

HW 3

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Problem 1

(a) We generate parameters for c and h in table 1 and table 2, respectively.

Table 1: value of c

c_1	c_2	c_3	c_4
31	28	40	44

Table 2: value of h

h_1	h_2	h_3
427	369	224

We formulate the LP as

$$\begin{aligned} \max \quad & 31x_1 + 28x_2 + 40x_3 + 44x_4 \\ \text{s. t.} \quad & 2x_1 + x_2 + 10x_3 + x_4 \leq 427 \\ & 10x_1 + x_2 + x_3 \leq 369 \\ & 2x_2 + x_3 + 3x_4 \leq 224 \\ & x_1 + x_2 \geq 10 \\ & x_1, x_2, x_3, x_4 \geq 0. \end{aligned} \tag{1}$$

We solve this problem using *LINDO* software. The solution is as follows

```
Global optimal solution found.
Objective value:                5083.394
Infeasibilities:                  0.000000
Total solver iterations:          3
Elapsed runtime seconds:        0.03
```

Model Class : LP

Total variables : 4
 Nonlinear variables : 0
 Integer variables : 0

 Total constraints : 5
 Nonlinear constraints : 0

 Total nonzeros : 16
 Nonlinear nonzeros : 0

Variable	Value	Reduced Cost
X1	33.95775	0.000000
X2	0.000000	4.746479
X3	29.42254	0.000000
X4	64.85915	0.000000

Row	Slack or Surplus	Dual Price
1	5083.394	1.000000
2	0.000000	2.348592
3	0.000000	2.630282
4	0.000000	13.88380
5	23.95775	0.000000

Therefore, the total profit is 5083.39 (in thousands of dollars), and the optimal value for each product is

Table 3: Production for each product

<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>
34.0	0	29.4	64.9

(b) The dual problem can be formulated as

$$\begin{aligned}
 \min \quad & 427y_1 + 369y_2 + 224y_3 + 10y_4 \\
 \text{s. t.} \quad & 2y_1 + 10y_2 + y_4 \geq 31 \\
 & y_1 + y_2 + 2y_3 + y_4 \geq 28 \\
 & 10y_1 + y_2 + y_3 \geq 40 \\
 & y_1 + 3y_3 \geq 44 \\
 & y_1, y_2, y_3 \geq 0, y_4 \leq 0
 \end{aligned} \tag{2}$$

or in *LINDO*

MIN= 427 * _2 + 369 * _3 + 224 * _4 + 10 * _5 ;

```

[X1] 2 * _2 + 10 * _3 + _5 >= 31;
[X2] _2 + _3 + 2 * _4 + _5 >= 28;
[X3] 10 * _2 + _3 + _4 >= 40;
[X4] _2 + 3 * _4 >= 44;
@BND( -1e+30, _5 , 0);
END

```

The optimal values for dual variables can be seen in the results printed in red rectangle in (a).

(c) Using the previous results,

- i. Machine III, because the shadow price for constraints III (Machine III) has the largest value, 13.88380, which means an additional 1 hour for h_3 would bring an extra increase in the total profit for about 13.88380. Therefore, I would be willing to pay for less than 13.88380 (in thousands of dollars) for an extra 1 hour for Machine III.
- ii. Yes. As can be seen, the slack for this contract (constraints 4 in our LP) is 23.95775, which means we can increase the minimum combined production of A and B by a number of 23.95775 while not decreasing the current total profit (or change the basis variables). Therefore, the current terms are profitable for the company and they should not renegotiate it.
- iii. As can be seen, we are now not producing product B. The reduced cost for product B is 4.746479, which means forcing the company to produce 1 ton of product B would make the company lose 4.746479 (in thousands of dollars). Therefore, if we enforce a production of 0.5 ton of product B, we would lose

$$4.746479 \times 0.5 = 2.3732395 \quad (3)$$

in thousand of dollars.

Problem 2

Solution

(a) We formulate the dual problem as

$$\begin{aligned}
 \max \quad & 4y_1 + 20y_2 + 6y_3 \\
 \text{s. t.} \quad & -y_1 + y_2 + y_3 \leq 0 \\
 & y_1 + y_3 \leq 1 \\
 & 2y_1 + 2y_3 \leq -1 \\
 & 2y_2 - 2y_3 \leq 2 \\
 & y_2 \leq 0.
 \end{aligned} \tag{4}$$

or in *LINDO*

MODEL:

```

MAX= 4 * _2 + 20 * _3 + 6 * _4 ;
[X2] _2 + _4 <= 1 ;
[X3] 2 * _2 + 2 * _4 <= - 1 ;
[X4] - 2 * _3 - 2 * _4 <= 2 ;
[X1] - _2 + _3 + _4 <= 0 ;
@FREE( _2 ) ; @BND( -1e+30, _3 , 0 ) ; @FREE( _4 ) ;
END

```

(b) To check the complementary slackness, we introduce slack variables $w_i \geq 0$ ($i = 1, 2, 3$) and $z_j \geq 0$ ($j = 1, 2, 3, 4$). The primal problem now becomes

$$\begin{aligned}
 \min \quad & x_2 - x_3 + 2x_4 \\
 \text{s. t.} \quad & -x_1 + x_2 + 2x_3 = 4 + w_1 \\
 & x_1 + 2x_4 + w_2 = 20 \\
 & x_1 + x_2 + 2x_3 - 2x_4 = 6 + w_3 \\
 & x_1, x_2, x_3, x_4 \geq 0.
 \end{aligned} \tag{5}$$

and its dual

$$\begin{aligned}
 \max \quad & 4y_1 + 20y_2 + 6y_3 \\
 \text{s. t.} \quad & -y_1 + y_2 + y_3 + z_1 = 0 \\
 & y_1 + y_3 + z_2 = 1 \\
 & 2y_1 + 2y_3 + z_3 = -1 \\
 & 2y_2 - 2y_3 + z_4 = 2 \\
 & y_2 \leq 0.
 \end{aligned} \tag{6}$$

Put $x = [2, 6, 0, 1]$ into (5) and (6) and use the complementary slackness, we get $y_3 = -1$ and $y_1 = -1$, which contradicts $y_1 + y_3 = 1$. Therefore, $x = [2, 6, 0, 1]$ is not optimal

(c)

Put $x = [1, 0, 2.5, 0]$ into (5) and (6), we get

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 2.5 \\ 0 \end{bmatrix}, \quad \begin{bmatrix} z_1 \\ z_2 \\ z_3 \\ z_4 \end{bmatrix} = \begin{bmatrix} 0 \\ 2 \\ 0 \\ 1 \end{bmatrix},$$

and

$$\begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} -\frac{1}{2} \\ 0 \\ -\frac{1}{2} \end{bmatrix}, \quad \begin{bmatrix} w_1 \\ w_2 \\ w_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 19 \\ 0 \end{bmatrix},$$

Since the complementary slackness holds, $[1, 0, 2.5, 0]$ is optimal.

We run a python code to test it. The solution $[1, 0, 2.5, 0]$ is indeed optimal.

```
from scipy.optimize import linprog
```

```
c=[0,1,-1,2]
```

```
A_ub=[[1,0,0,2]]
```

```
A_eq=[[-1,1,2,0],[1,1,2,-2]]
```

```
b_ub=[20]
```

```
b_eq=[4,6]
```

```
x0_bnds = [0, None]
```

```
x1_bnds = [0, None]
```

```
x2_bnds = [0, None]
```

```
x3_bnds = [0, None]
```

```
b_ub=[20]
```

```
res = linprog(c, A_ub, b_ub, A_eq, b_eq, bounds=(x0_bnds, x1_bnds, x2_bnds, x3_bnds,))  
print(res)
```