

# Applications of Linear Programming

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Lecture 2

# Solving LP: slack variables

Let us consider the following LP:

$$\begin{aligned} \text{Max } z &= 3x_1 + 2x_2 \\ x_1 + x_2 &\leq 80 \\ 2x_1 + x_2 &\leq 100 \\ x_1 &\leq 40 \\ x_1, x_2 &\geq 0 \end{aligned}$$

To change an inequality to an equation, we add a new **non-negative variable** called a **slack variable**.

$$\begin{aligned} \text{Max } z &= 3x_1 + 2x_2 \\ x_1 + x_2 + x_3 &= 80 \\ 2x_1 + x_2 + x_4 &= 100 \\ x_1 + x_5 &= 40 \\ x_1, \dots, x_5 &\geq 0 \end{aligned}$$

## Solving LP:

Now we have

$$\begin{aligned}
 \text{Max } z &= 3x_1 + 2x_2 \\
 x_1 + x_2 + x_3 &= 80 \\
 2x_1 + x_2 + x_4 &= 100 \\
 x_1 + x_5 &= 40 \\
 x_1, \dots, x_5 &\geq 0
 \end{aligned}$$

Express the slack variables from the individual equations

$x_3$	=	80	-	$x_1$	-	$x_2$
$x_4$	=	100	-	$2x_1$	-	$x_2$
$x_5$	=	40	-	$x_1$		
$z$	=	0	+	$3x_1$	+	$2x_2$

This is called a **dictionary**.

# Solving LP: dictionary

In the general case we have...

$$\begin{array}{rcccccccc}
 a_{11}x_1 & + & a_{12}x_2 & + & \dots & + & a_{1n}x_n & + & x_{n+1} & = & b_1 \\
 a_{21}x_1 & + & a_{22}x_2 & + & \dots & + & a_{2n}x_n & + & x_{n+2} & = & b_2 \\
 & & & & & & & & \vdots & & \\
 a_{m1}x_1 & + & a_{m2}x_2 & + & \dots & + & a_{mn}x_n & + & x_{n+m} & = & b_m \\
 \hline
 c_1x_1 & + & c_2x_2 & + & \dots & + & c_nx_n & & & = & z
 \end{array}$$

...and the dictionary is

$$\begin{array}{rcccccccc}
 x_{n+1} & = & b_1 & - & a_{11}x_1 & - & a_{12}x_2 & - & \dots & - & a_{1n}x_n \\
 x_{n+2} & = & b_2 & - & a_{21}x_1 & - & a_{22}x_2 & - & \dots & - & a_{2n}x_n \\
 & & & & & & & & \vdots & & \\
 x_{n+m} & = & b_m & - & a_{m1}x_1 & - & a_{m2}x_2 & - & \dots & - & a_{mn}x_n \\
 \hline
 z & = & & & c_1x_1 & + & c_2x_2 & + & \dots & + & c_nx_n
 \end{array}$$

# Terminology

**Decision variables:** variables of the original LP (given in standard form)  
( $x_1, x_2, \dots, x_n$ )

**Slack variables:** new non-negative variables to construct the dictionary  
( $x_{n+1}, x_{n+2}, \dots, x_{n+m}$ )

**Basic variables** variables on the left-hand side of the constraint equalities of the dictionary

**Non-basic variables** variables on the right-hand side of the constraint equalities of the dictionary

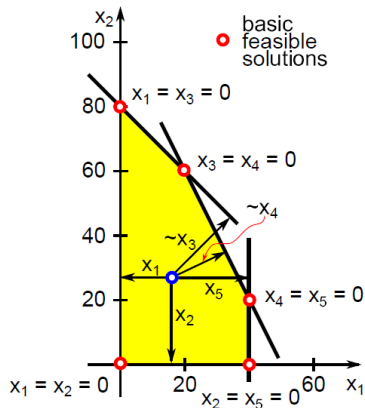
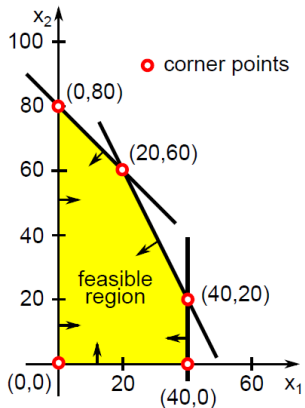
**Basic solution** vector  $x$ , such that the non-basic-variables are zero (and the basic variables are constant values of the respective equations on the right-hand side)

**Feasible basic solution** a basic solution which is feasible, i.e.  $b_i \geq 0$   
 $i = 1, 2, \dots, m$  is satisfied.

# Dictionary: basic solution

$$\begin{aligned}
 \max \quad & 3x_1 + 2x_2 \\
 \text{s.t.} \quad & x_1 + x_2 \leq 80 \\
 & 2x_1 + x_2 \leq 100 \\
 & x_1 \leq 40 \\
 & x_1, x_2 \geq 0
 \end{aligned}
 \longrightarrow$$

$$\begin{aligned}
 \max \quad & 3x_1 + 2x_2 \\
 \text{s.t.} \quad & x_1 + x_2 + x_3 = 80 \\
 & 2x_1 + x_2 + x_4 = 100 \\
 & x_1 + x_5 = 40 \\
 & x_1, x_2, x_3, x_4, x_5 \geq 0
 \end{aligned}$$



## Dictionary: basic solution

**Important remark:** Basic solutions are precisely the corner points of the feasible region.

Recall that we have discussed that to find an optimal solution to an LP, it suffices to find a best solution among all corner points. The above tells us how to compute them – they are the **basic feasible solutions**.

Now we suppose that  $b_1 \geq 0, \dots, b_m \geq 0$ , therefore the **basic solution** of the initial dictionary is

$$x = (0, 0, \dots, 0, b_1, b_2, \dots, b_m)$$

is a **feasible solution**.

# Simplex algorithm

- Algorithm: **iterative search of the optimal solution**
- **Tries to find a new dictionary** such that:
  - ① Every two consecutive **dictionaries are equivalent**
  - ② In each iteration the **value of the objective function is larger or equal**, than in case the previous dictionary
  - ③ The **basic solution is feasible** in each iteration
- How to find a new dictionary?
- How to find it, such that the assumptions above would be satisfied?
- How do we know that a basic solution is optimal?
- Is there an optimal solution for every LP?



# Simplex algorithm

**Pivot step:** calculate a new feasible by changing the roles of a basic and non-basic variable (i.e. re-arranging a constraint equation)

**Incoming variable:** a non-basic variable that become a basic variable of the new dictionary in a simplex iteration

**Outgoing variable:** a basic variable that become a non-basic variable of the new dictionary in a simplex iteration

**Two dictionaries are equivalent:** if the two have the same set of solutions with equal objective function values for the same solutions

**Proposition 1.** *Each dictionary is equivalent to the original system.*

# Simplex algorithm

**How to find a new dictionary? Example:**

$$\begin{array}{rcllcl} x_4 & = & 4 & - & x_1 & - & 2x_2 \\ x_5 & = & 6 & - & x_1 & & - 4x_3 \\ x_6 & = & 2 & & & - & 2x_2 + 2x_3 \\ \hline z & = & 10 & + & 3x_1 & + & 4x_2 - x_3 \end{array}$$

- The value of the objective function increases if  $x_1$  or  $x_2$  increases in the basic solution
- The value of the objective function decreases if  $x_3$  increases in the basic solution
- We can increase either  $x_1$  or  $x_2$ ...

# Simplex algorithm

How to find a dictionary, such that the assumptions above would be satisfied?

$$\begin{array}{rccccccc} x_4 & = & 4 & - & x_1 & - & 2x_2 \\ x_5 & = & 6 & - & x_1 & & - 4x_3 \\ x_6 & = & 2 & & & - & 2x_2 + 2x_3 \\ \hline z & = & 10 & + & 3x_1 & + & 4x_2 - x_3 \end{array}$$

- How much can we increase  $x_2$  before a (dependent) variable becomes negative?

$$x_4 \geq 0 \Rightarrow 4 - 2x_2 \geq 0 \Leftrightarrow 2 \geq x_2$$

$$x_6 \geq 0 \Rightarrow 2 - 2x_2 \geq 0 \Leftrightarrow 1 \geq x_2$$

- This can be streamlined into the simple “ratio” test.

# Simplex algorithm

How do we know that a basic solution is optimal?

$$\begin{array}{rccccr} x_4 & = & 4 & - & x_1 & - & 2x_2 & & & \\ x_5 & = & 6 & - & x_1 & & & - & 4x_3 & \\ x_6 & = & 2 & & & - & 2x_2 & + & 2x_3 & \\ \hline z & = & & - & 3x_1 & - & 4x_2 & - & x_3 & \end{array}$$

No more improvement possible → **optimal solution**

**Theorem 1.** *If there is no positive  $c_j$  ( $j = 1, 2, \dots, n + m$ ) constant in the objective function and there is no negative  $b_i$  ( $i = 1, 2, \dots, m$ ) constant in the constraint equations, then the basic solution of the dictionary is optimal.*

**Proof.** Homework



# Simplex algorithm

Is there an optimal solution for every LP?

$$\begin{array}{rccccccc} x_4 & = & 4 & + & x_1 & - & 2x_2 & & & \\ x_5 & = & 6 & + & x_1 & & & - & 4x_3 & \\ x_6 & = & 2 & & & - & 2x_2 & + & 2x_3 & \\ \hline z & = & & & 3x_1 & - & 4x_2 & - & x_3 & \end{array}$$

We can make  $x_1$  arbitrarily large and thus make  $z$  arbitrarily large  $\rightarrow$  unbounded LP

**Unbounded LP:** if the objective function of the maximization (minimization) LP problem can be arbitrarily large (small)

# Simplex algorithm

**Preparation:** find a starting feasible solution/dictionary

1. Convert to the canonical form (constraints are equalities) by adding slack variables  $x_{n+1}, \dots, x_{n+m}$
2. Construct a starting dictionary - express slack variables and objective function  $z$
3. If the resulting dictionary is feasible, then we are done with preparation  
If not, try to find a feasible dictionary using the **Phase I. method** (next lecture).

# Simplex algorithm

**Simplex step (maximization LP):** try to improve the solution

- (Optimality test):** If **no variable** appears with a **positive** coefficient in the equation for  $z$ 
  - STOP, current solution is **optimal**
  - set non-basic variables to zero
  - read off the values of the basic variables and the objective function  $z$ 
    - Hint: the values are the constant terms in respective equations
  - report this (optimal) solution
- Else pick a variable  $x_i$  having positive coefficient in the equation for  $z$ 
  - $x_i \equiv$  *incoming* variable
- Ratio test: in the dictionary, find an equation for a variable  $x_j$  in which
  - $x_i$  appears with a negative coefficient  $-a$
  - the ratio  $\frac{b}{a}$  is smallest possible  
(where  $b$  is the constant term in the equation for  $x_j$ )
- If no such  $x_j$  exists → stop, no optimal solution, report that **LP is unbounded**
- Else  $x_j \equiv$  *outgoing* variable → construct a new dictionary by *pivoting*:
  - express  $x_j$  from the equation for  $x_j$ ,
  - add this as a new equation,
  - remove the equation for  $x_j$ ,
  - substitute  $x_i$  to all other equations (including the one for  $z$ )
- Repeat from 1.

# Degeneracy

$$\begin{array}{r} x_4 = 1 \qquad \qquad \qquad - 2x_3 \\ x_5 = 3 - 2x_1 + 4x_2 - 6x_3 \\ x_6 = 2 + x_1 - 3x_2 - 4x_3 \\ \hline z = \qquad \qquad 2x_1 - x_2 + 8x_3 \end{array}$$

Pivoting:  $x_3$  enters, ratio test:  $x_4 : \frac{1}{2} = 1/2$ ,  $x_5 : \frac{3}{6} = 1/2$ ,  $x_6 : \frac{2}{4} = 1/2 \rightarrow$  any of  $x_4, x_5, x_6$  can be chosen

$\rightarrow$  we choose  $x_4$  to leave,  $x_3 = \frac{1}{2} - \frac{1}{2}x_4$

$$\begin{array}{r} x_3 = \frac{1}{2} \qquad \qquad \qquad - \frac{1}{2}x_4 \\ x_5 = \qquad - 2x_1 + 4x_2 + 3x_4 \\ x_6 = \qquad \qquad x_1 - 3x_2 + 2x_4 \\ \hline z = 4 + 2x_1 - x_2 - 4x_4 \end{array}$$

setting  $x_1 = x_2 = x_4 = 0$  yields  $x_3 = \frac{1}{2}$ ,  $x_5 = 0$ ,  $x_6 = 0$

now  $x_1$  enters, and  $x_5$  leaves (the only choice),  $x_1 = 2x_2 - \frac{3}{2}x_4 - \frac{1}{2}x_5$

$$\begin{array}{r} x_1 = \qquad 2x_2 + \frac{3}{2}x_4 - \frac{1}{2}x_5 \\ x_3 = \frac{1}{2} \qquad \qquad - \frac{1}{2}x_4 \\ x_6 = \qquad - x_2 + \frac{7}{2}x_4 - \frac{1}{2}x_5 \\ \hline z = 4 + 3x_2 - x_4 - x_5 \end{array}$$

setting  $x_2 = x_4 = x_5 = 0$  yields  $x_1 = 0$ ,  $x_3 = \frac{1}{2}$ ,  $x_6 = 0$

$\rightarrow$  **same solution** as before

if some basic variable is zero, then the basic solution is **degenerate**



# Degeneracy

**Problem:** several dictionaries may correspond to the same (degenerate) solution.

The **simplex rule may cycle**, it is possible to go back to the same dictionary if we are not careful enough when choosing the incoming/outgoing variables.

**Bland's rule:** From possible options, choose an incoming (outgoing) variable  $x_k$  with smallest subscript  $k$ .

**Theorem.** Simplex method using Bland's rule is guaranteed to terminate in a finite number of steps.

Alternative: **lexicographic rule** – choose as outgoing variable one whose row is lexicographically smallest (when divided by the constant term) – the coefficients in the objective function are guaranteed to strictly increase lexicographically

# Two-phase simplex method

The dictionary for **LP in standard (maximization) form**

$$x_{n+i} = b_i - \sum_{j=1}^n a_{ij}x_j \quad i = 1, 2, \dots, m$$

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

- If every  $b_i \geq 0$   $i = 1, 2, \dots, m$ , then we can run the simplex algorithm
- **But if not?**

## Phase I.

$$\begin{array}{rcll}
 \text{Max } z = & x_1 & - & x_2 & + & x_3 & & \\
 & 2x_1 & - & x_2 & + & x_3 & \leq & 4 \\
 & 2x_1 & - & 3x_2 & + & x_3 & \leq & -5 \\
 & -x_1 & + & x_2 & - & 2x_3 & \leq & -1 \\
 & & & & & x_1, x_2, x_3 & \geq & 0
 \end{array}$$

The corresponding dictionary:

$$\begin{array}{rcll}
 x_4 & = & 4 & - & 2x_1 & + & x_2 & - & 2x_3 \\
 x_5 & = & -5 & - & 2x_1 & + & 3x_2 & - & x_3 \\
 x_6 & = & -1 & + & x_1 & - & x_2 & + & 2x_3 \\
 \hline
 z & = & & & x_1 & - & x_2 & + & x_3
 \end{array}$$

This is **not feasible** since  $x_5, x_6 < 0$  in the basic solution.



## Phase I.

- consider the inequality whose right-hand side is most negative (in this case 2nd inequality)
- this inequality has an associated slack variable ( $x_5$ ), remove this variable from our set  $\rightarrow \{x_4, x_6\}$
- add  $x_0$  in place of the removed variable  $\rightarrow \{x_0, x_4, x_6\}$

$$x_0 = 5 + 2x_1 - 3x_2 + x_3 + x_5$$

$$x_4 = 9 - 2x_2 + x_5$$

$$x_6 = 4 + 3x_1 - 4x_2 + 3x_3 + x_5$$

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$$w = -5 - 2x_1 + 3x_2 - x_3 - x_5$$

and this is **feasible!**

# Phase I.

**Theorem** *The standard LP has a feasible solution if and only if  $w = 0$  is the optimal solution of the auxiliary problem.*

## Proof.

- Suppose that  $x$  is a feasible solution of the original problem.
- Then  $(x_0 = 0, x)$  is an optimal solution of the auxiliary problem and  $w((x_0 = 0, x)) = 0$
- Conversely, suppose that 0 is the optimum of the auxiliary problem taken in  $x^*$
- Then  $x_0^* = 0$ , and leaving  $x_0^*$  from  $x^*$  we obtain a feasible solution of the original problem



# Two-phase simplex method

## Steps of Phase I.

- 1 If the basic solution of the corresponding dictionary of the LP in is feasible, then go to Phase II. (simplex algorithm)
- 2 If not, associate the auxiliary problem and prepare its initial feasible dictionary
- 3 Solve the auxiliary problem with the simplex algorithm
- 4 If the optimum  $< 0$ , then there is no feasible solution of the original problem
- 5 If the optimum  $= 0$ , then we get a dictionary that is equivalent with the dictionary of the original LP and its basic solution is feasible

## Steps of Phase II

- 1 Run the simplex algorithm starting from the feasible dictionary obtained by Phase I.

# Two-phase simplex method: example

The initial dictionary:

$$\begin{array}{rcllcl}
 x_3 & = & -5 & - & x_1 & + & x_2 \\
 x_4 & = & 6 & - & x_1 & - & x_2 \\
 \hline
 z & = & & & 2x_1 & + & x_2
 \end{array}$$

The auxiliary problem:

$$\begin{array}{rcllclcl}
 x_3 & = & -5 & - & x_1 & + & x_2 & + & x_0 \\
 x_4 & = & 6 & - & x_1 & - & x_2 & + & x_0 \\
 \hline
 w & = & & & & & & - & x_0
 \end{array}$$



# Two-phase simplex method: example

right-hand side is most negative:  $x_3 = \dots \Rightarrow$ : use  $x_0$  here as basic variable

$$\begin{array}{rcccc}
 x_0 & = & 5 & + & x_1 & - & x_2 & + & x_3 \\
 x_4 & = & 11 & & & - & 2x_2 & + & x_3 \\
 \hline
 w & = & -5 & - & x_1 & + & x_2 & - & x_3
 \end{array}$$

Incoming:  $x_2$ , outgoing:  $x_0$ :

$$\begin{array}{rcccc}
 x_2 & = & 5 & + & x_1 & - & x_0 & + & x_3 \\
 x_4 & = & 1 & - & 2x_1 & + & 2x_0 & - & x_3 \\
 \hline
 w & = & & & & - & x_0
 \end{array}$$

Optimum:  $x_0 = 0$ ,  $w = 0 \Rightarrow$  leave  $x_0$ -t, back to the original objective function

## Two-phase simplex method: example

The obtained dictionary is equivalent to the initial one:

$$\begin{array}{rcl} x_2 & = & 5 + x_1 + x_3 \\ x_4 & = & 1 - 2x_1 - x_3 \\ \hline z & = & 2x_1 + x_2 \end{array}$$

We get ( $x_2$  is changed to  $5 + x_1 + x_3$  in the right-hand side)

$$\begin{array}{rcl} x_2 & = & 5 + x_1 + x_3 \\ x_4 & = & 1 - 2x_1 - x_3 \\ \hline z & = & 5 + 3x_1 + x_3 \end{array}$$

Running the simplex algorithm, the solution can be read from

$$\begin{array}{rcl} x_2 & = & 5.5 - 0.5x_4 + 0.5x_3 \\ x_1 & = & 0.5 - 0.5x_4 - 0.5x_3 \\ \hline z & = & 6.5 - 1.5x_4 - 0.5x_3 \end{array}$$

# Fundamental theorem of linear programming

**Theorem.** *For an arbitrary linear program in standard form, the following statements are true:*

- *If there is no optimal solution, then the problem is either infeasible or unbounded.*
- *If a feasible solution exists, then a basic feasible solution exists.*
- *If an optimal solution exists, then a basic optimal solution exists.*

**Proof.**

- The Phase I algorithm either proves that the problem is infeasible or produces a basic feasible solution
- The Phase II algorithm either discovers that the problem is unbounded or finds a basic optimal solution
- These statements depend, of course, on applying a variant of the simplex method that does not cycle, which we now know to exist.



# Summary

The process consists of two steps

1. Find a **feasible** solution (or determine that **none exists**).
2. Improve the feasible solution to an **optimal** solution.

