

Duality

Lecture 12-2 - CMSE 382

Prof. Elizabeth Munch

Michigan State University

::

Dept of Computational Mathematics, Science & Engineering

Mon, April 13, 2026

Topics:

- Strong duality in the convex case
- Dual for linear programming

Announcements:

- Homework 6 is due on Friday, April 17, 2026 at 11:59pm.

Section 1

Last Time

Dual objective function

Consider the general model referred to as the **primal model**

$$f^* = \min f(\mathbf{x})$$

such that $g_i(\mathbf{x}) \leq 0, i = 1, 2, \dots, m,$

$$h_j(\mathbf{x}) = 0, j = 1, 2, \dots, p,$$

$\mathbf{x} \in X$, where $X \subseteq \mathbb{R}^n$,

and f, g_i, h_j are functions defined on X .

The Lagrangian of the problem is

$$L(\mathbf{x}, \boldsymbol{\lambda}, \boldsymbol{\mu}) = f(\mathbf{x}) + \sum_{i=1}^m \lambda_i g_i(\mathbf{x}) + \sum_{j=1}^p \mu_j h_j(\mathbf{x}),$$

The **dual objective function**

$q : \mathbb{R}_+^m \times \mathbb{R}^p \rightarrow \mathbb{R} \cup \{-\infty\}$ is

$$q(\boldsymbol{\lambda}, \boldsymbol{\mu}) = \min_{\mathbf{x} \in X} L(\mathbf{x}, \boldsymbol{\lambda}, \boldsymbol{\mu}),$$

Weak duality theorem

Primal Problem

$$f^* = \min f(\mathbf{x})$$

such that $g_i(\mathbf{x}) \leq 0, i = 1, 2, \dots, m,$

$$h_j(\mathbf{x}) = 0, j = 1, 2, \dots, p,$$

$\mathbf{x} \in X$, where $X \subseteq \mathbb{R}^n$,

and f, g_i, h_j are functions defined on X .

Dual Problem

$$q^* = \max q(\boldsymbol{\lambda}, \boldsymbol{\mu})$$

such that $(\boldsymbol{\lambda}, \boldsymbol{\mu}) \in \text{dom}(q),$

where $\text{dom}(q) = \{(\boldsymbol{\lambda}, \boldsymbol{\mu}) \in \mathbb{R}_+^m \times \mathbb{R}^p : q(\boldsymbol{\lambda}, \boldsymbol{\mu}) > -\infty\}$, and

$$q(\boldsymbol{\lambda}, \boldsymbol{\mu}) = \min_{\mathbf{x} \in X} L(\mathbf{x}, \boldsymbol{\lambda}, \boldsymbol{\mu}).$$

Theorem (Weak duality theorem)

Consider the primal problem and its dual. Then $q^* \leq f^*$, where q^* , f^* are the optimal *dual* and *primal* values, respectively.

► Desmos example

Section 2

Strong Duality

Strong duality of convex problems with equality & inequality constraints

Primal Problem (P)

$$\begin{aligned} f^* &= \min f(\mathbf{x}) \\ \text{such that } & g_i(\mathbf{x}) \leq 0, i = 1, 2, \dots, m, \\ & h_j(\mathbf{x}) \leq 0, j = 1, 2, \dots, p, \\ & s_k(\mathbf{x}) = 0, k = 1, 2, \dots, q, \\ & \mathbf{x} \in X, \end{aligned}$$

- For (P): X is a convex set and f, g_1, \dots, g_m are convex functions over X . The functions $h_1, \dots, h_p, s_1, \dots, s_q$ are affine.

Dual Problem (D)

$$\begin{aligned} q^* &= \max q(\boldsymbol{\lambda}, \boldsymbol{\mu}) \\ \text{such that } & (\boldsymbol{\lambda}, \boldsymbol{\mu}) \in \text{dom}(q), \end{aligned}$$

where $\text{dom}(q) = \{(\boldsymbol{\lambda}, \boldsymbol{\mu}) \in \mathbb{R}_+^m \times \mathbb{R}^p : q(\boldsymbol{\lambda}, \boldsymbol{\mu}) > -\infty\}$, and $q(\boldsymbol{\lambda}, \boldsymbol{\mu}) = \min_{\mathbf{x} \in X} L(\mathbf{x}, \boldsymbol{\lambda}, \boldsymbol{\mu})$.

Theorem (Strong duality under equality & inequality constraints)

If the generalized Slater's condition is satisfied in (P) and f^ has a finite optimal value, then the optimal value of (D) is attained, and the optimal values of the primal and dual problems are the same $f^* = q^*$.*

Definition (Duality gap)

The duality gap is the difference between the optimal primal value f^* and the optimal dual value q^* given by

$$\Delta = f^* - q^*$$

- $\Delta \geq 0$.
- $\Delta = 0$ if and only if strong duality holds.

Section 3

Dual for Linear Programming

Dual for linear programming

Primal

$$f^* = \min \quad \mathbf{c}^\top \mathbf{x}$$

such that $A\mathbf{x} \leq \mathbf{b}$,

Dual

$$q^* = \max \quad -\mathbf{b}^\top \boldsymbol{\lambda}$$

such that $A^\top \boldsymbol{\lambda} = -\mathbf{c}$,
 $\boldsymbol{\lambda} \geq 0$.

Strong duality holds

If the primal problem is feasible (*meaning the constraint set is not empty*) and has a finite solution, then the **optimal dual value** is equal to the **optimal primal value**:

$$q^* = f^*.$$

Groups - Round 5

Group 1

Michal, Kyle, Daniel,
Purvi

Group 2

Joseph, Jack, Scott,
Breena

Group 3

Saitej, Dori, Noah,
Tianjian

Group 4

Dev, Shanze, Lowell,
Andrew

Group 5

Lora, Aidan, Arjun,
Monirul Amin

Group 6

Anthony, Abigail,
Atticus, Yousif

Group 7

Luis, Vinod, Morgan,
Dominic

Group 8

Jay, Jonid, Alice, Aaron

Group 9

Arya, Jake, K M Tausif,
Lauryn

Group 10

Maye, Ha, Zheng, Sai

Group 11

Jamie, Karen, Brandon,
Quang Minh

Group 12

Long, Sanskaar,
Braedon, Igor