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A SYMMETRIC DUAL THEOREM FOR
QUADRATIC PROGRAMS

W. S. DORN

New York University
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Duality relationships have been established previously [1, 2] for classes of quadratic programming problems. The present note presents a pair of dual quadratic programs, symmetric in the sense that the dual of the dual is clearly the original program again, and proves the duality relationship. As in [2] the proof rests on the dual theorem of linear programming [3].

In what follows capital letters A, Q, \dots denote matrices and small letters x, λ, \dots are column vectors. Prime denotes transpose and $x'y$ is the inner product of the vectors x and y . A vector inequality applies to each component of the vector, i. e., $x \geq 0$ implies that each component of x is non-negative.

The dual theorem for linear programs [3] states that if a solution to

$$\text{Minimize } p'x$$

subject to

$$\begin{aligned} Ax &\geq b \\ x &\geq 0 \end{aligned}$$

exists and is finite, then a solution also exists to its dual problem

$$\text{Maximize } b'v$$

subject to

$$\begin{aligned} A'v &\leq p \\ v &\geq 0 \end{aligned}$$

and moreover

$$\text{Minimum } p'x = \text{Maximum } b'v.$$

Consider now the following pair of quadratic programming problems

$$\left. \begin{aligned} \text{Minimize } f(z) &= z' \Omega z + \lambda' z \\ (\Omega + \Gamma)z &\geq \mu \\ z &\geq 0 \end{aligned} \right\} \quad \text{(I)}$$

$$\left. \begin{aligned} \text{Maximize } g(w) &= w'(-\Omega^{**})w + \mu'w \\ (-\Omega^* + \Gamma')w &\leq \lambda \\ w &\geq 0 \end{aligned} \right\} \quad \text{(II)}$$

where the matrices and vectors are partitioned as follows

$$\begin{aligned} \Omega &= \begin{Bmatrix} \frac{1}{2}C, & 0 \\ 0, & 0 \end{Bmatrix} & \Gamma &= \begin{Bmatrix} \frac{1}{2}C, & -C \\ 0, & A \end{Bmatrix} \\ \lambda &= \begin{pmatrix} 0 \\ p \end{pmatrix} & \mu &= \begin{pmatrix} 0 \\ b \end{pmatrix} \\ z &= \begin{pmatrix} y \\ x \end{pmatrix} & w &= \begin{pmatrix} u \\ v \end{pmatrix} \end{aligned}$$

and C is a symmetric, positive semi-definite, $n \times n$ matrix¹⁾, A is an $m \times n$ matrix, y, x, u, p are n dimensional vectors, and v, b are m dimensional vectors.

Notice that Ω is $2n \times 2n$, Γ is $(m+n) \times 2n$, λ and z are $2n \times 1$, μ and w are $(m+n) \times 1$. The matrix Ω^* in Problem II is equal to Ω with an appropriate number of columns of zeroes added or subtracted so that Ω^* is $2n \times (m+n)$. Similarly Ω^{**} is equal to Ω^* with the same number of rows of zeroes added or subtracted so that Ω^{**} is $(m+n) \times (m+n)$.

Dual Theorem: If a solution exists and is finite for either Problem I or Problem II, then a solution also exists for the other problem and

$$\text{Minimum } f(z) = \text{Maximum } g(w)$$

Proof: Assume a solution exists to Problem I which can be re-written.

1) The Symmetry restriction results in no loss of generality while the positive semi-definiteness assures that C and hence Ω is convex and thus that a local minimum of $f(z)$ is also a global one.

$$\begin{aligned} \text{Minimize } f^*(y, x) &= \frac{1}{2} y' C y + p' x \\ C y - C x &\geq 0 \\ A x &\geq b \\ y &\geq 0 \\ x &\geq 0 \end{aligned}$$

Let $y = \bar{y}$, $x = \bar{x}$ be the minimizing vector, and consider the linear problem

$$\text{Minimize } F(y, x) = -\frac{1}{2} y' C \bar{y} + \bar{y}' C y + p' x \tag{1}$$

$$C y - C x \geq 0 \tag{2}$$

$$A x \geq b \tag{3}$$

$$y \geq 0 \tag{4}$$

$$x \geq 0 \tag{5}$$

Denote this as Problem I'. Clearly \bar{y} , \bar{x} is a feasible solution to this problem. Assume there exists another feasible solution y^* , x^* such that

$$F(y^*, x^*) < F(\bar{y}, \bar{x}),$$

i. e.,

$$\bar{y}' C (y^* - \bar{y}) + p' (x^* - \bar{x}) < 0 \tag{6}$$

Define y_0 , x_0 as

$$\begin{aligned} y_0 &= k y^* + (1 - k) \bar{y} \\ 0 &< k < 1 \end{aligned}$$

$$x_0 = k x^* + (1 - k) \bar{x}$$

Then y_0 , x_0 is also a feasible solution and

$$f(y_0, x_0) - f(\bar{y}, \bar{x}) = k [\bar{y}' C (y^* - \bar{y}) + p' (x^* - \bar{x}) + \frac{1}{2} k (y^* - \bar{y})' C (y^* - \bar{y})]$$

By (6) the first term in square brackets is negative and from the positive semi-definiteness of C the second term is non-negative. If this second term vanishes, then the right hand member is negative. If the second term is positive then if k is sufficiently small, in fact, choosing k such that

$$k < \text{Min} \left\{ 1, -\frac{\bar{y}' C (y^* - \bar{y}) + p' (x^* - \bar{x})}{\frac{1}{2} (y^* - \bar{y})' C (y^* - \bar{y})} \right\}$$

the right hand member is negative. Thus, it is always possible to choose k so that

$$f(y_0, x_0) - f(\bar{y}, \bar{x}) < 0.$$

But \bar{y} , \bar{x} minimizes $f(y, x)$ in contradiction to this last inequality. Therefore,

$$F(\bar{y}, \bar{x}) \leq F(y^*, x^*)$$

and \bar{y} , \bar{x} minimizes Problem I.

From the dual theorem of linear programming, a solution exists to the dual to Problem I. The dual linear problem is

$$\begin{aligned} \text{Maximize } G(v) &= -\frac{1}{2} \bar{y}' C \bar{y} + b'v \\ Cu &\leq C \bar{y} \\ -Cu + A'v &\leq p \\ u &\geq 0 \\ v &\geq 0 \end{aligned}$$

which may be rephrased

$$\text{Maximize } G(v) = -\frac{1}{2} \bar{y}' C \bar{y} + b'v \quad (7)$$

$$A'v \leq p + C \bar{y} \quad (8)$$

$$v \geq 0 \quad (9)$$

Denote this as Problem II'. If $v = \bar{v}$ is the maximizing solution then by the dual theorem

$$b' \bar{v} = \bar{y}' C \bar{y} + p' \bar{x} \quad (10)$$

Now Problem II may be rewritten as

$$\text{Maximize } g^*(u, v) = -\frac{1}{2} u' C u + b'v \quad (11)$$

$$-Cu + A'v \leq p \quad (12)$$

$$u \geq 0 \quad (13)$$

$$v \geq 0 \quad (14)$$

Clearly $u = \bar{y}$, $v = \bar{v}$ is a feasible solution to Problem II.

Now since C is positive semi-definite

$$\frac{1}{2} (u - \bar{y})' C (u - \bar{y}) \geq 0$$

or

$$\frac{1}{2} u' C u \geq \bar{y}' C u - \frac{1}{2} \bar{y}' C \bar{y}$$

Subtracting $1/2 \bar{y}' C \bar{y}$ from both sides

$$\frac{1}{2} u' C u - \frac{1}{2} \bar{y}' C \bar{y} \geq \bar{y}' C (u - \bar{y})$$

So

$$\begin{aligned} g^*(\bar{y}, \bar{v}) - g^*(u, v) &= -\frac{1}{2}\bar{y}'C\bar{y} + \frac{1}{2}u'Cu + b'\bar{v} - b'v \\ &\geq \bar{y}'C(u - \bar{y}) + b'\bar{v} - b'v \end{aligned}$$

From (10) then

$$g^*(\bar{y}, \bar{v}) - g^*(u, v) \geq \bar{y}'Cu + p'\bar{x} - b'v \tag{15}$$

By (3) and (14)

$$-b'v \geq -\bar{x}'Av$$

and by (5) and (12)

$$p'\bar{x} \geq v'Ax - u'C\bar{x}$$

Putting these in (15)

$$g^*(\bar{y}, \bar{v}) - g^*(u, v) \geq u'(C\bar{y} - C\bar{x})$$

and from (2) and (13)

$$g^*(\bar{y}, \bar{v}) - g^*(u, v) \geq 0$$

Thus \bar{y}, \bar{v} is the maximizing solution to Problem II.

Finally from (10)

$$g^*(\bar{y}, \bar{v}) = -\frac{1}{2}\bar{y}'C\bar{y} + b'\bar{v} = \frac{1}{2}\bar{y}'C\bar{y} + p'\bar{x} = f^*(\bar{y}, \bar{x})$$

which verifies the equality of the objective functions.

From the symmetry of Problems I and II, it becomes obvious that the existence of a solution to Problem II also implies that a solution to Problem I exists.

Finally it should be noted that since C is symmetric, positive semi-definite, it may be taken to be a matrix all of whose elements are zero. In this special case the dual quadratic programs reduce to the corresponding dual linear programs previously stated.

REFERENCES

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