

## Linear Programming Algorithms - Simplex method

Source: Chapters 4,5 of Matoušek

Assume linear program  $(P)$  in *equational* form:

$$(P) \begin{cases} \text{maximize} & \mathbf{c}^T \mathbf{x} \\ \text{subject to} & A\mathbf{x} = \mathbf{b} \\ & \mathbf{x} \geq \mathbf{0} \end{cases}$$

where  $A \in \mathbb{R}^{m \times n}$ ,  $\mathbf{b} \in \mathbb{R}^m$  and  $\mathbf{c} \in \mathbb{R}^n$ .

**1:** *Can we assume that rows of  $A$  are linearly independent? Can we assume  $n \geq m$ ?*

**Solution:** Yes - if two rows are linearly dependent, then either there is no solution or one is useless. If the rows are linearly independent, we get for free that  $n \geq m$ .

A solution  $\mathbf{x}$  is called *basic feasible solution* if  $n - m$  entries of  $\mathbf{x}$  are zero and the columns of  $A$  corresponding to these remaining  $m$  entries are linearly independent.

**2:** *Is it possible to find two different basic feasible solutions with the same positions of  $n - m$  zero entries?*

**Solution:** No - if the solution has  $m$  nonzeros. The system of equations  $m \times m$  has a unique solution since the columns of  $A$  are linearly independent. Yes - if the actual optimal solution has more zeros, then some could be replaced by different columns.

**Theorem 1.** *If program  $(P)$  has an optimal solution, it also has an optimal solution that is a basic feasible solution.*

Corollary: Optimal solution to  $(P)$  can be found by examining all  $\binom{n}{m}$  subsets of columns of  $A$ .

A set  $B \subset \{1, \dots, n\}$  is *base* if columns of  $A$  indexed by  $B$  give a basic feasible solution (denoted by  $A_B$ ).

Simplex method: Start with a base and alter the base by changing one entry at a time.

Example of simplex method:

$$(P) \begin{cases} \text{maximize} & x_1 + x_2 \\ \text{subject to} & -x_1 + x_2 \leq 1 \\ & x_1 \leq 3 \\ & x_2 \leq 2 \\ & x_1, x_2 \geq 0 \end{cases}$$

**3:** Rewrite the program to equational form. (*Hint: use slack variables - that is add 3 more variables*)

**Solution:**

$$(P) \begin{cases} \text{maximize} & x_1 + x_2 \\ \text{subject to} & -x_1 + x_2 + x_3 = 1 \\ & x_1 + x_4 = 3 \\ & x_2 + x_5 = 2 \\ & x_1, x_2, x_3, x_4, x_5 \geq 0 \end{cases}$$

**4:** Is there some obvious basic feasible (not necessarily optimal solution)?

**Solution:**  $x_3 = 1$ ,  $x_4 = 3$ ,  $x_5 = 2$ , the base is nice identity matrix.

We create a thing called *simplex tableau* for base  $B = \{3, 4, 5\}$ :

$$\begin{array}{rclcl} x_3 & = & 1 & + & x_1 & - & x_2 \\ x_4 & = & 3 & - & x_1 & & \\ x_5 & = & 2 & & & - & x_2 \\ \hline z & = & 0 & + & x_1 & + & x_2 \end{array}$$

Features:  $A_B$  is identity matrix, solution is obvious, if non-basis variables are  $= 0$ , we keep only “ $A - A_B$ ”.  $z$  stands for the value of the objective function

**5:** Will  $z$  increase if we increase  $x_1$  or  $x_2$ ? How much we can increase  $x_2$  if  $x_1$  is kept at zero?

**Solution:** Both  $x_1$  and  $x_2$  have positive coefficient in the objective function. So increasing any of them will result in increase of the objective function.

If we are increasing  $x_2$ , then  $x_3$  and  $x_5$  are decreasing. So we cannot make  $x_2 > 1$ , otherwise  $x_3 < 0$ , which is not a feasible solution. Note that increasing  $x_2$  to 1 will make  $x_3 = 0$ .

**6:** Increase  $x_2$  as much as you can put it in the base. Use steps like in Gauss elimination to have  $x_2$  instead of  $x_3$  in the left top corner and nowhere else in the tableau. Note that the base will change to  $B = \{2, 4, 5\}$ .

**Solution:**

$$\begin{array}{rclcl} x_3 & = & 1 & + & x_1 & - & x_2 \\ x_4 & = & 3 & - & x_1 & & \\ x_5 & = & 2 & & & - & x_2 \\ \hline z & = & 0 & + & x_1 & + & x_2 \end{array} \sim \begin{array}{rclcl} x_3 & = & 1 & + & x_1 & - & x_2 \\ x_4 & = & 3 & - & x_1 & & \\ x_5 - x_3 & = & 1 & - & x_1 & & \\ \hline z + x_3 & = & 1 & + & 2x_1 & & \end{array} \sim \begin{array}{rclcl} x_2 & = & 1 & + & x_1 & - & x_3 \\ x_4 & = & 3 & - & x_1 & & \\ x_5 & = & 1 & - & x_1 & + & x_3 \\ \hline z & = & 1 & + & 2x_1 & - & x_3 \end{array}$$

The process when one variable is entering the base and another is leaving is called the **pivot step**.

**7:** Do another pivot step. Which of the variables in the objective function could be increased next? Increase it as much as possible and do a swap in the tableau as happened for  $x_2$  and  $x_3$ .

**Solution:** Since  $x_3$  has negative coefficient, we should not increase it. But  $x_2$  could be increased. The equation that limits the increase of  $x_5$  is the third one. Hence we take  $x_1$  from there.

$$\begin{array}{rclcl} x_2 & = & 1 & + & x_1 & - & x_3 \\ x_4 & = & 3 & - & x_1 & & \\ x_5 & = & 1 & - & x_1 & + & x_3 \\ \hline z & = & 1 & + & 2x_1 & - & x_3 \end{array} \sim \begin{array}{rclcl} x_2 + x_5 & = & 2 \\ x_4 - x_5 & = & 2 & & - & x_3 \\ x_5 & = & 1 & - & x_1 & + & x_3 \\ \hline z + 2x_5 & = & 3 & & + & 2x_3 \end{array} \sim \begin{array}{rclcl} x_2 & = & 2 & - & x_5 \\ x_4 & = & 2 & + & x_5 & - & x_3 \\ x_1 & = & 1 & - & x_5 & + & x_3 \\ \hline z & = & 3 & - & 2x_5 & + & x_3 \end{array}$$

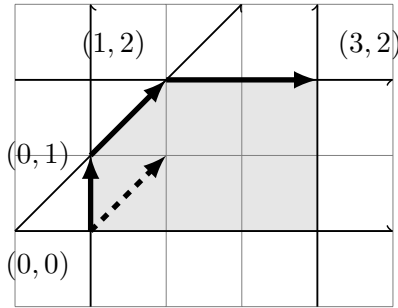
**8:** Can you do more pivot steps or is this the optimal solution? When is solution optimal?

**Solution:** We can still increase  $x_3$  and increase  $z$ . The bound is  $x_4$ .

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

Now we have the optimal solution, because increase in any of the variables will not result in increase of the objective function.

**9:** Draw the polytope of feasible solutions of program  $(P)$  (the original program in 2 variables  $x_1$  and  $x_2$ . Mark points that correspond to the steps of the solutions using simplex method and draw the direction of the objective function.



**10:** Use simplex method on the following example:

$$(P) \begin{cases} \text{maximize} & x_2 \\ \text{subject to} & -x_1 + x_2 \leq 0 \\ & x_1 \leq 2 \\ & x_1, x_2 \geq 0 \end{cases}$$

That is, convert to the equational form and do iterations until optimum solution is reached.

**Solution:**

$$\begin{array}{rcl} x_3 = 0 & x_1 - x_2 & x_2 = 0 \\ x_4 = 2 - x_1 & & x_1 - x_3 \\ \hline z = 0 & x_2 & \end{array} \sim \begin{array}{rcl} x_2 = 0 & x_1 - x_3 & x_1 = 2 \\ x_4 = 2 - x_1 & & - x_4 \\ \hline z = 0 & x_1 - x_3 & \end{array} \sim \begin{array}{rcl} x_1 = 2 & & - x_4 \\ x_2 = 2 - x_3 - x_4 & & \\ \hline z = 2 - x_3 - x_4 & & \end{array}$$

**11:** What happens with the objective function in the first step?

**Solution:** Nothing. It is staying zero.

- Simplex tableau is usually written in a matrix form (more condensed).
- There are versions - revised simplex method, dual simplex method for minimization, ...
- It is possible to construct an example that simplex method will cycle and never find a solution, if the pivot is chosen badly.
- Smart choice of pivot (Band's pivot rule - lexicographic rule) avoids cycling.
- There are many choices of pivot rules.
- polytopes may have many vertices (see cyclic polytope) but there is a chance of short path between any two vertices (initial solution and optimal solution) - recall Hirsh's conjecture.
- pivot rules can be tricked to walk through all vertices of cube (Klee-Minty cube)