

## 1. Sum of Squares

Given a polynomial  $p(t)$  in a single variable  $t$ , we are interested in knowing if we can write  $p(t)$  as a sum of squares of polynomials, i.e. whether we can write

$$p(t) = \sum_{j=1}^k (q_j(t))^2,$$

for some  $k \geq 1$  and some polynomials  $q_1(t), \dots, q_k(t)$ . This is an interesting question in many contexts, because if we could do this then we would know that  $p(t)$  is nonnegative for all values of  $t$ .<sup>1</sup>

First observe that if  $p(t)$  can be written as a sum of squares of polynomials then it must have even degree. We will therefore assume that  $p(t)$  has degree  $2d$  for some integer  $d \geq 1$  (the case  $d = 0$  corresponds to  $p(t)$  being a constant).

Let  $\vec{z} := \begin{bmatrix} 1 & t & \dots & t^d \end{bmatrix}^\top$ . Note that  $\vec{z}$  is a  $(d+1)$ -dimensional vector whose entries are polynomials in  $t$ .

- (a) Show that the polynomial  $p(t)$  of degree  $2d$  can be written as a sum of squares of polynomials if and only if there is a symmetric positive semidefinite matrix  $Q$  such that

$$p(t) = \vec{z}^\top Q \vec{z}.$$

(The equality here is an equality between polynomials in  $t$ .)

**Hint:** Every symmetric positive semidefinite matrix  $Q$  can be written as a sum of dyads, i.e.  $Q = \sum_{i=1}^n \vec{u}_i \vec{u}_i^\top$  if  $Q \in \mathbb{S}_+^n$ .

**Solution:** Suppose we can write

$$p(t) = \sum_{j=1}^k (q_j(t))^2,$$

for some  $k \geq 1$ . Note that each  $q_j(t)$  must have degree at most  $d$ . Let  $\vec{u}_j \in \mathbb{R}^{d+1}$  denote the vector of its coefficients, in the sense that

$$q_j(t) = \vec{u}_j^\top \vec{z}, \quad j = 1, \dots, k,$$

or equivalently,

$$q_j(t) = u_{j1} + u_{j2}t + \dots + u_{j(d+1)}t^d, \quad j = 1, \dots, k,$$

where  $\vec{u}_j = \begin{bmatrix} u_{j1} & \dots & u_{j(d+1)} \end{bmatrix}^\top \in \mathbb{R}^{d+1}$ . Then we have

$$p(t) = \sum_{j=1}^k (q_j(t))^2 = \sum_{j=1}^k (\vec{u}_j^\top \vec{z})^2 = \vec{z}^\top \left( \sum_{j=1}^k \vec{u}_j \vec{u}_j^\top \right) \vec{z}.$$

Defining  $Q := \sum_{j=1}^k \vec{u}_j \vec{u}_j^\top$ , note that  $Q$  is symmetric positive semidefinite and we have  $p(t) = \vec{z}^\top Q \vec{z}$ .

Conversely, if we could write  $p(t) = \vec{z}^\top Q \vec{z}$  for some symmetric positive semidefinite matrix  $Q$ , then, since we can write  $Q$  as a sum of  $k$  dyads, where  $k = \text{rank}(Q)$ , i.e.  $Q = \sum_{j=1}^k \vec{u}_j \vec{u}_j^\top$  for some  $\vec{u}_j \in \mathbb{R}^{d+1}$ ,  $1 \leq j \leq k$ , we would have

$$p(t) = \vec{z}^\top \left( \sum_{j=1}^k \vec{u}_j \vec{u}_j^\top \right) \vec{z} = \sum_{j=1}^k (\vec{u}_j^\top \vec{z})^2 = \sum_{j=1}^k (q_j(t))^2,$$

<sup>1</sup>In fact, it is known that if a polynomial  $p(t)$  in a single variable is nonnegative for all values of  $t$  then it can be written as a sum of squares of two polynomials, i.e.  $p(t) = r(t)^2 + s(t)^2$  for some polynomials  $r(t)$  and  $s(t)$ , but we do not need this fact.

where we define the polynomial  $q_j(t) := \vec{u}_j^\top \vec{z}$  for  $1 \leq j \leq k$ . This shows that  $p(t)$  can be written as a sum of squares of polynomials. We have thus shown that  $p(t)$  can be written as a sum of squares of polynomials if and only if we can write  $p(t) = \vec{z}^\top Q \vec{z}$  for some symmetric positive semidefinite matrix  $Q$ .

- (b) Show that we can pose the question of whether a given polynomial  $p(t)$  of degree  $2d$  can be written as a sum of squares of polynomials as a feasibility question for an SDP in standard form.

**Remark:** Recall that an SDP in standard form looks like:

$$\begin{aligned} \min_{X \in \mathbb{R}^{n \times n}} \quad & \text{trace}(CX) \\ \text{s.t.} \quad & \text{trace}(A_i X) = b_i, \text{ for each } i \in \{1, \dots, m\}, \\ & X \succeq 0. \end{aligned}$$

Here the minimization is over matrices  $X \in \mathbb{S}^n$ . The matrices  $C, A_1, \dots, A_m \in \mathbb{S}^n$  as well as the scalars  $b_1, \dots, b_m \in \mathbb{R}$  are given. The constraint  $X \succeq 0$  is the constraint that  $X$  should be symmetric positive semidefinite.

Also recall that to pose a minimization problem as a feasibility problem, we can just take the objective to be the constant 0 (so the question then just becomes whether the value of the problem is 0, in which case the problem is feasible, or  $\infty$ , in which case the problem is infeasible). For an SDP in standard form to be a feasibility problem, therefore, we could just take  $C$  to be the zero matrix.

**Solution:** Suppose

$$p(t) = b_1 + b_2 t + \dots + b_{2d+1} t^{2d},$$

where  $b_1, b_2, \dots, b_{2d+1} \in \mathbb{R}$ , and suppose

$$Q = \begin{bmatrix} q_{11} & q_{12} & \cdots & q_{1(d+1)} \\ q_{21} & q_{22} & & q_{2(d+1)} \\ \vdots & & & \vdots \\ q_{(d+1)1} & q_{(d+1)2} & \cdots & q_{(d+1)(d+1)} \end{bmatrix}.$$

The equation

$$p(t) = \vec{z}^\top Q \vec{z}$$

is equivalent to the equations

$$\text{trace}(A_i Q) = b_i, \text{ for each } i \in \{1, \dots, 2d+1\},$$

where  $A_i \in \mathbb{R}^{(d+1) \times (d+1)}$  is the matrix with zeros everywhere except for the coordinates  $(u, v)$  that satisfy  $u + v = i + 1$ .

The question of whether  $p(t)$  can be written as a sum of squares of polynomials can therefore be posed as the following SDP feasibility problem in standard form:

$$\begin{aligned} \min_X \quad & 0 \\ \text{s.t.} \quad & \text{trace}(A_i X) = b_i, \text{ for each } i \in \{1, \dots, 2d+1\}, \\ & X \succeq 0. \end{aligned}$$

Indeed, this problem is feasible if and only if there is a symmetric positive semidefinite  $Q$  such that  $p(t) = \vec{z}^\top Q \vec{z}$ .

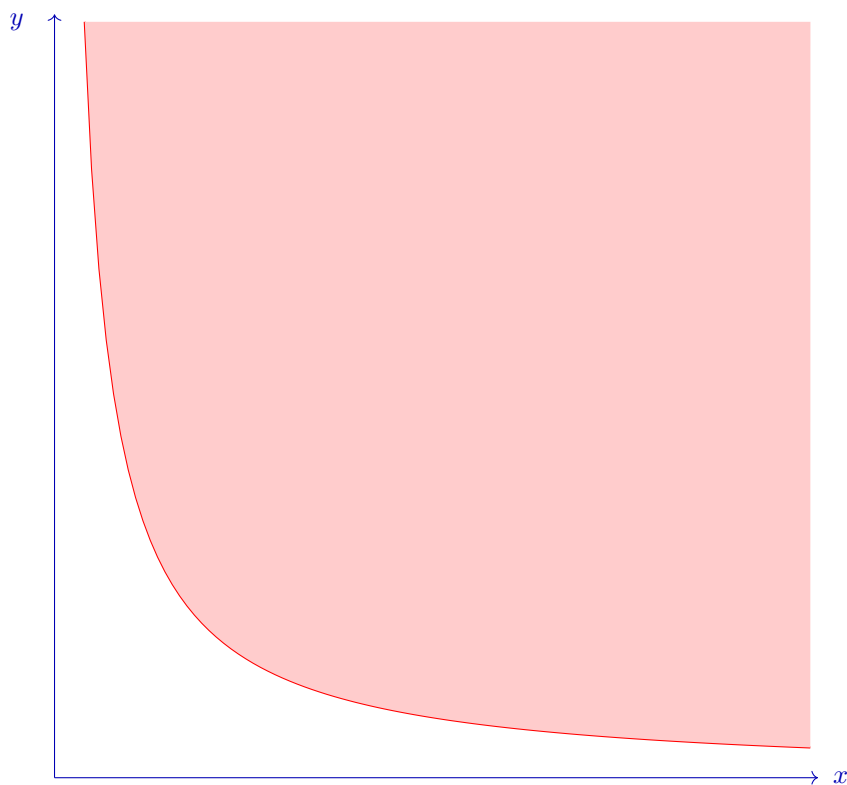
## 2. SDP Duality

Consider the following SDP in inequality form:

$$\begin{aligned} \min_{(x,y) \in \mathbb{R}^2} \quad & x \\ \text{s.t.} \quad & \begin{bmatrix} x & 1 \\ 1 & y \end{bmatrix} \succeq 0. \end{aligned} \tag{1}$$

(a) Draw the feasible set. Is it convex?

**Solution:** The symmetric matrix  $\begin{bmatrix} x & 1 \\ 1 & y \end{bmatrix}$  is symmetric positive semidefinite if and only if its diagonal entries and its determinant are nonnegative, i.e.  $x \geq 0$ ,  $y \geq 0$  and  $xy \geq 1$ . The region of  $(x, y)$  defined by these conditions is the epigraph of the convex function  $y = \frac{1}{x}$  with domain  $\mathbb{R}_{++} := \{(x, y) \in \mathbb{R}^2 | x, y > 0\}$ . Being the epigraph of a convex function, this set is a convex subset of  $\mathbb{R}^2$ . See Figure 1.



**Figure 1:** The domain of the primal SDP is the closed region above the curve  $xy = 1$  in the nonnegative quadrant. This is a convex set.

(b) Write the conic dual SDP in standard form.

**Solution:** If the primal SDP in inequality form is given by

$$\begin{aligned} \min_{\vec{x} \in \mathbb{R}^m} \quad & \vec{c}^\top \vec{x} \\ \text{s.t.} \quad & F_0 + \sum_{i=1}^m x_i F_i \succeq 0, \end{aligned}$$

where  $F_0, F_1, \dots, F_m \in \mathbb{S}^n$  and  $c \in \mathbb{R}^n$ , its dual is the SDP in standard form given by

$$\max_{X \in \mathbb{S}^n} \text{trace}(-F_0 X)$$

$$\begin{aligned} \text{s.t. } & \text{trace}(F_i X) = c_i, \text{ for each } i \in \{1, \dots, n\}, \\ & X \succeq 0. \end{aligned}$$

Here we have  $n = 2$ , with the vectors in  $\mathbb{R}^2$  written as  $\begin{bmatrix} x & y \end{bmatrix}^\top$ , and

$$c = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, F_0 = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, F_1 = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, F_2 = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}.$$

The dual of the primal SDP in (1) is given by

$$\begin{aligned} \max_{x_{11}, x_{12}, x_{22}} \quad & -2x_{12} \\ \text{s.t. } \quad & x_{11} = 1, \\ & x_{22} = 0, \\ & \begin{bmatrix} x_{11} & x_{12} \\ x_{12} & x_{22} \end{bmatrix} \succeq 0. \end{aligned} \tag{2}$$

To make it clear that it is in standard form, the dual of the primal SDP can be written explicitly as below if desired, although writing it in a more straightforward way as above would also be fine:

$$\begin{aligned} \max_{X \in \mathbb{S}^2} \quad & \text{trace} \left( \begin{bmatrix} 0 & -1 \\ -1 & 0 \end{bmatrix} X \right) \\ \text{s.t. } \quad & \text{trace} \left( \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} X \right) = 1, \\ & \text{trace} \left( \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} X \right) = 0, \\ & X \succeq 0. \end{aligned}$$

(c) Is the primal SDP feasible? Is it strictly feasible?

**Remark:** The SDP in inequality form

$$\begin{aligned} \min_{\vec{x} \in \mathbb{R}^m} \quad & \vec{c}^\top \vec{x} \\ \text{s.t. } \quad & F_0 + \sum_{i=1}^m x_i F_i \succeq 0, \end{aligned}$$

where  $F_0, F_1, \dots, F_m \in \mathbb{S}^n$ ,  $\vec{c} \in \mathbb{R}^n$ , is said to be strictly feasible if there is some  $\vec{x} \in \mathbb{R}^n$  such that  $F(\vec{x}) \in \mathbb{S}_{++}^n$ , i.e.  $F(\vec{x})$  is symmetric positive definite. Here  $F(\vec{x})$  denotes  $F_0 + \sum_{i=1}^m x_i F_i$ .

**Solution:** Since we can choose  $x$  and  $y$  such that  $\begin{bmatrix} x & 1 \\ 1 & y \end{bmatrix}$  is symmetric positive definite, the primal SDP in (1) is strictly feasible (and hence also feasible).

(d) Is the dual SDP feasible? Is it strictly feasible?

**Remark:** The SDP in standard form

$$\min_{X \in \mathbb{S}^n} \text{trace}(CX)$$

$$\begin{aligned} \text{s.t. } \text{trace}(A_i X) &= b_i, \text{ for each } i \in \{1, \dots, m\}, \\ X &\succeq 0, \end{aligned}$$

where  $C, A_1, \dots, A_m \in \mathbb{S}^m$ ,  $b_1, \dots, b_m \in \mathbb{R}$ , is said to be strictly feasible if there is some  $X \in \mathbb{S}_{++}^n$  (symmetric positive definite) satisfying the equality constraints  $\text{trace}(A_i X) = b_i$  for each  $i \in \{1, \dots, m\}$ .

**Solution:** From the dual SDP given in (2), we see that since we must have  $x_{22} = 0$ , the condition  $\begin{bmatrix} x_{11} & x_{12} \\ x_{12} & x_{22} \end{bmatrix} \succeq 0$  forces  $x_{12} = 0$ . Since we also require that  $x_{11} = 1$ , there is only one matrix  $X$  in the feasible set, namely  $\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$ . This matrix is not symmetric positive definite. We conclude that the dual SDP is feasible but not strictly feasible.

(e) Find the optimal primal value  $p^*$  and the optimal dual value  $d^*$ . Does strong duality hold?

**Solution:** From Figure 1 it is clear that the optimal primal value is  $p^* = 0$  (which is not attained, i.e. there is no optimal point). To see this analytically, first note that we must have  $x \geq 0$  at every feasible point. Further, for any  $\epsilon > 0$ , however small, the matrix  $\begin{bmatrix} \epsilon & 1 \\ 1 & \epsilon^{-1} \end{bmatrix}$  is symmetric positive semidefinite, so the objective function of the primal problem can be made to equal  $\epsilon$ .

As for the dual problem, it has only one feasible point and the value of the dual objective at that feasible point is 0, so we have  $d^* = 0$ .

Since  $p^* = d^*$ , strong duality holds.