

Conclusion from last time:

We know how to compute all solutions to  $Ax=b$ ,  $A$  a wide matrix, in the specific form  $x = x^{\text{basic}} + x^{\text{hom}}$ , where  $x^{\text{hom}}$  solves  $Ax^{\text{hom}}=0$ , and  $x^{\text{basic}}$  has at least as many 0 entries as "number of columns" minus "number of pivots".

= total number of variables in standard form LP problem, i.e.,  $x_1, x_2, \dots, s_1, s_2, \dots, u, v, \dots$

Now back to the standard form of LP problems: • minimize  $z = c^T x$

• constraints:  $Ax = b$  and  $x \geq 0$

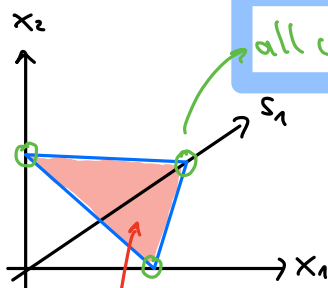
How does the feasible region look like?

$Ax=b$  describes an affine subspace, e.g., a plane in 3d.

The feasible region is that part of the subspace with  $x \geq 0$ !

= "quadrant" where all components are nonnegative (i.e., positive or zero)

E.g.:



all corner points are basic feasible solutions

feasible region has shape of a "simplex"

Important insight: If there is an optimal solution, we can always find one at a cornerpoint. And the cornerpoints correspond to the basic solutions of  $Ax=b$ .

These we can find with Gaussian elimination.

A feasible region like in the picture could arise from, e.g.,  $A = \underbrace{(5, 3, 4)}_{1 \times 3 \text{ matrix}}$

and  $b = \underline{2}$ .  
vector with 1 component

$\Rightarrow$  augmented matrix:  $\begin{matrix} x_1 & x_2 & s_1 \\ \downarrow & \downarrow & \downarrow \\ (5 & 3 & 4 \mid 2) \end{matrix}$

The possibilities for basic solutions are:

• pivot in column 1:  $R_1/5 \Rightarrow \left( \begin{array}{ccc|c} 1 & \frac{3}{5} & \frac{4}{5} & \frac{2}{5} \end{array} \right)$

$\Rightarrow x_1^{\text{basic}} = \begin{pmatrix} \frac{2}{5} \\ 0 \\ 0 \end{pmatrix}$

• pivot in column 2:  $R_1/3 \Rightarrow \left( \begin{array}{ccc|c} \frac{5}{3} & 1 & \frac{4}{3} & \frac{2}{3} \end{array} \right)$

$\Rightarrow x_2^{\text{basic}} = \begin{pmatrix} 0 \\ \frac{2}{3} \\ 0 \end{pmatrix}$

• pivot in column 3:  $R_1/4 \Rightarrow \left( \begin{array}{ccc|c} \frac{5}{4} & \frac{3}{4} & 1 & \frac{1}{2} \end{array} \right)$

$\Rightarrow x_3^{\text{basic}} = \begin{pmatrix} 0 \\ 0 \\ \frac{1}{2} \end{pmatrix}$

Now suppose  $c = \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix}$ .

Then we check:  $z_1 = c^T x_1^{\text{basic}} = (1, 2, 1) \begin{pmatrix} \frac{2}{5} \\ 0 \\ 0 \end{pmatrix} = 1 \cdot \frac{2}{5} + 2 \cdot 0 + 1 \cdot 0 = \frac{2}{5}$

$z_2 = c^T x_2^{\text{basic}} = (1, 2, 1) \begin{pmatrix} 0 \\ \frac{2}{3} \\ 0 \end{pmatrix} = 1 \cdot 0 + 2 \cdot \frac{2}{3} + 1 \cdot 0 = \frac{4}{3}$

$z_3 = c^T x_3^{\text{basic}} = (1, 2, 1) \begin{pmatrix} 0 \\ 0 \\ \frac{1}{2} \end{pmatrix} = 1 \cdot 0 + 2 \cdot 0 + 1 \cdot \frac{1}{2} = \frac{1}{2}$

$\Rightarrow z_1$  is the smallest, i.e.,  $x_1^{\text{basic}}$  is our optimal solution!

Let us write down our main insight more precisely:

Theorem: If a standard form LP problem has optimal solutions, then there is an optimal basic solution (i.e., an optimal solution that is a vertex/corner of the feasible region).

Proof idea:

Suppose  $x$  is optimal, but not a cornerpoint. Then there is always a vector  $v$  such that both  $x+v$  and  $x-v$  are still feasible. In fact, since  $x$  minimizes  $c^T x$ :

$$\left. \begin{array}{l} \cdot c^T x \leq c^T(x+v) \Rightarrow c^T v \geq 0 \\ \cdot c^T x \leq c^T(x-v) \Rightarrow c^T v \leq 0 \end{array} \right\} \Rightarrow c^T v = 0,$$

so  $x+v$  and  $x-v$  are also optimal!

Then: go along either  $v$  or  $-v$  until we hit the boundary of the feasible region, i.e., until one component becomes negative.

$\Rightarrow$  Repeat until we get stuck at a corner (where either  $v$  or  $-v$  will lead out of the feasible region). □