

# A linear algebra perspective on folding

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July 3, 2024

## 1 Main result

Consider the optimization problem

$$\begin{aligned} (\mathcal{P}_{init}) : \quad & \min c^T x \\ & \text{s.t. } Ax = b \\ & x \geq 0, \end{aligned}$$

with  $A \in \mathbb{R}^{m \times n}$ ,  $b \in \mathbb{R}^m$ ,  $c \in \mathbb{R}^n$ .

**Assumption 1** *There exist matrices  $P_1, P_2 \in \mathbb{R}^{m \times m'}$ ,  $Q \in \mathbb{R}^{n \times n'}$  and a matrix  $A' \in \mathbb{R}^{m' \times n'}$  such that*

- (i)  $AQ = P_1 A'$
- (ii)  $P_2^T A = A' Q^T$
- (iii)  $P_1 P_2^T b = b$
- (iv)  $QQ^T c = c$
- (v)  $Q \geq 0$ .

Using the notations introduced in Assumption 1, we can define another linear optimization problem

$$\begin{aligned} (\mathcal{P}_{small}) : \quad & \min c^T Qx' \\ & \text{s.t. } A'x' = P_2^T b \\ & x' \geq 0. \end{aligned}$$

**Lemma 1** *If  $x'$  is feasible for  $(\mathcal{P}_{small})$  then  $x = Qx'$  is feasible for  $(\mathcal{P}_{init})$ .*

*Proof:* The nonnegativity of  $x$  is implied by the fact that  $Q \geq 0$  and  $x' \geq 0$ . We then check that the constraints of  $(\mathcal{P}_{init})$ . To this end, we replace  $x$  by  $Qx'$  in the left-hand-sides of the constraints. We obtain successively

$$AQx' = P_1 A'x' = P_1 P_2^T b = b,$$

where the first equality is obtained by Assumption 1(i), the second equality by the feasibility of  $x'$  for  $(\mathcal{P}_{small})$  and the last equality by Assumption 1(iii).  $\square$

We now turn to the dual problems. Let us first write the problems. The dual of  $(\mathcal{P}_{init})$  reads

$$\begin{aligned} (\mathcal{D}_{init}) : \max & b^T u \\ \text{s.t.} & A^T u \leq c \\ & u \text{ free.} \end{aligned}$$

The dual of  $(\mathcal{P}_{small})$  reads

$$\begin{aligned} (\mathcal{D}_{small}) : \max & b^T P_2 u' \\ \text{s.t.} & A'^T u' \leq Q^T c \\ & u' \text{ free.} \end{aligned}$$

**Lemma 2** *If  $u'$  is feasible for  $(\mathcal{D}_{small})$  then  $u = P_2 u'$  is dual feasible for  $(\mathcal{D}_{init})$ .*

*Proof:* We replace  $u$  by  $P_2 u'$  in the constraints of  $(\mathcal{D}_{init})$  and we obtain

$$A^T P_2 u' = Q A'^T u' \leq Q Q^T c = c,$$

where the first equality is obtained by Assumption 1(ii), the inequality is obtained by the dual feasibility of  $u'$  and the fact that  $Q \geq 0$  and the last equality by Assumption 1(iv).  $\square$

**Lemma 3** *The objective value of a feasible solution  $x'$  of  $(\mathcal{P}_{small})$  is equal to the corresponding objective value of  $x = Qx'$ .*

*The objective value of a dual-feasible solution  $u'$  of  $(\mathcal{D}_{small})$  is equal to the corresponding objective value of  $u = P_2 u'$ .*

*Proof:* This follows from the fact that  $c^T x = c^T Qx'$  and  $b^T u = b^T P_2 u'$ .  $\square$

We conclude with the following corollary.

**Corollary 1** *If  $(x', u')$  is an optimal primal-dual pair for  $(\mathcal{P}_{small}) - (\mathcal{D}_{small})$  then  $(x, u) = (Qx', P_2 u')$  is an optimal primal-dual pair for  $(\mathcal{P}_{init}) - (\mathcal{D}_{init})$ .*

## 2 One way to construct a matrix $Q$

One way to construct a valid (orthogonal) matrix  $Q$  that satisfies Assumption 1(iv) is explained in the following. Assume that we split the vector  $c$  in  $r$  blocks  $c = (c^{(1)} \dots c^{(r)})$ . Furthermore, we assume that the sign of every element in the same block  $c^{(i)}$  is the same.

**Lemma 4** *The matrix*

$$Q = \left( \begin{array}{c|c|c} \frac{|c^{(1)}|}{\|c^{(1)}\|} & & \\ \hline & \frac{|c^{(2)}|}{\|c^{(2)}\|} & \\ \hline & & \ddots \end{array} \right)$$

*satisfies part (iv) and (v) of Assumption 1. Furthermore  $Q^T Q = I$ .*

*Proof:* The nonnegativity of  $Q$  is straightforward.

We compute

$$QQ^T = \left( \begin{array}{c|c|c} \frac{c^{(1)}(c^{(1)})^T}{\|c^{(1)}\|^2} & & \\ \hline & \frac{c^{(2)}(c^{(2)})^T}{\|c^{(2)}\|^2} & \\ \hline & & \ddots \end{array} \right).$$

Observe that the absolute value can be dropped because all components of  $c^{(i)}$  have the same sign. Hence

$$QQ^T c = \begin{pmatrix} c^{(1)} \frac{(c^{(1)})^T c^{(1)}}{\|c^{(1)}\|^2} \\ c^{(2)} \frac{(c^{(2)})^T c^{(2)}}{\|c^{(2)}\|^2} \\ \vdots \end{pmatrix} = c.$$

Similarly, it can be readily verified that  $Q^T Q = I$ . □

### 3 Fixing $Q$

**Assumption 2** *We assume that  $Q$  is fixed as in Lemma 4 and has the form*

$$Q = \left( \begin{array}{c|c|c} \frac{|c^{(1)}|}{\|c^{(1)}\|} & & \\ \hline & \frac{|c^{(2)}|}{\|c^{(2)}\|} & \\ \hline & & \ddots \end{array} \right).$$

We explore the complexity of finding  $P_1, P_2$  that allow us to satisfy conditions (i)-(iii) of Assumption 1.

**Lemma 5** *If Assumption 2 is satisfied, then the columns of  $Q$  are eigenvectors of  $QQ^T$  with eigenvalue 1 and all other eigenvalues are 0.*

*Proof:* This follows from the fact that  $QQ^T Q = Q$  and the fact that  $QQ^T$  is of rank  $n'$ . □

If  $Q^T Q = I$ , and if we multiply condition (ii) of 1 by  $Q$ , we observe that  $A'$  must take the form  $A' = P_2^T A Q$ . However, assuming  $A' = P_2^T A Q$  does not necessarily imply that condition (ii) of Assumption 1 is satisfied. The following lemmas show under which conditions this holds true.

**Lemma 6** We consider (a) the linear space  $L \subseteq \mathbb{R}^{m+n'}$  defined by  $L := \ker([A^T \mid -Q])$ , (b) its projection  $\text{proj}_{\mathbb{R}^m} L \subseteq \mathbb{R}^m$  be on the first  $m$  components, (c) a basis  $\{\beta_1, \dots, \beta_k\} \subset \mathbb{R}^m$  of this projection, where  $k = \dim(\text{proj}_{\mathbb{R}^m} L \subseteq \mathbb{R}^m)$  and (d) the matrix  $B := [\beta_1 \mid \dots \mid \beta_k] \in \mathbb{R}^{m \times k}$ . Under Assumption 2, for all  $U \in \mathbb{R}^{k \times m'}$ , letting  $P_2 = BU$  implies that Assumption 1(ii) is met.

*Proof:* We want to prove that assuming  $P_2 = BU$  implies that  $P_2^T A = P_2^T A Q Q^T$ . By (c), (d) and the construction of  $P_2 = BU$ , every column  $p$  of  $P_2$  satisfies  $p \in \text{proj}_{\mathbb{R}^m} L$ . Using (b), we know that for every column  $p \in \mathbb{R}^m$  of  $P_2$ , there exists a vector  $v \in \mathbb{R}^{n'}$  such that  $\begin{pmatrix} p \\ v \end{pmatrix} \in L$ . As a result, by (a), we have  $([A^T \mid -Q]) \begin{pmatrix} p \\ v \end{pmatrix} = 0$ . Writing this for every column  $p$  of  $P_2$ , we get that there exists a matrix  $V \in \mathbb{R}^{n' \times m'}$  such that  $([A^T \mid -Q]) \begin{pmatrix} P_2 \\ V \end{pmatrix} = 0$ , i.e.  $A^T P_2 = QV$ . So, regardless of the values in  $V$ , the columns of  $A^T P_2$  belong to the column space of  $Q$ . By Lemma 5, they belong to the eigenspace of  $Q Q^T$  for the eigenvalue 1, hence  $Q Q^T A^T P_2 = A^T P_2$  which is exactly the transpose of what we want to prove.  $\square$

In the following, we want to find  $P_1$  such that conditions (i) and (iii) of Assumption 1 are met. To present the next results, we need a few definitions.

**Definition 1** For  $A \in \mathbb{R}^{m \times n}$  and  $Q \in \mathbb{R}^{n \times n'}$ ,  $b \in \mathbb{R}^m$ , we define:

- $B \in \mathbb{R}^{m \times k}$  to be the projection on the first  $m$  coordinates of a basis of  $\ker([A^T \mid -Q])$ ,
- $M \in \mathbb{R}^{m \times (m-l)}$  be an orthogonal basis of the column space of  $(AQ \mid b)$ ,
- $C \in \mathbb{R}^{m \times l}$  to be a basis of the orthogonal space of the column space of  $(AQ \mid b)$ .

**Lemma 7** Under Assumption 2, assuming  $P_2 = BU$ , conditions (i) and (iii) of Assumption 1 are equivalent to finding matrices  $P \in \mathbb{R}^{m \times k}$  and  $V \in \mathbb{R}^{l \times m}$  such that

$$PB^T = MM^T + V^T C^T. \quad (1)$$

*Proof:* Using Lemma 6 and Assumption 2, we know that  $A' = P_2 A^T Q$ . If we rewrite conditions (i) and (iii) using  $A' = P_2^T A Q$ , we obtain respectively  $P_1 P_2^T A Q = A Q$  and  $P_1 P_2^T b = b$  which can be reinterpreted as the fact that the columns of  $AQ$  and  $b$  must be eigenvectors of  $P_1 P_2^T$  with respect to the eigenvalue 1. Since  $P_2 = BU$ , it means that  $P_1 P_2^T$  can be written  $P_1 U^T B^T$  which in turn must have the columns of  $AQ$  and  $b$  as eigenvectors with eigenvalue 1. This can be rewritten as

$$P_1 U^T B^T = MM^T + V^T C^T,$$

where  $V \in \mathbb{R}^{l \times m}$ .  $\square$

Observe that (1) is a linear system with  $m(k + l)$  unknowns and  $m^2$  equations. Finding a solution with  $P$  with a lower rank allows us to find a folded problem with a lower size.