

## SE524/EC524 Optimization Theory and Methods

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### Lecture 9: Outline

- ➊ Complementary slackness.
- ➋ Geometry of duality.
- ➌ Economic interpretation: shadow prices.
- ➍ Dual simplex.
- ➎ Geometry of dual simplex.

## Complementary slackness

### Theorem

**[Complementary slackness (CS)]** Let  $\mathbf{x}$  primal feasible and  $\mathbf{p}$  dual feasible. Then  $\mathbf{x}, \mathbf{p}$  optimal iff

$$p_i(\mathbf{a}'_i \mathbf{x} - b_i) = 0, \quad \forall i$$

$$x_j(c_j - \mathbf{p}'\mathbf{A}_j) = 0, \quad \forall j$$

## Dual simplex

In simplex method  $\mathbf{p}' = \mathbf{c}'_B \mathbf{B}^{-1}$  (by proof of strong duality).

**Primal optimality condition** is

$$\mathbf{c}' - \mathbf{c}'_B \mathbf{B}^{-1} \mathbf{A} \geq \mathbf{0}$$

which is the same as **dual feasibility**.

This is a **primal algorithm**: maintains **primal feasibility** and works towards **dual feasibility**.

**Dual algorithm**: maintains **dual feasibility** and works towards **primal feasibility**.

**Dual simplex**: Move from dual bfs to dual bfs.

## Dual simplex (cont.)

$-\mathbf{c}'_B \mathbf{x}_B$	$\bar{c}_1$	...	$\bar{c}_n$
$x_{B(1)}$			
$\vdots$	$\mathbf{B}^{-1} \mathbf{A}_1$	...	$\mathbf{B}^{-1} \mathbf{A}_n$
$x_{B(m)}$			

Do not require  $\mathbf{B}^{-1} \mathbf{b} \geq 0$ , i.e., we are at a basic solution of the primal (not bfs).

Assume  $\bar{\mathbf{c}} \geq \mathbf{0}$  (dual feasibility).

Dual cost is

$$\mathbf{p}' \mathbf{b} = \mathbf{c}'_B \mathbf{B}^{-1} \mathbf{b} = \mathbf{c}'_B \mathbf{x}_B$$

i.e., the negative of upper left corner.

If  $\mathbf{B}^{-1} \mathbf{b} \geq 0$  then both dual feasibility and primal feasibility, and also same cost  $\Rightarrow$  **optimality**.

Otherwise, change basis.

## An iteration of dual simplex

- ➊ Start with basis matrix  $\mathbf{B}$  and all reduced costs  $\geq 0$ .
- ➋ If  $\mathbf{B}^{-1} \mathbf{b} \geq 0$  have an optimal solution, else choose  $l$  s.t.  $x_{B(l)} < 0$ .
- ➌ Consider the  $l$ th row (pivot row)  $x_{B(l)}, v_1, \dots, v_n$ . If  $\forall i v_i \geq 0$  then dual optimal cost =  $+\infty$  and algorithm terminates.
- ➍ Else, let  $j$  s.t.

$$\frac{\bar{c}_j}{|v_j|} = \min_{\{i | v_i < 0\}} \frac{\bar{c}_i}{|v_i|}$$

Pivot element  $v_j$ .  $\mathbf{A}_j$  enters the basis and  $\mathbf{A}_{B(l)}$  exits.

## Anticycling

If  $\bar{c}_j = 0$  then we don't move and algorithm can cycle. To avoid this

### Lexicographic pivoting

To find pivot element, for each column with  $v_j < 0$  divide all its entries by  $|v_j|$  and choose lexicographically smallest column. If there is a tie choose column with smallest index.

**When should we use the dual simplex ?** When is more convenient. Suppose that we used primal simplex to solve a problem, and we found the optimal. Say we now want to re-solve the problem with a new  $\mathbf{b}$ . Reduced costs of the original problem provide dual bfs for the new problem.