



Linear algebra

Exercise sheet 2 / Model solutions

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1. (a) Show that $\|\cdot\|_1$ and $\|\cdot\|_\infty$ satisfy the three conditions in the definition of a norm. (See Definition 2.1 in the Lecture notes.)

- (b) Let $\mathbf{x} = [1, 2, 3]^T \in \mathbb{R}^3$. Calculate $\|\mathbf{x}\|_1$, $\|\mathbf{x}\|_2$ and $\|\mathbf{x}\|_\infty$.

Solution.

- (a) For the 1-norm, $\|\mathbf{x}\|_1$, being the sum of the absolute values of the components of \mathbf{x} , is clearly never negative and zero if and only if $\mathbf{x} = \mathbf{0}$. Moreover,

$$\|\alpha \mathbf{x}\|_1 = \sum_{i=1}^n |\alpha x_i| = |\alpha| \sum_{i=1}^n |x_i| = |\alpha| \|\mathbf{x}\|_1.$$

Lastly,

$$\|\mathbf{x} + \mathbf{y}\|_1 = \sum_{i=1}^n |x_i + y_i| \leq \sum_{i=1}^n |x_i| + |y_i| = \|\mathbf{x}\|_1 + \|\mathbf{y}\|_1.$$

For the ∞ -norm, $\|\mathbf{x}\|_\infty$, being the maximum of the absolute values of the components of \mathbf{x} , is clearly never negative and zero if and only if $\mathbf{x} = \mathbf{0}$. Moreover,

$$\|\alpha \mathbf{x}\|_\infty = \max_{i=1, \dots, n} |\alpha x_i| = |\alpha| \max_{i=1, \dots, n} |x_i| = |\alpha| \|\mathbf{x}\|_\infty.$$

Lastly,

$$\|\mathbf{x} + \mathbf{y}\|_\infty = \max_{i=1, \dots, n} |x_i + y_i| \leq \max_{i=1, \dots, n} |x_i| + |y_i| = \|\mathbf{x}\|_\infty + \|\mathbf{y}\|_\infty.$$

- (b) We have

$$\|\mathbf{x}\|_1 = |1| + |2| + |3| = 6,$$

$$\|\mathbf{x}\|_2 = \sqrt{1^2 + 2^2 + 3^2} = \sqrt{14},$$

$$\|\mathbf{x}\|_\infty = \max\{|1|, |2|, |3|\} = 3.$$

2. Let $W_1, W_2 \subset \mathbb{R}^n$ be subspaces with bases

$$\{\mathbf{q}_1, \dots, \mathbf{q}_k\} \quad \text{and} \quad \{\mathbf{v}_1, \dots, \mathbf{v}_m\}, \quad \text{respectively.}$$

Denote $Q = [\mathbf{q}_1, \dots, \mathbf{q}_k]$ and $V = [\mathbf{v}_1, \dots, \mathbf{v}_m]$.

- (a) Show that $W_1 + W_2 = R([Q \ V])$, where we define $W_1 + W_2$ as the vector subspace $\{\mathbf{x} + \mathbf{y} \mid \mathbf{x} \in W_1, \mathbf{y} \in W_2\}$ of \mathbb{R}^n .

(b) Let $T : \mathbb{R}^{k+m} \rightarrow \mathbb{R}^k$ be such that

$$(T\mathbf{x})_i = x_i \quad \text{for } i = 1, \dots, k.$$

Show that $W_1 \cap W_2 = \{QT\mathbf{x} \mid \mathbf{x} \in N([Q \ V])\}$.

Solution.

(a) By definition, the range of a matrix is the linear span of its columns. The sum $W_1 + W_2$ is equal to the span of $W_1 \cup W_2$. This proves the equality.

(b) If $\mathbf{y} \in W_1 \cap W_2$ then $\mathbf{y} = Q\mathbf{a} = V\mathbf{b}$ for some vectors \mathbf{a} and \mathbf{b} . If we set

$$\mathbf{x} = \begin{bmatrix} \mathbf{a} \\ -\mathbf{b} \end{bmatrix},$$

then we get $[Q \ V]\mathbf{x} = \mathbf{0}$, if and only if $\mathbf{x} \in N([Q \ V])$. Clearly, $T\mathbf{x} = \mathbf{a}$ and so $QT\mathbf{x} = \mathbf{y}$. This shows that $W_1 \cap W_2 \subseteq N([Q \ V])$.

Conversely, let $\mathbf{x} \in N([Q \ V])$ and write

$$\mathbf{x} = \begin{bmatrix} T\mathbf{x} \\ \mathbf{y} \end{bmatrix},$$

for some $\mathbf{y} \in \mathbb{R}^m$. Then we have that $QT\mathbf{x} = -V\mathbf{y}$, and hence $QT\mathbf{x} \in W_1 \cap W_2$. This proves $N([Q \ V]) \subseteq W_1 \cap W_2$ and concludes the proof.

3. Let $A \in \mathbb{R}^{m \times n}$.

(a) Show that $\mathbf{y}^T \mathbf{x} = 0$ for any $\mathbf{x} \in N(A)$ and $\mathbf{y} \in R(A^T)$. (Hint: $\mathbf{y} = A^T \mathbf{z}$ for some $\mathbf{z} \in \mathbb{R}^m$.)

(b) Let $\mathbf{x} \in \mathbb{R}^n$ be such that $\mathbf{y}^T \mathbf{x} = 0$ for any $\mathbf{y} \in R(A^T)$. Show that $\mathbf{x} \in N(A)$. (Hint: Choose $\mathbf{y} = A^T A\mathbf{x}$.)

Solution.

(a) By definition of $R(A^T)$ there exists $\mathbf{z} \in \mathbb{R}^m$ such that $\mathbf{y} = A^T \mathbf{z}$. Hence,

$$\mathbf{y}^T \mathbf{x} = (A^T \mathbf{z})^T \mathbf{x}.$$

Using the calculation rules for the transpose $(AB)^T = (B^T A^T)$ and $(A^T)^T = A$, one gets

$$(A^T \mathbf{z})^T \mathbf{x} = \mathbf{z}^T (A^T)^T \mathbf{x} = \mathbf{z}^T A\mathbf{x}.$$

By noticing that $\mathbf{x} \in N(A)$, so that $A\mathbf{x} = \mathbf{0}$, the proof is complete.

(b) Choosing $\mathbf{y} = A^T A \mathbf{x}$ gives

$$\mathbf{y}^T \mathbf{x} = (A^T A \mathbf{x})^T \mathbf{x}.$$

Using calculation rules of the transpose,

$$(A^T A \mathbf{x})^T \mathbf{x} = \mathbf{x}^T A^T A \mathbf{x} = \|A \mathbf{x}\|_2^2.$$

Hence, by assumption, $\|A \mathbf{x}\|_2^2 = 0$. By the properties of a norm, this implies that $A \mathbf{x} = \mathbf{0}$, thus $\mathbf{x} \in N(A)$.

4. Use Matlab to visualize the set

$$S := \{\mathbf{x} \in \mathbb{R}^2 \mid \|\mathbf{x}\|_* = 1\},$$

for $*$ = 1, 2 or ∞ . You may modify the function `plot_norm.m` found at the MyCourses page. Return both the script that you wrote and a printout of the resulting figure.

When $\|\mathbf{x}\|_2 = 1$, are $\|\mathbf{x}\|_1$ and $\|\mathbf{x}\|_\infty$ larger or smaller than one? Which four vectors $\mathbf{x} \in \mathbb{R}^2$ satisfy

$$\|\mathbf{x}\|_1 = \|\mathbf{x}\|_2 = \|\mathbf{x}\|_\infty = 1?$$

Justify your answer based on the figure that you draw.

Solution: The challenge in the visualization of the set S is to find a parametric presentation for it. Any $\mathbf{x} \in \mathbb{R}^2$ can be written as

$$\mathbf{x} = r \mathbf{v}_\theta, \quad \text{where} \quad \mathbf{v}_\theta = \begin{bmatrix} \sin \theta \\ \cos \theta \end{bmatrix}, \quad r \in \mathbb{R}, r \geq 0 \text{ and } \theta \in [0, 2\pi].$$

Now, one has

$$S = \{r \mathbf{v}_\theta \mid r \in \mathbb{R}, r \geq 0, \theta \in [0, 2\pi] \text{ and } \|r \mathbf{v}_\theta\|_* = 1\}.$$

By the properties of a norm, $r \|\mathbf{v}_\theta\|_* = 1$. This implies that $r = 1/\|\mathbf{v}_\theta\|_*$ and thus

$$S = \left\{ \frac{\mathbf{v}_\theta}{\|\mathbf{v}_\theta\|_*} \mid \theta \in [0, 2\pi] \right\}.$$

The set S is now easy to draw. See the file `norms.eps`. Here follows the code.

```
N = 1000;
t = linspace(0, 2*pi, N);

x = cos(t);
y = sin(t);

figure;

for p=[1, 2, Inf]
```

```
for i=1:N  
  
    v = [x(i) ; y(i)];  
  
    rho = 1./norm(v,p);  
  
    xplot(i) = rho*x(i);  
    yplot(i) = rho*y(i);  
  
end
```

```
hold on;  
plot(xplot,yplot);  
  
end
```

By the figure one gets $\|\mathbf{x}\|_\infty \leq \|\mathbf{x}\|_2 \leq \|\mathbf{x}\|_1$, where the equality holds for $\mathbf{x} = \pm \mathbf{e}_1$ and $\mathbf{x} = \pm \mathbf{e}_2$.