

# BV CH5: DUALITY

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## • LAGRANGIAN

Opt problem in standard form:

$$\begin{array}{ll} \text{minimize} & f_0(x), \quad x \in \mathbb{R}^n \\ \text{subject to} & f_i(x) \leq 0, \quad i=1 \dots m \\ & h_i(x) = 0, \quad i=1 \dots p \end{array} \quad \left. \begin{array}{l} (5.1) \\ \text{do NOT assume} \\ \text{convex!} \end{array} \right\}$$

$$D = \bigcap_{i=1}^m \text{dom } f_i \cap \bigcap_{i=1}^p \text{dom } h_i$$

$$L: \mathbb{R}^n \times \mathbb{R}^m \times \mathbb{R}^p \rightarrow \mathbb{R}$$

$$\rightarrow L(x, \lambda, v) = f_0(x) + \sum_{i=1}^m \lambda_i f_i(x) + \sum_{i=1}^p v_i h_i(x)$$

$$\text{dom } L = D \times \mathbb{R}^m \times \mathbb{R}^p$$

## • LAGRANGE DUAL FUNCTION

The Dual function finds the minimum value of the Lagrangian function over  $x$ .

$$g: \mathbb{R}^m \times \mathbb{R}^p \rightarrow \mathbb{R}, \quad \lambda \in \mathbb{R}^m, \quad v \in \mathbb{R}^p$$

$$g(\lambda, v) = \inf_{x \in D} L(x, \lambda, v) = \inf_{x \in D} \left( f_0(x) + \sum_{i=1}^m \lambda_i f_i(x) + \sum_{i=1}^p v_i f_i(x) \right)$$

When the Lagrangian is unbounded below in  $x$ , then the Dual function takes on the value  $-\infty$ .  
The Dual function is concave, even when the problem is not convex.

The Dual functions yields a lower bound on the optimal value of the problem.

$$\text{For } \lambda \geq 0 \text{ and any } v, \quad g(\lambda, v) \leq p^*$$

$$\tilde{x} \in D \quad f_i(\tilde{x}) \leq 0, \quad h_i(\tilde{x}) \leq 0, \quad \lambda \geq 0$$

$$\sum_{i=1}^m \lambda_i f_i(\tilde{x}) + \sum_{i=1}^p v_i h_i(\tilde{x}) \leq 0$$

$$L(\tilde{x}, \lambda, v) = f_0(\tilde{x}) + \sum_{i=1}^m \lambda_i f_i(\tilde{x}) + \sum_{i=1}^p v_i h_i(\tilde{x}) \leq f_0(\tilde{x})$$

$$L(\tilde{x}, \lambda, \nu) = f_0(\tilde{x}) + \sum_{i=1}^m \lambda_i f_i(\tilde{x}) + \sum_{i=1}^p \nu_i h_i(\tilde{x}) \leq f_0(\tilde{x})$$

$$g(\lambda, \nu) = \inf_{x \in D} L(x, \lambda, \nu) \leq L(\tilde{x}, \lambda, \nu) \leq f_0(\tilde{x})$$