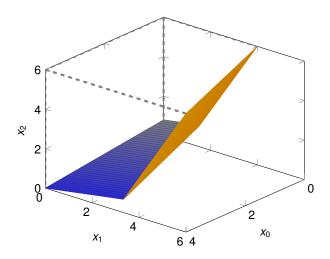


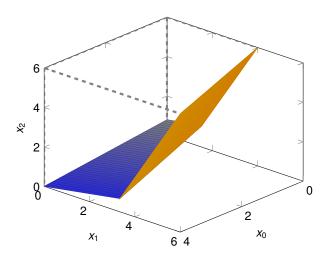


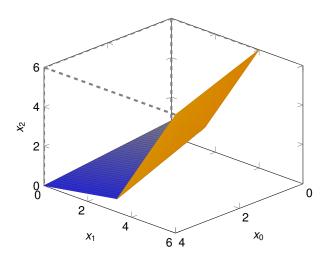


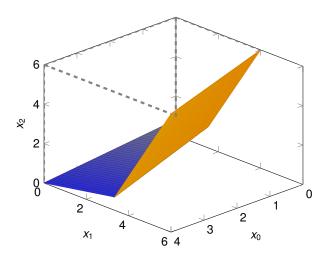


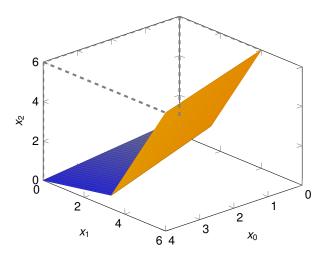


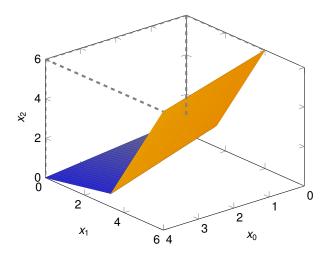


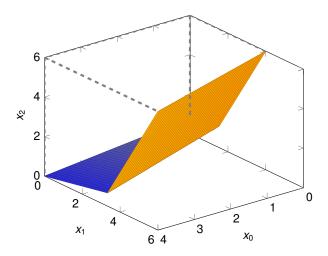


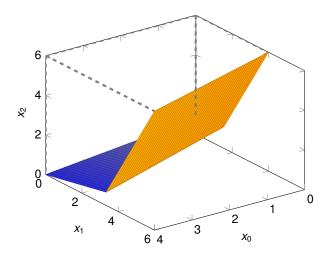


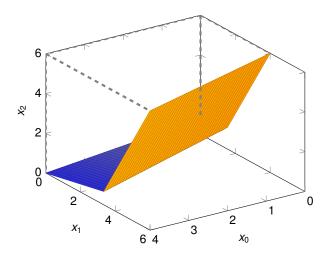


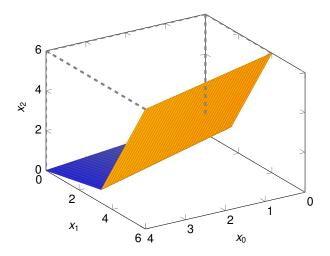


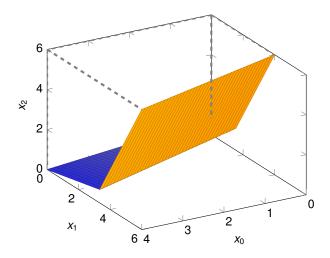


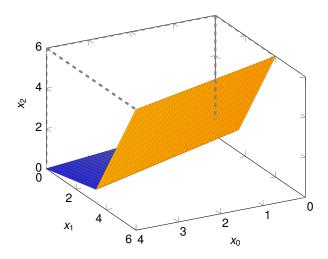


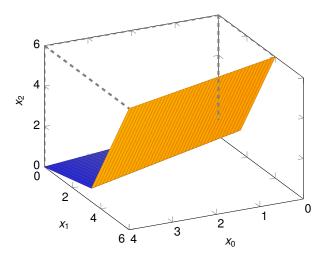


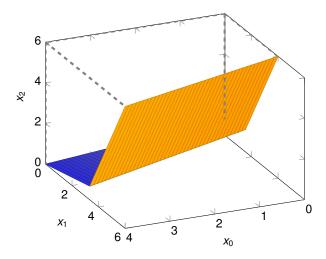


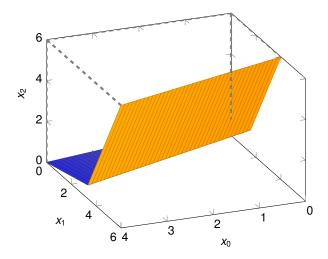


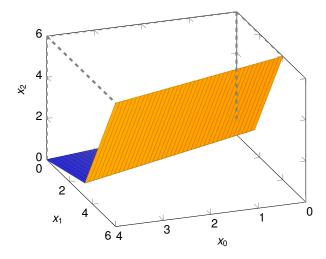


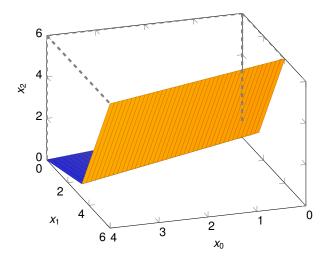


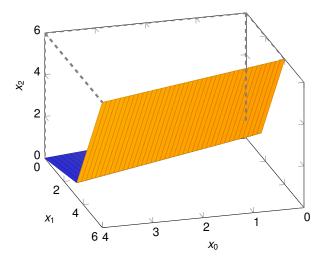


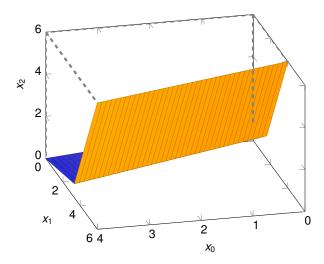


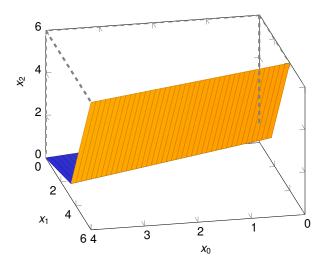


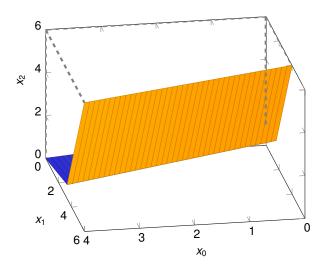


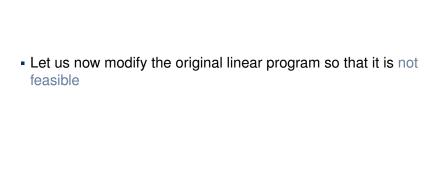










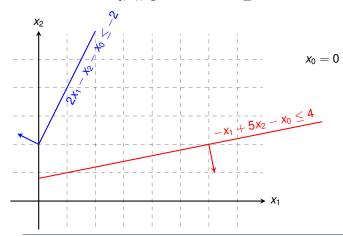


- Let us now modify the original linear program so that it is not feasible
- \Rightarrow Hence the auxiliary linear program has only a solution for a sufficiently large $x_0 > 0$!

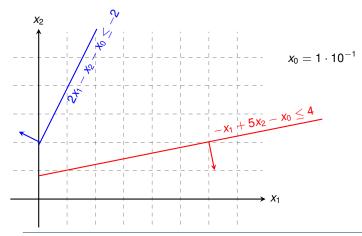
$$-x_0$$

$$x_0 \leq x_0 \leq x_0$$

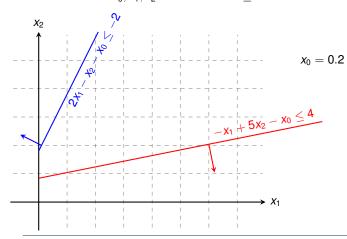
$$x_0 \leq$$



$$-x_0$$

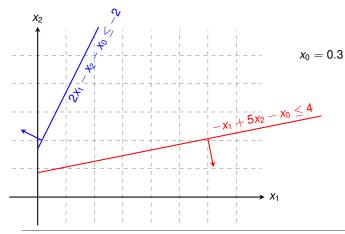


$$-x_0$$

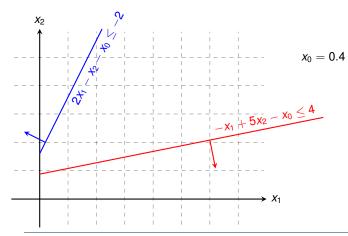


$$-x_0$$

$$x_0 \leq$$



$$-x_0$$

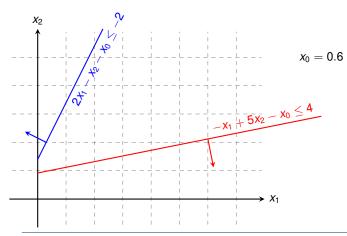


$$-x_0$$

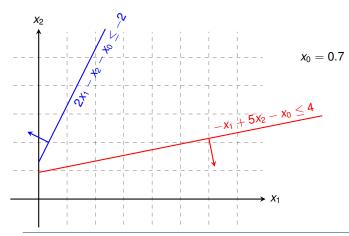
$$x_0 = 0.5$$

$$x_0 = 0.5$$

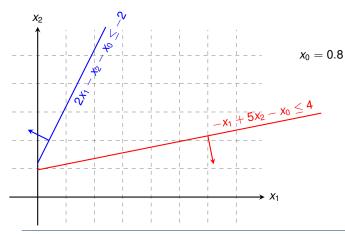
$$-x_0$$



$$-x_0$$



$$-x_0$$

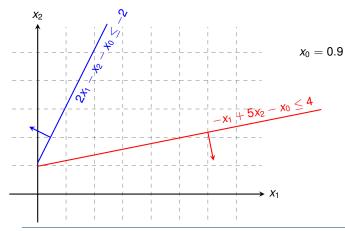


$$-x_0$$

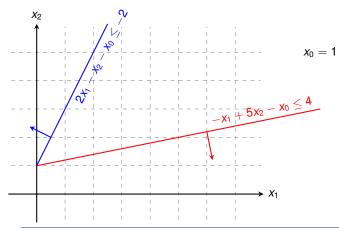
$$x_0 \leq x_0 \leq x_0$$

$$\leq 4$$

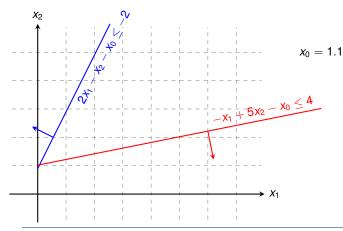
 ≥ 0



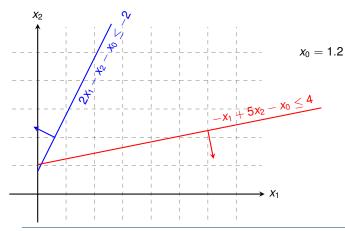
$$-x_0$$



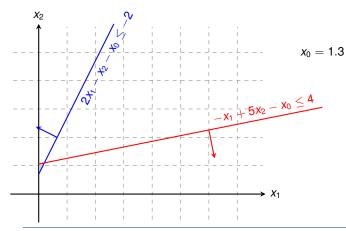
$$-x_0$$



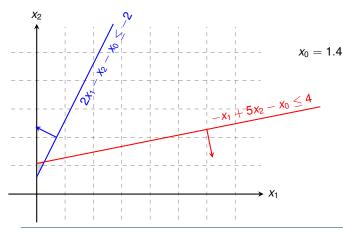
$$-x_0$$



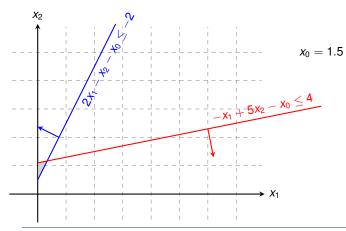
$$-x_0$$



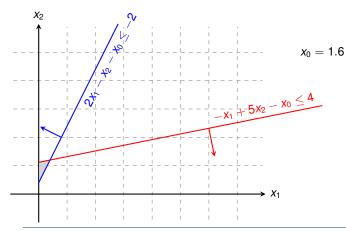
$$-x_0$$



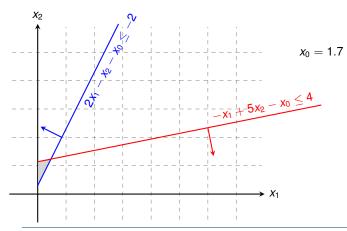
$$-x_0$$



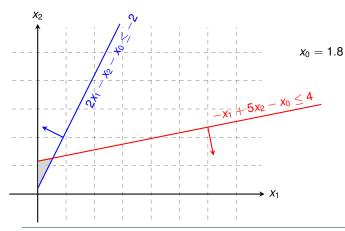
$$-x_0$$



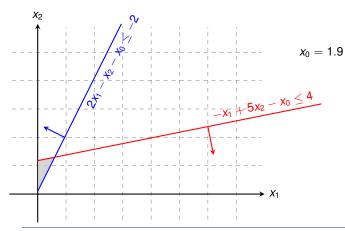
$$-x_0$$



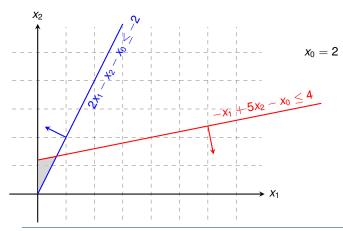
$$-x_0$$



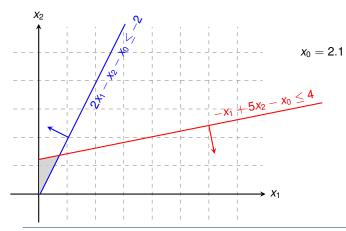
$$-x_0$$



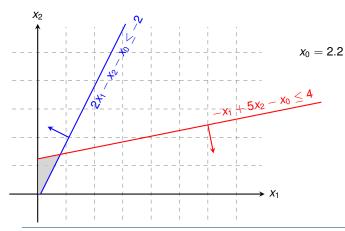
$$-x_0$$



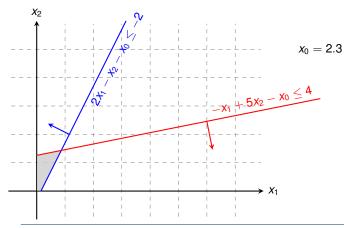
$$-x_0$$



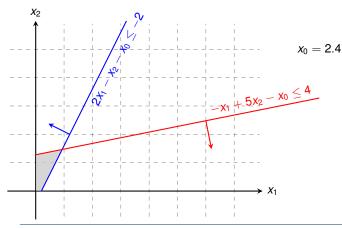
$$-x_0$$



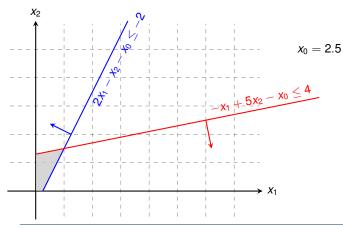
$$-x_0$$



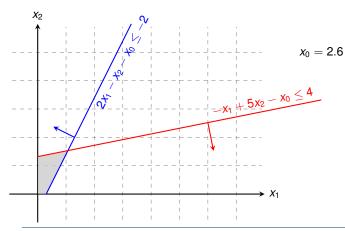
$$-x_0$$



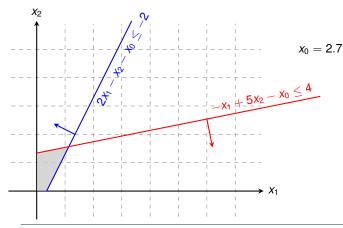
$$-x_0$$



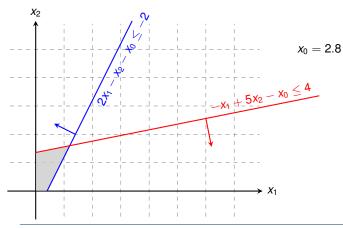
$$-x_0$$



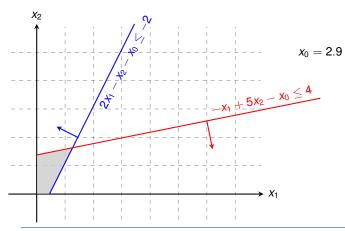
$$-x_0$$



$$-x_0$$



$$-x_0$$



maximise subject to

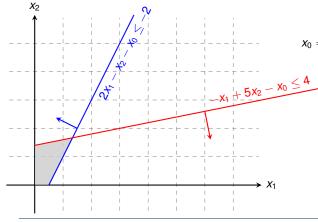
$$-x_0$$

$$\leq$$
 -2

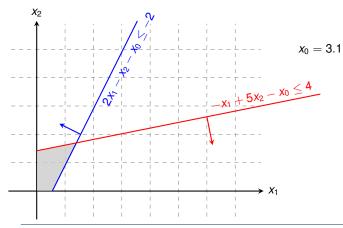
$$X_0, X_1, X_2$$

$$x_0$$

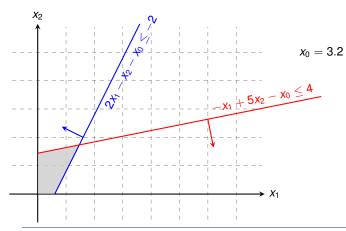
 $x_0 = 3$



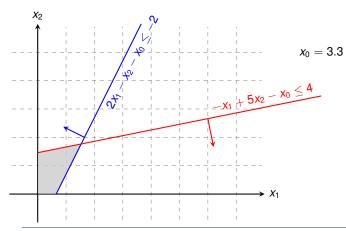
$$-x_0$$



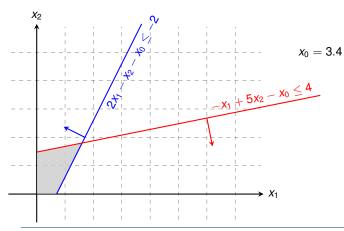
$$-x_0$$



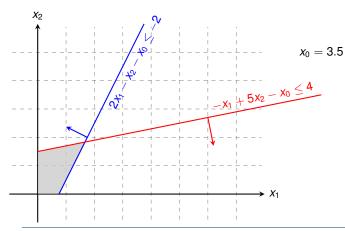
$$-x_0$$



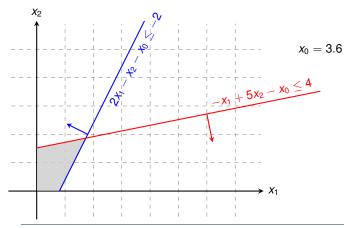
$$-x_0$$



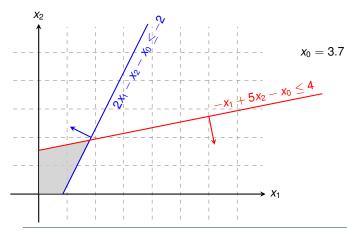
$$-x_0$$



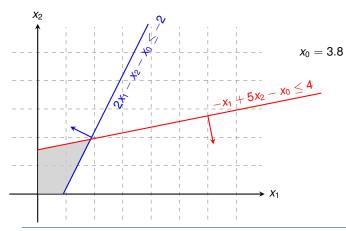
$$-x_0$$



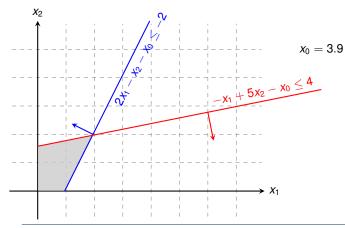
$$-x_0$$



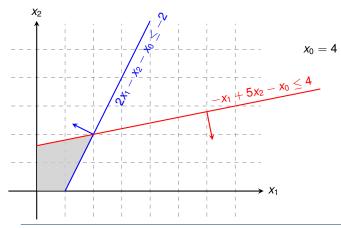
$$-x_0$$



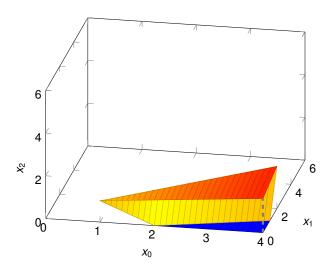
$$-x_0$$

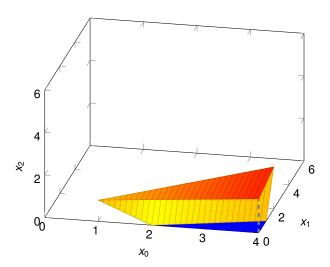


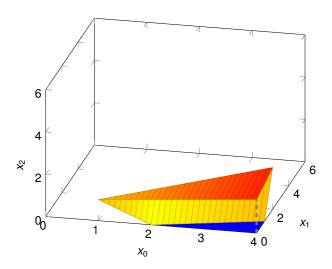
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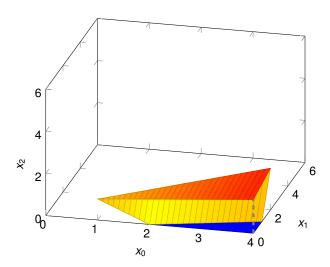


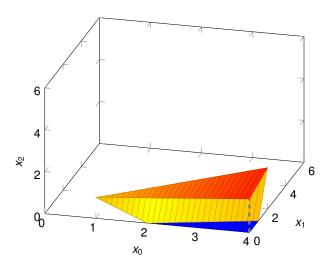
Now the Feasible Region of the Auxiliary LP in 3D

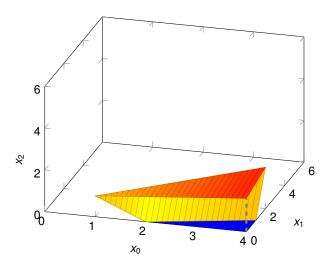


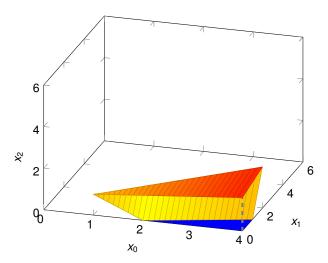


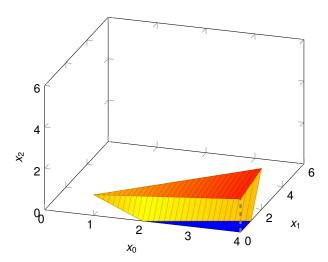


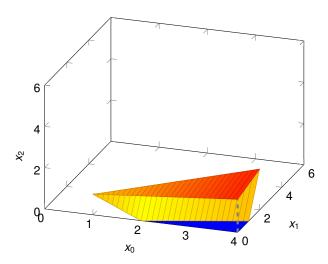


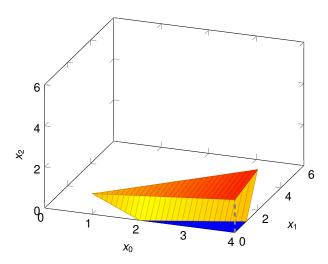


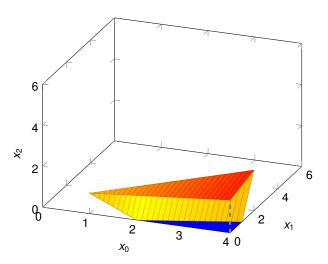


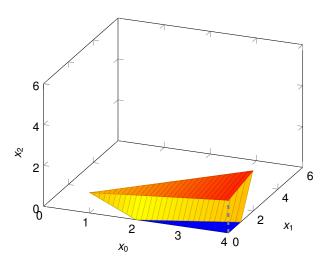


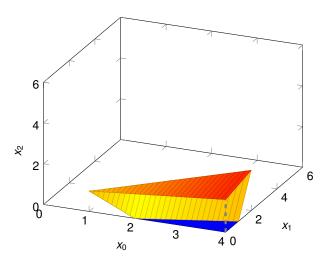


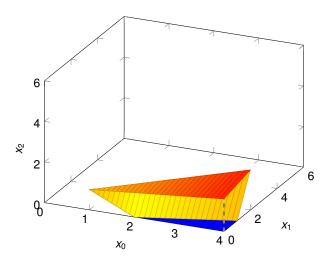


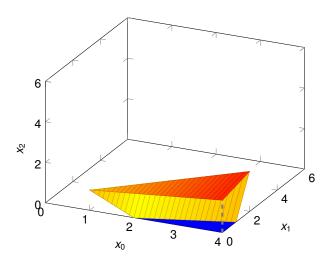


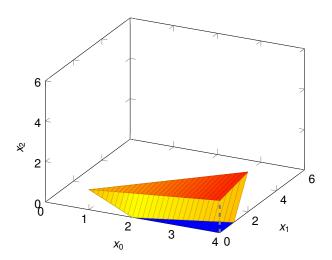


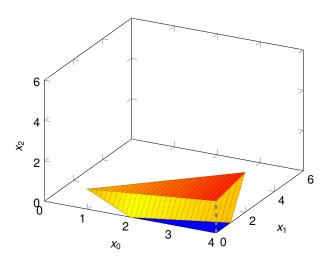


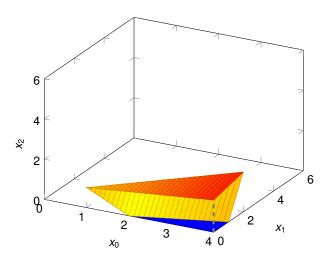


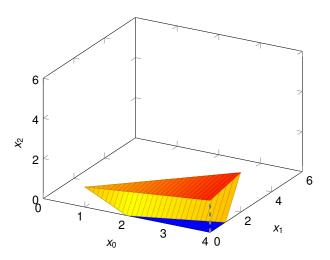


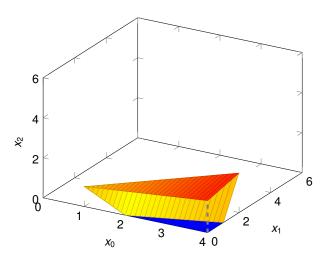


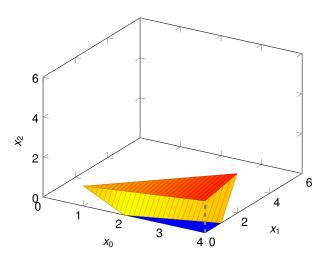


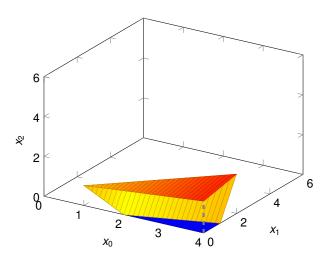


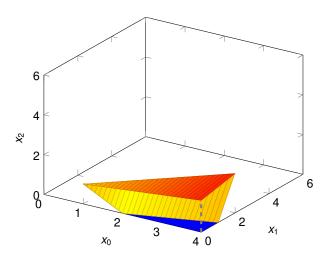


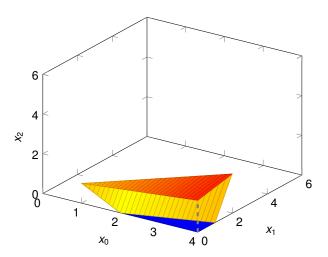


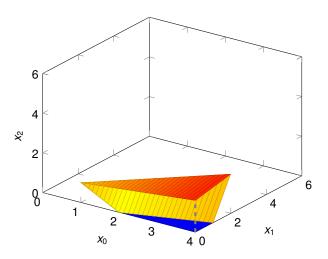


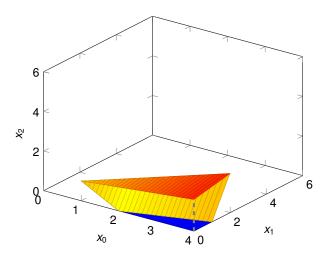


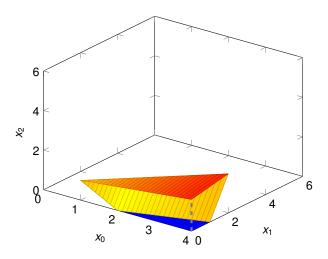


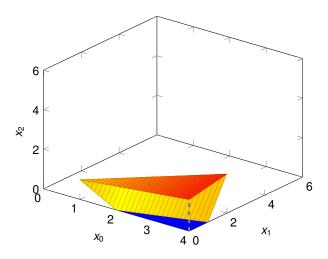


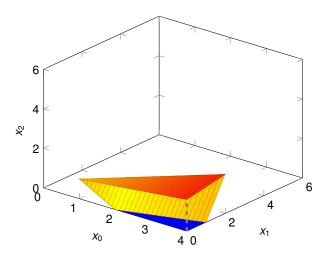


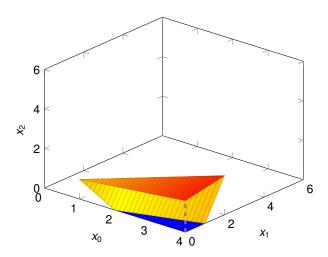


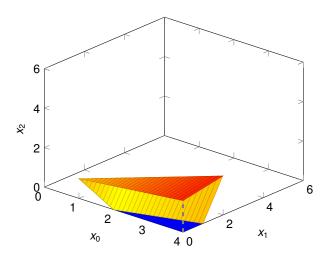


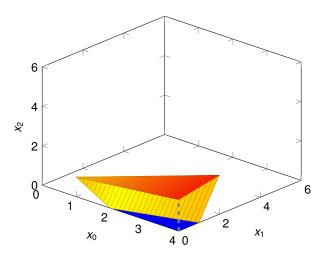


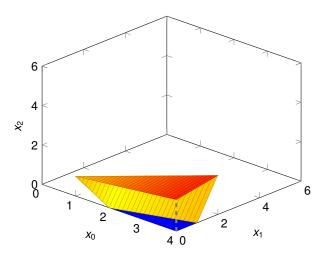


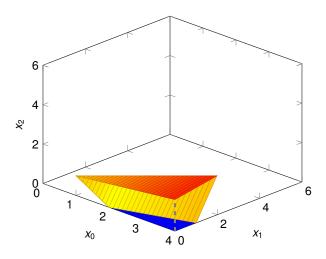


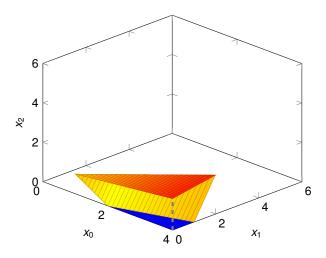


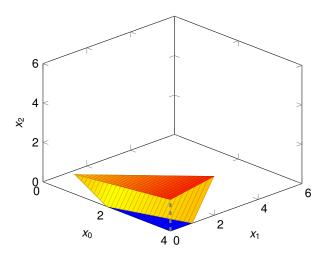


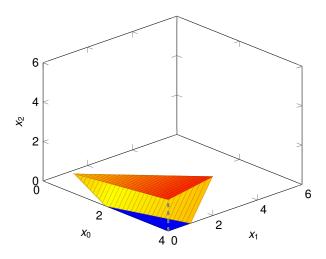


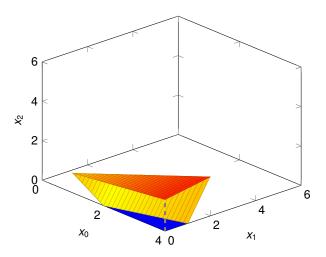


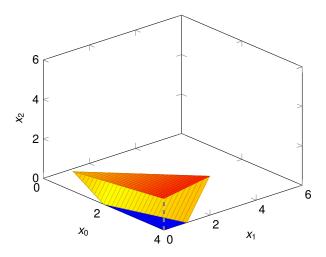


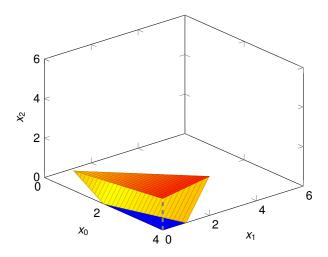


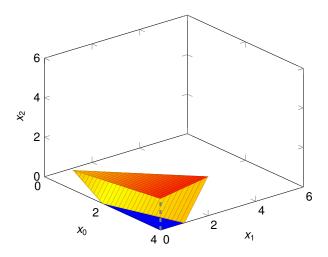


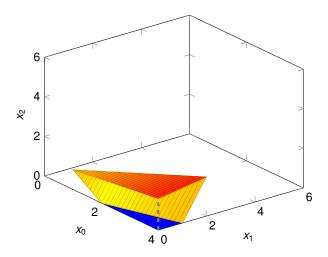


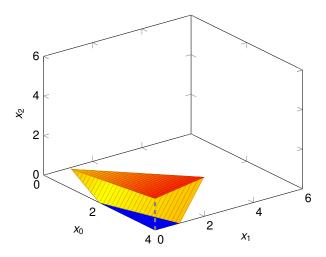


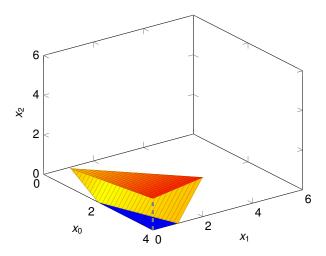


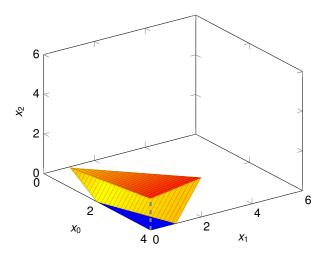


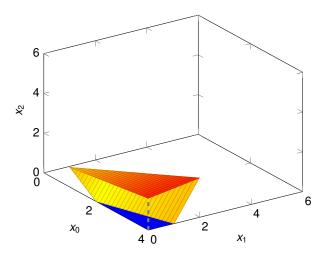


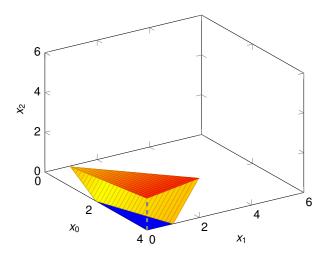


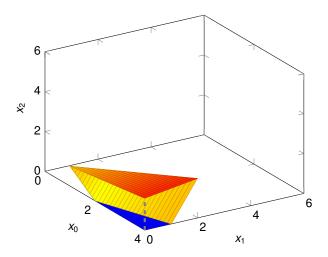


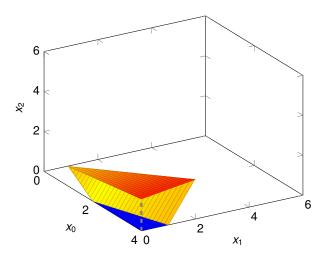


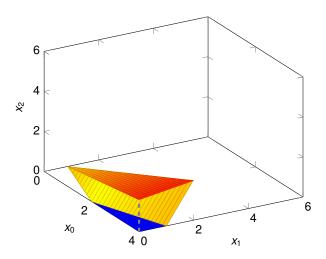


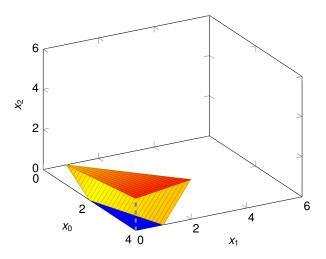


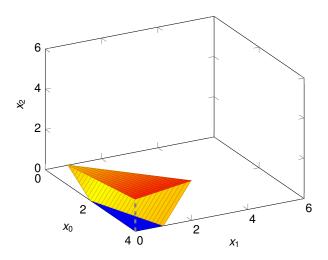


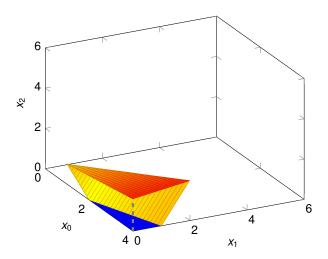


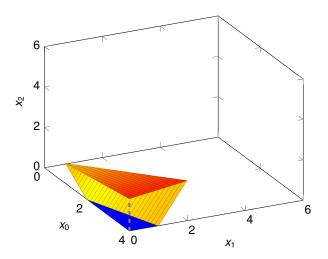


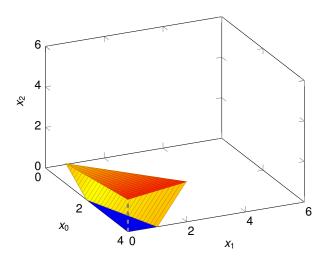


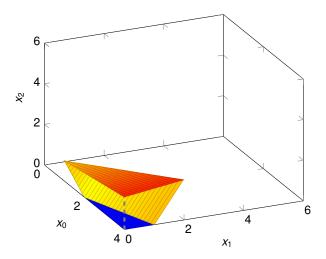


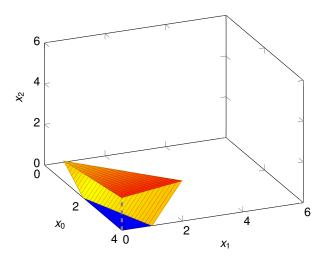


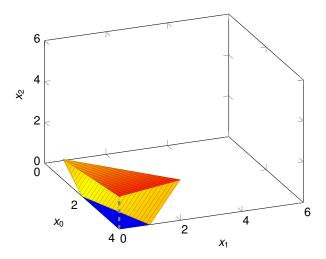


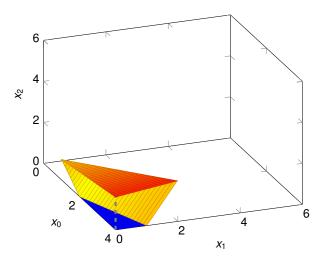


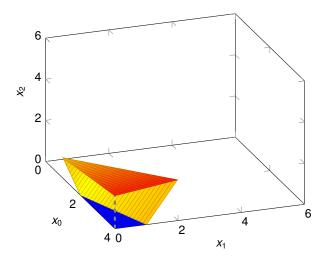


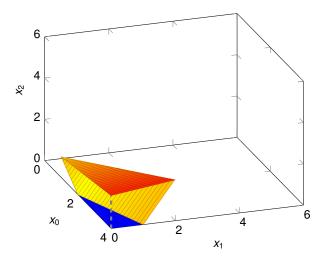


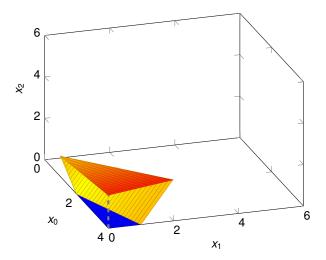


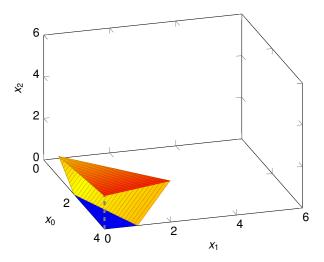


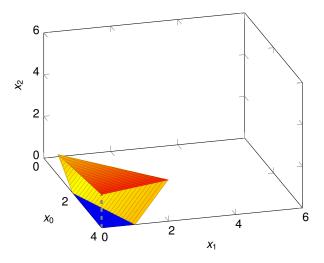


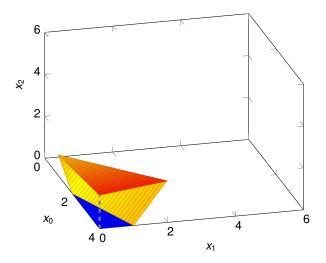


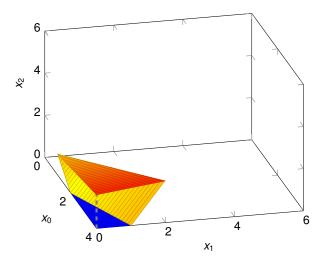


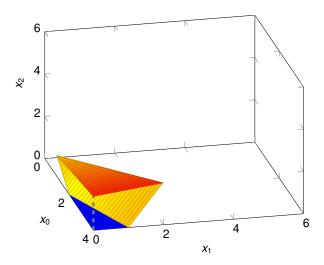


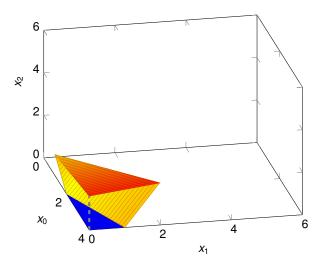


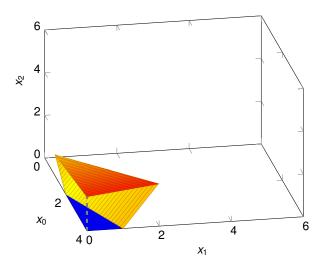


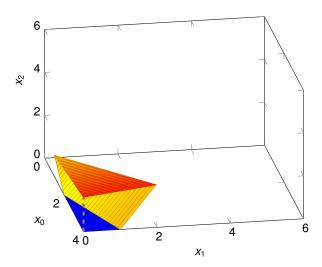


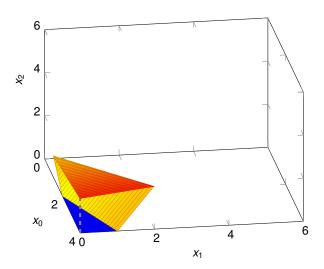


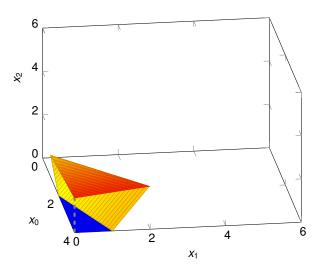


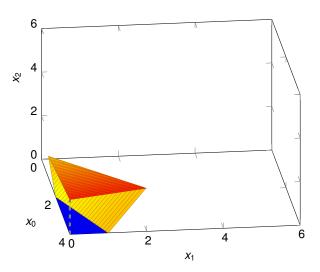


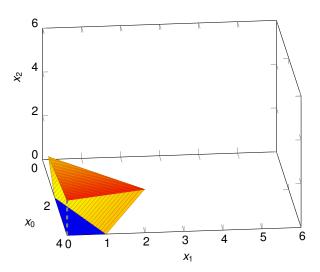


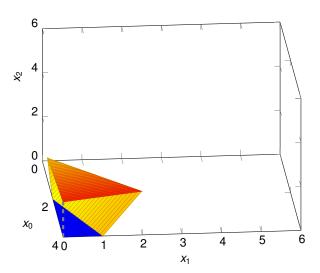


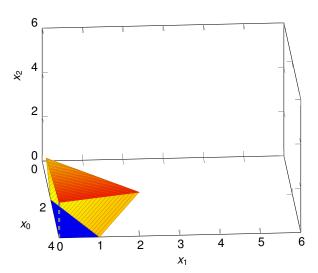


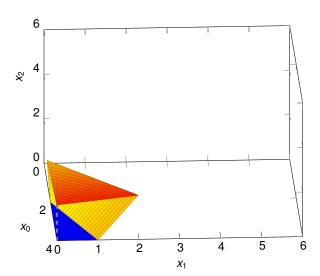


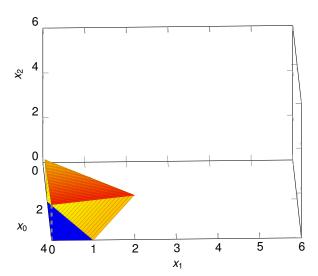


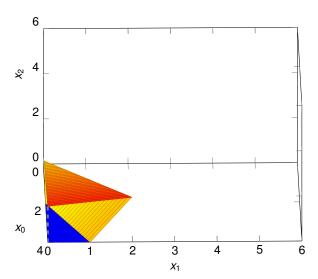












```
INITIALIZE-SIMPLEX (A, b, c)
     let k be the index of the minimum b_i
                                   // is the initial basic solution feasible?
 2 if b_{\nu} > 0
          return (\{1, 2, ..., n\}, \{n + 1, n + 2, ..., n + m\}, A, b, c, 0)
     form L_{\text{any}} by adding -x_0 to the left-hand side of each constraint
          and setting the objective function to -x_0
 5 let (N, B, A, b, c, \nu) be the resulting slack form for L_{aux}
    l = n + k
    //L_{\text{aux}} has n+1 nonbasic variables and m basic variables.
 8 (N, B, A, b, c, v) = PIVOT(N, B, A, b, c, v, l, 0)
 9 // The basic solution is now feasible for L_{\text{aux}}.
10 iterate the while loop of lines 3-12 of SIMPLEX until an optimal solution
          to L_{\text{any}} is found
     if the optimal solution to L_{\text{aux}} sets \bar{x}_0 to 0
12
          if \bar{x}_0 is basic
               perform one (degenerate) pivot to make it nonbasic
13
14
          from the final slack form of L_{\text{aux}}, remove x_0 from the constraints and
               restore the original objective function of L, but replace each basic
               variable in this objective function by the right-hand side of its
               associated constraint
15
          return the modified final slack form
     else return "infeasible"
```

```
Test solution with N = \{1, 2, ..., n\}, B = \{n + 1, n + 1\}
INITIALIZE-SIMPLEX (A, b, c)
                                                    \{2,\ldots,n+m\},\ \overline{x}_i=b_i\ \text{for}\ i\in B,\ \overline{x}_i=0\ \text{otherwise}.
     let k be the index of the minimum b_k
                                   // is the initial basic solution feasible?
 2 if b_{\nu} > 0
          return (\{1, 2, ..., n\}, \{n + 1, n + 2, ..., n + m\}, A, b, c, 0)
     form L_{\text{any}} by adding -x_0 to the left-hand side of each constraint
          and setting the objective function to -x_0
     let (N, B, A, b, c, v) be the resulting slack form for L_{aux}
    l = n + k
     //L_{\text{aux}} has n+1 nonbasic variables and m basic variables.
 8 (N, B, A, b, c, v) = PIVOT(N, B, A, b, c, v, l, 0)
    // The basic solution is now feasible for L_{\text{aux}}.
10 iterate the while loop of lines 3-12 of SIMPLEX until an optimal solution
          to L_{\text{any}} is found
     if the optimal solution to L_{\text{aux}} sets \bar{x}_0 to 0
12
          if \bar{x}_0 is basic
               perform one (degenerate) pivot to make it nonbasic
13
14
          from the final slack form of L_{\text{aux}}, remove x_0 from the constraints and
               restore the original objective function of L, but replace each basic
               variable in this objective function by the right-hand side of its
               associated constraint
15
          return the modified final slack form
```

else return "infeasible"

INITIALIZE-SIMPLEX (A, b, c)

if $b_{\nu} > 0$

Test solution with $N = \{1, 2, \dots, n\}$, $B = \{n + 1, n + 1\}$ $2, \ldots, n+m$, $\overline{x}_i = b_i$ for $i \in B$, $\overline{x}_i = 0$ otherwise. let k be the index of the minimum b_k // is the initial basic solution feasible? **return** $(\{1, 2, ..., n\}, \{n + 1, n + 2, ..., n + m\}, A, b, c, 0)$ form L_{any} by adding $-x_0$ to the left-hand side of each constraint and setting the objective function to $-x_0$ ℓ will be the leaving variable so let (N, B, A, b, c, v) be the resulting slack form for L_{aux} that x_{ℓ} has the most negative value.

```
l = n + k
     //L_{\text{aux}} has n+1 nonbasic variables and m basic variables.
   (N, B, A, b, c, v) = PIVOT(N, B, A, b, c, v, l, 0)
    // The basic solution is now feasible for L_{\text{aux}}.
   iterate the while loop of lines 3-12 of SIMPLEX until an optimal solution
          to L_{\text{any}} is found
     if the optimal solution to L_{\text{aux}} sets \bar{x}_0 to 0
12
          if \bar{x}_0 is basic
               perform one (degenerate) pivot to make it nonbasic
13
14
          from the final slack form of L_{\text{aux}}, remove x_0 from the constraints and
               restore the original objective function of L, but replace each basic
               variable in this objective function by the right-hand side of its
               associated constraint
15
          return the modified final slack form
     else return "infeasible"
```

```
Test solution with N = \{1, 2, \dots, n\}, B = \{n + 1, n + 1\}
INITIALIZE-SIMPLEX (A, b, c)
                                                   2, \ldots, n+m, \overline{x}_i = b_i for i \in B, \overline{x}_i = 0 otherwise.
     let k be the index of the minimum b_k
                                  // is the initial basic solution feasible?
    if b_{\nu} > 0
          return (\{1, 2, ..., n\}, \{n + 1, n + 2, ..., n + m\}, A, b, c, 0)
     form L_{\text{any}} by adding -x_0 to the left-hand side of each constraint
          and setting the objective function to -x_0
                                                                                \ell will be the leaving variable so
     let (N, B, A, b, c, v) be the resulting slack form for L_{aux}
    l = n + k
                                                                            that x_{\ell} has the most negative value.
     //L_{\text{aux}} has n+1 nonbasic variables and m basic variables.
    (N, B, A, b, c, v) = PIVOT(N, B, A, b, c, v, l, 0)
                                                                 Pivot step with x_{\ell} leaving and x_0 entering.
    // The basic solution is now feasible for L_{\text{aux}}.
    iterate the while loop of lines 3–12 of SIMPLEX until an optimal solution
          to L_{\text{any}} is found
     if the optimal solution to L_{\text{aux}} sets \bar{x}_0 to 0
12
          if \bar{x}_0 is basic
               perform one (degenerate) pivot to make it nonbasic
13
14
          from the final slack form of L_{\text{aux}}, remove x_0 from the constraints and
               restore the original objective function of L, but replace each basic
               variable in this objective function by the right-hand side of its
               associated constraint
15
          return the modified final slack form
     else return "infeasible"
```

```
Test solution with N = \{1, 2, \dots, n\}, B = \{n + 1, n + 1\}
INITIALIZE-SIMPLEX (A, b, c)
                                                   \{2,\ldots,n+m\},\ \overline{x}_i=b_i\ \text{for}\ i\in B,\ \overline{x}_i=0\ \text{otherwise}.
     let k be the index of the minimum b_k
                                  // is the initial basic solution feasible?
    if b_{\nu} > 0
          return (\{1, 2, ..., n\}, \{n + 1, n + 2, ..., n + m\}, A, b, c, 0)
     form L_{\text{any}} by adding -x_0 to the left-hand side of each constraint
          and setting the objective function to -x_0
                                                                                \ell will be the leaving variable so
     let (N, B, A, b, c, v) be the resulting slack form for L_{aux}
    l = n + k
                                                                            that x_{\ell} has the most negative value.
     //L_{\text{aux}} has n+1 nonbasic variables and m basic variables.
    (N, B, A, b, c, v) = PIVOT(N, B, A, b, c, v, l, 0)
                                                                 Pivot step with x_{\ell} leaving and x_0 entering.
    // The basic solution is now feasible for L_{\text{aux}}.
    iterate the while loop of lines 3–12 of SIMPLEX until an optimal solution
          to L_{\text{any}} is found
                                                                             This pivot step does not change
     if the optimal solution to L_{\text{aux}} sets \bar{x}_0 to 0
12
          if \bar{x}_0 is basic
                                                                                 the value of any variable.
               perform one (degenerate) pivot to make it nonbasic
13
14
          from the final slack form of L_{\text{aux}}, remove x_0 from the constraints and
               restore the original objective function of L, but replace each basic
               variable in this objective function by the right-hand side of its
               associated constraint
15
          return the modified final slack form
     else return "infeasible"
```

maximise
$$2x_1 - x_2$$
 subject to $2x_1 - x_2 \le 2$ $x_1 - 5x_2 \le -4$ $x_1, x_2 \ge 0$

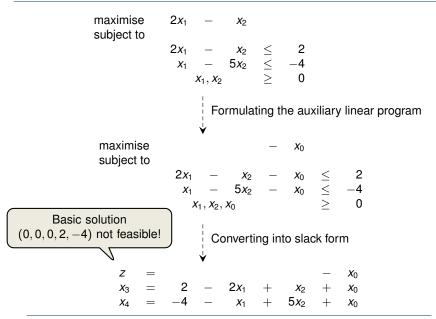
maximise subject to
$$2x_1 - x_2 \leq 2$$

$$2x_1 - 5x_2 \leq -4$$

$$x_1, x_2 \geq 0$$
Formulating the auxiliary linear program
$$- x_0$$
subject to
$$2x_1 - x_2 - x_0 \leq 2$$

$$x_1 - 5x_2 - x_0 \leq -4$$

$$x_1, x_2, x_0 \geq 0$$
Converting into slack form



$$z = x_3 = 2 - 2x_1 + x_2 + x_0$$

 $x_4 = -4 - x_1 + 5x_2 + x_0$
Pivot with x_0 entering and x_4 leaving

Example of Initialize-SIMPLEX (2/3)

Basic solution (4,0,0,6,0) is feasible!

$$\begin{array}{rclcrcl}
z & = & - & x_0 \\
x_2 & = & \frac{4}{5} & - & \frac{x_0}{5} & + & \frac{x_1}{5} & + & \frac{x_5}{5} \\
x_3 & = & \frac{14}{5} & + & \frac{4x_0}{5} & - & \frac{9x_1}{5} & + & \frac{x_5}{5}
\end{array}$$

$$z = -x_0$$

$$x_2 = \frac{4}{5} - \frac{x_0}{5} + \frac{x_1}{5} + \frac{x_4}{5}$$

$$x_3 = \frac{14}{5} + \frac{4x_0}{5} - \frac{9x_1}{5} + \frac{x_4}{5}$$

$$\int \text{Set } x_0 = 0 \text{ and express objective function}$$
by non-basic variables

$$\begin{array}{rcl}
z & = & - & x_0 \\
x_2 & = & \frac{4}{5} & - & \frac{x_0}{5} & + & \frac{x_1}{5} & + & \frac{x_2}{5} \\
x_3 & = & \frac{14}{5} & + & \frac{4x_0}{5} & - & \frac{9x_1}{5} & + & \frac{x_2}{5}
\end{array}$$

$$2x_1 - x_2 = 2x_1 - \left(\frac{4}{5} - \frac{x_0}{5} + \frac{x_1}{5} + \frac{x_4}{5}\right)$$

Set $x_0 = 0$ and express objective function by non-basic variables

$$z = -\frac{4}{5} + \frac{9x_1}{5} - \frac{x_4}{5}$$

$$x_2 = \frac{4}{5} + \frac{x_1}{5} + \frac{x_4}{5}$$

$$x_3 = \frac{14}{5} - \frac{9x_1}{5} + \frac{x_4}{5}$$

$$\begin{array}{rclcrcr}
z & = & - & x_0 \\
x_2 & = & \frac{4}{5} & - & \frac{x_0}{5} & + & \frac{x_1}{5} & + & \frac{x_2}{5} \\
x_3 & = & \frac{14}{5} & + & \frac{4x_0}{5} & - & \frac{9x_1}{5} & + & \frac{x_2}{5}
\end{array}$$

$$2x_1 - x_2 = 2x_1 - \left(\frac{4}{5} - \frac{x_0}{5} + \frac{x_1}{5} + \frac{x_4}{5}\right)$$

Set $x_0 = 0$ and express objective function by non-basic variables

$$z = -\frac{4}{5} + \frac{9x_1}{5} - \frac{x_2}{5}$$

$$x_2 = \frac{4}{5} + \frac{x_1}{5} + \frac{x_2}{5}$$

$$x_3 = \frac{14}{5} - \frac{9x_1}{5} + \frac{x_2}{5}$$

Basic solution $(0, \frac{4}{5}, \frac{14}{5}, 0)$, which is feasible!

$$\begin{array}{rclcrcr}
z & = & - & x_0 \\
x_2 & = & \frac{4}{5} & - & \frac{x_0}{5} & + & \frac{x_1}{5} & + & \frac{x_2}{5} \\
x_3 & = & \frac{14}{5} & + & \frac{4x_0}{5} & - & \frac{9x_1}{5} & + & \frac{x_2}{5}
\end{array}$$

Set
$$x_0 = 0$$
 and express objective function by non-basic variables
$$z = -\frac{4}{5} + \frac{9x_1}{x_1} - \frac{x_4}{x_4}$$

Basic solution $(0, \frac{4}{5}, \frac{14}{5}, 0)$, which is feasible!

Lemma 29.12

If a linear program L has no feasible solution, then INITIALIZE-SIMPLEX returns "infeasible". Otherwise, it returns a valid slack form for which the basic solution is feasible.

Fundamental Theorem of Linear Programming

Theorem 29.13 (Fundamental Theorem of Linear Programming)

Any linear program *L*, given in standard form, either

- 1. has an optimal solution with a finite objective value,
- 2. is infeasible, or
- 3. is unbounded.

If L is infeasible, SIMPLEX returns "infeasible". If L is unbounded, SIMPLEX returns "unbounded". Otherwise, SIMPLEX returns an optimal solution with a finite objective value.

Fundamental Theorem of Linear Programming

Theorem 29.13 (Fundamental Theorem of Linear Programming)

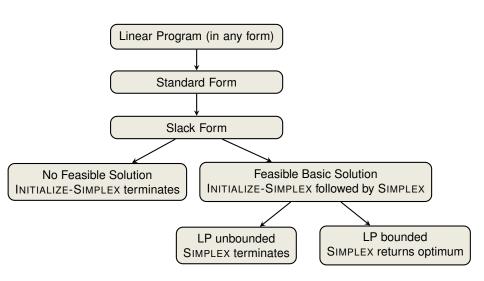
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Proof requires the concept of duality, which is not covered in this course (for details see CLRS3, Chapter 29.4)

Workflow for Solving Linear Programs



Linear Programming and Simplex: Summary and Outlook Linear Programming —

extremely versatile tool for modelling problems of all kinds

Linear Programming —

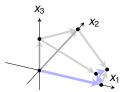
- extremely versatile tool for modelling problems of all kinds
- basis of Integer Programming, to be discussed in later lectures

Linear Programming —

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- Simplex Algorithm -

• In practice: usually terminates in polynomial time, i.e., O(m+n)

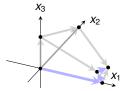


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- Simplex Algorithm -

- In practice: usually terminates in polynomial time, i.e., O(m+n)
- In theory: even with anti-cycling may need exponential time



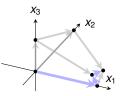
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Simplex Algorithm

- In practice: usually terminates in polynomial time, i.e., O(m+n)
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Research Problem: Is there a pivoting rule which makes SIMPLEX a polynomial-time algorithm?

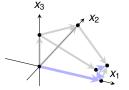


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Polynomial-Time Algorithms

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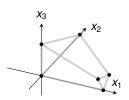
x₂

Xз

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Polynomial-Time Algorithms -

 Interior-Point Methods: traverses the interior of the feasible set of solutions (not just vertices!)



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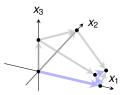
- Simplex Algorithm -

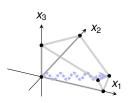
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Polynomial-Time Algorithms -

 Interior-Point Methods: traverses the interior of the feasible set of solutions (not just vertices!)





Test your Understanding



Which of the following statements are true?

- In each iteration of the Simplex algorithm, the objective function increases.
- 2. There exist linear programs that have exactly two optimal solutions.
- 3. There exist linear programs that have infinitely many optimal solutions.
- 4. The Simplex algorithm always runs in worst-case polynomial time.