The University of Texas at Austin Department of Electrical and Computer Engineering

EE381K: Convex Optimization — Fall 2019

Lecture 20

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Goal: In this lecture, we talk about using dual problem for solving problem, tricks for writing different versions (possibly simpler) of the dual of a problem, and the dual of SOCP and SDP.

1 Solving the primal problem via the dual problem

In cases that strong duality holds, if we find optimal dual solutions (λ^*, ν^*) , then we know that the minimizer of $L(\mathbf{x}, \lambda^*, \nu^*)$ is primal optimal. Using this connection we can first write \mathbf{x}^* as a function of λ^*, ν^* and then solve the dual problem and find λ^*, ν^* and use the expression for \mathbf{x}^* to compute the optimal solution of the primal problem.

Least-norm solution of linear equations: Consider the following problem

$$\min : \mathbf{x}^{\top} \mathbf{x}$$
 s.t. $: \mathbf{A} \mathbf{x} = \mathbf{b}.$

We know that for this problem Slater's condition holds (if the problem is feasible) and we have strong duality. The Lagrangian in this case is defined as

$$L(\mathbf{x}, \boldsymbol{\lambda}, \boldsymbol{\nu}) = L(\mathbf{x}, \boldsymbol{\nu}) = \mathbf{x}^{\top}\mathbf{x} + \boldsymbol{\nu}^{\top}(\mathbf{A}\mathbf{x} - \mathbf{b}) = \mathbf{x}^{\top}\mathbf{x} + \mathbf{x}^{\top}\mathbf{A}^{\top}\boldsymbol{\nu} - \boldsymbol{\nu}^{\top}\mathbf{b}$$

According to the KKT conditions, we know that optimal \mathbf{x}^* and $\boldsymbol{\nu}^*$ satisfy

$$\nabla_{\mathbf{x}} L(\mathbf{x}^*, \boldsymbol{\nu}^*) = \mathbf{0} \quad \Leftrightarrow \quad 2\mathbf{x}^* + \mathbf{A}^{\top} \boldsymbol{\nu}^* = \mathbf{0} \quad \Leftrightarrow \quad \mathbf{x}^* = -\frac{1}{2} \mathbf{A}^{\top} \boldsymbol{\nu}^*$$

Further note that in this case, the dual problem is

$$\max : -\frac{1}{4} \boldsymbol{\nu}^{\top} \mathbf{A} \mathbf{A}^{\top} \boldsymbol{\nu} - \mathbf{b}^{\top} \boldsymbol{\nu}$$

Hence, if we solve the dual problem and find $\boldsymbol{\nu}^*$, we can use the relation $\mathbf{x}^* = -\frac{1}{2}\mathbf{A}^{\top}\boldsymbol{\nu}^*$ to find an optimal solution for the primal problem. For instance, when $\mathbf{A}\mathbf{A}^{\top}$ is invertible, we know that the optimal solution for the dual problem is $\boldsymbol{\nu}^* = -2(\mathbf{A}\mathbf{A}^{\top})^{-1}\mathbf{b}$ and as a result $\mathbf{x}^* = \mathbf{A}^{\top}(\mathbf{A}\mathbf{A}^{\top})^{-1}\mathbf{b}$ is an optimal solution for the primal problem.

2 Introducing new variables and equality constraints

Equivalent formulations of a problem can lead to very different duals. Therefore, reformulating the primal problem can be useful when the dual is difficult to derive, or uninteresting.

Common reformulations techniques:

- Introduce new variables and equality constraints (will use this trick for the dual of an SOCP)
- Make explicit constraints implicit or vice-versa

2.1 Example

Consider the following LP with box constraints:

$$\begin{aligned} \min : & \mathbf{c}^{\top} \mathbf{x} \\ \text{s.t.} : & \mathbf{A} \mathbf{x} = \mathbf{b} \\ & -1 \le \mathbf{x} \le \mathbf{1} \end{aligned}$$

Indeed, the dual problem can be written as an LP too and has the following form:

$$\begin{aligned} \max: & -\mathbf{b}^{\top} \boldsymbol{\nu} - \mathbf{1}^{\top} \boldsymbol{\lambda}_1 - \mathbf{1}^{\top} \boldsymbol{\lambda}_2 \\ \text{s.t.}: & \mathbf{c} + \mathbf{A}^{\top} \boldsymbol{\nu} + \boldsymbol{\lambda}_1 - \boldsymbol{\lambda}_2 = \mathbf{0} \\ & \boldsymbol{\lambda}_1 \geq \mathbf{0}, \quad \boldsymbol{\lambda}_2 \geq \mathbf{0}. \end{aligned}$$

But, we can also write the dual problem in a different way. Basically, we do not dualize the box constraint. To do so we rewrite the primal objective function in a way that is well defined if the box-constrained is satisfied.

min:
$$f_0(\mathbf{x}) = \begin{cases} \mathbf{c}^{\top} \mathbf{x} & \text{if } -1 \leq \mathbf{x} \leq 1 \\ \infty & \text{otherwise} \end{cases}$$

s.t.: $\mathbf{A}\mathbf{x} = \mathbf{b}$

Considering this new formulation the Lagrangian can be written as

$$L(\mathbf{x}, \boldsymbol{\nu}) = \inf_{-\mathbf{1} \le \mathbf{x} \le \mathbf{1}} \left(\mathbf{c}^{\top} \mathbf{x} + \boldsymbol{\nu}^{\top} (\mathbf{A} \mathbf{x} - \mathbf{b}) \right)$$
$$= -\boldsymbol{\nu}^{\top} \mathbf{b} + \inf_{-\mathbf{1} \le \mathbf{x} \le \mathbf{1}} \left((\mathbf{c} + \mathbf{A}^{\top} \boldsymbol{\nu})^{\top} \mathbf{x} \right)$$
$$= -\boldsymbol{\nu}^{\top} \mathbf{b} - \| \mathbf{c} + \mathbf{A}^{\top} \boldsymbol{\nu} \|_{1}$$

Hence, the dual problem is unconstrained and can be written as

$$\max: \quad -\boldsymbol{\nu}^{\top}\mathbf{b} - \|\mathbf{c} + \mathbf{A}^{\top}\boldsymbol{\nu}\|_{1}$$

3 Dual of SOCP

Recall that an SOCP in general can be written as

min:
$$\mathbf{q}^{\top} \mathbf{x}$$

s.t.: $\|\mathbf{A}_i \mathbf{x} + \mathbf{b}_i\|_2 < \mathbf{c}_i^{\top} \mathbf{x} + d_i, \quad i = 1, \dots, m.$

with primal variable $\mathbf{x} \in \mathbb{R}^n$ and $\mathbf{A}_i \in \mathbb{R}^{n_i \times n}$.

In this subsection, we study the dual of this problem. To do so, we introduce new variables $\mathbf{y}_i \in \mathbb{R}^{n_i}$ defined as $\mathbf{y}_i := \mathbf{A}_i \mathbf{x} + \mathbf{b}_i$ and $t_i \in \mathbb{R}$ defined as $t_i := \mathbf{c}_i^{\mathsf{T}} \mathbf{x} + d_i$. Then the original SOCP problem can be written as

min:
$$\mathbf{q}^{\top} \mathbf{x}$$

s.t.: $\|\mathbf{y}_i\|_2 \le t_i$, $i = 1, \dots, m$.
 $\mathbf{y}_i = \mathbf{A}_i \mathbf{x} + \mathbf{b}_i$, $i = 1, \dots, m$.
 $t_i = \mathbf{c}_i^{\top} \mathbf{x} + d_i$, $i = 1, \dots, m$.

Note that the primal variables for the new problem are $\mathbf{x} \in \mathbb{R}^n$, $\mathbf{y}_1 \in \mathbb{R}^{n_1}$, ..., $\mathbf{y}_m \in \mathbb{R}^{n_m}$ and $\mathbf{t} \in \mathbb{R}^m$. The Lagrangian of the new problem can be written as

$$L(\mathbf{x}, \mathbf{y}_{1}, \dots, \mathbf{y}_{m}, \mathbf{t}, \boldsymbol{\lambda}, \boldsymbol{\nu}, \boldsymbol{\mu})$$

$$= \mathbf{q}^{\top} \mathbf{x} + \sum_{i=1}^{m} \lambda_{i} (\|\mathbf{y}_{i}\|_{2} - t_{i}) + \sum_{i=1}^{m} \boldsymbol{\nu}_{i}^{\top} (\mathbf{y}_{i} - \mathbf{A}_{i} \mathbf{x} - \mathbf{b}_{i}) + \sum_{i=1}^{m} \mu_{i} (t_{i} - \mathbf{c}_{i}^{\top} \mathbf{x} - d_{i})$$

$$= \left(\mathbf{q} - \sum_{i=1}^{m} \mathbf{A}_{i}^{\top} \boldsymbol{\nu}_{i} - \sum_{i=1}^{m} \mu_{i} \mathbf{c}_{i} \right)^{\top} \mathbf{x} + \sum_{i=1}^{m} \left(\lambda_{i} \|\mathbf{y}_{i}\|_{2} + \boldsymbol{\nu}_{i}^{\top} \mathbf{y}_{i} \right) + \sum_{i=1}^{m} \left(-\lambda_{i} + \mu_{i} \right) t_{i} - \sum_{i=1}^{m} \left(\boldsymbol{\nu}_{i}^{\top} \mathbf{b}_{i} + \mu_{i} d_{i} \right)$$

Now to find the dual function we need to minimize the Lagrangian with respect to \mathbf{x} , \mathbf{y}_1 , ..., \mathbf{y}_m , t_1 , ..., t_m .

The minimum over \mathbf{x} is bounded below if and only if

$$\mathbf{q} - \sum_{i=1}^{m} \mathbf{A}_i^{\top} \boldsymbol{\nu}_i - \sum_{i=1}^{m} \mu_i \mathbf{c}_i = \mathbf{0} \quad \Leftrightarrow \quad \mathbf{q} = \sum_{i=1}^{m} (\mathbf{A}_i^{\top} \boldsymbol{\nu}_i + \mu_i \mathbf{c}_i)$$

To minimize over \mathbf{y}_i , we note that for any given λ_i and $\boldsymbol{\nu}_i$ we have that

$$\inf_{\mathbf{y}_{i}} \left(\lambda_{i} \| \mathbf{y}_{i} \|_{2} + \boldsymbol{\nu}_{i}^{\top} \mathbf{y}_{i} \right) = \inf_{r \geq 0, \|\mathbf{q}_{i}\| = 1} \left(\lambda_{i} \| r \mathbf{q}_{i} \|_{2} + r \boldsymbol{\nu}_{i}^{\top} \mathbf{q}_{i} \right) \\
= \inf_{r \geq 0, \|\mathbf{q}_{i}\| = 1} \left(\lambda_{i} r + r \boldsymbol{\nu}_{i}^{\top} \mathbf{q}_{i} \right) \\
= \inf_{r \geq 0, \|\mathbf{q}_{i}\| = 1} r \left(\lambda_{i} + \boldsymbol{\nu}_{i}^{\top} \mathbf{q}_{i} \right) \\
= \inf_{r \geq 0} r \left(\lambda_{i} + \boldsymbol{\nu}_{i}^{\top} \frac{-\boldsymbol{\nu}_{i}}{\|\boldsymbol{\nu}_{i}\|} \right) \\
= \inf_{r \geq 0} r \left(\lambda_{i} - \|\boldsymbol{\nu}_{i}\| \right) \\
= \begin{cases} 0 & \text{if } \|\boldsymbol{\nu}_{i}\| \leq \lambda_{i} \\ -\infty & \text{otherwise} \end{cases}$$

The minimum over t_i is bounded below if and only if

$$\lambda_i = \mu_i$$

Considering these results we obtain that the dual function $g(\lambda, \nu, \mu)$ is given by

$$g(\boldsymbol{\lambda}, \boldsymbol{\nu}, \boldsymbol{\mu}) = \begin{cases} -\sum_{i=1}^{m} \left(\boldsymbol{\nu}_{i}^{\top} \mathbf{b}_{i} + \mu_{i} d_{i}\right) & \text{if } \mathbf{q} = \sum_{i=1}^{m} (\mathbf{A}_{i}^{\top} \boldsymbol{\nu}_{i} + \mu_{i} \mathbf{c}_{i}), \ \|\boldsymbol{\nu}_{i}\| \leq \lambda_{i}, \ \lambda_{i} = \mu_{i}, \\ -\infty & \text{otherwise} \end{cases}$$

As a result, the dual problem of SCOP can be written as

$$\max : -\sum_{i=1}^{m} \left(\boldsymbol{\nu}_{i}^{\top} \mathbf{b}_{i} + \mu_{i} d_{i} \right)$$
s.t.:
$$\sum_{i=1}^{m} (\mathbf{A}_{i}^{\top} \boldsymbol{\nu}_{i} + \mu_{i} \mathbf{c}_{i}) = \mathbf{q}$$

$$\|\boldsymbol{\nu}_{i}\| \leq \lambda_{i}, \quad i = 1, \dots, m.$$

$$\lambda_{i} = \mu_{i}, \quad i = 1, \dots, m.$$

which can be simplified as

$$\max : -\sum_{i=1}^{m} \left(\boldsymbol{\nu}_{i}^{\top} \mathbf{b}_{i} + \lambda_{i} d_{i} \right)$$
s.t.:
$$\sum_{i=1}^{m} (\mathbf{A}_{i}^{\top} \boldsymbol{\nu}_{i} + \lambda_{i} \mathbf{c}_{i}) = \mathbf{q}$$

$$\|\boldsymbol{\nu}_{i}\| \leq \lambda_{i}, \qquad i = 1, \dots, m.$$

4 Dual of SDP

Recall the general formulation of SDP in inequality form

min:
$$\mathbf{c}^{\top}\mathbf{x}$$

s.t.: $x_1\mathbf{F}_1 + x_2\mathbf{F}_2 + \dots + x_n\mathbf{F}_n \prec \mathbf{G}$,

where $\mathbf{G}, \mathbf{F}_1, \dots, \mathbf{F}_n \in \mathbf{S}^k$ and $\mathbf{A} \in \mathbb{R}^{p \times n}$.

In this case, the Lagrange multiplier is matrix $\mathbf{Z} \in \mathbf{S}^k$ and the Lagrangian is

$$L(\mathbf{x}, \mathbf{Z}) = \mathbf{c}^{\top} \mathbf{x} + \mathbf{tr}(\mathbf{Z}(x_1 \mathbf{F}_1 + x_2 \mathbf{F}_2 + \dots + x_n \mathbf{F}_n - \mathbf{G}))$$

= $(c_1 + \mathbf{tr}(\mathbf{Z}\mathbf{F}_1))x_1 + \dots + (c_n + \mathbf{tr}(\mathbf{Z}\mathbf{F}_n))x_n - \mathbf{tr}(\mathbf{Z}\mathbf{G})$

Hence, the dual function is

$$g(\mathbf{Z}) = \begin{cases} -\mathbf{tr}(\mathbf{ZG}) & \text{if } c_i + \mathbf{tr}(\mathbf{ZF}_i) = 0 \text{ for } i = 1, \dots, n \\ -\infty & \text{otherwise} \end{cases}$$

and therefore the dual problem is

max:
$$-\mathbf{tr}(\mathbf{ZG})$$

s.t.: $\mathbf{Z} \succeq \mathbf{0}$, $\mathbf{tr}(\mathbf{ZF}_i) + c_i = 0$, $i = 1, ..., n$.

which is an SDP in standard form. Similar as LP, when we start with a problem in inequality form the dual is in standard form.

Note: Strong duality holds if primal SDP is strictly feasible $(\exists \mathbf{x} : x_1 \mathbf{F}_1 + x_2 \mathbf{F}_2 + \cdots + x_n \mathbf{F}_n \prec \mathbf{G})$.

If we write the primal problem in standard form as

min:
$$\mathbf{tr}(\mathbf{CX})$$

s.t.: $\mathbf{tr}(\mathbf{A}_i\mathbf{X}) = b_i$, for $i = 1, \dots, p$, $\mathbf{X} \succeq \mathbf{0}$,

then the Lagrange multipliers are $\Lambda \in \S^{n \times n}$ and $\nu \in \mathbb{R}^p$, and the Lagrangian is defined as

$$L(\mathbf{X}, \boldsymbol{\Lambda}, \boldsymbol{\nu}) = \mathbf{tr}(\mathbf{C}\mathbf{X}) - \mathbf{tr}(\boldsymbol{\Lambda}\mathbf{X}) + \sum_{i=1}^{p} \nu_i (\mathbf{tr}(\mathbf{A}_i\mathbf{X}) - b_i)$$

$$= \mathbf{tr}\left(\left(\mathbf{C} - \boldsymbol{\Lambda} + \sum_{i=1}^{p} \nu_i \mathbf{A}_i\right) \mathbf{X}\right) - \sum_{i=1}^{p} \nu_i b_i$$

$$= \mathbf{tr}\left(\left(\mathbf{C} - \boldsymbol{\Lambda} + \sum_{i=1}^{p} \nu_i \mathbf{A}_i\right) \mathbf{X}\right) - \boldsymbol{\nu}^{\top} \mathbf{b}$$

Therefore, the dual function is

$$g(\mathbf{Z}) = \begin{cases} -\boldsymbol{\nu}^{\top} \mathbf{b} & \text{if } \mathbf{C} - \boldsymbol{\Lambda} + \sum_{i=1}^{p} \nu_{i} \mathbf{A}_{i} = \mathbf{0} \\ -\infty & \text{otherwise} \end{cases}$$

and the dual problem is

$$\begin{aligned} & \max: & & -\boldsymbol{\nu}^{\top}\mathbf{b} \\ & \text{s.t.}: & & \mathbf{C} - \boldsymbol{\Lambda} + \sum_{i=1}^{p} \nu_{i} \mathbf{A}_{i} = \mathbf{0}, & & \boldsymbol{\Lambda} \succeq \mathbf{0} \end{aligned}$$

which can be simplified as

$$\max : -\boldsymbol{\nu}^{\top} \mathbf{b}$$
s.t.:
$$\sum_{i=1}^{p} \nu_{i} \mathbf{A}_{i} + \mathbf{C} \succeq \mathbf{0}$$

which is an SDP in inequality form.