ISE 426 Optimization models and applications

Lecture 9 — September 26, 2014

Duality, continued

Reading:

- W.&V. Sections 6.5–6.7, pages 295-308
- ► H.&L. Section 6.1–6.4, pages 151-169

Reminders:

Quiz on 10/14, practice on 10/09.

Primal problem, dual problem

The **primal** has n variables and m constraints \Rightarrow The **dual** has m variables and n constraints

Properties of duality in LP

Weak duality: Given a primal $\min\{c^{\top}x: Ax \geq b, x \geq 0\}$ and its dual $\max\{b^{\top}u: A^{\top}u \leq c, u \geq 0\}$,

$$b^{\top}\bar{u} \leq c^{\top}\bar{x}$$

for any \bar{x} and \bar{u} feasible for their respective problems.

Strong duality: If a problem $\min\{c^{\top}x : Ax \geq b, x \geq 0\}$ is bounded and its dual $\max\{b^{\top}u : A^{\top}u \leq c, u \geq 0\}$ is bounded, their optimal solutions \bar{x} and \bar{u} coincide in value:

$$c^{\top}\bar{x} = b^{\top}\bar{u}$$

Properties of duality in LP (cont.)

Consequence: solving the dual or the primal doesn't matter: we get the same objective function value.

What if the primal (or the dual) is infeasible or unbounded?

Four cases:

- Primal bounded, dual bounded;
- Primal infeasible, dual infeasible;
- ▶ Primal unbounded ($c^{\top}x = -\infty$), dual infeasible;
- ▶ Primal infeasible, dual unbounded ($b^{\top}u = +\infty$).

		Dual		
		bounded	unbounded	infeasible
	bounded	Possible	_	_
Primal	unbounded	_	_	Possible
	infeasible	_	Possible	Possible

Unbounded LP problem

Consider the following **minimization** problem:

min
$$-5x_1 - 4x_2$$

 $2x_1 - x_2 \ge 1$
 $-x_1 + 2x_2 \ge 1$
 $x_1, x_2 \ge 0$

- ▶ How to show that the solution is unbounded?
- ▶ Consider direction (d_1, d_2) in which one can move infinitely while decreasing the objective function.

Unbounded LP problem \Rightarrow Infeasible dual

$$\begin{array}{ll}
-5d_1 - 4d_2 < 0 \\
2d_1 - d_2 & \geq 0 \\
-d_1 + 2d_2 & \geq 0 \\
d_1, d_2 \geq 0
\end{array}$$

For example $(d_1, d_2) = (1, 1)$.

The dual

$$\begin{array}{rcl}
\max & u_1 + u_2 \\
2u_1 - u_2 & \leq -5 \\
-u_1 + 2u_2 & \leq -4 \\
u_1, u_2 \geq 0
\end{array}$$

$$0 \le u_1(2d_1 - d_2) + u_2(-d_1 + 2d_2) = d_1(2u_1 - u_2) + d_2(-u_1 + 2u_2) \le -5d_1 - 4d_2 < 0$$

primal unboundedness - dual infeasibility

Primal - unbounded

$$\begin{array}{ll}
\min & c^{\top} x \\
 & Ax \ge b \\
 & x \ge 0
\end{array}$$

$$c^{\top}d < 0$$

$$Ad \ge 0$$

$$d > 0$$

Dual - infeasible

$$\max \quad b^{\top} u \\ A^{\top} u \le c \\ u \ge 0$$

for any feasible
$$u$$

 $0 \le d^{\top}A^{\top}u \le d^{\top}c < 0$
 $u \ge 0$

Primal problem and dual problem with equality constraints

	Primal		Dual
min	$3x_1 + 4x_2$ $5x_1 + 6x_2 \ge 7$ $8x_1 + 9x_2 \ge 10$ $-8x_1 - 9x_2 \ge -10$ $11x_1 + 12x_2 \ge 13$ $x_1, x_2 \ge 0$	max	$7u_1 + 10u_2' - 10u_2'' + 13u_3$ $5u_1 + 8u_2' - 8u_2'' + 11u_3 \le 3$ $6u_1 + 9u_2' - 9u_2'' + 12u_3 \le 4$ $u_1, u_2', u_2'', u_3 \ge 0$
min	$3x_1 + 4x_2$ $5x_1 + 6x_2 \ge 7$ $8x_1 + 9x_2 = 10$ $11x_1 + 12x_2 \ge 13$ $x_1, x_2 \ge 0$	max	$7u_1 + 10u_2 + 13u_3$ $5u_1 + 8u_2 + 11u_3 \le 3$ $6u_1 + 9u_2 + 12u_3 \le 4$ $u_1, u_3 \ge 0, u_2$ – unrstrtd

Primal problem and dual problem with unrestricted variables

Primal

min
$$3x_1 + 4(x'_2 - x''_2)$$

 $5x_1 + 6(x'_2 - x''_2) \ge 7$
 $8x_1 + 9(x'_2 - x''_2) \ge 10$
 $11x_1 + 12(x'_2 - x''_2) \ge 13$
 $x_1, x'_2, x''_2 \ge 0$

min
$$3x_1 + 4x_2$$

 $5x_1 + 6x_2 \ge 7$
 $8x_1 + 9x_2 \ge 10$
 $11x_1 + 12x_2 \ge 13$
 $x_1 > 0, x_2$ — unrstrtd

Dual

$$\max \quad 7u_1 + 10u_2 + 13u_3$$

$$5u_1 + 8u_2 + 11u_3 \le 3$$

$$6u_1 + 9u_2 + 12u_3 \le 4$$

$$-6u_1 - 9u_2 - 12u_3 \le -4$$

$$u_1, u_2, u_3 \ge 0$$

$$\begin{array}{ll} \max & 7u_1 + 10u_2 + 13u_3 \\ & 5u_1 + 8u_2 + 11u_3 \leq 3 \\ & 6u_1 + 9u_2 + 12u_3 = 4 \\ & u_1, u_2, u_3 \geq 0 \end{array}$$

How to construct the dual of an LP

Variable	Constraint
Constraint	Variable
Minimize	Maximize
Variable ≥ 0	Constraint ≤
Variable ≤ 0	Constraint \geq
Var. Unrestricted	Constraint =
Constraint ≤	Variable ≤ 0
Constraint \geq	Variable ≥ 0
Constraint =	Var. Unrestricted

LP primal and dual problem, standard form

$$\begin{array}{cc} \text{Dual} \\ \max & b^\top u \\ & A^\top u \leq c \end{array}$$

LP primal and dual problem, standard form

$$\begin{array}{ll}
\text{Primal} \\
\text{min} & c^{\top} x \\
Ax = b \\
x \ge 0
\end{array}$$

$$\begin{array}{c} \text{Dual} \\ \max \quad b^{\top} u \\ A^{\top} u + s = c \\ s \ge 0 \end{array}$$

What is the dual of the dual in standard form?

Complementary slackness

- Given a primal-dual pair, now we know how to solve one and get the optimal objective function of the other.
- e.g. Solve primal \Rightarrow get optimal obj.f. $c^{\top}\bar{x}$, an optimal solution \bar{x} , and the optimal dual obj.f. $b^{\top}\bar{u}$. **How do we get** \bar{u} ?

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Complementary Slackness: If the primal problem \min\{c^{\top}x: \sum_{i=1}^{n} a_{ji}x_{i} \geq b_{j} \ \forall j=1,2\dots,m,x\geq 0\} is bounded and admits optimum \bar{x}, and its dual \max\{b^{\top}u: \sum_{j=1}^{m} a_{ji}u_{j} \leq c_{i} \ \forall i=1,2\dots,n,u\geq 0\} is bounded and admits optimal solution \bar{u}, then \bar{u}_{j}(\sum_{i=1}^{n} a_{ji}\bar{x}_{i} - b_{j}) = 0 \quad \forall j=1,2\dots,m; \bar{x}_{i}(\sum_{j=1}^{m} a_{ji}\bar{u}_{j} - c_{i}) = 0 \quad \forall i=1,2\dots,n
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So if we solve the primal and get \bar{x} , we can get \bar{u} by solving a system of equations.

LP primal and dual problem, standard form

Primal Dual max
$$b^{\top}u$$

$$Ax = b$$

$$x \ge 0$$

$$A^{\top}u + s = c$$

$$s \ge 0$$

$$s_{i}x_{i} = 0 \implies (c_{i} - \sum_{i=1}^{m} a_{ji}u_{j})x_{i} = 0$$

Either the primal variable is zero of the dual constraint is tight.

Example

$$\begin{array}{llll} \min & 3x_1 + 4x_2 \\ & 5x_1 + 6x_2 & \geq 7 \\ & 8x_1 + 9x_2 & \geq 10 \\ & 11x_1 + 12x_2 & \geq 13 \\ & x_1, x_2 \geq 0 \end{array} \qquad \begin{array}{lll} \max & 7u_1 + 10u_2 + 13u_3 \\ & 5u_1 + 8u_2 + 11u_3 & \leq 3 \\ & 6u_1 + 9u_2 + 12u_3 & \leq 4 \\ & u_1, u_2, u_3 \geq 0 \end{array}$$

Solve the dual (with AMPL+CPLEX): get $(u_1, u_2, u_3) = (0.6, 0, 0)$. Find (x_1, x_2) with complementary slackness:

$$\begin{array}{l} u_1(5x_1+6x_2-7)=0 \\ u_2(8x_1+9x_2-10)=0 \\ u_3(11x_1+12x_2-13)=0 \\ x_1(5u_1+8u_2+11u_3-3)=0 \\ x_2(6u_1+9u_2+12u_3-4)=0 \end{array} \Rightarrow \begin{array}{l} 0.6(5x_1+6x_2-7)=0 \\ 0(8x_1+9x_2-10)=0 \\ 0(11x_1+12x_2-13)=0 \\ x_1(5\cdot 0.6+8\cdot 0+11\cdot 0-3)=0 \\ x_2(6\cdot 0.6+9\cdot 0+12\cdot 0-4)=0 \end{array}$$

$$5x_1 + 6x_2 = 7 x_1 \cdot 0 = 0 x_2 \cdot (-0.4) = 0$$
 \Rightarrow
$$5x_1 + 6x_2 = 7 x_1 \cdot 0 = 0 x_2 = 0$$
 \Rightarrow $x_1 = \frac{7}{5} x_2 = 0$

Another example

Consider the following LP problem:

- 1. Write its dual.
- 2. Solve the dual through the graphical method.
- After finding the optimal value of the dual variables, use complementary slackness to find the optimal value of the primal variables.

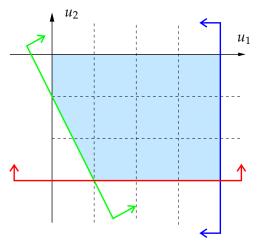
Another example: solution

The dual is

$$\begin{array}{cccc} \max & u_1 & +2u_2 \\ \text{s.t.} & -2u_1 & -u_2 & \leq 1 \\ & u_2 & \leq 2 \\ & -u_1 & -u_2 & \leq 3 \\ & -u_1 & & \geq -4 \\ & u_2 & \geq -3 \\ & u_1 \geq 0, u_2 \leq 0 \end{array}$$

Another example: solution

The second and third dual constraints are ignored as they are redundant to solve the problem.



The solution is clearly $(u_1, u_2) = (4, 0)$, corresponding to a value of 4 of the objective function.

Another example: solution

Complementarity slackness implies that:

$$x_1(-2u_1 - u_2 - 1) = 0$$

$$x_2(u_2 - 2) = 0$$

$$x_2(u_2 - 2) = 0$$

$$x_3(-u_1 - u_2 - 3) = 0$$

$$x_4(-u_1 + 4) = 0$$

$$x_4(u_1 + 4) = 0$$

$$x_5(u_2 + 3) = 0$$

$$u_1(-2x_1 - x_3 - x_4 - 1) = 0$$

which reduces, once we know the values of
$$u_1$$
 and u_2 , to:
$$x_1(-8-0-1)=0$$
$$x_2(0-2)=0$$

 $x_3(-4-0-3)=0$ $x_4(-4+4)=0$ $x_5(0+3)=0$

 $x_A = -2 \cdot 0 - 0 - 1 = -1.$

which implies $(x_1, x_2, x_3, x_5) = (0, 0, 0, 0)$, while

 $u_2(-x_1+x_2-x_3+x_5-2)=0$

 $-2x_1 - x_3 - x_4 = 1$

Maximum Flow

$$\begin{array}{ll} \max & \sum_{j \in V: (j,t) \in A} x_{jt} \\ \text{s.t.} & \sum_{j \in V: (i,j) \in A} x_{ij} = \sum_{j \in V: (j,i) \in A} x_{ji} & \forall i \in V: s \neq i \neq t \\ & 0 \leq x_{ij} \leq c_{ij} & \forall (i,j) \in A \end{array}$$

- \triangleright Variables for each node u_i for flow conservation constraints
- ▶ Variables for each arc z_{ij} for capacity constraints

$$\begin{array}{ll} \min & \sum_{(i,j) \in A} c_{ij} z_{ij} \\ \text{s.t.} & z_{ji} \geq u_j - u_i \quad \forall (i,j) \in A, \ i \neq s \ j \neq t \\ & z_{si} \geq u_i \quad \forall (s,i) \in A \\ & z_{it} \geq 1 - u_{it} \quad \forall (i,t) \in A \\ & 0 \leq z_{ij} \quad \forall (i,j) \in A \end{array}$$

From complementarity slackness: $z_{ij}(x_{ij} - c_{ij}) = 0$. What does it mean?

Transportation problem

Variables: qty of product from producer $i \in P$ to distributor $j \in D$: x_{ij} (non-negative)

Constraints:

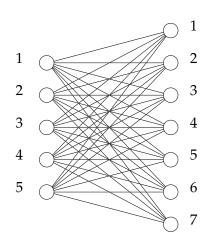
1. capacity:

$$\sum_{j\in D} x_{ij} \le p_i \quad \forall i\in P$$

2. demand:

$$\sum_{i\in P} x_{ij} \ge d_j \quad \forall j\in D$$

Objective function: total transportation cost, $\min \sum_{i \in P} \sum_{j \in D} c_{ij} x_{ij}$



Dual of the transportation problem

- ightharpoonup Variables for each supplier u_i for each supplier capacity constraints
- ightharpoonup Variables for each distributer v_j for each distributer demand constraints

$$\max \sum_{j \in D} d_j v_j - \sum_{i \in P} p_i u_i$$
s.t.
$$v_j - u_i \le c_{ij} \ \forall i \in P, \ j \in D$$

$$0 \le u_i, v_j \qquad \forall i \in P, \ j \in D$$

From complementarity slackness: $u_i(\sum_{j\in D} x_{ij} - p_i) = 0$ and $v_j(\sum_{i\in P} x_{ij} - d_j) = 0$.

From complementarity slackness: $x_{ij}(c_{ij} - v_j + u_i) = 0$. What does it mean? Only send product from i to j if the difference between the "fair market" buy price for i and cell price for j equals the transportation cost.

Shadow prices

Consider an LP problem $\min\{c^Tx : Ax \le b\}$. Suppose we solved it to the optimum and an optimal solution is x^* .

- associated with constraints
- ▶ if nonzero, the constraint is active : for inequality $a^T x \le b$, we have $a^T x^* = b$ (equality constraints are always active)
- it can be interpreted as the "marginal value" of the constraint (or of the resource/budget/limit/...the constraint is associated with)

Reduced costs

- associated with variables
- ightharpoonup if nonzero, the variable x_i is at its *lower* or *upper* bound
- ▶ gives an estimate of the "marginal value" of x_i
- ▶ i.e., if the coefficient of x_i in the objective function were lowered by that amount, the optimal solution would have $x_i \neq 0$.