

Q1

2) standardizing, we get

$$\begin{aligned} \min z &= -x_1 - x_2 \\ \text{st } &\begin{cases} -x_1 + x_2 + x_3 = 2 \\ -2x_1 + 3x_2 - x_4 = 1 \\ x_1, x_2, x_3, x_4 \geq 0 \end{cases} \quad \text{--- (1)} \end{aligned}$$

Adding an artificial variable  $z_1$  to the LHS of (1), we solve the auxiliary LP by minimizing

$$Z = z_1 = 1 + 2x_1 - 3x_2 + x_4$$

Initial basic representation:

	$x_1$	$x_2$	$x_3$	$x_4$	$z_1$	RHS	Ratio
$z$	-2	3		-1		1	
$x_3$	-1	1	1			2	2
$z_1$	-2	(3)		-1	1	1	$\frac{1}{3}$
$z$					-1	0	
$x_3$	$-\frac{1}{3}$		1	$\frac{1}{3}$	$-\frac{1}{3}$	$\frac{5}{3}$	
$x_2$	$-\frac{2}{3}$	1		$-\frac{1}{3}$	$\frac{1}{3}$	$\frac{1}{3}$	

$\therefore$  LP feasible  $\Rightarrow$  ~~start~~ start phase 2 using index set  $I = \{2, 3\}$   
we want to minimize  $z = -x_1 - x_2 = -x_1 - (\frac{1}{3} + \frac{2}{3}x_1 + \frac{1}{3}x_4)$

$$= -\frac{1}{3} - \frac{5}{3}x_1 - \frac{1}{3}x_4$$

	$x_1$	$x_2$	$x_3$	$x_4$	RHS	Ratio
$z$	$\frac{5}{3}$			$\frac{1}{3}$	$-\frac{1}{3}$	
$x_3$	$-\frac{1}{3}$		1	$\frac{1}{3}$	$\frac{5}{3}$	
$x_2$	$-\frac{2}{3}$	1		$-\frac{1}{3}$	$\frac{1}{3}$	

Now the coefficient of  $x_1$  in the objective row is  $\frac{5}{3} > 0$ , but its coefficient in the  $x_3$  &  $x_2$  rows are negative

$\Rightarrow$  by the simplex algorithm, we know that the LP is unbounded.

$\therefore$  No optimal solution exists for the LP. ~~and~~  
(max of  $y$  tends to  $\infty$ ).

b) Let  $x_a = \sum_{k=1}^N \alpha_k x^{(k)}$  be an arbitrary weighted average of the  $N$  optimal basic feasible solutions. (BFS).

### Feasibility

Every  $x^{(k)}$  satisfies  $x^{(k)} \geq 0$  &  $Ax^{(k)} = b \quad \forall k \in \{1, \dots, N\}$

$$\begin{aligned} \Rightarrow Ax_a &= A \sum_{k=1}^N \alpha_k x^{(k)} = \sum_{k=1}^N \alpha_k [Ax^{(k)}] \\ &= \sum_{k=1}^N \alpha_k b \\ &= b \quad \text{since } \left( \sum \alpha_k = 1 \right). \end{aligned}$$

$$\begin{aligned} \text{Also, } x^{(k)} \geq 0 &\Rightarrow \alpha_k x^{(k)} \geq 0 \quad \forall k \in \{1, \dots, N\} \quad \text{since } \alpha_k \geq 0. \\ &\Rightarrow x_a = \sum_{k=1}^N \alpha_k x^{(k)} \geq 0 \end{aligned}$$

$\therefore x_a$  is in the feasible set of the LP.

### Optimality

We know the optimal objective value is  $c^T x^{(1)} = c^T x^{(2)} = \dots = c^T x^{(N)}$ .

$$\begin{aligned} \text{Now } c^T x_a &= c^T \sum_{k=1}^N \alpha_k x^{(k)} = \sum_{k=1}^N \alpha_k [c^T x^{(k)}] \\ &= \sum_{k=1}^N \alpha_k [c^T x^{(1)}] \\ &= c^T x^{(1)} \quad \left( \text{since } \sum \alpha_k = 1 \right) \end{aligned}$$

$\therefore$  objective value corresponding to  $x_a$  is the same as optimal objective value  $\Rightarrow x_a$  is an optimal solution!

$\therefore$  shown that ~~any~~ an arbitrary ~~way~~  $x_a$  is both feasible & optimal  $\Rightarrow$  every such  $x_a$  is feasible & optimal!

c) i) True This is demonstrated by the Klee-Minty problem of  $n$  variables, where the ~~any~~ number of iterations ~~of the~~ the simplex algorithm required in the worst case = number of 'vertices' of the Klee-Minty cube

$$\Rightarrow \text{worst case complexity exponential in } n. = 2^n,$$

ii) ~~True~~ let  $n = \dots$

ii) True. Let <sup>an</sup> ~~the~~ index set be  $I$ , decision variables  $= x_1, \dots, x_n$ ,  
 reduced cost of NBV  $x_j = r_j$  ( $j \notin I$ ), objective function  $= z$ .

Then we have:

$$z = a_0 + \sum_{j \notin I} r_j x_j, \text{ where } a_0 = \text{some constant}$$

If  $r_q = 0 \Rightarrow$  objective cannot change (i.e. increase/decrease)  
 by changing  $x_q$ !

iii) False

Consider the LP

$$\begin{aligned} \min \quad & x_1 \\ \text{s.t.} \quad & \begin{cases} x_1 \leq 1 \\ x_1 \geq 0 \end{cases} \end{aligned}$$

Then standardizing gives

$$\begin{aligned} \min \quad & x_1 \\ \text{s.t.} \quad & \begin{cases} x_1 + x_2 = 1 \\ x_1, x_2 \geq 0 \end{cases} \end{aligned}$$

where  $x_2 =$  slack ~~variable (only slack)~~ variable (the only 1)

Then taking index set  $= \{2\}$ , we have the basic solution  $(0, 1)$ ,  
 but ~~not~~ the value of the slack here is  $\neq 0$ !

iv) False.

It can be the case where the objective function of the phase-1  
 simplex algorithm is 0 with some artificial variables, being  
 basic (refer to Q5 of Tutorial 4) still

In this case, we would need to do additional pivots to  
 remove the artificial variables from the basic variables

(this process does not change the phase-1 objective value ~~if~~  
 if it was 0 to begin with)

~~phase-1 objective value~~

d)

$x_1, x_2 \geq 0$  &  $x_1 + 3x_2 = 1 \Rightarrow$  denominator  $x_1 + 3x_2$  is always positive in the feasible set.

Obviously (\*) implies feasible set is bounded as well.

$\Rightarrow$  Introduce  $y_0, y_1, y_2$  st  $x_1 = \frac{y_1}{y_0}, x_2 = \frac{y_2}{y_0}$  where  $y_0 > 0, y_1, y_2 \geq 0$

~~Then we have Homogenization~~ Homogenizing, we get the equivalent problem:  
~~Then, Normalizing the denominator, we get the equivalent problem:~~

$$\min \frac{\max \{y_1 + 3y_2, y_0 - 2y_2\}}{y_1 + y_2}$$

$$\text{st } \begin{cases} y_1 + 3y_2 = y_0 \\ y_0 > 0, y_1, y_2 \geq 0 \end{cases}$$

(using  $y_0 > 0$ )

Normalizing, we get the equivalent problem:

$$\min \left( \max \{y_1 + 3y_2, y_0 - 2y_2\} \right)$$

$$\text{st } y_1 + y_2 = 1$$

$$y_1 + 3y_2 = y_0$$

$$y_0 \geq 0, y_1, y_2 \geq 0$$

noting that we can loosen the constraint on  $y_0$ .

This is equivalent to the LP:

$$\min z$$

$$\text{st } \begin{cases} z \geq y_1 + 3y_2 \\ z \geq y_0 - 2y_2 \end{cases}$$

$$y_1 + y_2 = 1$$

$$y_1 + 3y_2 = y_0$$

$$y_0 \geq 0, z \text{ free}$$

Standardizing & letting  $z = z^+ - z^-$ , where  $z^+, z^- \geq 0$ , we get the equivalent LP:

$$\min z^+ - z^-$$

$$\text{st } y_1 + y_2 = 1$$

$$-y_0 + y_1 + 3y_2 = 0$$

$$y_1 + 3y_2 - z^+ + z^- + x_1 = 0$$

$$y_0 - 2y_2 - z^+ + z^- + x_2 = 0$$

$$x \geq 0, y \geq 0, z^+, z^- \geq 0$$

e)

let  $\delta_x = \begin{cases} 1 & \text{if } x \text{ is undertaken} \\ 0 & \text{otherwise} \end{cases}$  for  $x \in \{A, B, \dots, G\}$ .

let  $c_x$  &  $p_x$  = capital & profit of investment  $x$ , <sup>respectively</sup>  $\forall x \in \{A, \dots, G\}$ .

We want to solve:

$$\max \sum_{x \in \{A, \dots, G\}} \delta_x p_x$$

$$\text{s.t. } \begin{cases} \sum_{x \in \{A, \dots, G\}} \delta_x c_x \leq 10 \end{cases}$$

$$\delta_A + \delta_C \leq 1$$

$$\delta_B + \delta_D + \delta_E \leq 1$$

$$\delta_A + \delta_C \geq \delta_B + \delta_D$$

$$\delta_A \geq \delta_G$$

$$\delta_x \in \{0, 1\} \quad \forall x \in \{A, \dots, G\}$$

(Q2)

~~First check that B is invertible:~~

$$\det B = 9 - 5 \neq 0 \Rightarrow B \text{ invertible}$$

Hence  $I = \{1, 2\}$  is a valid index set, and its associated basis is B

~~Now suffice to check optimality: using notation in notes,~~

a) i) First standardize:

$$\min y = -5x_1 - 6x_2$$

$$\text{s.t. } \begin{cases} x_1 + x_2 + x_3 = p \\ 5x_1 + 9x_2 + x_4 = 45 \\ x \geq 0 \end{cases}$$

Note that B corresponds to the index set  $\{1, 2\} =: I$ .

$$\text{Now } \det B = 9 - 5 \neq 0 \Rightarrow B \text{ invertible}$$

$\Rightarrow I$  valid index set & B valid basis.

$\Rightarrow$  Suffice to check optimality. Using notation in notes,

$$\begin{aligned} \text{reduced cost } r &= c_N - N^T B^{-T} c_B \\ &= \begin{pmatrix} 0 \\ 0 \end{pmatrix} - \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \left[ \begin{pmatrix} 1 & -1 \\ 9 & -5 \end{pmatrix} \right]^T \begin{pmatrix} -5 \\ -6 \end{pmatrix} \\ &= -\frac{1}{4} \begin{pmatrix} 9 & -5 \\ -1 & 1 \end{pmatrix} \begin{pmatrix} -5 \\ -6 \end{pmatrix} \\ &= -\frac{1}{4} \begin{pmatrix} -15 \\ -1 \end{pmatrix} \\ &= \begin{pmatrix} 15/4 \\ 1/4 \end{pmatrix} \geq 0. \Rightarrow x_B \text{ optimal} \end{aligned}$$

$\Rightarrow$  Also, when  $p \geq 6$ ,

$$\begin{aligned} x_B &= \frac{1}{4} \begin{pmatrix} 9 & -1 \\ -5 & 1 \end{pmatrix} \begin{pmatrix} 6 \\ 45 \end{pmatrix} \\ &= \begin{pmatrix} 9/4 \\ 15/4 \end{pmatrix} \geq 0. \Rightarrow x_B \text{ feasible} \end{aligned}$$

$\Rightarrow I$  gives rise to an optimal basis feasible solution  
 $\Rightarrow \Delta_n$  optimal basis matrix when  $p \geq 6$  is B.

ii) For  $5 \leq p < 9$ ,

$$\begin{aligned} B^{-1} \begin{pmatrix} p \\ 45 \end{pmatrix} &= \frac{1}{4} \begin{pmatrix} 9 & -1 \\ -5 & 1 \end{pmatrix} \begin{pmatrix} p \\ 45 \end{pmatrix} \\ &= \frac{1}{4} \begin{pmatrix} 9p - 45 \\ 45 - 5p \end{pmatrix} \geq 0 \end{aligned}$$



Defining  $V'_{LP}(p) = \min -5x_1 - 6x_2$   
 st  $\begin{cases} x_1 + x_2 \leq p \\ 5x_1 + 9x_2 \leq 45 \\ x \geq 0 \end{cases}$ ,

we then have (by a result in the notes)

$$V'_{LP}(p) = V'_{LP}(6) + \Pi_1 (p-6), \text{ since } B^{-1} \begin{pmatrix} p \\ 45 \end{pmatrix} \geq 0.$$

where  $\Pi_1 =$  first component of the shadow price  $\Pi = (B^{-1})^T c_B$

$$= \frac{1}{4} \begin{pmatrix} 9 & -5 \\ -1 & 1 \end{pmatrix} \begin{pmatrix} -5 \\ -6 \end{pmatrix}$$

$$= \begin{pmatrix} -15/4 \\ -1/4 \end{pmatrix}.$$

~~$\Rightarrow V'_{LP}(p) =$~~

Also,  $V'_{LP}(6) = -5(9/4) - 6(15/4)$   
 $= -\frac{135}{4}.$

$\Rightarrow V'_{LP}(p) = -\frac{135}{4} - \frac{15}{4}(p-6)$   
 $= -\frac{1}{4}(15p + 45)$

Now  $V_{LP}(p) = -V'_{LP}(p)$

$= \frac{1}{4}(15p + 45).$  ~~for  $6 \leq p \leq 9$~~

b) i) ~~True~~ True We know for  $5 \leq p \leq 9$ , ~~the optimal~~  $\exists$  optimal solution for the LP, by (a).

Assuming the integral version of (1) ~~is~~ feasible for  $5 \leq p \leq 9$ , then since in the integral version we are taking a maximum over a smaller set compared to the LP, we have  $V_{LP}(p) \geq V_{zp}(p)$ .  $\forall p \in [5, 9]$ .

ii) True

Using the formula from (a), if  $5 \leq p_1 \leq p_2 \leq 9$

$\Rightarrow V_{LP}(p_2) = \frac{1}{4}(15p_2 + 45) \geq \frac{1}{4}(15p_1 + 45) = V_{LP}(p_1).$

iii) True

Taking  $p_1 = 6$  &  $p_2 = 9$ , we know  $V_{LP}(p_1) = V_{LP}(6)$   
 $= \frac{135}{4},$

and  $(9, 0)$  is an element in the feasible set of the zLP when  $p_2 = 9$  is  $(9, 0)$ , with objective value 45.

$\Rightarrow V_{zp}(9) \geq 45 > \frac{135}{4} = V_{LP}(6) \Rightarrow \text{True!}$

c) ~~Using (a), we know~~  
 ~~$x_2 = \frac{15}{4}$~~

Using notation from (a), we know

$$B^{-1}N = \frac{1}{4} \begin{pmatrix} 9 & -1 \\ -5 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \\ = \frac{1}{4} \begin{pmatrix} 9 & -1 \\ -5 & 1 \end{pmatrix}$$

$\Rightarrow$  Since  $x_B = B^{-1}b - B^{-1}N x_N$ , we know

$$x_2 = \frac{15}{4} + \frac{5}{4} x_3 - \frac{1}{4} x_4$$

$$\Leftrightarrow x_2 - \frac{5}{4} x_3 + \frac{1}{4} x_4 = \frac{15}{4}$$

$\therefore$  Gomory cut corresponding to  $x_2$  is

$$\left(-\frac{5}{4} - \left[-\frac{5}{4}\right]\right) x_3 + \left(\frac{1}{4} - \left[\frac{1}{4}\right]\right) x_4 \geq \left\lfloor \frac{15}{4} - \left[\frac{15}{4}\right] \right\rfloor$$

$$\Leftrightarrow \frac{3}{4} x_3 + \frac{1}{4} x_4 \geq \frac{3}{4}$$

$$\Leftrightarrow 3x_3 + x_4 \geq 3$$

Now  $x_3 = 6 - x_1 - x_2$  &  $x_4 = 45 - 5x_1 - 9x_2$ , so substituting:

$$3(6 - x_1 - x_2) + (45 - 5x_1 - 9x_2) \geq 3$$

$$\Leftrightarrow 8x_1 + 12x_2 \leq 60$$

$$\Leftrightarrow 2x_1 + 3x_2 \leq 15$$

$\uparrow$  Gomory cut in  $x_1, x_2$ .



d) i) It is evident that  $x^{LP} = (\frac{1}{2}, \alpha+1)$  satisfies

$$\begin{cases} x_1 \leq \frac{1}{2} \\ -2(\alpha+1)x_1 + x_2 \leq 0 \\ x_1, x_2 \geq 0 \end{cases}$$

Also,  $x_2 \leq 2(\alpha+1)x_1 \leq 2(\alpha+1)(\frac{1}{2}) = \alpha+1$

$\Rightarrow$  maximum value <sup>that</sup> of the objective can take  $= \alpha+1$ , which is satisfied by  $x^{LP}$ .

$\Rightarrow x^{LP}$  optimal.

Now suppose  $x^* = (x_1^*, x_2^*)$  is another optimal solution.

Then objective value corresponding to  $x^* = \alpha+1$

$$\Rightarrow x_2^* = \alpha+1$$

$$\Rightarrow -2(\alpha+1)x_1^* + (\alpha+1) \leq 0$$

$$\Rightarrow x_1^* \geq \frac{1}{2}$$

$$\text{But } x_1^* \leq \frac{1}{2} \Rightarrow x_1^* = \frac{1}{2} \Rightarrow x^* = x^{LP}.$$

Thus  $x^{LP}$  is the unique optimal solution

ii)  $x_1 \leq \frac{1}{2}$  &  $x_1 \in \mathbb{N} \Rightarrow x_1 = 0.$

$$-2(\alpha+1)x_1 + x_2 \leq 0 \Rightarrow x_2 \leq 0 \quad \text{since } x_1 = 0.$$

$$\Rightarrow x_2 = 0 \quad \text{since } x_2 \in \mathbb{N}.$$

Hence the only element in the  $\mathbb{Z}P$  is  $(0,0)$ .  
feasible set of the

$$\Rightarrow x^{IP} = (0,0).$$

iii) Direct computation:  $x_2^{LP} - x_2^{IP} = \alpha+1 - 0 = \alpha+1.$

iv) No.  $\|x^{LP} - x^{IP}\| = \sqrt{(\alpha+1)^2 + \frac{1}{4}}$

$$= \sqrt{(\alpha+1)^2 + \frac{1}{4}}$$

$$\approx \alpha+1 \quad \text{when } \alpha \text{ is large}$$

Thus when  $\alpha = 10000$ , distance from  $x^{LP}$  to  $x^{IP}$  on the plane  $\approx 10001$ , and only increases as  $\alpha \rightarrow \infty$ .

using Euclidean metric.

~~we cannot deduce anything about~~

d) ::) This shows that we cannot just search for <sup>an</sup> ~~the~~ optimal ~~solution~~ <sup>solution</sup> of an ILP within ~~a~~ a hypersphere (circle in this case) centered about the optimal solution of its LP relaxation.

(As shown in this case, as  $\alpha \rightarrow \infty$ , we need to set ~~the~~ the radius of this ~~sphere~~ hypersphere  $\rightarrow \infty$ !)  
 $\Rightarrow$  impractical.