

## (Mostly) Convex Duality

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Problem Set 9 — Tuesday, May 11th

The exercises for this week will not count toward your grade, but you are highly encouraged to solve them all.

**Exercise 1: Different Duals**

Let  $G = (V, E)$  be a directed graph with capacities  $\mathbf{c} \in \mathbb{R}^E \geq \mathbf{0}$ , and edge-vertex incidence matrix  $\mathbf{B}$ , and consider a demand vector  $\mathbf{d} \in \mathbb{R}^V$  with  $\mathbf{d} \perp \mathbf{1}$ .

Let us try to minimize  $\|\mathbf{C}^{-1}\mathbf{f}\|_\infty$ , where  $\mathbf{C} = \text{diag}_{e \in E} \mathbf{c}(e)$ . This leads to an optimization problem

$$\begin{aligned} \min_{\mathbf{f} \in \mathbb{R}_{\geq 0}^E} \quad & \|\mathbf{C}^{-1}\mathbf{f}\|_\infty \\ \text{s.t.} \quad & \mathbf{B}\mathbf{f} = \mathbf{d} \end{aligned}$$

- Compute the dual of this problem. Note that we encoded the “constraint”  $\mathbf{f} \geq \mathbf{0}$  in the domain of the variable  $\mathbf{f}$ . This means that there will *not* be a dual variable associated with this constraint.
- Does Strong Duality hold for this pair of primal and dual problems?

Consider also a variant of this problem, where we instead treat  $\mathbf{f} \geq \mathbf{0}$  as an explicit constraint:

$$\begin{aligned} \min_{\mathbf{f} \in \mathbb{R}^E} \quad & \|\mathbf{C}^{-1}\mathbf{f}\|_\infty \\ \text{s.t.} \quad & \mathbf{B}\mathbf{f} = \mathbf{d} \\ & \mathbf{f} \geq \mathbf{0} \end{aligned}$$

This problem has a different dual problem because we made  $\mathbf{f} \geq \mathbf{0}$  an explicit constraint.

- Compute the dual of this variant of the problem.
- Explain how, given an optimal solution to the first dual problem, we can compute an optimal solution to the second dual problem.

**Exercise 2: A “Broken” Dual**

Consider the following optimization problem

$$\begin{aligned} \min_{x \in \mathbb{R}, y \in \mathbb{R}_{>0}} \quad & e^{-x} \\ \text{s.t.} \quad & x^2/y \leq 0. \end{aligned}$$

- Compute the dual program.
- What is the optimal value of the primal program? And of the dual program? Does strong duality hold? Does Slater's condition hold?

### Exercise 3: Norms and a Lagrangian

Suppose  $1 < q < p < \infty$ . In this exercise, we want to prove that for  $\mathbf{x} \in \mathbb{R}^n$ , we have

$$\|\mathbf{x}\|_q \leq n^{1/q-1/p} \|\mathbf{x}\|_p. \quad (1)$$

Consider the following optimization problem:

$$\max_{\mathbf{x} \in \mathbb{R}^n: \|\mathbf{x}\|_p \leq 1} \|\mathbf{x}\|_q^q$$

- Is this a convex optimization problem?
- Informally, explain why, at any maximizing  $\mathbf{x}$  for the above problem, there should exist a  $\lambda > 0$  such that

$$\nabla_{\mathbf{x}} \left( \|\mathbf{x}\|_q^q - \lambda \|\mathbf{x}\|_p^p \right) = \mathbf{0}.$$

- Use the existence of such a  $\lambda$  to prove Equation (??).

### Exercise 4: Flows and Voltages and Other Powers

Consider a connected undirected graph  $G = (V, E)$  with resistances  $\mathbf{r} \in \mathbb{R}^E$  and edge-vertex incidence matrix  $\mathbf{B}$ , and a demand vector  $\mathbf{d} \in \mathbb{R}^V$  with  $\mathbf{d} \perp \mathbf{1}$ .

Given some  $p > 1$ , we'll look at the flow optimization problem

$$\begin{aligned} \min_{\mathbf{f} \in \mathbb{R}^E} \sum_e \mathbf{r}(e) \frac{1}{p} |\mathbf{f}(e)|^p \\ \text{s.t. } \mathbf{B}\mathbf{f} = \mathbf{d}. \end{aligned}$$

- Is the above optimization problem convex?
- Does Slater's condition hold for the problem?
- What is the dual problem for the problem? Define  $q > 0$  to be the number such that  $1 = \frac{1}{q} + \frac{1}{p}$ . Try to find a clean expression of the dual. Writing the expression in terms of  $q$  instead of  $p$  will simplify it.

Suppose our instance of the optimization problem as an optimal flow solution  $\mathbf{f}^*$ . Let  $\alpha = \sum_e \mathbf{r}(e) \frac{1}{p} |\mathbf{f}^*(e)|^p$  be the optimal problem value. Suppose that for some particular edge  $\hat{e}$  we have

$$\gamma = \frac{\mathbf{r}(\hat{e}) \frac{1}{p} |\mathbf{f}^*(\hat{e})|^p}{\alpha}.$$

Now, consider a modified instance with resistances  $\tilde{\mathbf{r}} \in \mathbb{R}^E$  given by

$$\tilde{\mathbf{r}}(e) = \begin{cases} \mathbf{r}(e) & \text{for } e \neq \hat{e} \\ 2^{p-1}\mathbf{r}(e) & \text{for } e = \hat{e} \end{cases}$$

That is, we increase the resistance on edge  $\hat{e}$  by a factor  $2^{p-1}$ . Let  $\tilde{\alpha}$  denote optimal value of program with the new resistances  $\tilde{\mathbf{r}}$ . Prove that

$$\tilde{\alpha} \geq \left(1 + \frac{p-1}{2}\gamma\right) \alpha.$$

*Hint: use the dual problem!*