

LINEAR PROGRAMMING

[V. CH5]: DUALITY THEORY

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MOTIVATION: FINDING UPPER BOUNDS

A first example

Resource Allocation Problem

THE DUAL PROBLEM

Duality Theorems

Associated with every linear program is another called its *dual*. The dual of this dual linear program is the *original linear program* (which is then referred to as the *primal* linear program).

- ↪ linear programs come in primal/dual pairs.
- ↪ every feasible solution for one of these two linear programs gives a bound on the optimal value for the other.

These ideas are important and form a subject called *duality theory*, the topic of this chapter.

We begin with an example:

$$\begin{array}{llll} \max_x & 4x_1 + & x_2 + & 3x_3 \\ \text{subject to} & x_1 + & 4x_2 & \leq 1 \\ & 3x_1 - & x_2 + & x_3 \leq 3 \\ & x_1, & x_2, & x_3 \geq 0 \end{array}$$

Observe that

→ every feasible solution to this LP provides a *lower bound* on the optimal value, ζ^* .

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To answer:

we need to give upper bounds

$$\begin{array}{llll} \max_x & 4x_1 + & x_2 + & 3x_3 \\ \text{subject to} & x_1 + & 4x_2 & \leq 1 \\ & 3x_1 - & x_2 + & x_3 \leq 3 \\ & x_1, & x_2, & x_3 \geq 0 \end{array}$$

We can find a bound as follows:

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We can find a bound as follows:

→ Multiply the first constraint by 2 and *add* that to 3 times the second constraint

$$\begin{array}{rcll}
 & 2 \times (& x_1 + & 4x_2 & \leq 1 &) \\
 + & 3 \times (& 3x_1 - & x_2 + & x_3 \leq 3 &) \\
 \hline
 & 11x_1 + & 5x_2 + & 3x_3 \leq & 11 &
 \end{array}$$

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Since each variable is nonnegative, we can compare the sum against the objective function

$$\zeta = 4x_1 + x_2 + 3x_3 \leq 11x_1 + 5x_2 + 3x_3 \leq 11$$

$$\Rightarrow \zeta^* \leq 11$$

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These bounds leave a gap, $9 \leq \zeta^* \leq 11$. Now,

better insight on the quality of feasible solutions!

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→ Multiply the first constraint by $y_1 (\geq 0)$ and *add* that to $y_2 (\geq 0)$ times the second constraint

$$\begin{array}{r}
 y_1 \times (1x_1 + \leq 1) \\
 + \quad y_2 \times (3x_1 - x_2 + x_3 \leq 3) \\
 \hline
 (y_1 + 3y_2)x_1 + \quad (4y_1 - y_2)x_2 + \quad (y_2)x_3 \leq y_1 + 3y_2
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→ Enforce that each of the coefficients of the x_i 's be *at least as large as* the corresponding coefficient in the objective function, i.e.

$$\begin{array}{r}
 y_1 + 3y_2 \geq 4 \\
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→ then we can compare the objective function against this sum (and its bound).

$$\zeta = 4x_1 + x_2 + 3x_3 \leq (y_1 + 3y_2)x_1 + (4y_1 - y_2)x_2 + (y_2)x_3 \leq y_1 + 3y_2$$

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We are naturally led to the following optimization problem:

$$\begin{array}{ll} \min_y & y_1 + 3y_2 \\ & y_1 + 3y_2 \geq 4 \\ & 4y_1 - y_2 \geq 1 \\ & y_2 \geq 3 \\ & y_1, y_2 \geq 0 \end{array}$$

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↪ This problem is called the *dual* LP associated with the given LP.

GMPL Code 1: Primal model.

```
1  var x1 >=0;
2  var x2 >=0;
3  var x3 >=0;
4
5  subject to con1:  x1 + 4*x2      <= 1;
6  subject to con2: 3*x1 -  x2 + x3 <= 3;
7
8  maximize z: 4*x1 + x2 + 3*x3;
9
10 solve;
11
12 display x1.val, x2.val, x3.val, z.val;
13
14 end;
```

```
1 GLPSOL: GLPK LP/MIP Solver, v4.65
2 Reading model section from ex1_inSlides.mod...
3 .
4 .
5 .
6 Model has been successfully generated
7 GLPK Simplex Optimizer, v4.65
8 .
9 .
10 .
11 OPTIMAL LP SOLUTION FOUND
12 Time used: 0.0 secs
13 Memory used: 0.1 Mb (102265 bytes)
14 Display statement at line 12
15 x1.val = 0
16 x2.val = 0.25
17 x3.val = 3.25
18 z.val = 10
19 Model has been successfully processed
```

GMPL Code 2: Dual model.

```
1  var y1 >=0;
2  var y2 >=0;
3
4  subject to con1:  y1 + 3*y2 >= 4  ;
5  subject to con2: 4*y1 -  y2 >= 1  ;
6  subject to con3:          y2 >= 3  ;
7
8  minimize z: y1 + 3*y2 ;
9
10 solve;
11
12 display y1.val, y2.val, z.val;
13
14 end;
```



```
1  GLPSOL: GLPK LP/MIP Solver, v4.65
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3  .
4  .
5  .
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7  GLPK Simplex Optimizer, v4.65
8  .
9  .
10 .
11 OPTIMAL SOLUTION FOUND BY LP PREPROCESSOR
12 Time used: 0.0 secs
13 Memory used: 0.1 Mb (94214 bytes)
14 Display statement at line 12
15 y1.val = 1
16 y2.val = 3
17 z.val = 10
18 Model has been successfully processed
```

As a another example, consider the Resource Allocation Problem.

Recall that

$$\begin{aligned} & \max_x \quad c_1x_1 + c_2x_2 + \cdots + c_nx_n \\ \text{subject to} \quad & a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n \leq b_1 \\ & \vdots \\ & a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n \leq b_m \\ & x_1, x_2, \cdots, x_n \geq 0 \end{aligned}$$

where

c_j = profit per unit of product j produced

b_i = unit of raw material i on hand

a_{ij} = units raw material i required to produce one unit of product j

If we produce one unit less of product j , then:

for each i , we free up a_{ij} units of raw material i .

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Selling these unused raw materials for y_1, y_2, \dots, y_m dollars/unit yields:

$a_{1j}y_1 + a_{2j}y_2 + \dots + a_{mj}y_m$ dollars.

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Only interested if this revenue exceeds lost profit on each product j :

$$a_{1j}y_1 + a_{2j}y_2 + \dots + a_{mj}y_m \geq c_j, \quad j = 1, 2, \dots, n.$$

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Now, consider a buyer offering to purchase our **entire inventory**. Subject to above constraints, buyer wants to minimize cost:

$$\min_y \quad b_1y_1 + b_2y_2 + \dots + b_my_m$$

And the following linear program needs to be solved

$$\begin{aligned} \min_y \quad & b_1 y_1 + b_2 y_2 + \cdots + b_m y_m \\ \text{subject to} \quad & a_{11} y_1 + a_{21} y_2 + \cdots + a_{m1} y_m \geq c_1 \\ & \vdots \\ & a_{1n} y_1 + a_{2n} y_2 + \cdots + a_{mn} y_n \geq c_n \\ & y_1, y_2, \cdots, y_m \geq 0 \end{aligned}$$

↔ This problem is called the *dual* LP associated with the given LP.

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A first example

Resource Allocation Problem

THE DUAL PROBLEM

Duality Theorems

Given a linear programming problem in standard form,

$$\begin{aligned} \max_x \quad & \sum_{j=1}^n c_j x_j \\ \text{subject to} \quad & \sum_{j=1}^n a_{ij} x_j \leq b_i, \quad i = 1, 2, \dots, m \\ & x_j \geq 0, \quad j = 1, 2, \dots, n \end{aligned}$$

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the associated *dual* linear program is given by

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Primal Problem

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↪ Original problem is called the primal problem.

Dual in "Standard" Form

$$\begin{aligned} - \max_y \quad & \sum_{i=1}^m -b_i y_i \\ \text{subject to} \quad & \sum_{i=1}^m -a_{ij} y_i \leq -c_j, \quad j = 1, 2, \dots, n \\ & y_i \geq 0, \quad i = 1, 2, \dots, m \end{aligned}$$

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Theorem *Dual of dual is primal.*

THEOREM (WEAK DUALITY)

if (x_1, \dots, x_n) is *feasible* for the primal problem and (y_1, \dots, y_m) is *feasible* for the dual problem, then

$$\sum_j c_j x_j \leq \sum_i b_i y_i$$

PROOF.

$$\begin{aligned}\sum_j c_j x_j &\leq \sum_j \left(\sum_i y_i a_{ij} \right) x_j \\ &= \sum_i \sum_j y_i a_{ij} x_j \\ &= \sum_i \left(\sum_j a_{ij} x_j \right) y_i \\ &\leq \sum_i b_i y_i\end{aligned}$$

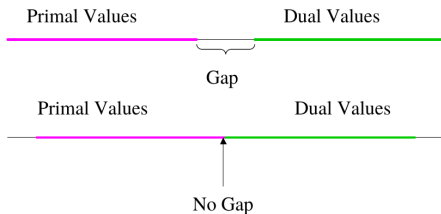
□

An important question:

Is there a gap between *the largest primal value* and *the smallest dual value*?

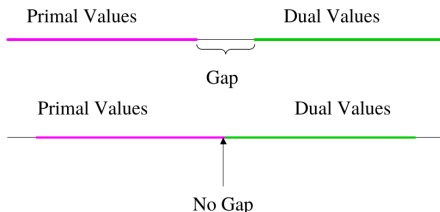
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↪ Answer is provided by the Strong Duality Theorem (coming later).

Consider the exmple:

Initial primal and dual dictionaries are provided.

Primal

$$\begin{array}{rcllcl} \zeta = & 0 - & 3x_1 + & 2 x_2 + & 1 x_3 \\ \hline w_1 = & 0 - & 0x_1 + & x_2 - & 2x_3 \\ w_2 = & 3 + & 3x_1 - & 4x_2 - & x_3 \end{array}$$

Dual

$$\begin{array}{rcll} -\xi = & 0 + & 0y_1 - & 3y_2 \\ \hline z_1 = & 3 - & 0y_1 - & 3y_2 \\ z_2 = & -2 - & 1y_1 + & 4y_2 \\ z_3 = & -1 + & 2y_1 + & 1y_2 \end{array}$$

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\rightsquigarrow Make analogous pivot in dual: z_2 leaves, y_2 enters.

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After first pivot

Primal

$$\zeta = \frac{3}{2} - \frac{3}{2}x_1 - \frac{1}{2}w_2 + \frac{1}{2}x_3$$

$$w_1 = \frac{3}{4} + \frac{3}{4}x_1 - \frac{1}{4}w_2 - \frac{9}{4}x_3$$

$$x_2 = \frac{3}{4} + \frac{3}{4}x_1 - \frac{1}{4}w_2 - \frac{1}{4}x_3$$

Dual

$$-\xi = -\frac{3}{2} - \frac{3}{4}y_1 - \frac{3}{4}z_2$$

$$z_1 = \frac{3}{2} - \frac{3}{4}y_1 - \frac{3}{4}z_2$$

$$y_2 = \frac{1}{2} + \frac{1}{4}y_1 + \frac{1}{4}z_2$$

$$z_3 = -\frac{1}{2} + \frac{9}{4}y_1 + \frac{1}{4}z_2$$

After first pivot

Primal

$$\begin{array}{r} \zeta = \frac{3}{2} - \frac{3}{2}x_1 - \frac{1}{2}w_2 + \frac{1}{2}x_3 \\ \hline w_1 = \frac{3}{4} + \frac{3}{4}x_1 - \frac{1}{4}w_2 - \frac{9}{4}x_3 \\ x_2 = \frac{3}{4} + \frac{3}{4}x_1 - \frac{1}{4}w_2 - \frac{1}{4}x_3 \end{array}$$

\rightsquigarrow negative transpose property intact.

Dual

$$\begin{array}{r} -\xi = -\frac{3}{2} - \frac{3}{4}y_1 - \frac{3}{4}z_2 \\ \hline z_1 = \frac{3}{2} - \frac{3}{4}y_1 - \frac{3}{4}z_2 \\ y_2 = \frac{1}{2} + \frac{1}{4}y_1 + \frac{1}{4}z_2 \\ z_3 = -\frac{1}{2} + \frac{9}{4}y_1 + \frac{1}{4}z_2 \end{array}$$

After first pivot

Primal

$$\begin{array}{r} \zeta = \frac{3}{2} - \frac{3}{2}x_1 - \frac{1}{2}w_2 + \frac{1}{2}x_3 \\ \hline w_1 = \frac{3}{4} + \frac{3}{4}x_1 - \frac{1}{4}w_2 - \frac{9}{4}x_3 \\ x_2 = \frac{3}{4} + \frac{3}{4}x_1 - \frac{1}{4}w_2 - \frac{1}{4}x_3 \end{array}$$

\rightsquigarrow negative transpose property intact.

\rightsquigarrow Primal is feasible, dual is not.

Dual

$$\begin{array}{r} -\xi = -\frac{3}{2} - \frac{3}{4}y_1 - \frac{3}{4}z_2 \\ \hline z_1 = \frac{3}{2} - \frac{3}{4}y_1 - \frac{3}{4}z_2 \\ y_2 = \frac{1}{2} + \frac{1}{4}y_1 + \frac{1}{4}z_2 \\ z_3 = -\frac{1}{2} + \frac{9}{4}y_1 + \frac{1}{4}z_2 \end{array}$$

After first pivot

Primal

$$\begin{array}{r} \zeta = \frac{3}{2} - \frac{3}{2}x_1 - \frac{1}{2}w_2 + \frac{1}{2}x_3 \\ \hline w_1 = \frac{3}{4} + \frac{3}{4}x_1 - \frac{1}{4}w_2 - \frac{9}{4}x_3 \\ x_2 = \frac{3}{4} + \frac{3}{4}x_1 - \frac{1}{4}w_2 - \frac{1}{4}x_3 \end{array}$$

\rightsquigarrow negative transpose property intact.

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\rightsquigarrow Use primal to pick pivot: x_3 enters, w_1 leaves.

Dual

$$\begin{array}{r} -\xi = -\frac{3}{2} - \frac{3}{4}y_1 - \frac{3}{4}z_2 \\ \hline z_1 = \frac{3}{2} - \frac{3}{4}y_1 - \frac{3}{4}z_2 \\ y_2 = \frac{1}{2} + \frac{1}{4}y_1 + \frac{1}{4}z_2 \\ z_3 = -\frac{1}{2} + \frac{9}{4}y_1 + \frac{1}{4}z_2 \end{array}$$

After first pivot

Primal

$$\begin{array}{r} \zeta = \frac{3}{2} - \frac{3}{2}x_1 - \frac{1}{2}w_2 + \frac{1}{2}x_3 \\ \hline w_1 = \frac{3}{4} + \frac{3}{4}x_1 - \frac{1}{4}w_2 - \frac{9}{4}x_3 \\ x_2 = \frac{3}{4} + \frac{3}{4}x_1 - \frac{1}{4}w_2 - \frac{1}{4}x_3 \end{array}$$

\rightsquigarrow negative transpose property intact.

\rightsquigarrow Primal is feasible, dual is not.

\rightsquigarrow Use primal to pick pivot: x_3 enters, w_1 leaves.

\rightsquigarrow Make analogous pivot in dual: z_3 leaves, y_1 enters.

Dual

$$\begin{array}{r} -\xi = -\frac{3}{2} - \frac{3}{4}y_1 - \frac{3}{4}z_2 \\ \hline z_1 = \frac{3}{2} - \frac{3}{4}y_1 - \frac{3}{4}z_2 \\ y_2 = \frac{1}{2} + \frac{1}{4}y_1 + \frac{1}{4}z_2 \\ z_3 = -\frac{1}{2} + \frac{9}{4}y_1 + \frac{1}{4}z_2 \end{array}$$

After second pivot

Primal

$$\begin{array}{r} \zeta = \frac{5}{3} - \frac{4}{3}x_1 - \frac{5}{9}w_2 - \frac{2}{9}w_1 \\ \hline x_3 = \frac{1}{3} + \frac{1}{3}x_1 - \frac{1}{9}w_2 - \frac{4}{9}w_1 \\ x_2 = \frac{2}{3} + \frac{2}{3}x_1 - \frac{2}{9}w_2 + \frac{1}{9}w_1 \end{array}$$

Dual

$$\begin{array}{r} -\xi = -\frac{5}{3} - \frac{1}{3}z_3 - \frac{2}{3}z_2 \\ \hline z_1 = \frac{4}{3} - \frac{1}{3}z_3 - \frac{2}{3}z_2 \\ y_2 = \frac{5}{9} + \frac{1}{9}z_3 + \frac{2}{9}z_2 \\ y_1 = \frac{2}{9} + \frac{4}{9}z_3 - \frac{1}{9}z_2 \end{array}$$

After second pivot

Primal

$$\begin{array}{r} \zeta = \frac{5}{3} - \frac{4}{3}x_1 - \frac{5}{9}w_2 - \frac{2}{9}w_1 \\ \hline x_3 = \frac{1}{3} + \frac{1}{3}x_1 - \frac{1}{9}w_2 - \frac{4}{9}w_1 \\ x_2 = \frac{2}{3} + \frac{2}{3}x_1 - \frac{2}{9}w_2 + \frac{1}{9}w_1 \end{array}$$

↪ negative transpose property remains intact.

Dual

$$\begin{array}{r} -\xi = -\frac{5}{3} - \frac{1}{3}z_3 - \frac{2}{3}z_2 \\ \hline z_1 = \frac{4}{3} - \frac{1}{3}z_3 - \frac{2}{3}z_2 \\ y_2 = \frac{5}{9} + \frac{1}{9}z_3 + \frac{2}{9}z_2 \\ y_1 = \frac{2}{9} + \frac{4}{9}z_3 - \frac{1}{9}z_2 \end{array}$$

After second pivot

Primal

$$\begin{aligned} \zeta &= \frac{5}{3} - \frac{4}{3}x_1 - \frac{5}{9}w_2 - \frac{2}{9}w_1 \\ x_3 &= \frac{1}{3} + \frac{1}{3}x_1 - \frac{1}{9}w_2 - \frac{4}{9}w_1 \\ x_2 &= \frac{2}{3} + \frac{2}{3}x_1 - \frac{2}{9}w_2 + \frac{1}{9}w_1 \end{aligned}$$

\rightsquigarrow negative transpose property remains intact.

\rightsquigarrow Primal and dual are both optimal.

Dual

$$\begin{aligned} -\xi &= -\frac{5}{3} - \frac{1}{3}z_3 - \frac{2}{3}z_2 \\ z_1 &= \frac{4}{3} - \frac{1}{3}z_3 - \frac{2}{3}z_2 \\ y_2 &= \frac{5}{9} + \frac{1}{9}z_3 + \frac{2}{9}z_2 \\ y_1 &= \frac{2}{9} + \frac{4}{9}z_3 - \frac{1}{9}z_2 \end{aligned}$$

After second pivot

Primal

$$\begin{aligned} \zeta &= \frac{5}{3} - \frac{4}{3}x_1 - \frac{5}{9}w_2 - \frac{2}{9}w_1 \\ x_3 &= \frac{1}{3} + \frac{1}{3}x_1 - \frac{1}{9}w_2 - \frac{4}{9}w_1 \\ x_2 &= \frac{2}{3} + \frac{2}{3}x_1 - \frac{2}{9}w_2 + \frac{1}{9}w_1 \end{aligned}$$

\rightsquigarrow negative transpose property remains intact.

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Dual

$$\begin{aligned} -\xi &= -\frac{5}{3} - \frac{1}{3}z_3 - \frac{2}{3}z_2 \\ z_1 &= \frac{4}{3} - \frac{1}{3}z_3 - \frac{2}{3}z_2 \\ y_2 &= \frac{5}{9} + \frac{1}{9}z_3 + \frac{2}{9}z_2 \\ y_1 &= \frac{2}{9} + \frac{4}{9}z_3 - \frac{1}{9}z_2 \end{aligned}$$

\rightsquigarrow Simplex Alg. applied to primal, solves both the primal and the dual.

THEOREM (STRONG DUALITY)

If the primal problem has an optimal solution,

$$x^* = (x_1^*, x_2^*, \dots, x_n^*)$$

then the dual also has an optimal solution,

$$y^* = (y_1^*, y_2^*, \dots, y_m^*)$$

and

$$\sum_j c_j x_j^* = \sum_i c_i y_i^*$$

↪ If primal has an optimal solution, then there is **no duality gap**.

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↪ If primal has an optimal solution, then there is **no duality gap**.

↪ Let's prove it.