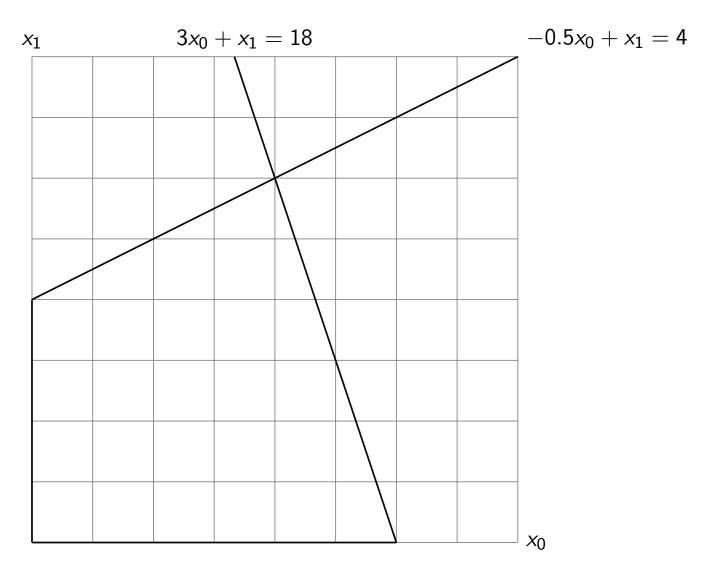
#### Contents Lecture 2

- Course lab theme and project: integer program solver
- You will implement this during labs 1 4, with labs 5 and 6 as reserve
- The labs have other contents as well, such as
  - the gdb debugger
  - valgrind
  - operf, gprof, gcov
  - POWER8 pipeline simulator
  - optimizing compilers

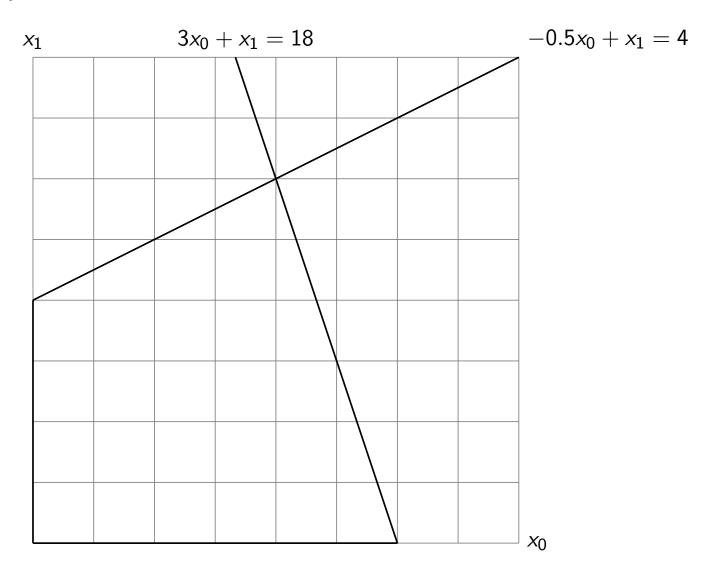
# Linear program: maximize a linear function in a region

- a region defined by lines including  $x_i \ge 0$ , i.e. linear constraints, and
- an objective function such as  $\max z = x_0 + 2x_1$



# Solving a linear program

- Find  $x_i \in \mathbb{R}$  which maximizes z
- Here  $x_i$  are called decision variables



## Our example

ullet Objective function and linear constraints using  $\leq$ 

$$\max \ z = x_0 + 2x_1$$

$$-0.5x_0 + x_1 \le 4$$
  
 $3x_0 + x_1 \le 18.$ 

- Also: implicitly  $x_i \ge 0$
- We can use e.g.  $x_i \ge 4$  or  $x_i = 5$  but they can be rewritten to use  $\le$

### Linear programs

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$$\max z = c_0 x_0 + c_1 x_1 + \dots + c_{n-1} x_{n-1} 
a_{0,0} x_0 + a_{0,1} x_1 + \dots + a_{0,n-1} \le b_0 
a_{1,0} x_0 + a_{1,1} x_1 + \dots + a_{1,n-1} \le b_1 
\dots 
a_{m-1,0} x_0 + a_{m-1,1} x_1 + \dots + a_{m-1,n-1} \le b_{m-1} 
x_0, x_1, \dots, x_{n-1} \ge 0$$

or simpler as

$$\begin{array}{ccc}
\max & z = cx \\
& Ax & \leq b \\
& x & \geq 0.
\end{array}$$

#### Solutions

- Each constraint defines a halfplane in *n* dimensions.
- The intersection of these halfplanes defines the **feasible region**, P, with **feasible solutions**  $x \in P$ .
- The feasible region is convex, and a point where halfplanes intersect is called a **vertex**.
- A linear program is either:
  - infeasible when P is empty,
  - unbounded when no finite solution exists, or
  - **feasible**, in which case we search for an optimal solution  $\mathbf{x}^* \in \mathbf{P}$  which maximizes z.
- There may exist more than one optimal solution.

# Local and global optimal solutions

- We denote by z(x) the value of the objective function z at point x.
- A solution x is **local optimum** for z(x) if there is an  $\epsilon > 0$  such that  $z(x) \ge z(y)$  for all  $y \in P$  with  $||x y|| \le \epsilon$ .

#### Theorem

A local optimum of a linear program is also a global optimum.

#### Theorem

For a bounded feasible linear program with feasible region P, at least one vertex is an optimal solution.

 So we only need to check z in the vertices and not the inner part of the region.

### Slack form

$$x_{n+0} = b_0 - \sum_{j=0}^{n-1} a_{0,j} x_j$$

$$x_{n+1} = b_1 - \sum_{j=0}^{n-1} a_{1,j} x_j$$

$$\dots$$

$$x_{n+m-1} = b_{m-1} - \sum_{j=0}^{n-1} a_{m-1,j} x_j$$

$$x_i \geq 0 \qquad 0 \leq i \leq n+m-1$$

$$(1)$$

- The variables on the left hand side are called basic variable and occur only once, i.e. neither in any sum on the right hand side, nor in the objective function.
- The other variables are called nonbasic variables.

2020

## Slack form of our example

We start with

$$\max z = x_0 + 2x_1$$
 $-0.5x_0 + x_1 \le 4$ 
 $3x_0 + x_1 \le 18$ 

 and then introduce two new variables, one for each constraint, and write it on slack form:

$$\max z = x_0 + 2x_1 + y$$

$$x_2 = 4 - (-0.5x_0 + x_1)$$

$$x_3 = 18 - (3x_0 + x_1).$$

- All  $x_i \ge 0$  and y is initially zero
- We rewrite the problem until all coefficients in the objective function become negative, and set all nonbasic variables to zero

## Entering and leaving basic variables

- Select a nonbasic variable with positive c<sub>i</sub> coefficient
- We take nonbasic variable  $x_0$  as the so called entering basic variable

max 
$$z = x_0 + 2x_1 + y$$

$$x_2 = 4 - (-0.5x_0 + x_1)$$
  
 $x_3 = 18 - (3x_0 + x_1)$ .

- Since  $c_0$  is positive, we want to increase  $x_0$  as much as possible
- The basic variables can limit how much  $x_0$  may be increased (if there is no restriction, then the linear program is unbounded)
- $x_3$  restricts increasing  $x_0$  to at most 6.
- Therefore we select  $x_3$  as the so called **leaving basic variable**.

### Rewritten linear program

- We rewrite the linear program by letting the entering and leaving basic variables switch roles.
- This is a tedious but simple algebraic manipulation
- Do this by hand at least once

$$\max z = -0.333x_3 + 1.667x_1 + 6$$

$$x_2 = 7 - (0.167x_3 + 1.167x_1)$$
  
 $x_0 = 6 - (0.333x_3 + 0.333x_1)$ 

- Next we must select  $x_1$  as entering basic variable
- $x_2$  is restricted by  $7-1.167x_1 \geq 0$
- $x_0$  is restricted by  $6-0.333x_1 \geq 0$
- x<sub>2</sub> is most restricted and becomes the leaving basic variable

#### Solution

- All  $c_i$  are negative so z cannot be increased with positive values of the nonbasic variables.
- By setting the nonbasic variables to zero the maximum becomes 16 in x = (4, 6) which indeed is a vertex.

max 
$$z = -0.6x_3 - 1.4x_2 + 16$$

$$x_1 = 6 - (0.1x_3 + 0.9x_2)$$
  
 $x_0 = 4 - (0.3x_3 - 0.3x_2)$ 

- Summary: we start in a vertex and then go to a neighboring vertex until all coefficients are negative, which gives the optimal solution.
- It was an open problem but George Dantzig was late for a lecture at Berkeley and mistook it for a home assignment (he got a PhD for it).

#### More issues

- The nonbasic variables are always set to zero
- The basic variables are always set to the corresponding  $b_i$
- If the point **0** is not in the feasible region we cannot use it as the start vertex.
- This happens if some  $b_i$  is negative
- We will look into that later in the course.

### Integer programming

- Integer programming is similar to linear programming with the extra condition that  $x_i \in \mathbb{N}$ .
- Some problems including this have no efficient algorithms
- One bad "method" to solve problems is to enumerate all solutions
- This does not sound good though
- We will use the algorithm design paradigm branch-and-bound to solve integer programs (not all due to integer programming is NP-complete)

#### Branch

- A relaxation makes a problem simpler (by solving another problem)
- For integer programming we solve the corresponding linear program, i.e. relaxing the integer requirement on the solution.
- Suppose we have an integer program and give it to the Simplex algorithm and  $x_k \notin \mathbb{N}$
- Assume the Simplex algorithm assigns  $x_k = u$
- We can then branch by creating two new linear programs:
  - one with the additional constraint constraint  $x_k \leq \lfloor u \rfloor$ , and
  - another with the additional constraint  $x_k \geq \lceil u \rceil$ .
- Each new problem is solved directly with the Simplex algorithm
- If it has an integer solution we can limit the search tree (bound)
- If it has a non-integer solution and it is better than best the integer solution we put it in the queue

### Bound

- If the Simplex algorithm found an integer solution we do the following
- We check if this is the best integer solution found so far, and remember it in that case
- We remove from the queue all unexplored linear programs whose optimal value is less than the value of the integer solution we just found

## Pseudo code and project

- The distributed pseudo code is in Appendix B in the book printed 2020
- It is as simple possible
- It is your task to translate it to C and then to optimize it.
- The only requirement is that it should work in some selected problems
- Not even commercial programs work perfectly
- At forsete.cs.lth.se you will be able to upload your C code
- Forsete is an automatic grader and will give you a score
- The score determines who win a coffee mug and does not affect your grade
- You are allowed to use any technique except running a different solver
- You are not allowed to use any open source code: only the C Standard library and code you have written yourself