DM545/DM871 Linear and Integer Programming

Lecture 6 More on Duality

Marco Chiarandini

Department of Mathematics & Computer Science University of Southern Denmark

Outline

1. Derivation Lagrangian Duality

2. Dual Simplex

3. Sensitivity Analysis

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Summary

- Derivation:
 - 1. economic interpretation
 - 2. bounding
 - 3. multipliers
 - 4. recipe
 - 5. Lagrangian
- Theory:
 - Symmetry
 - Weak duality theorem
 - Strong duality theorem
 - Complementary slackness theorem
- Dual Simplex
- Sensitivity Analysis, Economic interpretation

Outline

1. Derivation

Lagrangian Duality

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Derivation Dual Simplex Sensitivity Analysis

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Lagrangian Duality

Relaxation: if a problem is hard to solve then find an easier problem resembling the original one that provides information in terms of bounds. Then, search for the strongest bounds.

min
$$13x_1 + 6x_2 + 4x_3 + 12x_4$$

 $2x_1 + 3x_2 + 4x_3 + 5x_4 = 7$
 $3x_1 + 2x_3 + 4x_4 = 2$
 $x_1, x_2, x_3, x_4 \ge 0$

We wish to reduce to a problem easier to solve, ie:

min
$$c_1x_1 + c_2x_2 + \ldots + c_nx_n$$

 $x_1, x_2, \ldots, x_n \ge 0$

solvable by inspection: if c < 0 then $x = +\infty$, if $c \ge 0$ then x = 0. measure of violation of the constraints:

$$7 - (2x_1 + 3x_2 + 4x_3 + 5x_4) 2 - (3x_1 + 2x_3 + 4x_4)$$

We relax these measures in obj. function with Lagrangian multipliers y_1 , y_2 . We obtain a family of problems:

$$PR(y_1, y_2) = \min_{x_1, x_2, x_3, x_4 \ge 0} \left\{ \begin{aligned} & 13x_1 + 6x_2 + 4x_3 + 12x_4 \\ & + y_1(7 - 2x_1 - 3x_2 - 4x_3 - 5x_4) \\ & + y_2(2 - 3x_1 - 2x_3 - 4x_4) \end{aligned} \right\}$$

- 1. for all $y_1, y_2 \in \mathbb{R} : opt(PR(y_1, y_2)) \le opt(P)$
- 2. $\max_{y_1, y_2 \in \mathbb{R}} \{ \text{opt}(PR(y_1, y_2)) \} \le \text{opt}(P)$

PR is easy to solve.

(It can be also seen as a proof of the weak duality theorem)

$$PR(y_1, y_2) = \min_{x_1, x_2, x_3, x_4 \ge 0} \begin{cases} (13 - 2y_2 - 3y_2) x_1 \\ + (6 - 3y_1) x_2 \\ + (4 - 4y_1 - 2y_2) x_3 \\ + (12 - 5y_1 - 4y_2) x_4 \\ + 7y_1 + 2y_2 \end{cases}$$

if coefficient of x is < 0 then bound is $-\infty$ then LB is useless

$$(13 - 2y_2 - 3y_2) \ge 0$$

$$(6 - 3y_1) \ge 0$$

$$(4 - 4y_1 - 2y_2) \ge 0$$

$$(12 - 5y_1 - 4y_2) \ge 0$$

If they all hold then we are left with $7y_1 + 2y_2$ because all go to 0.

$$\max 7y_1 + 2y_2$$

$$2y_2 + 3y_2 \le 13$$

$$3y_1 \le 6$$

$$4y_1 + 2y_2 \le 4$$

$$5y_1 + 4y_2 \le 12$$

General Formulation

$$\begin{array}{ll} \min & z = \mathbf{c}^T \mathbf{x} & \quad \mathbf{c} \in \mathbb{R}^n \\ & A\mathbf{x} = \mathbf{b} & \quad A \in \mathbb{R}^{m \times n}, \mathbf{b} \in \mathbb{R}^m \\ & \quad \mathbf{x} \geq \mathbf{0} & \quad \mathbf{x} \in \mathbb{R}^n \end{array}$$

$$\begin{aligned} & \max_{\mathbf{y} \in \mathbb{R}^m} \{ \min_{\mathbf{x} \in \mathbb{R}^n_+} \{ \mathbf{c}^T \mathbf{x} + \mathbf{y}^T (\mathbf{b} - A \mathbf{x}) \} \} \\ & \max_{\mathbf{y} \in \mathbb{R}^m} \{ \min_{\mathbf{x} \in \mathbb{R}^n_+} \{ (\mathbf{c}^T - \mathbf{y}^T A) \mathbf{x} + \mathbf{y}^T \mathbf{b} \} \} \end{aligned}$$

$$\max_{A^T y} b^T y \\ y \in \mathbb{R}^m \le c$$

Derivation

Dual Simplex

Sensitivity Analysis

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Dual Simplex

 Dual simplex (Lemke, 1954): apply the simplex method to the dual problem and observe what happens in the primal tableau:

$$\max\{c^{T}x \mid Ax \le b, x \ge 0\} = \min\{b^{T}y \mid A^{T}y \ge c^{T}, y \ge 0\}$$
$$= -\max\{-b^{T}y \mid -A^{T}y \le -c^{T}, y \ge 0\}$$

• We obtain a new algorithm for the primal problem: the dual simplex lt corresponds to the primal simplex applied to the dual

Primal simplex on primal problem:

- 1. pivot > 0
- 2. $col c_j$ with wrong sign
- 3. row: min $\left\{\frac{b_i}{a_{ij}}: a_{ij} > 0, i = 1, ..., m\right\}$

Dual simplex on primal problem:

- 1. pivot < 0
- 2. row $b_i < 0$ (condition of feasibility)
- 3. col: min $\left\{ \left| \frac{c_j}{a_{ij}} \right| : a_{ij} < 0, j = 1, 2, ..., n + m \right\}$ (least worsening solution)

Dual Simplex

- 1. (primal) simplex on primal problem (the one studied so far)
- 2. Now: dual simplex on primal problem ≡ primal simplex on dual problem (implemented as dual simplex, understood as primal simplex on dual problem)

Uses of 1.:

- The dual simplex can work better than the primal in some cases.
 Eg. since running time in practice between 2m and 3m, then if m = 99 and n = 9 then better the dual
- Infeasible start
 Dual based Phase I algorithm (Dual-primal algorithm)

Dual Simplex for Phase I

Primal:

$$\begin{array}{ll} \max & -x_1 - x_2 \\ -2x_1 - x_2 \leq & 4 \\ -2x_1 + 4x_2 \leq -8 \\ -x_1 + 3x_2 \leq -7 \\ x_1, x_2 \geq & 0 \end{array}$$

Initial tableau

			x1	1	x2	1	w 1	1	w2		wЗ	1	-z	1	b	-
		-+-		+		+		+		+		+		+		-
			-2		-1	1	1		0		0	1	0		4	-
		-	-2	1	4	1	0	1	1	-	0	1	0	1	-8	
		-	- 1		3	-	0		0	-	1	-	0		-7	
++																
		- 1	- 1	1	- 1	1	0	1	0	-1	0	1	1	1	0	1

infeasible start

• x_1 enters, w_2 leaves

Dual:

$$\begin{array}{lll} \min & 4y_1 - 8y_2 - 7y_3 \\ -2y_1 - 2y_2 - & y_3 \ge -1 \\ -y_1 + 4y_2 + 3y_3 \ge -1 \\ & y_1, y_2, y_3 \ge & 0 \end{array}$$

• Initial tableau (min $bv \equiv -\max -bv$)

feasible start (thanks to $-x_1 - x_2$)

y₂ enters, z₁ leaves

• x_1 enters, w_2 leaves

1	1	x 1	1	x2	1	w 1	1	w 2	1	w3	1	-z	1	b	l
++															
		0	1	-5	1	1	1	- 1	1	0	1	0	1	12	l
		1	1	-2	1	0	1	-0.5	-	0	1	0	1	4	l
		0		1		0	1	-0.5		1		0		-3	l
++															
i	- 1	Λ	1	2	1	Λ	1	0.5	1	٥	1	- 1	1	1	Ĺ

• w₂ enters, w₃ leaves



(note that we kept $c_j < 0$, ie, optimality)

• y_2 enters, z_1 leaves

														b	
İ	1	1	1	1	I	0.5	1	0.5	1	0	1	0	l	0.5	i
	-+-		+		+		+		+		+		+	 -4	Ì

y₃ enters, y₂ leaves



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Economic Interpretation

$$\max 5x_0 + 6x_1 + 8x_2$$

$$6x_0 + 5x_1 + 10x_2 \le 60$$

$$8x_0 + 4x_1 + 4x_2 \le 40$$

$$4x_0 + 5x_1 + 6x_2 \le 50$$

$$x_0, x_1, x_2 \ge 0$$

final tableau:

- Which values have the variables, the reduced costs, the shadow prices (or marginal prices), the dual variables?
- If one slack variable > 0 then overcapacity: $s_2 = 2$ then the second constraint is not tight
- How many products can be produced at most? at most m
- How much more expensive a product not selected should be? look at reduced costs: $c_j + \pi a_j > 0$
- What is the value of extra capacity of manpower? In +1 out +1/5

Economic Interpretation

Game: Suppose two economic operators:

- P owns the factory and produces goods
- D is in the market buying and selling raw material and resources
- D asks P to close and sell him/her all resources
- P considers if the offer is convenient
- D wants to spend least possible
- y are prices that D offers for the resources
- $\sum y_i b_i$ is the amount D has to pay to have all resources of P
- $\sum y_i a_{ij} \ge c_i$ total value to make j > price per unit of product
- P either sells all resources $\sum y_i a_{ij}$ or produces product $j(c_j)$
- ullet without \geq there would not be negotiation because P would be better off producing and selling
- ▶ at optimality the situation is indifferent (strong th.)
- ▶ resource 2 that was not totally utilized in the primal has been given value 0 in the dual. (complementary slackness th.) Plausible, since we do not use all the resource, likely to place not so much value on it.
- ▶ for product $0 \sum y_i a_{ij} > c_j$ hence not profitable producing it. (complementary slackness th.)

Sensitivity Analysis aka Postoptimality Analysis

Instead of solving each modified problem from scratch, exploit results obtained from solving the original problem.

$$\max\{c^T x \mid Ax = b, l \le x \le u\}$$
 (*)

- (I) changes to coefficients of objective function: $\max\{\tilde{c}^Tx\mid Ax=b, l\leq x\leq u\}$ (primal) x^* of (*) remains feasible hence we can restart the simplex from x^*
- (II) changes to RHS terms: $\max\{c^Tx \mid Ax = \tilde{b}, l \leq x \leq u\}$ (dual) x^* optimal feasible solution of (*) basic sol \bar{x} of (II): $\bar{x}_N = x_N^*$, $A_B\bar{x}_B = \tilde{b} A_N\bar{x}_N$ \bar{x} is dual feasible and we can start the dual simplex from there. If \tilde{b} differs from b only slightly it may be we are already optimal.

Derivation Dual Simplex Sensitivity Analysis (primal)

(dual)

(III) introduce a new variable:

$$\max \quad \sum_{j=1}^{6} c_j x_j$$

$$\sum_{j=1}^{6} a_{ij} x_j = b_i, \ i = 1, \dots, 3$$

$$l_j \leq x_j \leq u_j, \ j = 1, \dots, 6$$

$$[x_1^*, \dots, x_6^*] \text{ feasible}$$

$$\max \sum_{j=1}^{7} c_j x_j$$

$$\sum_{j=1}^{7} a_{ij} x_j = b_i, \ i = 1, \dots, 3$$

$$l_j \leq x_j \leq u_j, \ j = 1, \dots, 7$$

$$[x_1^*, \dots, x_6^*, 0] \text{ feasible}$$

(IV) introduce a new constraint:

$$\sum_{i=0}^{6} a_{4j}x_j = b_4$$

$$[x_1^*,\ldots,x_6^*]$$
 optimal $[x_1^*,\ldots,x_6^*,x_7^*,x_8^*]$ dual feasible

$$\sum_{j=1}^{6} a_{5j} x_j = b_5$$

 $I_i < x_i < u_i$

$$b_5$$
 $i = 7.8$

$$x_7^* = b_4 - \sum_{j=1}^6 a_{4j} x_j^*$$

Examples

(I) Variation of reduced costs:

The last tableau gives the possibility to estimate the effect of variations

For a variable in basis the perturbation goes unchanged in the red. costs. Eg:

$$\max (6 + \delta)x_1 + 8x_2 \implies \bar{c}_1 = 1(6 + \delta) - \frac{2}{5} \cdot 5 - 1 \cdot 4 = \delta$$

then need to bring in canonical form and hence δ changes the obj value. For a variable not in basis, if it changes the sign of the reduced cost \Longrightarrow worth bringing in basis \Longrightarrow the δ term propagates to other columns

(II) Changes in RHS terms

(It would be more convenient to augment the second. But let's take $\epsilon=0$.) If $60+\delta \Longrightarrow$ all RHS terms change and we must check feasibility Which are the multipliers for the first row? $k_1=\frac{1}{6}, k_2=-\frac{1}{4}, k_3=0$

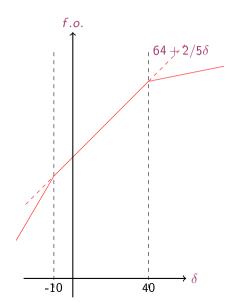
I:
$$1/5(60 + \delta) - 1/4 \cdot 40 + 0 \cdot 0 = 12 + \delta/5 - 10 = 2 + \delta/5$$

II:
$$-1/5(60 + \delta) + 1/2 \cdot 40 + 0 \cdot 0 = -60/5 + 20 - \delta/5 = 8 - 1/5\delta$$

Risk that RHS becomes negative

Eg: if $\delta = -10$ \Longrightarrow tableau stays optimal but not feasible \Longrightarrow apply dual simplex

Graphical Representation



(III) Add a variable

$$\begin{array}{ll} \max 5x_0 + 6x_1 + 8x_2 \\ 6x_0 + 5x_1 + 10x_2 \leq 60 \\ 8x_0 + 4x_1 + 4x_2 \leq 40 \\ x_0, x_1, x_2 \geq 0 \end{array}$$

Reduced cost of
$$x_0$$
? $c_j + \sum \pi_i a_{ij} = +1 \cdot 5 - \frac{2}{5} \cdot 6 + (-1)8 = -\frac{27}{5}$

To make worth entering in basis:

- increase its cost
- decrease the technological coefficient in constraint II: $5-2/5 \cdot 6 a_{20} > 0$

(IV) Add a constraint

Final tableau not in canonical form, need to iterate with dual simplex

(V) change in a technological coefficient:

- first effect on its column
- then look at c
- finally look at b

Relevance of Sensistivity Analysis

- The dominant application of LP is mixed integer linear programming.
- In this context it is extremely important being able to begin with a model instantiated in one form followed by a sequence of problem modifications
 - row and column additions and deletions,
 - variable fixings

interspersed with resolves

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