

2.1 Let  $n \geq 2$ . Given the matrix  $A = (a_{ij}) \in \mathbb{R}^{n \times n}$ , the permutation matrix  $Q \in \mathbb{R}^{n \times n}$  reverses the order of the rows of  $A$ , so that  $(QA)_{i,j} = a_{n+1-i,j}$ . If  $L \in \mathbb{R}^{n \times n}$  is a lower triangular matrix, what is the structure of the matrix  $QLQ$ ?

Show how to factorise  $A \in \mathbb{R}^{n \times n}$  in the form  $A = UL$ , where  $U \in \mathbb{R}^{n \times n}$  is unit upper triangular and  $L \in \mathbb{R}^{n \times n}$  is lower triangular. What conditions on  $A$  will ensure that the factorisation exists? Give an example of a square matrix  $A$  which cannot be factorised in this way.

Since  $Q$  is a permutation matrix, then it is orthogonal. Thus  $Q^{-1}$  exists and  $Q^{-1} = Q^T$ . Since reversing the rows twice is the identity operation, then  $Q^{-1} = Q$ . It follows that  $Q^T = Q$  so the matrix  $Q$  is symmetric.

First note that  $L^T$  is upper triangular. Thus

$$(LQ)^T = Q^T L^T = QL^T$$

is an upper triangular matrix with the order of the rows reversed. Diagrammatically it looks like

$$L^T = \begin{bmatrix} x & x & \dots & x \\ & x & & \vdots \\ \ominus & & \ddots & \vdots \\ & & & x \end{bmatrix} \quad QL^T = \begin{bmatrix} \ominus & & & x \\ & x & & \vdots \\ x & x & \dots & x \end{bmatrix}$$

where the  $x$ 's represent possibly non-zero entries. Transposing this matrix does not change the structure, thus

$$LQ = (LQ)^{TT}$$

Also looks like

$$LQ = \begin{bmatrix} \mathcal{O} & & & x \\ & x & & \vdots \\ x & x & \dots & x \end{bmatrix}$$

Finally  $QLQ$  is obtained by reversing the rows of  $LQ$  to obtain a matrix that looks like

$$QLQ = \begin{bmatrix} x & x & \dots & x \\ & x & & \vdots \\ \mathcal{O} & & & x \end{bmatrix}$$

which is upper triangular.

To see how to factorize  $A = UL$  consider  $A^T$  and factor this matrix so  $QAQ = LU$  using the usual method by Gaussian elimination. This method succeeds if no row swaps need to be made while performing the elimination steps.

Since  $QQ = I$  it follows that

$$A = QLUL = QL I UL = QL(QQ)UL = (QLQ)(QUQ)$$

By what we've just shown  $QLQ$  is upper triangular. Similar reasoning implies  $QUQ$  is lower triangular. Setting

$$\tilde{U} = QLQ \quad \text{and} \quad \tilde{L} = QUQ$$

now yields the factorization  $A = \tilde{U}\tilde{L}$  where  $\tilde{U}$  is upper triangular and  $\tilde{L}$  is lower triangular.

Suppose  $Q A Q = \begin{bmatrix} 0 & 1 \\ 2 & 3 \end{bmatrix}$ . Then there is a 0 in the pivot position so  $Q A Q$  can't be factored as LU.

Noting that  $Q = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$  and solving back for  $A$  yields

$$A = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 2 & 3 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 3 & 2 \end{bmatrix} = \begin{bmatrix} 3 & 2 \\ 1 & 0 \end{bmatrix}$$

It follows that  $\begin{bmatrix} 3 & 2 \\ 1 & 0 \end{bmatrix}$  can't be factored as  $\tilde{U}\tilde{L}$ .

2.8 (i) Show that, for any vector  $\mathbf{v} = (v_1, \dots, v_n)^T \in \mathbb{R}^n$ ,

$$\|\mathbf{v}\|_\infty \leq \|\mathbf{v}\|_2 \quad \text{and} \quad \|\mathbf{v}\|_2^2 \leq \|\mathbf{v}\|_1 \|\mathbf{v}\|_\infty.$$

In each case give an example of a nonzero vector  $\mathbf{v}$  for which equality is attained. Deduce that  $\|\mathbf{v}\|_\infty \leq \|\mathbf{v}\|_2 \leq \|\mathbf{v}\|_1$ . Show also that  $\|\mathbf{v}\|_2 \leq \sqrt{n} \|\mathbf{v}\|_\infty$ .

(ii) Show that, for any matrix  $A \in \mathbb{R}^{m \times n}$ ,

$$\|A\|_\infty \leq \sqrt{n} \|A\|_2 \quad \text{and} \quad \|A\|_2 \leq \sqrt{m} \|A\|_\infty.$$

In each case give an example of a matrix  $A$  for which equality is attained. (See the footnote following Definition 2.12 for the meaning of  $\|A\|_1$ ,  $\|A\|_2$  and  $\|A\|_\infty$  when  $A \in \mathbb{R}^{m \times n}$ .)

(i) Let  $\mathbf{v} \in \mathbb{R}^n$ . Now for each  $i$  we have

$$|v_i|^2 \leq |v_1|^2 + |v_2|^2 + \dots + |v_n|^2,$$

Thus

$$|v_i| \leq \sqrt{|v_1|^2 + \dots + |v_n|^2} = \|\mathbf{v}\|_2.$$

Then

$$\begin{aligned} \|\mathbf{v}\|_\infty &= \max\{|v_1|, |v_2|, \dots, |v_n|\} \\ &\leq \max\{\|\mathbf{v}\|_2, \|\mathbf{v}\|_2, \dots, \|\mathbf{v}\|_2\} = \|\mathbf{v}\|_2 \end{aligned}$$

Therefore  $\|\mathbf{v}\|_\infty \leq \|\mathbf{v}\|_2$ .

To see the other inequality estimate

$$\|v\|_2^2 = |v_1|^2 + |v_2|^2 + \dots + |v_n|^2 = \sum_{i=1}^n |v_i| |v_i|$$

$$\leq \sum_{i=1}^n \max\{|v_1|, |v_2|, \dots, |v_n|\} |v_i| = \sum_{i=1}^n \|v\|_\infty |v_i|$$

$$\Rightarrow \|v\|_\infty \sum_{i=1}^n |v_i| = \|v\|_\infty \|v\|_1.$$

Therefore

$$\|v\|_2^2 \leq \|v\|_\infty \|v\|_1.$$

Since  $\|v\|_\infty \leq \|v\|_2$  from before, substituting yields

$$\|v\|_2^2 \leq \|v\|_\infty \|v\|_1 \leq \|v\|_2 \|v\|_1.$$

Now if  $\|v\|_2 = 0$  then  $v=0$  and clearly

$$\|v\|_\infty \leq \|v\|_2 \leq \|v\|_1$$

If  $\|v\|_2 \neq 0$  we can cancel and obtain  $\|v\|_2 \leq \|v\|_1$ . In this case it again follows that

$$\|v\|_\infty \leq \|v\|_2 \leq \|v\|_1.$$

Finally we show that  $\|v\|_2 \leq \sqrt{n} \|v\|_\infty$ . To see this note that

$$\begin{aligned} \|v\|_2^2 &= \sum_{i=1}^n |v_i|^2 \leq \sum_{i=1}^n \max\{|v_1|^2, |v_2|^2, \dots, |v_n|^2\} \\ &= \sum_{i=1}^n \|v\|_\infty^2 = n \|v\|_\infty^2. \end{aligned}$$

Taking square roots now yields that

$$\|v\|_2 \leq \sqrt{n} \|v\|_\infty.$$

(ii) Let  $A \in \mathbb{R}^{m \times n}$ . By definition

$$\begin{aligned} \|A\|_\infty &= \max \{ \|Ax\|_\infty : \|x\|_\infty = 1 \} \\ &\leq \max \{ \|Ax\|_2 : \|x\|_\infty = 1 \} \end{aligned}$$

since the vector norms  $\|Ax\|_\infty \leq \|Ax\|_2$ . Now

$$\|x\|_2 \leq \sqrt{n} \|x\|_\infty$$

implies

$$\{x : \|x\|_\infty = 1\} \subseteq \{x : \|x\|_2 \leq \sqrt{n}\}$$

Therefore taking the max. over a larger set yields

$$\|A\|_\infty \leq \max \{ \|Ax\|_2 : \|x\|_2 \leq \sqrt{n} \}$$

On the other hand

$$\|A\|_2 = \max \{ \|Ax\|_2 : \|x\|_2 = 1 \}$$

Now if  $0 < \|x\|_2 < 1$  then  $\frac{x}{\|x\|_2} = 1$  and

$$\begin{aligned}\|Ax\|_2 &= \|Ax\|_2 \frac{\|x\|_2}{\|x\|_2} = \left\| A \frac{x}{\|x\|_2} \right\|_2 \|x\|_2 \\ &\leq \|A\|_2 \|x\|_2 \leq \|A\|_2\end{aligned}$$

Consequently

$$\max \{ \|Ax\|_2 : \|x\|_2 \leq 1 \} \leq \|A\|_2$$

and so

$$\|A\|_2 = \max \{ \|Ax\|_2 : \|x\|_2 \leq 1 \}.$$

To finish note that  $\|Ax\|_2 \frac{\sqrt{n}}{2\sqrt{n}} = \sqrt{n} \|A \frac{x}{\sqrt{n}}\|_2$ . Thus

$$\max \{ \|Ax\|_2 : \|x\|_2 \leq 1 \} = \max \left\{ \frac{1}{\sqrt{n}} \|A \sqrt{n}x\|_2 : \|x\|_2 \leq 1 \right\}$$

and taking  $y = \sqrt{n}x$  yields  $x = \frac{y}{\sqrt{n}}$  so that

$$\begin{aligned}\|A\|_2 &= \max \left\{ \frac{1}{\sqrt{n}} \|Ay\|_2 : \left\| \frac{y}{\sqrt{n}} \right\|_2 \leq 1 \right\} \\ &= \frac{1}{\sqrt{n}} \max \{ \|Ay\|_2 : \|y\|_2 \leq \sqrt{n} \}\end{aligned}$$

Combining this with the bound for  $\|A\|_\infty$  yield

$$\|A\|_\infty \leq \max \{ \|Ax\|_2 : \|x\|_2 \leq \sqrt{n} \} = \sqrt{n} \|A\|_2$$

as desired.

Now bound  $\|A\|_2$  in terms of  $\|A\|_\infty$ . Since  $Ax \in \mathbb{R}^m$  then part (i) shows  $\|Ax\|_2 \leq \sqrt{m} \|Ax\|_\infty$ . It follows that

$$\|A\|_2 = \max \{ \|Ax\|_2 : \|x\|_2 = 1 \} \leq \max \{ \sqrt{m} \|Ax\|_\infty : \|x\|_2 = 1 \}$$

Now  $\|x\|_\infty \leq \|x\|_2$  implies

$$\{x : \|x\|_2 = 1\} \subseteq \{x : \|x\|_\infty \leq 1\}$$

therefore, taking maximum over the larger set yields

$$\max \{ \sqrt{m} \|Ax\|_\infty : \|x\|_2 = 1 \} \leq \max \{ \sqrt{m} \|Ax\|_\infty : \|x\|_\infty \leq 1 \}.$$

As before  $\|x\|_\infty \leq 1$  implies  $\|Ax\|_\infty \leq \|A\|_\infty \|x\|_\infty \leq \|A\|_\infty$  so

$$\|A\|_\infty = \max \{ \|Ax\|_\infty : \|x\|_\infty = 1 \} = \max \{ \|Ax\|_\infty : \|x\|_\infty \leq 1 \}$$

Consequently

$$\begin{aligned} \|A\|_2 &\leq \max \{ \sqrt{m} \|Ax\|_\infty : \|x\|_\infty \leq 1 \} \\ &= \sqrt{m} \max \{ \|Ax\|_\infty : \|x\|_\infty \leq 1 \} = \sqrt{m} \|A\|_\infty. \end{aligned}$$

which was to be shown.