# LINEAR PROGRAMMING

[V. CH2]: THE SIMPLEX METHOD

# Phillip Keldenich Ahmad Moradi

Department of Computer Science Algorithms Department TU Braunschweig

November 7, 2022

### SOME EXAMPLES FIRST

THE SIMPLEX ALGORITHM

INITIALIZATION/INFEASIBILITY

UNBOUNDEDNESS

GEOMETRY

### SIMPLEX ALGORITHM

In this chapter, we are going to learn a *method to solve* general linear programs. The method, called *Simplex algorithm*, will be developed for a general linear program (LP) in *standard form*.

#### SIMPLEX ALGORITHM

In this chapter, we are going to learn a *method to solve* general linear programs. The method, called *Simplex algorithm*, will be developed for a general linear program (LP) in *standard form*.

### Consider a simple example:

#### **EXAMPLE**

$$\max_{x} \quad 5x_{1} + \quad 4x_{2} + \quad 3x_{3}$$
s.t. 
$$2x_{1} + \quad 3x_{2} + \quad x_{3} \leq 5$$

$$4x_{1} + \quad x_{2} + \quad 2x_{3} \leq 11$$

$$3x_{1} + \quad 4x_{2} + \quad 2x_{3} \leq 8$$

$$x_{1}, \quad x_{2}, \quad x_{3} \geq 0$$

Start by adding the so-called slack variables and convert *inequality* constraints to *equality* ones.

For each of the less-than inequalities: Introduce a slack variable that represents the difference between the right-hand side and the left-hand side.

 $\rightsquigarrow$  Introducing slack variable  $w_1$ 

$$2x_1 + 3x_2 + x_3 \le 5 \iff w_1 = 5 - 2x_1 - 3x_2 - x_3, \quad w_1 \ge 0$$

 $\rightsquigarrow$  Introducing  $w_2$ 

$$4x_1 + x_2 + 2x_3 \le 11 \iff w_2 = 11 - 4x_1 - x_2 - 2x_3, \quad w_2 \ge 0$$

 $\rightsquigarrow$  Introducing  $w_3$ 

$$3x_1 + 4x_2 + 2x_3 \le 8 \iff w_3 = 8 - 3x_1 - 4x_2 - 2x_3, \quad w_3 \ge 0$$

We get the following *equivalent* LP

The simplex method is an iterative process in which:

 $\rightsquigarrow$  we start with a less-than-optimal solution  $(\dot{x}_1, \dot{x}_2, \cdots, \dot{w}_3)$  that satisfies the equations and non-negativities and then

We get the following equivalent LP

The simplex method is an iterative process in which:

- $\leadsto$  we start with a less-than-optimal solution  $(\dot{x}_1,\dot{x}_2,\cdots,\dot{w}_3)$  that satisfies the equations and non-negativities and then
- $\leadsto$  we look for a new solution  $(\bar{x}_1,\bar{x}_2,\cdots,\bar{w}_3)$ , which is better in the sense that it has a larger objective function value:

$$5\bar{x}_1 + 4\bar{x}_2 + 3\bar{x}_3 > 5\dot{x}_1 + 4\dot{x}_2 + 3\dot{x}_3$$

We get the following equivalent LP

The simplex method is an iterative process in which:

- $\rightsquigarrow$  we start with a less-than-optimal solution  $(\dot{x}_1, \dot{x}_2, \cdots, \dot{w}_3)$  that satisfies the equations and non-negativities and then
- $\leadsto$  we look for a new solution  $(\bar{x}_1,\bar{x}_2,\cdots,\bar{w}_3)$ , which is better in the sense that it has a larger objective function value:

$$5\bar{x}_1 + 4\bar{x}_2 + 3\bar{x}_3 > 5\dot{x}_1 + 4\dot{x}_2 + 3\dot{x}_3$$

→ We continue this process until we arrive at a solution that cannot be improved.

We get the following equivalent LP

The simplex method is an iterative process in which:

- $\rightsquigarrow$  we start with a less-than-optimal solution  $(\dot{x}_1,\dot{x}_2,\cdots,\dot{w}_3)$  that satisfies the equations and non-negativities and then
- $\leadsto$  we look for a new solution  $(\bar{x}_1,\bar{x}_2,\cdots,\bar{w}_3)$ , which is better in the sense that it has a larger objective function value:

$$5\bar{x}_1 + 4\bar{x}_2 + 3\bar{x}_3 > 5\dot{x}_1 + 4\dot{x}_2 + 3\dot{x}_3$$

→ We continue this process until we arrive at a solution that cannot be improved.

This final solution is then an optimal solution.

# **INITIAL SOLUTION**

Consider our example problem.

$$w_1 = 5 - 2x_1 - 3x_2 - x_3$$
  
 $w_2 = 11 - 4x_1 - x_2 - 2x_3$   
 $w_3 = 8 - 3x_1 - 4x_2 - 2x_3$ 

To start the iterative process, we need an *initial feasible solution*.

### INITIAL SOLUTION

Consider our example problem.

$$w_1 = 5 - 2x_1 - 3x_2 - x_3$$
  
 $w_2 = 11 - 4x_1 - x_2 - 2x_3$   
 $w_3 = 8 - 3x_1 - 4x_2 - 2x_3$ 

To start the iterative process, we need an *initial feasible solution*.

Simply set all the *original* variables to *zero*:

$$x_1 = 0$$
,  $x_2 = 0$ ,  $x_3 = 0$ .

Now, use the equations to determine the slack variables:

$$w_1 = 5$$
,  $w_2 = 11$ ,  $w_3 = 8$ .

### INITIAL SOLUTION

Consider our example problem.

$$w_1 = 5 - 2x_1 - 3x_2 - x_3$$
  
 $w_2 = 11 - 4x_1 - x_2 - 2x_3$   
 $w_3 = 8 - 3x_1 - 4x_2 - 2x_3$ 

To start the iterative process, we need an *initial feasible solution*.

Simply set all the *original* variables to *zero*:

$$x_1 = 0$$
,  $x_2 = 0$ ,  $x_3 = 0$ .

Now, use the equations to determine the slack variables:

$$w_1 = 5$$
,  $w_2 = 11$ ,  $w_3 = 8$ .

Luckily, we found a feasible solution:

$$(\dot{x}_1, \dot{x}_2, \dot{x}_3, \dot{w}_1, \dot{w}_2, \dot{w}_3) = (0, 0, 0, 5, 11, 8)$$

with objective function value  $\zeta = 0$ .

#### SOLUTION IMPROVEMENT

We now ask whether this solution can be improved.

$$\max_{x} \quad \zeta = 0 + 5 x_{1} + 4x_{2} + 3x_{3}$$
s.t. 
$$w_{1} = 5 - 2x_{1} - 3x_{2} - x_{3}$$

$$w_{2} = 11 - 4x_{1} - x_{2} - 2x_{3}$$

$$w_{3} = 8 - 3x_{1} - 4x_{2} - 2x_{3}$$

$$x_{1}, x_{2}, x_{3}, w_{1}, w_{2}, w_{3} \ge 0$$

#### SOLUTION IMPROVEMENT

We now ask whether this solution can be improved.

$$\max_{x} \quad \zeta = 0 + 5 x_{1} + 4x_{2} + 3x_{3}$$
s.t. 
$$w_{1} = 5 - 2x_{1} - 3x_{2} - x_{3}$$

$$w_{2} = 11 - 4x_{1} - x_{2} - 2x_{3}$$

$$w_{3} = 8 - 3x_{1} - 4x_{2} - 2x_{3}$$

$$x_{1}, x_{2}, x_{3}, w_{1}, w_{2}, w_{3} \ge 0$$

#### Observation.

Since the coefficient of  $x_1$  in the objective function is *positive*, if we increase the value of  $x_1$  from zero to some positive value, we will increase  $\zeta$ .

#### SOLUTION IMPROVEMENT

We now ask whether this solution can be improved.

$$\max_{x} \quad \zeta = 0 + 5 x_{1} + 4x_{2} + 3x_{3}$$
s.t. 
$$w_{1} = 5 - 2x_{1} - 3x_{2} - x_{3}$$

$$w_{2} = 11 - 4x_{1} - x_{2} - 2x_{3}$$

$$w_{3} = 8 - 3x_{1} - 4x_{2} - 2x_{3}$$

$$x_{1}, x_{2}, x_{3}, w_{1}, w_{2}, w_{3} \ge 0$$

#### Observation.

Since the coefficient of  $x_1$  in the objective function is *positive*, if we increase the value of  $x_1$  from zero to some positive value, we will increase  $\zeta$ .

#### Observation.

As we change  $x_1$ 's value, the values of the slack variables will also change. We must make sure that we do not let any of them go negative.

$$\max_{x} \quad \zeta = 0 + 5 x_{1} + 4x_{2} + 3x_{3}$$
s.t. 
$$w_{1} = 5 - 2x_{1} - 3x_{2} - x_{3}$$

$$w_{2} = 11 - 4x_{1} - x_{2} - 2x_{3}$$

$$w_{3} = 8 - 3x_{1} - 4x_{2} - 2x_{3}$$

$$x_{1}, x_{2}, x_{3}, w_{1}, w_{2}, w_{3} \ge 0$$

 $x_2$  and  $x_3$  are currently set to 0, we see that

$$w_1 = 5 - 2x_1$$
,

and so keeping  $w_1$  non-negative imposes

$$w_1 \ge 0 \iff 5 - 2x_1 \ge 0 \iff x_1 \le \frac{5}{2}.$$

$$\max_{x} \zeta = 0 + \underbrace{5}_{x_{1}} x_{1} + 4x_{2} + 3x_{3}$$
s.t. 
$$w_{1} = 5 - 2x_{1} - 3x_{2} - x_{3}$$

$$w_{2} = 11 - 4x_{1} - x_{2} - 2x_{3}$$

$$w_{3} = 8 - 3x_{1} - 4x_{2} - 2x_{3}$$

$$x_{1}, x_{2}, x_{3}, w_{1}, w_{2}, w_{3} \ge 0$$

 $x_2$  and  $x_3$  are currently set to 0, we see that

$$w_1 = 5 - 2x_1$$

and so keeping  $w_1$  non-negative imposes

$$w_1 \ge 0 \iff 5 - 2x_1 \ge 0 \iff x_1 \le \frac{5}{2}.$$

- $\rightarrow$  Non-negativity of  $w_2$  imposes the bound that  $x_1 \leq \frac{11}{4}$ .
- $\rightsquigarrow$  Non-negativity of  $w_3$  imposes the bound that  $x_1 \leq \frac{8}{3}$ .

$$\max_{x} \zeta = 0 + \underbrace{5}_{x_{1}} x_{1} + 4x_{2} + 3x_{3}$$
s.t. 
$$w_{1} = 5 - 2x_{1} - 3x_{2} - x_{3}$$

$$w_{2} = 11 - 4x_{1} - x_{2} - 2x_{3}$$

$$w_{3} = 8 - 3x_{1} - 4x_{2} - 2x_{3}$$

$$x_{1}, x_{2}, x_{3}, w_{1}, w_{2}, w_{3} \ge 0$$

 $x_2$  and  $x_3$  are currently set to 0, we see that

$$w_1 = 5 - 2x_1,$$

and so keeping  $w_1$  non-negative imposes

$$w_1 \ge 0 \iff 5 - 2x_1 \ge 0 \iff x_1 \le \frac{5}{2}.$$

- $\rightarrow$  Non-negativity of  $w_2$  imposes the bound that  $x_1 \leq \frac{11}{4}$ .
- $\rightarrow$  Non-negativity of  $w_3$  imposes the bound that  $x_1 \leq \frac{8}{3}$ .

Since all of these non-negativity conditions must be met, we see that  $x_1$  cannot be made larger than the smallest of these bounds:  $x_1 \leq \frac{5}{2}$ .

Now we can be sure raising  $x_1$  up to  $\frac{5}{2}$  will not destroy non-negativity of variables.

Now we can be sure raising  $x_1$  up to  $\frac{5}{2}$  will not destroy non-negativity of variables. Set  $x_1=\frac{5}{2}$  and re-compute slack values according to the defining equations

$$w_1 = 5 - 2x_1$$
  
 $w_2 = 11 - 4x_1$   
 $w_3 = 8 - 3x_1$ 

we get

$$w_1 = 0 , w_2 = 1 , w_3 = \frac{1}{2}.$$

Now we can be sure raising  $x_1$  up to  $\frac{5}{2}$  will not destroy non-negativity of variables. Set  $x_1 = \frac{5}{2}$  and re-compute slack values according to the defining equations

$$w_1 = 5 - 2x_1$$

$$w_2 = 11 - 4x_1$$

$$w_3 = 8 - 3x_1$$

we get

$$w_1 = 0 , \ w_2 = 1 , \ w_3 = \frac{1}{2}.$$

Our new solution then is

$$(\bar{x}_1, \bar{x}_2, \bar{x}_3, \bar{w}_1, \bar{w}_2, \bar{w}_3) = (\frac{5}{2}, 0, 0, 0, 1, \frac{1}{2})$$

with objective function value

$$\zeta = 5\bar{x}_1 + 4\bar{x}_2 + 3\bar{x}_3 = \frac{25}{2} > 0$$

Now we can be sure raising  $x_1$  up to  $\frac{5}{2}$  will not destroy non-negativity of variables. Set  $x_1 = \frac{5}{2}$  and re-compute slack values according to the defining equations

$$w_1 = 5 - 2x_1$$

$$w_2 = 11 - 4x_1$$

$$w_3 = 8 - 3x_1$$

we get

$$w_1 = 0$$
,  $w_2 = 1$ ,  $w_3 = \frac{1}{2}$ .

Our new solution then is

$$(\bar{x}_1, \bar{x}_2, \bar{x}_3, \bar{w}_1, \bar{w}_2, \bar{w}_3) = (\frac{5}{2}, 0, 0, 0, 1, \frac{1}{2})$$

with objective function value

$$\zeta = 5\bar{x}_1 + 4\bar{x}_2 + 3\bar{x}_3 = \frac{25}{2} > 0$$

We found an improved solution!

Lets capture what we have done up to now.

- We considered the following special layout

$\zeta =$	0 +	$5x_1 + $	$4x_2 + $	$3x_3$
$\overline{w_1} =$	5 —	$2x_1 - $	$3x_2$ -	$x_3$
$w_2 =$	11 -	$4x_1 -$	$x_2$ -	$2x_3$
$w_3 =$	8 —	$3x_1 - $	$4x_2 - $	$2x_3$

Lets capture what we have done up to now.

- We considered the following special layout

- Then, we found an initial feasible solution by setting variables on the right  $(x_i)$  to zero and reading off variables on the left  $(w_i)$ .

Lets capture what we have done up to now.

- We considered the following special layout

- Then, we found an initial feasible solution by setting variables on the right  $(x_i)$  to zero and reading off variables on the left  $(w_i)$ .
- Then, we looked at the objective function and found a variable  $(x_1)$  with positive coefficient. Increasing  $x_1$  will improve objective function value.

Lets capture what we have done up to now.

- We considered the following special layout

- Then, we found an initial feasible solution by setting variables on the right  $(x_i)$  to zero and reading off variables on the left  $(w_i)$ .
- Then, we looked at the objective function and found a variable  $(x_1)$  with positive coefficient. Increasing  $x_1$  will improve objective function value.
- Then, we used the layout to compute maximum possible increase in  $x_1$  and thus improved the objective function while keeping variables on the left non-negative. This way, we constructed a new improved feasible solution.

Lets capture what we have done up to now.

- We considered the following special layout

- Then, we found an initial feasible solution by setting variables on the right  $(x_i)$  to zero and reading off variables on the left  $(w_i)$ .
- Then, we looked at the objective function and found a variable  $(x_1)$  with positive coefficient. Increasing  $x_1$  will improve objective function value.
- Then, we used the layout to compute maximum possible increase in  $x_1$  and thus improved the objective function while keeping variables on the left non-negative. This way, we constructed a new improved feasible solution.

Only this easy because of the special layout!

# CONTINUING

But how to proceed?

### **CONTINUING**

### But how to proceed?

#### Observation.

What made the first step easy was the fact that we had one group of variables that were initially zero and we had the rest explicitly expressed in terms of these.

#### CONTINUING

#### But how to proceed?

#### Observation.

What made the first step easy was the fact that we had one group of variables that were initially zero and we had the rest explicitly expressed in terms of these.

- This special layout is called a dictionary.
- In a dictionary, objective and variables on the left are defined by variables on the right.
- Dependent variables (on the left) are called basic variables.
- Independent variables (on the right) are called nonbasic variables.
- Setting variables on the right to zero and reading off the values of the variables on the left gives us a dictionary solution.

We need to retain this layout/structure after moving to the new solution.

We need to retain this layout/structure after moving to the new solution.

#### Observation.

Raising  $x_1$  up to  $\frac{5}{2}$ , decreases  $w_1$  to zero. It seems now that (in the new solution):  $x_1$  is a basic variable and  $w_1$  is a non-basic variable.

We need to retain this layout/structure after moving to the new solution.

### Observation.

Raising  $x_1$  up to  $\frac{5}{2}$ , decreases  $w_1$  to zero. It seems now that (in the new solution):  $x_1$  is a basic variable and  $w_1$  is a non-basic variable.

Lets rewrite  $w_1$ 's defining equation as

$$w_1 = 5 - 2x_1 - 3x_2 - x_3 \iff x_1 = \frac{5}{2} - \frac{1}{2}w_1 - \frac{3}{2}x_2 - \frac{1}{2}x_3$$
.

We need to retain this layout/structure after moving to the new solution.

### Observation.

Raising  $x_1$  up to  $\frac{5}{2}$ , decreases  $w_1$  to zero. It seems now that (in the new solution):  $x_1$  is a basic variable and  $w_1$  is a non-basic variable.

Lets rewrite  $w_1$ 's defining equation as

$$w_1 = 5 - 2x_1 - 3x_2 - x_3 \iff x_1 = \frac{5}{2} - \frac{1}{2}w_1 - \frac{3}{2}x_2 - \frac{1}{2}x_3$$
.

Now, use the r.h.s. to describe  $w_2, w_3$  and  $\zeta$  only with the new set of independent variables:  $w_1, x_2$  and  $x_3$  as

$$\begin{array}{cccccccccccccc} \zeta = & 12.5 - & 2.5w_1 - & 3.5x_2 + & 0.5x_3 \\ \hline x_1 = & 2.5 - & 0.5w_1 - & 1.5x_2 - & 0.5x_3 \\ \hline w_2 = & 1 + & 2w_1 + & 5x_2 \\ \hline w_3 = & 0.5 + & 1.5w_1 + & 0.5x_2 - & 0.5x_3 \\ \end{array}$$

#### But how to proceed?

We need to retain this layout/structure after moving to the new solution.

#### Observation.

Raising  $x_1$  up to  $\frac{5}{2}$ , decreases  $w_1$  to zero. It seems now that (in the new solution):  $x_1$  is a basic variable and  $w_1$  is a non-basic variable.

Lets rewrite  $w_1$ 's defining equation as

$$w_1 = 5 - 2x_1 - 3x_2 - x_3 \iff x_1 = \frac{5}{2} - \frac{1}{2}w_1 - \frac{3}{2}x_2 - \frac{1}{2}x_3$$
.

Now, use the r.h.s. to describe  $w_2, w_3$  and  $\zeta$  only with the new set of independent variables:  $w_1, x_2$  and  $x_3$  as

$$\begin{array}{cccccccccccccc} \frac{\zeta = & 12.5 - & 2.5w_1 - & 3.5x_2 + & 0.5x_3}{x_1 = & 2.5 - & 0.5w_1 - & 1.5x_2 - & 0.5x_3} \\ w_2 = & 1 + & 2w_1 + & 5x_2 \\ w_3 = & 0.5 + & 1.5w_1 + & 0.5x_2 - & 0.5x_3 \end{array}$$

#### Note.

We can recover our current solution by setting the *independent* (non-basic) variables to zero and using the equations to read off the values for the dependent (basic) variables.

Having the current (dictionary) solution and its corresponding dictionary, we can look for any further improvement.

Having the current (dictionary) solution and its corresponding dictionary, we can look for any further improvement.

Now  $x_3$  is the only variable with a positive coefficient.

Having the current (dictionary) solution and its corresponding dictionary, we can look for any further improvement.

Now  $x_3$  is the only variable with a positive coefficient.

Again, we need to determine how much  $x_3$  can be increased without violating the requirement that all the dependent variables remain nonnegative.

Having the current (dictionary) solution and its corresponding dictionary, we can look for any further improvement.

Now  $x_3$  is the only variable with a positive coefficient.

Again, we need to determine how much  $x_3$  can be increased without violating the requirement that all the dependent variables remain nonnegative.

This time, we see that the equation for  $w_2$  is not affected by changes in  $x_3$ , but the equations for  $x_1$  and  $w_3$  do impose bounds, namely  $x_3 \le 5$  and  $x_3 \le 1$ , respectively.

Having the current (dictionary) solution and its corresponding dictionary, we can look for any further improvement.

Now  $x_3$  is the only variable with a positive coefficient.

Again, we need to determine how much  $x_3$  can be increased without violating the requirement that all the dependent variables remain nonnegative.

This time, we see that the equation for  $w_2$  is not affected by changes in  $x_3$ , but the equations for  $x_1$  and  $w_3$  do impose bounds, namely  $x_3 \le 5$  and  $x_3 \le 1$ , respectively.

 $\rightarrow x_3$  could be increased up to 1.

Set  $x_3=1$  and re-compute dependent (basic) variable values according to the defining equations:

$$x_1 = 2.5 - 0.5x_3$$
  
 $w_2 = 1$   
 $w_3 = 0.5 - 0.5x_3$ 

we get

$$x_1 = 2, \quad w_2 = 1, \quad w_3 = 0.$$

Set  $x_3=1$  and re-compute dependent (basic) variable values according to the defining equations:

$$x_1 = 2.5 - 0.5x_3$$
  
 $w_2 = 1$   
 $w_3 = 0.5 - 0.5x_3$ 

we get

$$x_1 = 2$$
,  $w_2 = 1$ ,  $w_3 = 0$ .

Our new solution then is

$$(\bar{x}_1, \bar{x}_2, \bar{x}_3, \bar{w}_1, \bar{w}_2, \bar{w}_3) = (2, 0, 1, 0, 1, 0)$$

with objective function value

$$\zeta = 5\bar{x}_1 + 4\bar{x}_2 + 3\bar{x}_3 = 13.$$

Set  $x_3=1$  and re-compute dependent (basic) variable values according to the defining equations:

$$x_1 = 2.5 - 0.5x_3$$
  
 $w_2 = 1$   
 $w_3 = 0.5 - 0.5x_3$ 

we get

$$x_1 = 2$$
,  $w_2 = 1$ ,  $w_3 = 0$ .

Our new solution then is

$$(\bar{x}_1, \bar{x}_2, \bar{x}_3, \bar{w}_1, \bar{w}_2, \bar{w}_3) = (2, 0, 1, 0, 1, 0)$$

with objective function value

$$\zeta = 5\bar{x}_1 + 4\bar{x}_2 + 3\bar{x}_3 = 13.$$

We found an improved solution!

In order to retain a dictionary layout for this solution, use  $w_2$ 's defining equation and re-write it as

$$w_3 = 0.5 + 1.5w_1 + 0.5x_2 - 0.5x_3 \iff x_3 = 1 + 3w_1 + x_2 - 2w_3$$
.

Now, use the right-hand side to describe  $x_1, w_2$  and  $\zeta$  only with the new set of independent variables:  $w_1, x_2$  and  $w_3$  as

In order to retain a dictionary layout for this solution, use  $w_2$ 's defining equation and re-write it as

$$w_3 = 0.5 + 1.5w_1 + 0.5x_2 - 0.5x_3 \iff x_3 = 1 + 3w_1 + x_2 - 2w_3$$
.

Now, use the right-hand side to describe  $x_1, w_2$  and  $\zeta$  only with the new set of independent variables:  $w_1, x_2$  and  $w_3$  as

#### Note.

There is *no independent variable* for which an increase in its value would produce a corresponding increase in  $\zeta$  and the algorithm stops.

In order to retain a dictionary layout for this solution, use  $w_2$ 's defining equation and re-write it as

$$w_3 = 0.5 + 1.5w_1 + 0.5x_2 - 0.5x_3 \iff x_3 = 1 + 3w_1 + x_2 - 2w_3$$
.

Now, use the right-hand side to describe  $x_1, w_2$  and  $\zeta$  only with the new set of independent variables:  $w_1, x_2$  and  $w_3$  as

#### Note.

There is *no independent variable* for which an increase in its value would produce a corresponding increase in  $\zeta$  and the algorithm stops.

Claim: The current dictionary solution is *optimal*! The objective value  $\zeta$  is at most 13. Why?

In order to retain a dictionary layout for this solution, use  $w_2$ 's defining equation and re-write it as

$$w_3 = 0.5 + 1.5w_1 + 0.5x_2 - 0.5x_3 \iff x_3 = 1 + 3w_1 + x_2 - 2w_3$$
.

Now, use the right-hand side to describe  $x_1$ ,  $w_2$  and  $\zeta$  only with the new set of independent variables:  $w_1$ ,  $x_2$  and  $w_3$  as

#### Note.

There is *no independent variable* for which an increase in its value would produce a corresponding increase in  $\zeta$  and the algorithm stops.

Claim: The current dictionary solution is *optimal*! The objective value  $\zeta$  is at most 13. Why? We got to the equation

$$\zeta = 13 - w_1 - 3x_2 - w_3$$

by equivalence-preserving steps using only the constraints of our linear program!

# ANOTHER EXAMPLE

#### Consider another example:

#### **EXAMPLE**

$$\max_{x} -x_{1} + 3x_{2} - 3x_{3}$$
s.t. 
$$3x_{1} - x_{2} - 2x_{3} \le 7$$

$$-2x_{1} - 4x_{2} + 4x_{3} \le 3$$

$$x_{1} - 2x_{3} \le 4$$

$$-2x_{1} + 2x_{2} + x_{3} \le 8$$

$$3x_{1} \le 5$$

$$x_{1}, x_{2}, x_{3} \ge 0$$

Rewrite examples with slack variables:

#### We obtain

an initial dictionary with
 as non-basic (independent) variables
 as basic (dependent) variables

Rewrite examples with slack variables:

#### We obtain

- an initial dictionary with
  - $x_1, x_2, x_3$  as non-basic (independent) variables on the right, and  $w_1, w_2, w_3, w_4, w_5$  as basic (dependent) variables on the left

Rewrite examples with slack variables:

#### We obtain

- an initial dictionary with
  - $x_1, x_2, x_3$  as non-basic (independent) variables on the right, and  $w_1, w_2, w_3, w_4, w_5$  as basic (dependent) variables on the left
- and dictionary solution

Rewrite examples with slack variables:

#### We obtain

- an initial dictionary with
  - $x_1, x_2, x_3$  as non-basic (independent) variables on the right, and  $w_1, w_2, w_3, w_4, w_5$  as basic (dependent) variables on the left
- and dictionary solution

$$(\dot{x}_1, \dot{x}_2, \dot{x}_3, \dot{w}_1, \dot{w}_2, \dot{w}_3, \dot{w}_4, \dot{w}_5) = (0, 0, 0, 7, 3, 4, 8, 5)$$

$\zeta =$	_	$x_1 +$	$3x_2 -$	$3x_3$
$\overline{w_1} =$	7 —	$3x_1 + $	$x_2 + $	$2x_3$
$w_2 =$	3 +	$2x_1 + $	$4x_2 -$	$4x_3$
$w_3 =$	4 -	$x_1$	+	$2x_3$
$w_4 =$	8 +	$2x_1$ -	$2x_2 - $	$x_3$
$w_5 =$	5-	$3x_1$		

– If  $x_2$  increases,  $\zeta$  goes up. Which variable reaches 0 first?

$\zeta =$	_	$x_1 +$	$3 x_2 -$	$3x_3$
$\overline{w_1} =$	7 —	$3x_1 + $	$x_2 + $	$2x_3$
$w_2 =$	3 +	$2x_1 + $	$4x_2 -$	$4x_3$
$w_3 =$	4 -	$x_1$	+	$2x_3$
$w_4 =$	8 +	$2x_1 - $	$2x_2 - $	$x_3$
$w_5 =$	5-	$3x_1$		

- If  $x_2$  increases,  $\zeta$  goes up. Which variable reaches 0 first?
- As  $x_2$  increases,  $w_4$  reaches zero; all other basic variables stay or go up! We say:  $x_2$  becomes basic and  $w_4$  becomes nonbasic.

$\zeta =$	_	$x_1 + $	$3 x_2 -$	$3x_3$
$\overline{w_1} =$	7 —	$3x_1 + $	$x_2 + $	$2x_3$
$w_2 =$	3 +	$2x_1 + $	$4x_2 -$	$4x_3$
$w_3 =$	4 -	$x_1$	+	$2x_3$
$w_4 =$	8 +	$2x_1$ -	$2x_2 - $	$x_3$
$w_5 =$	5-	$3x_1$		

- If  $x_2$  increases,  $\zeta$  goes up. Which variable reaches 0 first?
- As  $x_2$  increases,  $w_4$  reaches zero; all other basic variables stay or go up! We say:  $x_2$  becomes basic and  $w_4$  becomes nonbasic.
- Algebraically rearrange equations to retain the corresponding dictionary. This is called a pivot.

$\zeta =$	_	$x_1 +$	$3 x_2 -$	$3x_3$
$\overline{w_1} =$	7 —	$3x_1 + $	$x_2 + $	$2x_3$
$w_2 =$	3 +	$2x_1 + $	$4x_2 -$	$4x_3$
$w_3 =$	4 -	$x_1$	+	$2x_3$
$w_4 =$	8 +	$2x_1$ -	$2x_2 - $	$x_3$
$w_5 =$	5-	$3x_1$		

- If  $x_2$  increases,  $\zeta$  goes up. Which variable reaches 0 first?
- As  $x_2$  increases,  $w_4$  reaches zero; all other basic variables stay or go up! We say:  $x_2$  becomes basic and  $w_4$  becomes nonbasic.
- Algebraically rearrange equations to retain the corresponding dictionary. This is called a pivot.
- This basically means: Rearrange the linear equation defining the leaving variable  $w_4$  to isolate the entering variable  $x_2$ , and substitute the new definition of  $x_2$  in all other equations.

### PIVOT STEP

A pivot:  $x_2$  gets basic (enters the basis) and  $w_4$  gets nonbasic (leaves the basis).

$\zeta =$	12 +	$2 x_1 -$	$1.5w_4 -$	$4.5x_3$
$\overline{w_1} =$	11 –	$2x_1$ -	$0.5w_4 +$	$1.5x_{3}$
$w_2 =$	19 +	$6x_1 -$	$2w_4$ -	$6x_3$
$w_3 =$	4 -	$x_1$	+	$2x_3$
$x_2 =$	4+	$x_1$ -	$0.5w_4$ -	$0.5x_{3}$
$w_5 =$	5 -	$3x_1$		

$\zeta =$	12 +	$2 x_1 -$	$1.5w_4 -$	$4.5x_3$
$\overline{w_1} =$	11 –	$2x_1$ -	$0.5w_4 +$	$1.5x_3$
$w_2 =$	19 +	$6x_1 -$	$2w_4$ -	$6x_3$
$w_3 =$	4-	$x_1$	+	$2x_3$
$x_2 =$	4+	$x_1$ -	$0.5w_4$ -	$0.5x_{3}$
$w_5 =$	5 -	$3x_1$		

- Now, let  $x_1$  increase. Which basic variable becomes nonbasic?

- Now, let  $x_1$  increase. Which basic variable becomes nonbasic?
- Of the basic variables,  $w_5$  hits zero first at  $x_1 = \frac{5}{3}$ .  $x_1$  enters and  $w_5$  leaves the basis.

- Now, let  $x_1$  increase. Which basic variable becomes nonbasic?
- Of the basic variables,  $w_5$  hits zero first at  $x_1 = \frac{5}{3}$ .  $x_1$  enters and  $w_5$  leaves the basis.
- Rearrange equations accordingly.

# RESULTING DICTIONARY

$\zeta =$	$\frac{46}{3}$ -	$\frac{2}{3}w_5$ -	$\frac{3}{2}w_4$ -	$\frac{9}{2}x_3$
$w_1 =$	$\frac{23}{3}$ +	$\frac{2}{3}w_5$ -	$\frac{1}{2}w_4 +$	$\frac{3}{2}x_{3}$
$w_2 =$	29 -	$2w_5 -$	$2w_4$ -	$6x_3$
$w_3 =$	$\frac{7}{3}$ +	$\frac{1}{3}w_5$	+	$2x_3$
$x_2 =$	$\frac{17}{3}$ -	$\frac{1}{3}w_5$ -	$\frac{1}{2}w_4$ -	$\frac{1}{2}x_3$
$x_1 =$	$\frac{5}{3}$ -	$\frac{1}{3}w_5$		

# RESULTING DICTIONARY

$$\frac{\zeta = \frac{46}{3} - \frac{2}{3}w_5 - \frac{3}{2}w_4 - \frac{9}{2}x_3}{w_1 = \frac{23}{3} + \frac{2}{3}w_5 - \frac{1}{2}w_4 + \frac{3}{2}x_3}$$

$$w_2 = 29 - 2w_5 - 2w_4 - 6x_3$$

$$w_3 = \frac{7}{3} + \frac{1}{3}w_5 + 2x_3$$

$$x_2 = \frac{17}{3} - \frac{1}{3}w_5 - \frac{1}{2}w_4 - \frac{1}{2}x_3$$

$$x_1 = \frac{5}{3} - \frac{1}{3}w_5$$

- no improvement is possible and the dictionary solution

# RESULTING DICTIONARY

$$\frac{\zeta = \frac{46}{3} - \frac{2}{3}w_5 - \frac{3}{2}w_4 - \frac{9}{2}x_3}{w_1 = \frac{23}{3} + \frac{2}{3}w_5 - \frac{1}{2}w_4 + \frac{3}{2}x_3}$$

$$\frac{w_2 = 29 - 2w_5 - 2w_4 - 6x_3}{w_3 = \frac{7}{3} + \frac{1}{3}w_5 + 2x_3}$$

$$\frac{x_2 = \frac{17}{3} - \frac{1}{3}w_5 - \frac{1}{2}w_4 - \frac{1}{2}x_3}{x_1 = \frac{5}{3} - \frac{1}{3}w_5}$$

- no improvement is possible and the dictionary solution

$$(\bar{x}_1, \bar{x}_2, \bar{x}_3, \bar{w}_1, \bar{w}_2, \bar{w}_3, \bar{w}_4, \bar{w}_5) = (\frac{5}{3}, \frac{17}{3}, 0, \frac{23}{3}, 29, \frac{7}{3}, 0, 0)$$

is optimal with  $\zeta = \frac{46}{3}$ .

#### SOME EXAMPLES FIRST

THE SIMPLEX ALGORITHM

INITIALIZATION/INFEASIBILITY

UNBOUNDEDNESS

GEOMETRY

### INPUT

We now try to describe the simplex algorithm to solve a general linear program. Given an LP in standard form:

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

$$s.t. \sum_{j=1}^{n} a_{ij}x_{j} \leq b_{i} \quad , \quad i = 1, 2, \cdots, m$$

$$x_{j} \geq 0 \quad , \quad j = 1, 2, \cdots, n.$$

Our first task is to introduce

#### INPUT

We now try to describe the simplex algorithm to solve a general linear program. Given an LP in standard form:

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

$$s.t. \sum_{j=1}^{n} a_{ij}x_{j} \leq b_{i} \quad , \quad i = 1, 2, \cdots, m$$

$$x_{j} \geq 0 \quad , \quad j = 1, 2, \cdots, n.$$

Our first task is to introduce slack variables and a name for the objective function value:

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

$$w_i = b_i - \sum_{j=1}^{n} a_{ij} x_j, \qquad i = 1, 2, \cdots, m$$

# FIRST DICTIONARY

As we saw in our examples, as the simplex method proceeds, the slack variables become *intertwined* with the original variables, and the whole collection is treated the same.

So lets rewrite

$$(x_1, \dots, x_n, w_1, \dots, w_m) = (x_1, \dots, x_n, x_{n+1}, \dots, x_{n+m})$$

That is, we let  $x_{n+i} = w_i$ ,  $i = 1, 2, \dots, m$ .

## FIRST DICTIONARY

As we saw in our examples, as the simplex method proceeds, the slack variables become *intertwined* with the original variables, and the whole collection is treated the same.

So lets rewrite

$$(x_1, \dots, x_n, w_1, \dots, w_m) = (x_1, \dots, x_n, x_{n+1}, \dots, x_{n+m})$$

That is, we let  $x_{n+i} = w_i$ ,  $i = 1, 2, \dots, m$ .

With this notation, our first dictionary looks like

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

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

# FIRST DICTIONARY

As we saw in our examples, as the simplex method proceeds, the slack variables become *intertwined* with the original variables, and the whole collection is treated the same.

So lets rewrite

$$(x_1, \dots, x_n, w_1, \dots, w_m) = (x_1, \dots, x_n, x_{n+1}, \dots, x_{n+m})$$

That is, we let  $x_{n+i} = w_i, \quad i = 1, 2, \dots, m$ .

With this notation, our first dictionary looks like

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

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

As the simplex method progresses, it moves from one dictionary to another in its search for an optimal solution. Each dictionary has m basic variables and n nonbasic variables.

Let

 ${\mathscr B}$  denote the set of indices, from  $\{1,2,\ldots,n+m\}$ , corresponding to the basic variables,

N

denote the set of indices corresponding to the nonbasic variables

Let

- ${\mathscr B}$  denote the set of indices, from  $\{1,2,\ldots,n+m\}$ , corresponding to the basic variables,
- ${\cal N}$  denote the set of indices corresponding to the nonbasic variables

Initially, we have

$$\mathcal{N} = \{1, 2, ..., n\}$$
  
 $\mathcal{B} = \{n + 1, n + 2, ..., n + m\}$ 

but this of course changes after the first iteration.

Let

- ${\mathscr B}$  denote the set of indices, from  $\{1,2,\ldots,n+m\}$ , corresponding to the basic variables,
- ${\mathcal N}$  denote the set of indices corresponding to the nonbasic variables

Initially, we have

$$\mathcal{N} = \{1, 2, ..., n\}$$
  
 $\mathcal{B} = \{n + 1, n + 2, ..., n + m\}$ 

but this of course changes after the first iteration.

Down the road, the current dictionary will look like:

$$\zeta = \bar{\zeta} + \sum_{j \in \mathcal{N}} \bar{c}_j x_j$$

$$x_i = \bar{b}_i - \sum_{j \in \mathcal{N}} \bar{a}_{ij} x_j, \quad i \in \mathscr{B}$$

Let

 ${\mathscr B}$  denote the set of indices, from  $\{1,2,\ldots,n+m\}$ , corresponding to the basic variables,

 ${\mathcal N}$  denote the set of indices corresponding to the nonbasic variables

### Initially, we have

$$\mathcal{N} = \{1, 2, ..., n\}$$
  
 $\mathcal{B} = \{n + 1, n + 2, ..., n + m\}$ 

but this of course changes after the first iteration.

Down the road, the current dictionary will look like:

$$\zeta = \bar{\zeta} + \sum_{j \in \mathcal{N}} \bar{c}_j x_j$$

$$x_i = \bar{b}_i - \sum_{j \in \mathcal{N}} \bar{a}_{ij} x_j, \quad i \in \mathcal{B}$$

#### Note.

We have put bars over the coefficients to indicate that they *change* as the algorithm progresses.

Within each iteration of the simplex method,

exactly one variable goes from nonbasic to basic.

Within each iteration of the simplex method,

exactly one variable goes from nonbasic to basic.

The variable that goes from nonbasic to basic is called the *entering variable*. It is chosen with the aim of increasing  $\zeta$ ; that is, one whose coefficient is *positive*:

$$\text{pick } k \text{ from } \{j \in \mathcal{N}: \bar{c}_j > 0\}.$$

Within each iteration of the simplex method,

exactly one variable goes from nonbasic to basic.

The variable that goes from nonbasic to basic is called the *entering variable*. It is chosen with the aim of increasing  $\zeta$ ; that is, one whose coefficient is *positive*:

$$\text{pick } k \text{ from } \{j \in \mathcal{N}: \bar{c}_j > 0\}.$$

#### Note.

If the set  $\{j \in \mathcal{N} : \bar{c}_j > 0\}$  is *empty*, then the current solution is optimal. If the set consists of *more than one element*, then we have a choice of which element to pick.

Within each iteration of the simplex method,

exactly one variable goes from nonbasic to basic.

The variable that goes from nonbasic to basic is called the *entering variable*. It is chosen with the aim of increasing  $\zeta$ ; that is, one whose coefficient is *positive*:

$$\text{pick } k \text{ from } \{j \in \mathcal{N}: \bar{c}_j > 0\}.$$

#### Note.

If the set  $\{j \in \mathcal{N} : \bar{c}_j > 0\}$  is *empty*, then the current solution is optimal. If the set consists of *more than one element*, then we have a choice of which element to pick.

- $\leadsto$  For now, suffice it to pick an index k having the largest coefficient (which again could leave us with a choice).
- $\leadsto$  Technically, any choice works; in practice, the choice has a strong influence on the number of steps we have to do.

Within each iteration of the simplex method,

exactly one variable goes from basic to nonbasic.

The variable that goes from basic to nonbasic is called the *leaving variable*. It is chosen to preserve non-negativity of the current basic variables.

Within each iteration of the simplex method,

exactly one variable goes from basic to nonbasic.

The variable that goes from basic to nonbasic is called the *leaving variable*. It is chosen to preserve non-negativity of the current basic variables.

Once we have decided that  $x_k$  will be the entering variable, its value will be increased from *zero* to a *positive* value. This increase will change the values of the basic variables as:

$$x_i = \bar{b}_i - \bar{a}_{ik} x_k, \quad i \in \mathscr{B}.$$

Within each iteration of the simplex method,

exactly one variable goes from basic to nonbasic.

The variable that goes from basic to nonbasic is called the *leaving variable*. It is chosen to preserve non-negativity of the current basic variables.

Once we have decided that  $x_k$  will be the entering variable, its value will be increased from *zero* to a *positive* value. This increase will change the values of the basic variables as:

$$x_i = \bar{b}_i - \bar{a}_{ik} x_k, \quad i \in \mathscr{B}.$$

We must ensure that each of these variables remains non-negative. Hence, we require that

$$0 \le \bar{b}_i - \bar{a}_{ik} x_k, \quad i \in \mathscr{B}.$$

Within each iteration of the simplex method,

exactly one variable goes from basic to nonbasic.

The variable that goes from basic to nonbasic is called the *leaving variable*. It is chosen to preserve non-negativity of the current basic variables.

Once we have decided that  $x_k$  will be the entering variable, its value will be increased from *zero* to a *positive* value. This increase will change the values of the basic variables as:

$$x_i = \bar{b}_i - \bar{a}_{ik} x_k, \quad i \in \mathscr{B}.$$

We must ensure that each of these variables remains non-negative. Hence, we require that

$$0 \le \bar{b}_i - \bar{a}_{ik} x_k, \quad i \in \mathscr{B}.$$

#### Note.

Of these expressions, the only ones that can go negative (as  $x_k$  increases) are those for which  $\bar{a}_{ik}$  is *positive*; the rest remain fixed or increase.

Hence, we can restrict our attention to those i's for which  $\bar{a}_{ik}$  is positive. And for such an i, the value of  $x_k$  at which the expression becomes zero is

$$x_k = \frac{\bar{b}_i}{\bar{a}_{ik}}, \quad i \in \mathscr{B} : \bar{a}_{ik} > 0.$$

Hence, we can restrict our attention to those i's for which  $\bar{a}_{ik}$  is positive. And for such an i, the value of  $x_k$  at which the expression becomes zero is

$$x_k = \frac{\bar{b}_i}{\bar{a}_{ik}}, \quad i \in \mathcal{B} : \bar{a}_{ik} > 0.$$

Since we do not want any of these to go negative, we must raise  $\boldsymbol{x}_k$  only to the smallest of all of these values:

$$x_k = \min_{i \in \mathcal{B} : \bar{a}_{ik} > 0} \quad \frac{\bar{b}_i}{\bar{a}_{ik}}$$

Hence, we can restrict our attention to those i's for which  $\bar{a}_{ik}$  is positive. And for such an i, the value of  $x_k$  at which the expression becomes zero is

$$x_k = \frac{\bar{b}_i}{\bar{a}_{ik}}, \quad i \in \mathcal{B} : \bar{a}_{ik} > 0.$$

Since we do not want any of these to go negative, we must raise  $x_k$  only to the smallest of all of these values:

$$x_k = \min_{i \in \mathscr{B} \; : \; \bar{a}_{ik} > 0} \quad \frac{\bar{b}_i}{\bar{a}_{ik}}$$

Therefore, the rule for selecting the leaving variable is:

pick 
$$l$$
 from  $\{i\in \mathscr{B}\,:\, \bar{a}_{ik}>0 \text{ and } \frac{\bar{b}_i}{\bar{a}_{ik}} \text{ is minimal}\}$ 

Hence, we can restrict our attention to those i's for which  $\bar{a}_{ik}$  is positive. And for such an i, the value of  $x_k$  at which the expression becomes zero is

$$x_k = \frac{\bar{b}_i}{\bar{a}_{ik}}, \quad i \in \mathcal{B} : \bar{a}_{ik} > 0.$$

Since we do not want any of these to go negative, we must raise  $x_k$  only to the smallest of all of these values:

$$x_k = \min_{i \in \mathscr{B} : \bar{a}_{ik} > 0} \quad \frac{\bar{b}_i}{\bar{a}_{ik}}$$

Therefore, the rule for selecting the leaving variable is:

pick 
$$l$$
 from  $\{i\in \mathscr{B}\,:\, \bar{a}_{ik}>0 \text{ and } \frac{\bar{b}_i}{\bar{a}_{ik}} \text{ is minimal}\}$ 

What special situation occurs if we have a choice regarding the leaving variable?

Hence, we can restrict our attention to those i's for which  $\bar{a}_{ik}$  is positive. And for such an i, the value of  $x_k$  at which the expression becomes zero is

$$x_k = \frac{\bar{b}_i}{\bar{a}_{ik}}, \quad i \in \mathcal{B} : \bar{a}_{ik} > 0.$$

Since we do not want any of these to go negative, we must raise  $x_k$  only to the smallest of all of these values:

$$x_k = \min_{i \in \mathscr{B} \ : \ \bar{a}_{ik} > 0} \quad \frac{\bar{b}_i}{\bar{a}_{ik}}$$

Therefore, the rule for selecting the leaving variable is:

pick 
$$l$$
 from  $\{i\in \mathcal{B}\,:\, \bar{a}_{ik}>0 \text{ and } \frac{\bar{b}_i}{\bar{a}_{ik}} \text{ is minimal}\}$ 

What special situation occurs if we have a choice regarding the leaving variable? We end up with a basic variable that is 0, a so-called *degeneracy*.

Hence, we can restrict our attention to those i's for which  $\bar{a}_{ik}$  is positive. And for such an i, the value of  $x_k$  at which the expression becomes zero is

$$x_k = \frac{\bar{b}_i}{\bar{a}_{ik}}, \quad i \in \mathcal{B} : \bar{a}_{ik} > 0.$$

Since we do not want any of these to go negative, we must raise  $x_k$  only to the smallest of all of these values:

$$x_k = \min_{i \in \mathscr{B} : \bar{a}_{ik} > 0} \quad \frac{\bar{b}_i}{\bar{a}_{ik}}$$

Therefore, the rule for selecting the leaving variable is:

pick 
$$l$$
 from  $\{i\in\mathcal{B}\,:\, \bar{a}_{ik}>0 \text{ and } \frac{\bar{b}_i}{\bar{a}_{ik}} \text{ is minimal}\}$ 

What special situation occurs if we have a choice regarding the leaving variable?

- → We end up with a basic variable that is 0, a so-called *degeneracy*.
- → This can cause problems! Can you think of any? We will deal with that in the next lecture!

Once the leaving basic and entering nonbasic variables have been selected,

- the move from the current dictionary to the new dictionary involves appropriate row operations to achieve the *interchange*.
- In general, this means rearranging the defining equation of the leaving variable to make it define the entering variable instead, and substituting the new definition of the entering variable into all right-hand sides.
- This step from one dictionary to the next is called a *pivot*.

Once the leaving basic and entering nonbasic variables have been selected,

- the move from the current dictionary to the new dictionary involves appropriate row operations to achieve the *interchange*.
- In general, this means rearranging the defining equation of the leaving variable to make it define the entering variable instead, and substituting the new definition of the entering variable into all right-hand sides.
- This step from one dictionary to the next is called a *pivot*.

In the algorithm, we usually pivot to improve our solution. What are the minimum requirements to do a pivot?

Once the leaving basic and entering nonbasic variables have been selected,

- the move from the current dictionary to the new dictionary involves appropriate row operations to achieve the *interchange*.
- In general, this means rearranging the defining equation of the leaving variable to make it define the entering variable instead, and substituting the new definition of the entering variable into all right-hand sides.
- This step from one dictionary to the next is called a *pivot*.

In the algorithm, we usually pivot to improve our solution. What are the minimum requirements to do a pivot?

Actually, we only need the entering variable to occur with a non-zero coefficient in the defining equation of the leaving variable.

Once the leaving basic and entering nonbasic variables have been selected,

- the move from the current dictionary to the new dictionary involves appropriate row operations to achieve the *interchange*.
- In general, this means rearranging the defining equation of the leaving variable to make it define the entering variable instead, and substituting the new definition of the entering variable into all right-hand sides.
- This step from one dictionary to the next is called a *pivot*.

In the algorithm, we usually pivot to improve our solution. What are the minimum requirements to do a pivot?

Actually, we only need the entering variable to occur with a non-zero coefficient in the defining equation of the leaving variable.

Of course, such general pivots might lead to infeasible dictionaries or make our solution worse.

Once the leaving basic and entering nonbasic variables have been selected,

- the move from the current dictionary to the new dictionary involves appropriate row operations to achieve the *interchange*.
- In general, this means rearranging the defining equation of the leaving variable to make it define the entering variable instead, and substituting the new definition of the entering variable into all right-hand sides.
- This step from one dictionary to the next is called a *pivot*.

In the algorithm, we usually pivot to improve our solution. What are the minimum requirements to do a pivot?

Actually, we only need the entering variable to occur with a non-zero coefficient in the defining equation of the leaving variable.

Of course, such general pivots might lead to infeasible dictionaries or make our solution worse.

As mentioned, there is *often* more than one choice for the entering variable (and sometimes also for the leaving variable). Particular rules that make the choice *unambiguous* are called *pivot rules*.

SOME EXAMPLES FIRST

THE SIMPLEX ALGORITHM

INITIALIZATION/INFEASIBILITY

UNBOUNDEDNESS

GEOMETRY

# NEW EXAMPLE

# Consider the following example:

# EXAMPLE

$$\begin{array}{lll} \max_{x} & -2x_{1} - & x_{2} \\ \text{s.t.} & -x_{1} + & x_{2} \leq -1 \\ & -x_{1} - & 2x_{2} \leq -2 \\ & & x_{2} \leq 1 \\ & x_{1}, & x_{2} \geq 0 \end{array}$$

→ The initial dictionary is *not* feasible. Why?

# NEW EXAMPLE

# Consider the following example:

# **EXAMPLE**

$$\begin{array}{lll} \max_{x} & -2x_{1} - & x_{2} \\ \text{s.t.} & -x_{1} + & x_{2} \leq -1 \\ & -x_{1} - & 2x_{2} \leq -2 \\ & & x_{2} \leq 1 \\ & x_{1}, & x_{2} \geq 0 \end{array}$$

- → The initial dictionary is *not* feasible. Why?
- → Up to now, we only considered problems for which the right-hand sides were all *non-negative*. This *ensured* that the *initial dictionary was feasible*. Now, we discuss what to do when this is not the case as the above example.

# THE PROBLEM IN GENERAL

Given an LP:

$$\max_{x} \quad \sum_{j=1}^{n} c_{j} x_{j}$$

$$s.t. \quad \sum_{j=1}^{n} a_{ij} x_{j} \leq b_{i} \quad , \quad i = 1, 2, \cdots, m$$

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

The initial dictionary looks like

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

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

- The solution associated with this dictionary is obtained by setting:  $x_j = 0$ ,  $w_i = b_i$
- This solution is feasible if and only if all  $b_i$ 's are non-negative.
- → But what if they are not?

We handle this difficulty by introducing an auxiliary problem for which

- (1) a feasible dictionary is easy to find, and
- (2) an optimal dictionary provides a feasible dictionary for the original problem,
- (3) or proves that no feasible solution exists.

The auxiliary problem is

We handle this difficulty by introducing an auxiliary problem for which

- (1) a feasible dictionary is easy to find, and
- (2) an optimal dictionary provides a feasible dictionary for the original problem,
- (3) or proves that no feasible solution exists.

The auxiliary problem is

$$\max_{x} -x_{0}$$

$$s.t. \quad \sum_{j=1}^{n} a_{ij}x_{j} - x_{0} \leq b_{i}, \quad i = 1, 2, \dots, m$$

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

We handle this difficulty by introducing an auxiliary problem for which

- (1) a feasible dictionary is easy to find, and
- (2) an optimal dictionary provides a feasible dictionary for the original problem,
- (3) or proves that no feasible solution exists.

The auxiliary problem is

$$\max_{x} -x_{0}$$

$$s.t. \quad \sum_{j=1}^{n} a_{ij}x_{j} - x_{0} \leq b_{i}, \quad i = 1, 2, \dots, m$$

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

The auxiliary problem is always feasible: Simply set  $x_j = 0$  for  $j = 1, \dots, n$ , and then pick  $x_0$  sufficiently large.

We handle this difficulty by introducing an auxiliary problem for which

- (1) a feasible dictionary is easy to find, and
- (2) an optimal dictionary provides a feasible dictionary for the original problem,
- (3) or proves that no feasible solution exists.

The auxiliary problem is

$$\max_{x} -x_{0}$$

$$s.t. \quad \sum_{j=1}^{n} a_{ij}x_{j} - x_{0} \leq b_{i}, \quad i = 1, 2, \dots, m$$

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

 $\rightarrow$  The auxiliary problem is always feasible: Simply set  $x_j = 0$  for  $j = 1, \dots, n$ , and then pick  $x_0$  sufficiently large.

#### Note.

The original problem has a feasible solution iff the auxiliary problem has a feasible solution with  $x_0 = 0$ . In other words, the original problem has a feasible solution iff the optimal solution to the auxiliary problem has zero objective value.

Even though the auxiliary problem clearly has feasible solutions, we have not yet shown that it has an easily obtained feasible dictionary. It is best to illustrate how to obtain a feasible dictionary with an example.

# Consider again the example

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

# The auxiliary problem is

$$\max_{x} -x_{0}$$
s.t.  $-x_{1} + x_{2} - x_{0} \le -1$ 
 $-x_{1} - 2x_{2} - x_{0} \le -2$ 
 $x_{2} - x_{0} \le 1$ 
 $x_{1}, x_{2}, x_{0} \ge 0$ 

Next, we introduce slack variables and write down an initial *infeasible* dictionary:

$\xi =$				$-1 x_0$
$\overline{w_1} =$	-1 +	$x_1$ -	$x_2 + $	$x_0$
$w_2 =$	-2 +	$x_1 +$	$2x_2 + $	$x_0$
$w_3 =$	1	_	$x_2 +$	$x_0$

Next, we introduce slack variables and write down an initial *infeasible* dictionary:

$\xi =$				$-1 x_0$
$\overline{w_1} =$	-1 +	$x_1$ -	$x_2 + $	$x_0$
$w_2 =$	-2 +	$x_1 +$	$2x_2 + $	$x_0$
$w_3 =$	1	_	$x_2 + $	$x_0$

To turn it feasible, *all we need* is to do a pivot with variable  $x_0$  entering and the *most infeasible* basic variable,  $w_2$ , leaving. Why?

Next, we introduce slack variables and write down an initial *infeasible* dictionary:

$\xi =$				$-1 x_0$
$\overline{w_1} =$	-1 +	$x_1$ -	$x_2 + $	$x_0$
$w_2 =$	-2 +	$x_1 +$	$2x_2 + $	$x_0$
$w_3 =$	1	_	$x_2 + $	$x_0$

To turn it feasible, *all we need* is to do a pivot with variable  $x_0$  entering and the *most infeasible* basic variable,  $w_2$ , leaving. Why?

$$\frac{\xi = -2 + 1x_1 + 2x_2 - 1w_2}{w_1 = 1} - 3x_2 + w_2$$

$$x_0 = 2 - x_1 - 2x_2 + w_2$$

$$w_3 = 3 - x_1 - 3x_2 + w_2$$

 $\leadsto$  Note that we now have a feasible dictionary, so we can apply the simplex method as defined earlier in this chapter.

### Consider our feasible dictionary:

$\xi =$	-2 +	$1 x_1 +$	$2 x_2 -$	$1w_2$
$\overline{w_1} =$	1	_	$3x_2 + $	$w_2$
$x_0 =$	2 -	$x_1$ -	$2x_2 + $	$w_2$
$w_3 =$	3 -	$x_1$ -	$3x_2 + $	$w_2$

#### Consider our feasible dictionary:

We pick  $x_2$  to enter and  $w_1$  to leave the basis. We get

#### Consider our feasible dictionary:

We pick  $x_2$  to enter and  $w_1$  to leave the basis. We get

Now, for the second step, we pick  $x_1$  to enter and  $x_0$  to leave the basis.

We get:

$\xi =$	_	$x_0$		
$x_2 =$	0.33	_	$0.33w_1 +$	$0.33w_{2}$
$x_1 =$	1.33 -	$x_0 + $	$0.67w_1 +$	$0.33w_{2}$
$w_3 =$	0.67 +	$x_0 +$	$0.33w_1 -$	$0.33w_{2}$

We get:

This dictionary is optimal for the auxiliary problem. Just note that

If optimal  $\xi < 0$ , the original LP is infeasible!

We get:

This dictionary is *optimal* for the auxiliary problem. Just note that

If optimal 
$$\xi < 0\,,$$
 the original LP is  $\it infeasible!$ 

We now drop  $x_0$  from the equations and *reintroduce the original objective function*:

$$\zeta = -2x_1 - x_2 = -3 - w_1 - w_2.$$

What did we do to get from the old definition of the objective to the new one?

We get:

This dictionary is *optimal* for the auxiliary problem. Just note that

If optimal 
$$\xi < 0\,,$$
 the original LP is  $\it infeasible!$ 

We now drop  $x_0$  from the equations and *reintroduce the original objective function*:

$$\zeta = -2x_1 - x_2 = -3 - w_1 - w_2.$$

What did we do to get from the old definition of the objective to the new one? Substitution!

We get:

This dictionary is *optimal* for the auxiliary problem. Just note that

If optimal 
$$\xi < 0$$
, the original LP is  $infeasible!$ 

We now drop  $x_0$  from the equations and *reintroduce the original objective function*:

$$\zeta = -2x_1 - x_2 = -3 - w_1 - w_2.$$

What did we do to get from the old definition of the objective to the new one? Substitution! Hence, the starting feasible dictionary for the *original problem* is

$$\begin{array}{ccccccc} \zeta = & -3 - & w_1 - & w_2 \\ \hline x_2 = & 0.33 - & 0.33w_1 + & 0.33w_2 \\ x_1 = & 1.33 + & 0.67w_1 + & 0.33w_2 \\ w_3 = & 0.67 + & 0.33w_1 - & 0.33w_2 \end{array}$$

We get:

This dictionary is *optimal* for the auxiliary problem. Just note that

If optimal 
$$\xi < 0\,,$$
 the original LP is  $infeasible!$ 

We now drop  $x_0$  from the equations and *reintroduce the original objective function*:

$$\zeta = -2x_1 - x_2 = -3 - w_1 - w_2.$$

What did we do to get from the old definition of the objective to the new one? Substitution! Hence, the starting feasible dictionary for the *original problem* is

$$\begin{array}{ccccccc} \zeta = & -3 - & w_1 - & w_2 \\ \hline x_2 = & 0.33 - & 0.33w_1 + & 0.33w_2 \\ x_1 = & 1.33 + & 0.67w_1 + & 0.33w_2 \\ w_3 = & 0.67 + & 0.33w_1 - & 0.33w_2 \end{array}$$

As it turns out, this dictionary is *optimal* for the *original* problem (since the coefficients of all the variables in the equation for  $\zeta$  are negative), but we *cannot* expect to be this lucky in general.

## TWO-PHASE SIMPLEX

→ All we normally can expect is that the dictionary so obtained will be *feasible* for the original problem, at which point we continue to apply the simplex method until an optimal solution is reached.

→ The process of solving the auxiliary problem to find an initial feasible solution is often referred to as Phase I, whereas the process of going from a feasible solution to an optimal solution is called Phase II. The overall algorithm is called Two-Phase Simplex Algorithm.

SOME EXAMPLES FIRST

THE SIMPLEX ALGORITHM

INITIALIZATION/INFEASIBILITY

**UNBOUNDEDNESS** 

GEOMETRY

## Consider the following dictionary:

$\zeta =$	0 +	$2 x_1 -$	$x_2 +$	$1 x_3$
$w_1 =$	4 +	$5x_1$ -	$3x_2 + $	$x_3$
$w_2 =$	10 +	$1x_1 + $	$5x_2$ –	$2x_3$
$w_3 =$	7 +		$4x_2 -$	$3x_3$
$w_4 =$	6 +	$2x_1 + $	$2x_2$ –	$4x_3$
$w_5 =$	6 +	$3x_1 + $		$3x_3$

## Consider the following dictionary:

$\zeta =$	0 +	$2x_1 -$	$x_2 +$	$1 x_3$
$w_1 =$	4 +	$5x_1$ -	$3x_2 + $	$x_3$
$w_2 =$	10 +	$1x_1 + $	$5x_2$ –	$2x_3$
$w_3 =$	7 +		$4x_2 -$	$3x_3$
$w_4 =$	6 +	$2x_1 + $	$2x_2$ -	$4x_3$
$w_5 =$	6 +	$3x_1 + $		$3x_3$

–  $x_1$  could be increased to improve  $\zeta$ 

# Consider the following dictionary:

$\zeta =$	0 +	$2x_1 -$	$x_2 +$	$1 x_3$
$w_1 =$	4 +	$5x_1$ -	$3x_2 + $	$x_3$
$w_2 =$	10 +	$1x_1 + $	$5x_2$ -	$2x_3$
$w_3 =$	7 +		$4x_2 -$	$3x_3$
$w_4 =$	6 +	$2x_1 + $	$2x_2$ –	$4x_3$
$w_5 =$	6 +	$3x_1 + $		$3x_3$

- $x_1$  could be increased to improve  $\zeta$
- Which basic variable decreases to 0 first?

## Consider the following dictionary:

$\zeta =$	0 +	$2x_1 -$	$x_2 +$	$1 x_3$
$w_1 =$	4 +	$5x_1 -$	$3x_2 + $	$x_3$
$w_2 =$	10 +	$1x_1 + $	$5x_2$ -	$2x_3$
$w_3 =$	7 +		$4x_2 -$	$3x_3$
$w_4 =$	6 +	$2x_1 + $	$2x_2 - $	$4x_3$
$w_5 =$	6 +	$3x_1 + $		$3x_3$

- $x_1$  could be increased to improve  $\zeta$
- Which basic variable decreases to 0 first?
- None of the basic variables will decrease.  $x_1$  can grow without bound,  $\zeta$  along with it.

## Consider the following dictionary:

$\zeta =$	0 +	$2 x_1 -$	$x_2 + $	$1 x_3$
$w_1 =$	4 +	$5x_1$ -	$3x_2 + $	$x_3$
$w_2 =$	10 +	$1x_1 + $	$5x_2$ -	$2x_3$
$w_3 =$	7 +		$4x_2 -$	$3x_3$
$w_4 =$	6 +	$2x_1 + $	$2x_2 - $	$4x_3$
$w_5 =$	6 +	$3x_1 + $		$3x_3$

- $x_1$  could be increased to improve  $\zeta$
- Which basic variable decreases to 0 first?
- None of the basic variables will decrease.  $x_1$  can grow without bound,  $\zeta$  along with it.

Unboundedness occurs!

## UNBOUNDEDNESS

#### Note.

Given a feasible dictionary, unboundedness occurs when there exists a non-basic variable with positive coefficient in the objective function whose increase is not bounded by any of the existing basic variables.

## Unboundedness

#### Note.

Given a feasible dictionary, unboundedness occurs when there exists a non-basic variable with positive coefficient in the objective function whose increase is not bounded by any of the existing basic variables.

→ As the non-basic variable goes up, the objective function increases without bound.

## Unboundedness

#### Note.

Given a feasible dictionary, unboundedness occurs when there exists a non-basic variable with positive coefficient in the objective function whose increase is not bounded by any of the existing basic variables.

- As the non-basic variable goes up, the objective function increases without bound.
- → Going back to the rule for selecting the leaving variable:

pick 
$$l$$
 from  $\{i\in \mathscr{B}\,:\, \bar{a}_{ik}>0 \text{ and } \frac{\bar{b}_i}{\bar{a}_{ik}} \text{ is minimal}\}$ 

unboundedness, will happen when  $\forall i \in \mathcal{B}: \; : \; \bar{a}_{ik} \leq 0.$ 

#### ANOTHER EXAMPLE

As another example consider the following dictionary

We have:  $k = 3, \mathcal{B} = \{2, 4, 5\}$  and

$$\bar{a}_{23} = -2, \bar{a}_{43} = 0, \bar{a}_{53} = 0$$

all non-positive.

Unboundedness occurs!

SOME EXAMPLES FIRST

THE SIMPLEX ALGORITHM

INITIALIZATION/INFEASIBILITY

UNBOUNDEDNESS

**GEOMETRY** 

When the number of variables in a linear programming problem is *three* or *less*,

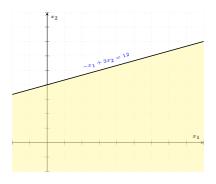
- → we can graph the set of feasible solutions
- $\rightarrow$  we can also graph the level sets of the objective function.

This way, finding the the optimal solution on this picture is usually trivial. To illustrate, consider the following problem:

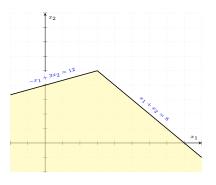
$$\begin{array}{llll} \max & + 3x_1 & + 2x_2 \\ \text{s.t.} & - x_1 & + 3x_2 & \leq 12 \\ & + x_1 & + x_2 & \leq 8 \\ & + 2x_1 & - x_2 & \leq 10 \\ & x_1, x_2 \geq 0 \end{array}$$

Each constraint (including the non-negativity constraints on the variables) is a half-plane.

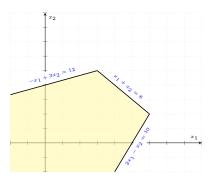
- → These half-planes can be determined by first graphing the equation one obtains by replacing the inequality with an *equality* and then check some specific point.
- → The set of feasible solutions is just the *intersection* of these half-planes.



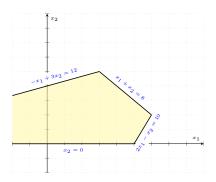
$$(x_1, x_2)$$
 :  $-x_1 + 3x_2 \le 12$ 



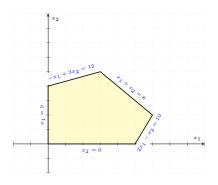
$$(x_1, x_2)$$
 : 
$$\begin{cases} -x_1 & + & 3x_2 \leq 12 \\ +x_1 & - & x_2 \leq 8 \end{cases}$$



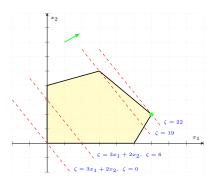
$$(x_1, x_2) : \begin{cases} -x_1 & + & 3x_2 \leq 12 \\ +x_1 & - & x_2 \leq 8 \\ +2x_1 & - & x_2 \leq 10 \end{cases}$$

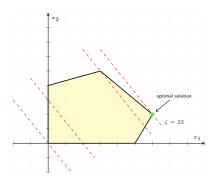


$$(x_1, x_2) : \begin{cases} -x_1 & + & 3x_2 \leq 12 \\ +x_1 & - & x_2 \leq 8 \\ +2x_1 & - & x_2 \leq 10 \\ & & x_2 \geq 0 \end{cases}$$



$$(x_1, x_2) : \begin{cases} -x_1 & + & 3x_2 & \leq & 12 \\ +x_1 & - & x_2 & \leq & 8 \\ +2x_1 & - & x_2 & \leq & 10 \\ & & x_2 & \geq & 0 \\ x_1 & & & \geq & 0 \end{cases}$$

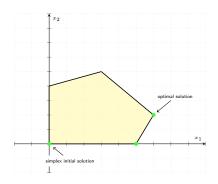




#### Observation.

Algorithms of this type do exist but in higher dimensions the algebra required to implement such an algorithm gets quite complicated.

→ It turns out that the *simplex method* is algebraically much simpler and, on average performs well.



## For the problem at hand:

 $\leadsto$  the simplex method starts at (0,0) and jumps to adjacent vertices (green dots) of the feasible set until it finds a vertex that is an optimal solution. Starting at (0,0), it only takes two simplex pivots to get to the optimal solution.