Solution (3.2.10). (a) It is not an inner product because it is not symmetric nor is it positive. But it is bilinear.

(b) We can use the relation  $\sin^2(\theta) = 1 - \cos^2(\theta)$  to solve this problem. First note that

$$(v \times w)^2 = ||v||^2 ||w||^2 - \langle v, w \rangle^2$$

by expanding both sides in coordinates. But then using the definition of cos,

$$(v \times w)^{2} = ||v||^{2} ||w||^{2} - ||v||^{2} ||w||^{2} \cos^{2}(\theta) = ||v||^{2} ||w||^{2} \sin^{2}(\theta).$$

Taking the square root of both sides we see that  $||v||||w||\sin(\theta) = \pm(v \times w)$ . But is  $v \times w$  positive or negative? We have to narrow it down to either + or -. Turns out to be the +. To prove this, it suffices to show that  $(v \times w) \ge 0$  when  $\theta \in [0, \pi]$ , which is the domain of cos.

Fix ||v|| and ||w|| and vary  $\theta$ . Since  $\sin(\theta) = 0$  only when v and w are parellel, i.e. when  $\theta = 0, \pi$ , it suffices to show that  $v \times w > 0$  for any one  $\theta \in (0, \pi)$  by the intermediate value theorem. We let  $\theta = \pi/2$ . Then  $v = (v_1, v_2)$  and

$$w = \frac{||w||}{||v||}(-v_2, v_1)$$

which is 90 degree from v. Then

$$v \times w = \frac{||w||}{||v||}(v_1^2 + v_2^2) > 0$$

Thus we can take the positive root and

$$v \times w = ||v||||w||\sin(\theta).$$

(c) This follows from the formula. We know that  $\sin(\theta) = 0$  if and only if  $\theta = 0, \pi$  which is the case if and only if v and w are parellel.

(d) Consider the following diagram.



From here we can see that the area of the parallelogram is

$$A = (\|w\| + \|v\|\cos\theta) \|v\|\sin\theta - 2\left(\frac{1}{2}\|v\|\cos\theta\|v\|\sin\theta\right) = \|w\|\|v\|\sin\theta = |v \times w|$$

as desired. You can also see that  $A = \left| \det \begin{pmatrix} v_1 & w_1 \\ v_2 & w_2 \end{pmatrix} \right|$  which is how you generalize this to  $\mathbb{R}^n$ . You can only take a cross product in  $\mathbb{R}^3$ , if you remember from multi.

Solution (3.2.16). This is a row reduction problem. The vectors orthogonal (1, 2, 3) and (-2, 0, 1) are exactly the kernel of the matrix

$$\begin{pmatrix} 1 & 2 & 3 \\ -2 & 0 & 1 \end{pmatrix}.$$

Row reduction tells you that the kernel is the span of the vector (2, -7, 4).

Solution (3.2.25). We can do one better. Let  $W \subseteq V$  be a subspace. Let  $W^{\perp}$  be the set of all vectors orthogonal to all vectors of W. We can show that  $W^{\perp}$  is a subspace. We have to show it satisfies the subspace axioms.

First  $W^{\perp} \neq \emptyset$  since  $0 \in W^{\perp}$ . For all  $w \in W$ , we know that  $\langle 0, w \rangle = 0$ . The zero vector is orthogonal to everything.

Second, let  $u, v \in W^{\perp}$ . We show that  $u + v \in W^{\perp}$  also. By bilinearity,

$$\langle u+v,w\rangle = \langle u,w\rangle + \langle v,w\rangle = 0 + 0 = 0.$$

Therefore u + v is orthogonal to all  $w \in W$  as well.

Finally,  $\langle cv, w \rangle = c \langle v, w \rangle = 0$  as well. So  $W^{\perp}$  is a subspace. In particular, when  $W = \operatorname{span}(v)$ , then  $W^{\perp}$  is all the vectors perpendicular to v, like the problem asks.

Solution (3.2.40). It's true!

$$||w|| = ||-v+(v+w)|| \leq ||-v|| + ||v+w|| = ||v|| + ||v+w||$$

Solution (3.3.10c). We show that  $||v|| = 2|v_1| + |v_2|$  is a norm on  $\mathbb{R}^2$ . First, it is positive by definition, since absolute values are always positive. Second, it is homogeneous since

$$||cv|| = 2|cv_1| + |cv_2| = |c|(2|v_1| + |v_2|) = |c| \cdot ||v||.$$

Finally the triangle inequality follows from the triangle inequality for the absolute value.

$$\begin{aligned} ||v + w|| &= 2|v_1 + w_1| + |v_2 + w_2| \\ &\leq 2(|v_1| + |w_1|) + |v_2| + |w_2| \\ &= 2|v_1| + |v_2| + 2|w_1| + |w_2| \\ &= ||v|| + ||w|| \end{aligned}$$

Solution (3.3.35i). We have to show that

$$||v||_2 \le ||v||_1 \le \sqrt{n} ||v||_2.$$

First, we know that by distributive property

$$||v||_{1}^{2} = \left(\sum_{i=1}^{n} |v_{i}|\right)^{2} = \sum_{i=1}^{n} |v_{i}|^{2} + \sum_{i \neq j}^{n} |v_{i}v_{j}| \ge \sum_{i=1}^{n} |v_{i}|^{2} = ||v||_{2}^{2}.$$

Taking square roots, this proves the first inequality. For the next inequality, we can do a clever trick using Cauchy-Schwartz.

$$||v||_1 = \sum_{i=1}^n |v_i| = (|v_1|, \dots, |v_n|) \cdot (1, 1, \dots, 1)$$
  
$$\leq ||(|v_1|, \dots, |v_n|)|| \cdot ||(1, \dots, 1)|| = ||v||_2 \sqrt{n}$$

Homework 5

Another way to show this inequality is to find

 $c = \min\{||u||_1 \mid ||u||_2 = 1\}$ 

and

 $d = \max\{||u||_1 \mid ||u||_2 = 1\}.$ 

The minimum value of the  $L^1$  norm on the  $L^2$  unit sphere is achieved at u = (1, 0, ..., 0), so that c = 1. The maximum is achieved at  $u = (1/\sqrt{n}, ..., 1/\sqrt{n})$  so that  $d = \sqrt{n}$ . Therefore  $||v||_2 \le ||v||_1 \le \sqrt{n}||v||_2$ .

Solution (3.3.39). Since  $||f||_{\infty}$  is the maximum value of |f(x)| on the interval [a, b], then we know that  $f(x)^2 \leq ||f||_{\infty}^2$  on the interval as well. We can integrate both sides of this inequality and get

$$||f||_{2}^{2} = \int_{a}^{b} f(x)^{2} dx \le \int_{a}^{b} ||f||_{\infty}^{2} dx = (b-a)||f||_{\infty}^{2}.$$

Taking the square root of both sides shows that

$$||f||_2 \le \sqrt{b-a}||f||_{\infty}.$$

Solution. It's false. I picked some random A and S matrices and it gave me a counterexample. Let

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} \quad S = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 1 & 1 & 0 \end{pmatrix} \quad B = S^{-1}AS = \frac{1}{2} \begin{pmatrix} 10 & 11 & 9 \\ 22 & 23 & 21 \\ -2 & -1 & -3 \end{pmatrix}.$$

Then  $||A||_{\infty} = 7 + 8 + 9 = 24$  but  $||B||_{\infty} = (22 + 23 + 21)/2 = 33$ .