

assumptions:  $A$  is an  $m \times n$  matrix,  $\text{rank}(A)=m$ ,  $A=[a_1, \dots, a_n]$  (so that  $a_i$  is the  $i^{\text{th}}$  column of  $A$ ).

A solution  $x$  to  $Ax=b$  is basic if  $\{a_i : x_i \neq 0\}$  is linearly independent.

Note: The equation  $Ax=b$  has at most  $\binom{n}{m}$  basic solutions (thus finitely many). Consider the linear programming problem

$$(LP) \quad \max c'x : Ax=b, x \geq 0.$$

$x \in \mathbf{R}^n$  is feasible for (LP) if  $Ax=b$  and  $x \geq 0$ .  $x \in \mathbf{R}^n$  is optimal for (LP) if  $x$  is feasible for (LP) and

$$(Ay=b \text{ and } y \geq 0) \Rightarrow c'x \geq c'y$$

Fundamental Theorem: If (LP) has an optimal solution then it has a basic optimal solution.

Proof: Suppose that (LP) does have an optimal solution. Let  $p$  be the smallest integer for which an optimal solution exists having  $p$  nonzero entries. Choose an optimal solution  $x$  having  $p$  nonzero entries. Without loss of generality, we can write  $x^t = (x_1, \dots, x_p, 0, \dots, 0)$  with  $x_1 > 0, \dots, x_p > 0$ . Let  $\mu = c'x$  be the optimal value of (LP).

Our goal is to show that (LP) has a basic optimal solution. If  $x$  itself is basic, there is nothing to show, so assume it is not. The proof will be complete when we have shown that this assumption leads us to a contradiction.

Since  $x$  is not basic,  $\{a_i : x_i \neq 0\}$  is linearly dependent, meaning that there are numbers  $y_1, \dots, y_p$ , not all zero, such that  $y_1 a_1 + \dots + y_p a_p = 0$ .

Equivalently, we can write  $Ay=0$ , where  $y^t = (y_1, \dots, y_p, 0, \dots, 0)$ .

There is some positive number  $\epsilon$  such that

$$x_1 + ty_1 > 0, \dots, x_p + ty_p > 0 \text{ whenever } -\epsilon < t < \epsilon.$$

Since  $Ay=0$  and  $Ax=b$ , it follows that

$$A(x+ty)=b \text{ for all } t \in \mathbf{R}.$$

Combining these two observations, it follows that

$$x+ty \text{ is feasible for (LP) whenever } -\epsilon < t < \epsilon.$$

Since  $x$  is optimal for (LP), we have

$$c'(x+ty) \leq c'x = \mu \text{ whenever } -\epsilon < t < \epsilon.$$

But then  $t(c'y) \leq 0$  whenever  $-\epsilon < t < \epsilon$ , which implies  $c'y=0$ .

Choose  $t^*$  as close to zero as possible so that  $x+t^*y$  has  $p-1$  nonzero entries. Then  $x+t^*y \geq 0$  and  $A(x+t^*y)=b$  so  $x+t^*y$  is feasible for (LP). Also,  $c'(x+t^*y) = c'x = \mu$ , so  $x+t^*y$  is an optimal solution to (LP). This, however, contradicts that  $p$  is the smallest number for which an optimal solution exists having  $p$  nonzero entries.