

ORF 522: Lecture 5

Linear Programming: Chapter 5 Duality

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Resource Allocation

Recall the resource allocation problem ($m = 2$, $n = 3$):

$$\begin{array}{ll}\text{maximize} & c_1x_1 + c_2x_2 + c_3x_3 \\ \text{subject to} & a_{11}x_1 + a_{12}x_2 + a_{13}x_3 \leq b_1 \\ & a_{21}x_1 + a_{22}x_2 + a_{23}x_3 \leq b_2 \\ & x_1, x_2, x_3 \geq 0,\end{array}$$

where

c_j = profit per unit of product j produced

b_i = units of raw material i on hand

a_{ij} = units raw material i required to produce 1 unit of prod j .

Closing Up Shop

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

- a_{1j} units of raw material 1 and
- a_{2j} units of raw material 2.

Selling these unused raw materials for y_1 and y_2 dollars/unit yields $a_{1j}y_1 + a_{2j}y_2$ dollars.

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

$$a_{1j}y_1 + a_{2j}y_2 \geq c_j, \quad j = 1, 2, 3.$$

Consider a buyer offering to purchase our entire inventory.

Subject to above constraints, buyer wants to minimize cost:

$$\begin{aligned} & \text{minimize} && b_1y_1 + b_2y_2 \\ & \text{subject to} && a_{11}y_1 + a_{21}y_2 \geq c_1 \\ & && a_{12}y_1 + a_{22}y_2 \geq c_2 \\ & && a_{13}y_1 + a_{23}y_2 \geq c_3 \\ & && y_1, y_2 \geq 0 . \end{aligned}$$

Duality

Every Problem:

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

Has a Dual:

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

Dual of Dual

Primal Problem:

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

Original problem is called the *primal problem*.

A problem is defined by its data (notation used for the variables is arbitrary).

Dual in “Standard” Form:

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

Dual is “negative transpose” of primal.

Theorem *Dual of dual is primal.*

Weak Duality Theorem

If (x_1, x_2, \dots, x_n) is feasible for the primal and (y_1, y_2, \dots, y_m) is feasible for the dual, then

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

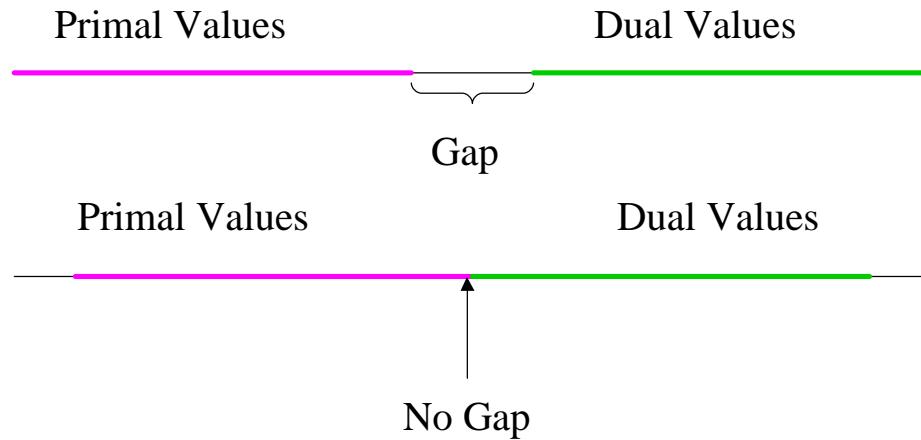
Proof.

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

Gap or No Gap?

An important question:

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



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

Simplex Method and Duality

A Primal Problem:

$$\begin{array}{ll} \text{obj} = & \boxed{0} + \boxed{-3} x_1 + \boxed{2} x_2 + \boxed{1} x_3 \\ w_1 = & \boxed{0} - \boxed{0} x_1 - \boxed{-1} x_2 - \boxed{2} x_3 \\ w_2 = & \boxed{3} - \boxed{-3} x_1 - \boxed{4} x_2 - \boxed{1} x_3 \end{array}$$

Its Dual:

$$\begin{array}{ll} \text{obj} = & \boxed{0} + \boxed{0} y_1 + \boxed{-3} y_2 \\ z_1 = & \boxed{3} - \boxed{0} y_1 - \boxed{3} y_2 \\ z_2 = & \boxed{-2} - \boxed{1} y_1 - \boxed{-4} y_2 \\ z_3 = & \boxed{-1} - \boxed{-2} y_1 - \boxed{-1} y_2 \end{array}$$

Notes:

- Dual is negative transpose of primal.
- Primal is feasible, dual is not.

Use primal to choose pivot: x_2 enters, w_2 leaves.
Make analogous pivot in dual: z_2 leaves, y_2 enters.

Second Iteration

After First Pivot:

Primal (feasible):

$$\begin{array}{lcl} \text{obj} & = & \boxed{\frac{3}{2}} + \boxed{-\frac{3}{2}} x_1 + \boxed{-\frac{1}{2}} w_2 + \boxed{\frac{1}{2}} x_3 \\ w_1 & = & \boxed{\frac{3}{4}} - \boxed{-\frac{3}{4}} \boxed{x_1} - \boxed{\frac{1}{4}} \boxed{w_2} - \boxed{\frac{9}{4}} \boxed{x_3} \\ x_2 & = & \boxed{\frac{3}{4}} - \boxed{-\frac{3}{4}} \boxed{x_1} - \boxed{\frac{1}{4}} \boxed{w_2} - \boxed{\frac{1}{4}} \boxed{x_3} \end{array}$$

Dual (still not feasible):

$$\begin{array}{lcl} \text{obj} & = & \boxed{-\frac{3}{2}} + \boxed{-\frac{3}{4}} y_1 + \boxed{-\frac{3}{4}} z_2 \\ z_1 & = & \boxed{\frac{3}{2}} - \boxed{\frac{3}{4}} \boxed{y_1} - \boxed{\frac{3}{4}} \boxed{z_2} \\ y_2 & = & \boxed{\frac{1}{2}} - \boxed{-\frac{1}{4}} \boxed{y_1} - \boxed{-\frac{1}{4}} \boxed{z_2} \\ z_3 & = & \boxed{-\frac{1}{2}} - \boxed{-\frac{9}{4}} \boxed{y_1} - \boxed{-\frac{1}{4}} \boxed{z_2} \end{array}$$

Note: *negative transpose property intact.*

Again, use primal to pick pivot: x_3 enters, w_1 leaves.

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

After Second Iteration

Primal:

- Is *optimal*.

$$\begin{array}{lcl} \text{obj} & = & \boxed{\frac{5}{3}} + \boxed{-\frac{4}{3}} x_1 + \boxed{-\frac{5}{9}} w_2 + \boxed{-\frac{2}{9}} w_1 \\ x_3 & = & \boxed{\frac{1}{3}} - \boxed{-\frac{1}{3}} x_1 - \boxed{\frac{1}{9}} w_2 - \boxed{\frac{4}{9}} w_1 \\ x_2 & = & \boxed{\frac{2}{3}} - \boxed{-\frac{2}{3}} x_1 - \boxed{\frac{2}{9}} w_2 - \boxed{-\frac{1}{9}} w_1 \end{array}$$

Dual:

- Negative transpose property remains intact.
- Is *optimal*.

$$\begin{array}{lcl} \text{obj} & = & \boxed{-\frac{5}{3}} + \boxed{-\frac{1}{3}} z_3 + \boxed{-\frac{2}{3}} z_2 \\ z_1 & = & \boxed{\frac{4}{3}} - \boxed{\frac{1}{3}} z_3 - \boxed{\frac{2}{3}} z_2 \\ y_2 & = & \boxed{\frac{5}{9}} - \boxed{-\frac{1}{9}} z_3 - \boxed{-\frac{2}{9}} z_2 \\ y_1 & = & \boxed{\frac{2}{9}} - \boxed{-\frac{4}{9}} z_3 - \boxed{\frac{1}{9}} z_2 \end{array}$$

Conclusion

Simplex method applied to primal problem (two phases, if necessary), solves both the primal and the dual.

Strong Duality Theorem

Conclusion on previous slide is the essence of the *strong duality theorem* which we now state:

Theorem. *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

$$\sum_j c_j x_j^* = \sum_i b_i y_i^*.$$

Paraphrase:

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

Duality Gap

Four possibilities:

- Primal optimal, dual optimal (no gap).
- Primal unbounded, dual infeasible (no gap).
- Primal infeasible, dual unbounded (no gap).
- Primal infeasible, dual infeasible (infinite gap).

Example of *infinite gap*:

$$\begin{array}{ll}\text{maximize} & 2x_1 - x_2 \\ \text{subject to} & x_1 - x_2 \leq 1 \\ & -x_1 + x_2 \leq -2 \\ & x_1, x_2 \geq 0.\end{array}$$

Complementary Slackness

Theorem. *At optimality, we have*

$$\begin{aligned} x_j z_j &= 0, && \text{for } j = 1, 2, \dots, n, \\ w_i y_i &= 0, && \text{for } i = 1, 2, \dots, m. \end{aligned}$$

Proof

Recall the proof of the Weak Duality Theorem:

$$\begin{aligned}\sum_j c_j x_j &\leq \sum_j (c_j + z_j)x_j = \sum_j \left(\sum_i y_i a_{ij} \right) x_j = \sum_{ij} y_i a_{ij} x_j \\&= \sum_i \left(\sum_j a_{ij} x_j \right) y_i = \sum_i (b_i - w_i) y_i \leq \sum_i b_i y_i,\end{aligned}$$

The inequalities come from the fact that

$$\begin{aligned}x_j z_j &\geq 0, \quad \text{for all } j, \\w_i y_i &\geq 0, \quad \text{for all } i.\end{aligned}$$

By Strong Duality Theorem, the inequalities are equalities at optimality.

Dual Simplex Method

When: dual feasible, primal infeasible (i.e., pinks on the left, not on top).

An Example. Showing both primal and dual dictionaries:

obj	=	0.0	+	-2.0	x1	+	-4.0	x2	+	0.0	x3	+	-6.0	x4
w1	=	-3.0	-	-1.0	x1	-	2.0	x2	-	0.0	x3	-	-1.0	x4
w2	=	-5.0	-	2.0	x1	-	-3.0	x2	-	0.0	x3	-	-2.0	x4
w3	=	8.0	-	2.0	x1	-	3.0	x2	-	3.0	x3	-	2.0	x4

obj	=	0.0	+	3.0	y1	+	5.0	y2	+	-8.0	y3
z1	=	2.0	-	1.0	y1	-	-2.0	y2	-	-2.0	y3
z2	=	4.0	-	-2.0	y1	-	3.0	y2	-	-3.0	y3
z3	=	0.0	-	0.0	y1	-	0.0	y2	-	-3.0	y3
z4	=	6.0	-	1.0	y1	-	2.0	y2	-	-2.0	y3

Looking at dual dictionary: y_2 enters, z_2 leaves.

On the primal dictionary: w_2 leaves, x_2 enters.

After pivot...

Dual Simplex Method: Second Pivot

Going in, we have:

obj =	-6.6667	+	-4.6667	x1	+	-1.3333	w2	+	0.0	x3	+	-3.3333	x4
w1 =	-6.3333	-	0.3333	x1	-	0.6667	w2	-	0.0	x3	-	-2.3333	x4
x2 =	1.6667	-	-0.6667	x1	-	-0.3333	w2	-	0.0	x3	-	0.6667	x4
w3 =	3.0	-	4.0	x1	-	1.0	w2	-	3.0	x3	-	0.0	x4

obj =	6.6667	+	6.3333	y1	+	-1.6667	z2	+	-3.0	y3
z1 =	4.6667	-	-0.3333	y1	-	0.6667	z2	-	-4.0	y3
y2 =	1.3333	-	-0.6667	y1	-	0.3333	z2	-	-1.0	y3
z3 =	0.0	-	0.0	y1	-	0.0	z2	-	-3.0	y3
z4 =	3.3333	-	2.3333	y1	-	-0.6667	z2	-	0.0	y3

Looking at dual: y_1 enters, z_4 leaves.

Looking at primal: w_1 leaves, x_4 enters.

Dual Simplex Method Pivot Rule

obj	=	-6.6667	+	-4.6667	x1	+	-1.3333	w2	+	0.0	x3	+	-3.3333	x4
w1	=	-6.3333	-	0.3333	x1	-	0.6667	w2	-	0.0	x3	-	-2.3333	x4
x2	=	1.6667	-	-0.6667	x1	-	-0.3333	w2	-	0.0	x3	-	0.6667	x4
w3	=	3.0	-	4.0	x1	-	1.0	w2	-	3.0	x3	-	0.0	x4

Referring to the primal dictionary:

- Pick leaving variable from those rows that are *infeasible*.
- Pick entering variable from a box with a negative value and which can be increased the least (on the dual side).

Next primal dictionary shown on next page...

Dual Simplex Method: Third Pivot

Going in, we have:

obj	=	-15.7143	+	-5.1429	x1	+	-2.2857	w2	+	0.0	x3	+	-1.4286	w1
x4	=	2.7143	-	-0.1429	x1	-	-0.2857	w2	-	0.0	x3	-	-0.4286	w1
x2	=	-0.1429	-	-0.5714	x1	-	-0.1429	w2	-	0.0	x3	-	0.2857	w1
w3	=	3.0	-	4.0	x1	-	1.0	w2	-	3.0	x3	-	0.0	w1

Which variable must leave and which must enter?

See next page...

Dual Simplex Method: Third Pivot—Answer

Answer is: x_2 leaves, x_1 enters.

Resulting dictionary is OPTIMAL:

obj =	-17.0	+	-9.0	x2 +	-1.0	w2 +	0.0	x3 +	-4.0	w1
x4 =	2.75	-	-0.25	x2 -	-0.25	w2 -	0.0	x3 -	-0.5	w1
x1 =	0.25	-	-1.75	x2 -	0.25	w2 -	0.0	x3 -	-0.5	w1
w3 =	2.0	-	7.0	x2 -	0.0	w2 -	3.0	x3 -	2.0	w1

Dual-Based Phase I Method

Example:

obj =	0.0	+	-4.0	x1 +	2.0	x2 +	3.0	x3		
w1 =	0.0	+	1.0	-	2.0	x1 -	-1.0	x2 -	3.0	x3
w2 =	0.0	+	1.0	-	3.0	x1 -	-3.0	x2 -	-4.0	x3
w3 =	-3.0	+	1.0	-	-1.0	x1 -	-1.0	x2 -	1.0	x3
w4 =	-1.0	+	1.0	-	-2.0	x1 -	0.0	x2 -	0.0	x3

Notes:

- Two objective functions: the true objective (on top), and a fake one (below it).
- For *Phase I*, use the fake objective—it's dual feasible.
- Two right-hand sides: the real one (on the left) and a fake (on the right).
- Ignore the fake right-hand side—we'll use it in another algorithm later.

Phase I—First Pivot: w_3 leaves, x_1 enters.

After first pivot...

Dual-Based Phase I Method—Second Pivot

Current dictionary:

obj	=	-12.0	+	-4.0	w3	+	6.0	x2	+	-1.0	x3
			+	-1.0	w3	+	0.0	x2	+	-2.0	x3
w1	=	-6.0	+	3.0	-	2.0	w3	-	-3.0	x2	-
w2	=	-9.0	+	4.0	-	3.0	w3	-	-6.0	x2	-
x1	=	3.0	+	-1.0	-	-1.0	w3	-	1.0	x2	-
w4	=	5.0	+	-1.0	-	-2.0	w3	-	2.0	x2	-

Dual pivot: w_2 leaves, x_2 enters.

After pivot:

obj	=	-3.0	+	-1.0	w3	+	1.0	w2	+	-2.0	x3
			+	-1.0	w3	+	0.0	w2	+	-2.0	x3
w1	=	-1.5	+	1.0	-	0.5	w3	-	-0.5	w2	-
x2	=	1.5	+	-0.6667	-	-0.5	w3	-	-0.1667	w2	-
x1	=	1.5	+	-0.3333	-	-0.5	w3	-	0.1667	w2	-
w4	=	2.0	+	0.3333	-	-1.0	w3	-	0.3333	w2	-

Dual-Based Phase I Method—Third Pivot

Current dictionary:

Dual pivot:
 w_1 leaves,
 w_2 enters.

obj	=	-3.0	+	-1.0	w3	+	1.0	w2	+	-2.0	x3
			+	-1.0	w3	+	0.0	w2	+	-2.0	x3
w1	=	-1.5	+	1.0	-	0.5	w3	-	-0.5	w2	-
x2	=	1.5	+	-0.6667	-	-0.5	w3	-	-0.1667	w2	-
x1	=	1.5	+	-0.3333	-	-0.5	w3	-	0.1667	w2	-
w4	=	2.0	+	0.3333	-	-1.0	w3	-	0.3333	w2	-
										-2.3333	x3

After pivot:

It's **feasible**!

Fourth Pivot—Phase II

Current dictionary:

obj	=	0.0	+	0.0	w3	+	2.0	w1	+	9.0	x3
w2	=	3.0	+	-2.0	-	-1.0	w3	-	-2.0	w1	-
x2	=	2.0	+	-1.0	-	-0.6667	w3	-	-0.3333	w1	-
x1	=	1.0	+	0.0	-	-0.3333	w3	-	0.3333	w1	-
w4	=	1.0	+	1.0	-	-0.6667	w3	-	0.6667	w1	-

It's feasible.

Ignore fake objective.

Use the real thing (top row).

Primal pivot: x_3 enters, w_4 leaves.

Final Dictionary

After pivot:

obj =	6.75	+	4.5	w3	+	-2.5	w1	+	-6.75	w4
		+	-2.0	w3	+	1.0	w1	+	1.5	w4
w2 =	11.25	+	6.25	-	-6.5	w3	-	3.5	w1	-
x2 =	3.25	+	0.25	-	-1.5	w3	-	0.5	w1	-
x1 =	0.5	+	-0.5	-	0.0	w3	-	0.0	w1	-
x3 =	0.75	+	0.75	-	-0.5	w3	-	0.5	w1	-

Problem is **unbounded!**