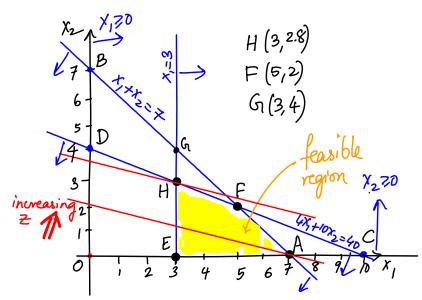
#### MATH 364: Lecture 10 (09/19/2024)

\* correspondence between bes's & corner points
\* Simplex method

#### Farmer Jones LP

max 
$$z = 30x_1 + 100x_2$$
  
S.t.  $x_1 + x_2 + s_1 = 7$   
 $4x_1 + 10x_2 + s_2 = 40$   
 $10x_1 - l_3 = 30$   
 $x_1, x_2, s_1, s_2, l_3 = 30$ 



We saw  $BV = \{x_1, x_2, 8, \}$ ,  $NBV = \{82, 83\}$  gives the  $BS \equiv H(3, 2.8)$ .

For BV =  $\{x_2, 8_2, e_3\}$ , NBV =  $\{x_1, 8_3\}$  gives the basic solution  $= \begin{cases} x_1 & \text{opposite solution} \\ x_1 & \text{opposite solution} \end{cases}$   $= \begin{cases} x_1 & \text{opposite solution} \\ x_2 & \text{opposite solution} \end{cases}$   $= \begin{cases} x_2 & \text{opposite solution} \\ x_2 & \text{opposite solution} \end{cases}$   $= \begin{cases} x_2 & \text{opposite solution} \\ x_2 & \text{opposite solution} \end{cases}$   $= \begin{cases} x_2 & \text{opposite solution} \\ x_2 & \text{opposite solution} \end{cases}$   $= \begin{cases} x_2 & \text{opposite solution} \\ x_2 & \text{opposite solution} \end{cases}$   $= \begin{cases} x_2 & \text{opposite solution} \\ x_2 & \text{opposite solution} \end{cases}$   $= \begin{cases} x_2 & \text{opposite solution} \\ x_2 & \text{opposite solution} \end{cases}$   $= \begin{cases} x_2 & \text{opposite solution} \\ x_2 & \text{opposite solution} \end{cases}$   $= \begin{cases} x_2 & \text{opposite solution} \\ x_2 & \text{opposite solution} \end{cases}$   $= \begin{cases} x_2 & \text{opposite solution} \\ x_2 & \text{opposite solution} \end{cases}$   $= \begin{cases} x_2 & \text{opposite solution} \\ x_2 & \text{opposite solution} \end{cases}$   $= \begin{cases} x_2 & \text{opposite solution} \\ x_2 & \text{opposite solution} \end{cases}$   $= \begin{cases} x_2 & \text{opposite solution} \\ x_2 & \text{opposite solution} \end{cases}$   $= \begin{cases} x_2 & \text{opposite solution} \\ x_2 & \text{opposite solution} \end{cases}$   $= \begin{cases} x_2 & \text{opposite solution} \\ x_2 & \text{opposite solution} \end{cases}$   $= \begin{cases} x_2 & \text{opposite solution} \\ x_2 & \text{opposite solution} \end{cases}$   $= \begin{cases} x_2 & \text{opposite solution} \\ x_2 & \text{opposite solution} \end{cases}$   $= \begin{cases} x_2 & \text{opposite solution} \\ x_2 & \text{opposite solution} \end{cases}$   $= \begin{cases} x_2 & \text{opposite solution} \\ x_2 & \text{opposite solution} \end{cases}$   $= \begin{cases} x_2 & \text{opposite solution} \\ x_2 & \text{opposite solution} \end{cases}$ 

$$\begin{cases} X_2 = 7 \\ 10X_2 + 8_2 = 40 \\ -\ell_3 = 30 \end{cases} \implies X_2 = 7, \ \ell_3 = -30, \ S_2 = -30.$$

We could identify the best corresponding to each corner point directly from the picture!

Let's consider f(5,2).

 $x_1=5$ ,  $x_2=2$  are in BV.

At F, the (land-avai) and (labor\_hrs)

constraints are binding (i.e., satisfied as equalities). Hence

 $B_1=0$  and  $B_2=0$ . But the

(min corn) constraint is non-binding

at F, hence ezzo. Hence

and the LR is  $\overline{X} = \begin{array}{c} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \\ x_7 \\ x_8 \\ x_9 \\ x_$ 

We can present the correspondence between corner points and best's in a table as shown below.

## Correspondence between Les's and correr points

Corner point	BV	NBV	[x, x <sub>2</sub> 8, s <sub>2</sub> e <sub>3</sub> ]
A (7,0)	X1, P3, 82	X2, B1	[7001240]
E (3,0)	X1,81,82		[3 0 4 28 0]
H(3, 2.8)	X1, X2, 8,	82, 63	[3 2.8 1.2 0 0]
F(5,2)	x,,x2,l3	81, 82	[5 2 0 0 20]

Let's summarize a bit. We have seen the following results.

\* If the LP in standard form has an optimal solution, there must be a corner point that is optimal.

\* corner points \$\implies bfs

Hence we get the following result.

Theorem If an LP in standard form has an optimal solution, then it has an optimal bfs.

The simplex method explores the corner points, or bfs. The idea is to start at one bfs, and move to a neighboring bfs (or corner point) at which the objective function is better. This procedure is equivalent to sliding the Z-line in 2D.

While the idea of a neighboring corner point is straightforward to imagine in 2D, we switch to an algebraic view in higher dimensions using the correspondence given above.

#### Simplex Method

\* Start at a best = corner point.

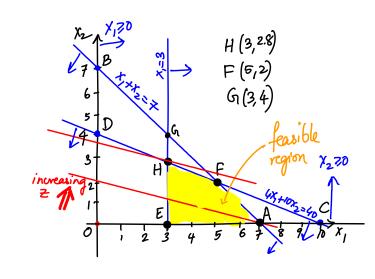
\* If not optimal, move to a "nearby" (adjacent) lofs so that the z-value improves

\* If no such "better" adjacent bls exists, the current corner point = bls is optimal.

### Adjacent bfs

Def for an LP in standard form with n variables and m constraints  $(m \le n, rank(A) = m)$ , two lots's are said to be adjacent if they have (m-1) common basic variables.

Corner point	₿V	NBV	BFS [x, x2 8, s2 e3]
A (7,0)	(X1) (3, (82)	X2, B1	[7001240]
E (3,0)	(X), 81, 82	X2, l3	[3 0 4 28 0]
H(3, 2.8)	X, X2, 8	82, 83	[3 2.8 1.2 0 0]
F (5,2)	X, X2, 63	81, 82	[5 2 0 0 20]



For instance, we could start at A(7,0), where Z=210, and more to the adjacent bfs F(5,2), where Z=350. Notice that the other adjacent bfs to A is E(3,0). But Z=90 at E, and hence we do not move to E.

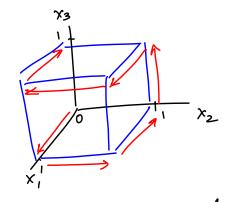
Repeating the same procedure, we move from F to the adjacent Corner points Lfs H(3, 2.8), where Z=370. From H, both adjacent corner points (F and E) have smaller 2-values, and hence we can conclude that H is an optimal solution.

We could, alternatively start at E and move to H directly.

Could some UP by evaluating 2 at every bls (corner point). What is the max # bls's an LP can have?

For each both we need to choose m basic variables out of n variables. Hence there are  $\binom{n}{m} = \frac{n!}{m!(n-m)!}$  choices, which would be quite large! But not all of these choices might lead to a bife

On the one hand, there are artificially constructed LP instances for which every version of the simplex algorithm would have to inspect all of (m) bigs (all of them being feasible, of course).



2" vertices in general!

But on the other hand, most LPs arisine out of applicatione tend not to exhibit such structure and applicatione tend not to exhibit such structure and the simplex method is usually very fact in solving most LPs.

# Simplex Method

# Simplex Algorithm for maximization Us

Step 1 Convert LP to standard form.

Step 2 Obtain a bits from the standard form.

Step 3 Find if current by is optimal.

If YES, STOP.

Step 4 9¢ current bles is not optimal, find which non-basic variable should become basic, and which basic variable should become non-basic in order to move to an adjacent bles with a higher objective function value.

Step 5 Use EROS to Obtain the adjacent bfs.
Return to Step 3.

We specify more details for each step as we illustrate the simplex algorithm on an example. We start with an LP where all constraints are '\(\leq'\). Step 2 becomes easy in this case. We will discuss how to deal with \(\geq'\) and \(\leq'\) constraints later on. We will assume also that all variables are non-negative for now.

## Solve the following LP noing the simplex method

$$\max_{S:t} z = 2x_1 + 3x_2$$

$$S:t \cdot x_1 + 2x_2 \le 6 \quad 8_1 = 0$$

$$2x_1 + x_2 \le 8 \quad 8_2 = 0$$

$$x_1, x_2 = 0$$

max 
$$z = 2x_1 + 3x_2$$
  
S.t.  $x_1 + 2x_2 + 8_1 = 6$   
 $2x_1 + x_2 + 8_2 = 8$   
 $x_1, x_2, s_1, s_2 = 0$ 

We first write the LP in a more organized manner:

0. 
$$(\frac{2}{x_1} - 2x_1 - 3x_2)$$
  
1.  $(\frac{2}{x_1} + 2x_2 + 3)$   
2  $(\frac{2}{x_1} + x_2)$  = 8 Canonical form

Def An LP is written in canonical form if each row including Row-0 has a variable (including Z) with coefficient 1 in that row and zero in every other row.

Here, 22,8,827 is the set of canonical variables.

We can choose z and the remaining m canonical variables in the starting logs. If the rhs (bi for the ith constraint) values are all =0, we can read off the lofs from the canonical form. Here, we set  $x_1=x_2=0$ , and get  $s_1=6$ ,  $s_2=8$ , and z=0.