

Solutions for Proposed Homework 7:

Part A:

4.3.23) $[T(1)]_{\mathfrak{U}} = [1]_{\mathfrak{U}} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$, $[T(t)]_{\mathfrak{U}} = [3]_{\mathfrak{U}} = \begin{bmatrix} 3 \\ 0 \\ 0 \end{bmatrix}$ and $[T(t^2)]_{\mathfrak{U}} = [9]_{\mathfrak{U}} = \begin{bmatrix} 9 \\ 0 \\ 0 \end{bmatrix}$, so we see

that

$$[T]_{\mathfrak{U}} = \begin{bmatrix} 1 & 3 & 9 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}.$$

T is not an isomorphism and $\{(t-3), (t-3)^2\}$ is a basis for $\ker(T)$, while $\{1\}$ is a basis for $\text{im}(T)$, so T has rank 1.

4.3.24) $[T(1)]_{\mathfrak{U}} = [1]_{\mathfrak{B}} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$, $[T(t-3)]_{\mathfrak{U}} = [0]_{\mathfrak{B}} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$ and $[T((t-3)^2)]_{\mathfrak{U}} = [0]_{\mathfrak{B}} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$, so

we see that

$$[T]_{\mathfrak{U}} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}.$$

T is not an isomorphism and $\{(t-3), (t-3)^2\}$ is a basis for $\ker(T)$, while $\{1\}$ is a basis for $\text{im}(T)$, so T has rank 1.

4.3.46) a) Since $[1]_{\mathfrak{U}} = [1 \ 0 \ 0]$, $[t-3]_{\mathfrak{U}} = [-3 \ 1 \ 0]$, and $[(t-3)^2]_{\mathfrak{U}} = [9 \ -6 \ 1]$, we see that

$$S_{\mathfrak{B} \rightarrow \mathfrak{U}} = \begin{bmatrix} 1 & -3 & 9 \\ 0 & 1 & -6 \\ 0 & 0 & 1 \end{bmatrix}.$$

b) Notice that

$$\begin{bmatrix} 1 & -3 & 9 \\ 0 & 1 & -6 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 3 & 9 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & -3 & 9 \\ 0 & 1 & -6 \\ 0 & 0 & 1 \end{bmatrix}.$$

c) Similarly,

$$S_{\mathfrak{U} \rightarrow \mathfrak{B}} = \begin{bmatrix} 1 & 3 & 9 \\ 0 & 1 & 6 \\ 0 & 0 & 1 \end{bmatrix}.$$

4.3.54) Since $\left[T \left(\begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix} \right) \right]_{\mathfrak{B}} = \left[\begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix} \right]_{\mathfrak{B}} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ and $\left[T \left(\begin{bmatrix} 5 \\ -4 \\ 1 \end{bmatrix} \right) \right]_{\mathfrak{B}} = \left[\begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \right]_{\mathfrak{B}} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$ (since $\begin{bmatrix} 5 \\ -4 \\ 1 \end{bmatrix}$ is orthogonal to $\begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix}$), we see that

$$[T]_{\mathfrak{B}} = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}.$$

5.1.12) The Triangle Inequality: If $\vec{v}, \vec{w} \in \mathbb{R}^n$, then

$$\|\vec{v} + \vec{w}\| \leq \|\vec{v}\| + \|\vec{w}\|.$$

Proof: Recall that, by definition,

$$\|\vec{v} + \vec{w}\|^2 = (\vec{v} + \vec{w}) \cdot (\vec{v} + \vec{w}) = \vec{v} \cdot \vec{v} + 2\vec{v} \cdot \vec{w} + \vec{w} \cdot \vec{w} = \|\vec{v}\|^2 + 2\vec{v} \cdot \vec{w} + \|\vec{w}\|^2.$$

The Cauchy-Schwartz inequality implies that $\vec{v} \cdot \vec{w} \leq \|\vec{v}\|\|\vec{w}\|$, so we conclude that

$$\|\vec{v} + \vec{w}\|^2 \leq \|\vec{v}\|^2 + 2\|\vec{v}\|\|\vec{w}\| + \|\vec{w}\|^2 = (\|\vec{v}\| + \|\vec{w}\|)^2.$$

Since, $\|\vec{v} + \vec{w}\| \geq 0$ and $\|\vec{v}\| + \|\vec{w}\| \geq 0$, this implies that

$$\|\vec{v} + \vec{w}\| \leq \|\vec{v}\| + \|\vec{w}\|.$$

5.1.28) Let $\mathfrak{B} = \left(\begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ -1 \\ -1 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \\ -1 \\ 1 \end{bmatrix} \right)$ and let V be the subspace of \mathbb{R} spanned by \mathfrak{B} . Notice that each element of \mathfrak{B} has length 2 and that distinct elements are orthogonal. Therefore if we scale each vector in \mathfrak{B} by $\frac{1}{2}$ we obtain an orthonormal basis $\mathfrak{C} = \left(\begin{bmatrix} \frac{1}{2} \\ \frac{1}{2} \\ \frac{1}{2} \\ \frac{1}{2} \end{bmatrix}, \begin{bmatrix} \frac{1}{2} \\ \frac{1}{2} \\ -\frac{1}{2} \\ -\frac{1}{2} \end{bmatrix}, \begin{bmatrix} \frac{1}{2} \\ -\frac{1}{2} \\ -\frac{1}{2} \\ \frac{1}{2} \end{bmatrix} \right)$ for V .

Therefore the projection of $\begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$ to V is given by

$$\left(\begin{bmatrix} \frac{1}{2} \\ \frac{1}{2} \\ \frac{1}{2} \\ \frac{1}{2} \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} \right) \begin{bmatrix} \frac{1}{2} \\ \frac{1}{2} \\ \frac{1}{2} \\ \frac{1}{2} \end{bmatrix} + \left(\begin{bmatrix} \frac{1}{2} \\ \frac{1}{2} \\ -\frac{1}{2} \\ -\frac{1}{2} \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} \right) \begin{bmatrix} \frac{1}{2} \\ \frac{1}{2} \\ -\frac{1}{2} \\ -\frac{1}{2} \end{bmatrix} + \left(\begin{bmatrix} \frac{1}{2} \\ -\frac{1}{2} \\ -\frac{1}{2} \\ \frac{1}{2} \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} \right) \begin{bmatrix} \frac{1}{2} \\ -\frac{1}{2} \\ -\frac{1}{2} \\ \frac{1}{2} \end{bmatrix} = \begin{bmatrix} \frac{3}{4} \\ \frac{1}{4} \\ -\frac{1}{4} \\ \frac{1}{4} \end{bmatrix}.$$

5.1.32) The 2×2 matrix $G = \begin{bmatrix} \vec{v}_1 \cdot \vec{v}_1 & \vec{v}_1 \cdot \vec{v}_2 \\ \vec{v}_2 \cdot \vec{v}_1 & \vec{v}_2 \cdot \vec{v}_2 \end{bmatrix}$ is not invertible if and only if it has rank less than 2, which occurs if and only if its determinant

$$(\vec{v}_1 \cdot \vec{v}_1)(\vec{v}_2 \cdot \vec{v}_2) - (\vec{v}_1 \cdot \vec{v}_2)^2 = \|\vec{v}_1\|^2\|\vec{v}_2\|^2 - (\vec{v}_1 \cdot \vec{v}_2)^2 = 0.$$

However, by the Cauchy-Schwartz inequality, this occurs if and only if \vec{v}_1 and \vec{v}_2 are parallel. Therefore, G is invertible if and only if \vec{v}_1 and \vec{v}_2 are not parallel.

Part B:

Problem 1:

Let V and W be vector spaces and let $\mathcal{B} = (b_1, \dots, b_n)$ be an ordered basis for V . Let $f : \mathcal{B} \rightarrow W$ is a function. If $\vec{v} \in V$ be arbitrary, then, since \mathcal{B} is an ordered basis of V , there exist unique scalars $c_1, \dots, c_n \in \mathbb{R}$ such that $\vec{v} = \sum_{i=1}^n c_i \vec{b}_i$. We then define $T : V \rightarrow W$ by setting

$$T(\vec{v}) = \sum_{i=1}^n c_i f(\vec{b}_i).$$

In particular, T is an extension of the function f since $T(b_i) = f(b_i)$ for all i .

Suppose that $\vec{x}, \vec{y} \in V$ and that $\vec{x} = \sum_{i=1}^n c_i \vec{b}_i$ and $\vec{y} = \sum_{i=1}^n d_i \vec{b}_i$. Then $\vec{x} + \vec{y} = \sum_{i=1}^n (c_i + d_i) \vec{b}_i$, so

$$\begin{aligned} T(\vec{v} + \vec{w}) &= T\left(\sum_{i=1}^n c_i \vec{b}_i + \sum_{i=1}^n d_i \vec{b}_i\right) = T\left(\sum_{i=1}^n (c_i + d_i) \vec{b}_i\right) \\ &= \sum_{i=1}^n (c_i + d_i) f(\vec{b}_i) = \left(\sum_{i=1}^n c_i f(\vec{b}_i)\right) + \left(\sum_{i=1}^n d_i f(\vec{b}_i)\right) = T(\vec{x}) + T(\vec{y}) \end{aligned}$$

so T respects addition. If $k \in \mathbb{R}$, $\vec{x} \in V$ and $\vec{x} = \sum_{i=1}^n c_i \vec{b}_i$, then $k\vec{x} = \sum_{i=1}^n k c_i \vec{b}_i$, so

$$T(k\vec{x}) = T\left(k \sum_{i=1}^n c_i \vec{b}_i\right) = T\left(\sum_{i=1}^n k c_i \vec{b}_i\right) = \sum_{i=1}^n k c_i f(\vec{b}_i) = k \sum_{i=1}^n c_i f(\vec{b}_i) = k T(\vec{x}),$$

which implies that T respects scalar multiplication. Therefore, T is a linear transformation.

Finally, to see that T is unique, suppose the linear map $S : V \rightarrow W$ also has the property that S agrees with f on \mathcal{B} . Then for all $\vec{v} = \sum_{i=1}^n c_i \vec{b}_i \in V$ we have

$$S(\vec{v}) = S\left(\sum_{i=1}^n c_i \vec{b}_i\right) = \sum_{i=1}^n c_i S(\vec{b}_i) = \sum_{i=1}^n c_i f(\vec{b}_i) = \sum_{i=1}^n c_i T(\vec{b}_i) = T\left(\sum_{i=1}^n c_i \vec{b}_i\right) = T(\vec{v}).$$

Problem 2:

Let $T : V \rightarrow W$ be a linear transformation and let \mathcal{B} and \mathcal{C} be ordered bases for V .

(a) TRUE If $T : V \rightarrow W$ is a linear transformation and $\ker[T]_{\mathcal{B}} = \{[\vec{v}]_{\mathcal{B}} : \vec{v} \in \ker(T)\}$.

Proof: Suppose that $\vec{x} \in \ker[T]_{\mathcal{B}}$. Let $\vec{v} \in V$ be the unique vector in V such that $[\vec{v}]_{\mathcal{B}} = \vec{x}$. Then by definition of $[T]_{\mathcal{B}}$ and of $\ker[T]_{\mathcal{B}}$, respectively,

$$[T(\vec{v})]_{\mathcal{B}} = [T]_{\mathcal{B}}[\vec{v}]_{\mathcal{B}} = [T]_{\mathcal{B}}\vec{x} = \vec{0}.$$

Since $[T(\vec{v})]_{\mathcal{B}} = \vec{0}$, $T(\vec{v}) = \vec{0}_W$. Thus, $\vec{x} = [\vec{v}]_{\mathcal{B}}$ for some $\vec{v} \in \ker T$, which proves that $\ker[T]_{\mathcal{B}} \subseteq \{[\vec{v}]_{\mathcal{B}} : \vec{v} \in \ker T\}$.

Now suppose that $\vec{x} \in \{[\vec{v}]_{\mathcal{B}} : \vec{v} \in \ker T\}$. Then there exists a vector $\vec{v} \in V$ such that $\vec{x} = [\vec{v}]_{\mathcal{B}}$ and

$$[T]_{\mathcal{B}}\vec{x} = [T]_{\mathcal{B}}[\vec{v}]_{\mathcal{B}} = [T(\vec{v})]_{\mathcal{B}} = [\vec{0}_W]_{\mathcal{B}} = \vec{0},$$

hence $\vec{x} \in \ker[T]_{\mathcal{B}}$. Therefore, $\{[\vec{v}]_{\mathcal{B}} : \vec{v} \in \ker T\} \subseteq \ker[T]_{\mathcal{B}}$ and the proof is complete. \square

(b) TRUE $\text{im}[T]_{\mathcal{B}} = \{[\vec{v}]_{\mathcal{B}} : \vec{v} \in \text{im}(T)\}$.

Proof: Suppose that $\vec{y} \in \text{im}[T]_{\mathcal{B}}$. Then there exists a vector $\vec{x} \in \mathbb{R}^n$ such that $[T]_{\mathcal{B}}\vec{x} = \vec{y}$, by definition of the image of $[T]_{\mathcal{B}}$. Let \vec{v} be the unique vector in V for which $[\vec{v}]_{\mathcal{B}} = \vec{x}$, and let $\vec{w} = T(\vec{v})$. Then $\vec{w} \in \text{im } T$, and

$$\vec{y} = [T]_{\mathcal{B}}\vec{x} = [T]_{\mathcal{B}}[\vec{v}]_{\mathcal{B}} = [T(\vec{v})]_{\mathcal{B}} = [\vec{w}]_{\mathcal{B}}$$

implying that $\vec{y} \in \{[\vec{w}]_{\mathcal{B}} : \vec{w} \in \text{im}(T)\}$. Thus, $\text{im}[T]_{\mathcal{B}} \subseteq \{[\vec{w}]_{\mathcal{B}} : \vec{w} \in \text{im}(T)\}$.

On the other hand, suppose that $\vec{y} \in \{[\vec{w}]_{\mathcal{B}} : \vec{w} \in \text{im}(T)\}$. Then $\vec{y} = [\vec{w}]_{\mathcal{B}}$ for some $\vec{w} \in \text{im } T$. By definition of the image of T , there exists a vector $\vec{v} \in V$ such that $T(\vec{v}) = \vec{w}$. Let $\vec{x} = [\vec{v}]_{\mathcal{B}} \in \mathbb{R}^n$. Then

$$\vec{y} = [\vec{w}]_{\mathcal{B}} = [T(\vec{v})]_{\mathcal{B}} = [T]_{\mathcal{B}}[\vec{v}]_{\mathcal{B}} = [T]_{\mathcal{B}}\vec{x}$$

and we conclude $\vec{y} \in \text{im}[T]_{\mathcal{B}}$. Therefore, $\{[\vec{w}]_{\mathcal{B}} : \vec{w} \in \text{im}(T)\} \subseteq \text{im } T_{\mathcal{B}}$ and our proof is complete. \square

(c) FALSE $\ker[T]_{\mathcal{B}}$ need not be the same as $\ker[T]_{\mathcal{C}}$.

Counterexample: Consider the map $T : \mathcal{P}_1 \rightarrow \mathcal{P}_1$ given by $T(ax + b) = b$. If $\mathcal{B} = (x, 1)$, then $\ker[T]_{\mathcal{B}} = \text{span} \left(\begin{bmatrix} 1 \\ 0 \end{bmatrix} \right)$, but if $\mathcal{C} = (1, x)$, then $\ker[T]_{\mathcal{C}} = \text{span} \left(\begin{bmatrix} 0 \\ 1 \end{bmatrix} \right)$.

(d) FALSE If $\vec{v}, \vec{w} \in V$, $[\vec{v}]_{\mathcal{B}} \cdot [\vec{w}]_{\mathcal{B}}$ need not be the same as $[\vec{v}]_{\mathcal{C}} \cdot [\vec{w}]_{\mathcal{C}}$.

Counterexample: Consider the ordered bases $\mathcal{B} = (1)$ and $\mathcal{C} = (2)$ for \mathbb{R} . Then if $\vec{v} = \vec{w} = 2$, then

$$[\vec{v}]_{\mathcal{B}} \cdot [\vec{w}]_{\mathcal{B}} = 2 \cdot 2 = 4 \neq 1 = 1 \cdot 1 = [\vec{v}]_{\mathcal{C}} \cdot [\vec{w}]_{\mathcal{C}}.$$

Problem 3:

a) Suppose that $A, B \in \mathbb{R}^{2 \times 2}$ and $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ and $B = \begin{bmatrix} e & f \\ g & h \end{bmatrix}$. Then $A + B = \begin{bmatrix} a+e & b+f \\ c+g & d+h \end{bmatrix}$ and

$$T(A+B) = \begin{bmatrix} a+e & 0 & b+f \\ 0 & 0 & 0 \\ c+g & 0 & d+h \end{bmatrix} = \begin{bmatrix} a & 0 & b \\ 0 & 0 & 0 \\ c & 0 & d \end{bmatrix} + \begin{bmatrix} e & 0 & f \\ 0 & 0 & 0 \\ g & 0 & h \end{bmatrix} = T(A) + T(B)$$

so T respects addition. Similarly, if $k \in \mathbb{R}$, $A \in \mathbb{R}^{2 \times 2}$ and $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$, then $kA = \begin{bmatrix} ka & kb \\ kc & kd \end{bmatrix}$ and

$$T(kA) = \begin{bmatrix} ka & 0 & kb \\ 0 & 0 & 0 \\ kc & 0 & kd \end{bmatrix} = k \begin{bmatrix} a & 0 & b \\ 0 & 0 & 0 \\ c & 0 & d \end{bmatrix} = kT(A)$$

so T respects scalar multiplication. Since T respects addition and scalar multiplication, it is a linear transformation.

b) $\left\{ \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 1 & 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \right\}$ is a basis for $\text{im}(T)$. So,

$$\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix} \right\} = \{\vec{e}_1, \vec{e}_3, \vec{e}_7, \vec{e}_9\}$$

is a basis for $[\text{im}(T)]_{\mathfrak{B}}$.

Since T is injective, $\ker T = \{\vec{0}\}$, and, by convention, its basis is empty. Similarly, the basis for $[\ker T]_{\mathfrak{B}}$ is empty.

c) Since $\left[T \left(\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \right) \right]_{\mathfrak{B}_3} = \left[\begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \right]_{\mathfrak{B}_3} = \vec{e}_1$, $\left[T \left(\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \right) \right]_{\mathfrak{B}_3} = \left[\begin{bmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \right]_{\mathfrak{B}_3} = \vec{e}_3$,
 $\left[T \left(\begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} \right) \right]_{\mathfrak{B}_3} = \left[\begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 1 & 0 & 0 \end{bmatrix} \right]_{\mathfrak{B}_3} = \vec{e}_7$, and $\left[T \left(\begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right) \right]_{\mathfrak{B}_3} = \left[\begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \right]_{\mathfrak{B}_3} = \vec{e}_9$, we see

that if

$$C = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix},$$

then $[T(A)]_{\mathcal{B}_3} = C[A]_{\mathcal{B}_2}$ for all $A \in \mathbb{R}^{2 \times 2}$.

Problem 4:

- (a) First notice that $\vec{0} \in V$, since V is a subspace of \mathbb{R}^n and that $\vec{0} \in V^\perp$, so $\vec{0} \in V \cap V^\perp$. On the other hand, if $\vec{v} \in V \cap V^\perp$, then $\vec{v} \cdot \vec{v} = 0$, so $\vec{v} = \vec{0}$. Therefore, $V \cap V^\perp = \{\vec{0}\}$.
- (b) First we show that $V \subseteq (V^\perp)^\perp$: Let $\vec{v} \in V$. Then, \vec{v} is perpendicular to every vector in V^\perp , so $\vec{v} \in (V^\perp)^\perp$. Now, we notice that by Theorem 5.1.8, part c of the book,

$$\dim(V^\perp)^\perp = n - \dim(V^\perp) = n - (n - \dim V) = \dim V.$$

Hence, since $V \subseteq (V^\perp)^\perp$ and they are subspaces of the same dimension, they must be equal.

- (c) Let's prove the forward direction. Assume $V \subseteq W$, and let $\vec{x} \in W^\perp$. Then, $\vec{x} \cdot \vec{w} = 0$ for all $\vec{w} \in W$, and since $V \subseteq W$, that implies that $\vec{x} \cdot \vec{v} = 0$ for all $\vec{v} \in V$. Hence, $\vec{x} \in V^\perp$.

Let's prove the backwards direction. Assume that $W^\perp \subseteq V^\perp$. By the forward direction, which we just proved, this implies that $(V^\perp)^\perp \subseteq (W^\perp)^\perp$, and the result follows from part (b).

- (d) Let $\vec{x} \in V^\perp \cap W^\perp$, we want to show that $\vec{x} \in (V + W)^\perp$, that is, that for all $\vec{y} \in V + W$, we have that $\vec{x} \cdot \vec{y} = 0$.

Let $\vec{y} \in V + W$. There exists $\vec{v} \in V$ and $\vec{w} \in W$ such that $\vec{y} = \vec{v} + \vec{w}$. Since $\vec{x} \in V^\perp$, we have that $\vec{x} \cdot \vec{v} = 0$, and since $\vec{x} \in W^\perp$, we also have that $\vec{x} \cdot \vec{w} = 0$. Hence,

$$\vec{x} \cdot \vec{y} = \vec{x} \cdot \vec{v} + \vec{x} \cdot \vec{w} = 0,$$

which shows $(V + W)^\perp \subseteq V^\perp \cap W^\perp$.

Conversely, we note that $V \subseteq V + W$ and $W \subseteq V + W$. By part (c), this implies that $(V + W)^\perp \subseteq V^\perp$ and $(V + W)^\perp \subseteq W^\perp$. Hence, $(V + W)^\perp \subseteq V^\perp \cap W^\perp$.

Therefore, $(V + W)^\perp = V^\perp \cap W^\perp$.

Problem 5:

- (a) TRUE For every subspace V of \mathbb{R}^n , the orthogonal projection $\text{proj}_V : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is a projection.

Proof: By Theorem 5.1.4 in the book, we know that proj_V is a linear transformation. Recall that $\text{proj}_V(\vec{x}) = \vec{x}^\parallel$, where \vec{x}^\parallel is the only vector in V such that $\vec{x} - \vec{x}^\parallel$ is in V^\perp . Since $\vec{x}^\parallel - \vec{x}^\parallel = \vec{0} \in V^\perp$, we have that

$$\text{proj}_V(\text{proj}_V(\vec{x})) = \text{proj}_V(\vec{x}^\parallel) = \vec{x}^\parallel = \text{proj}_V(\vec{x})$$

for all $\vec{x} \in \mathbb{R}^n$. □

- (b) TRUE For every projection $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$ and basis \mathcal{B} of \mathbb{R}^n , the \mathcal{B} -matrix $[T]_{\mathcal{B}}$ of T is a *projection matrix*, meaning that $([T]_{\mathcal{B}})^2 = [T]_{\mathcal{B}}$.

Proof: Let A be the $n \times n$ matrix such that $T(\vec{x}) = A\vec{x}$ for all $\vec{x} \in \mathbb{R}^n$. Since T is a projection, we have that $A^2 = A$. Let S be the change of basis matrix from \mathfrak{B} to the standard basis \mathcal{E} . Then $[T]_{\mathfrak{B}} = S^{-1}AS$, so

$$([T]_{\mathfrak{B}})^2 = (S^{-1}AS)^2 = S^{-1}ASS^{-1}AS = S^{-1}AS = [T]_{\mathfrak{B}}.$$

□

- (c) FALSE If $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is a projection and $V = \text{im } T$, T need not agree with proj_V .

Counterexample: Let $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ be the linear transformation given by the matrix

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

We have that $A^2 = A$, so T is a projection. We have that $\text{im}(T)$ is the x -axis. However, the projection onto the x -axis has standard matrix

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

so T is not the orthogonal projection onto $\text{im}(T)$.

Alternatively, we can see that $\ker(T) = \text{span}\left(\begin{bmatrix} 1 \\ -1 \end{bmatrix}\right)$, but $\begin{bmatrix} 1 \\ -1 \end{bmatrix}$ is not perpendicular to the x -axis, so T is not the orthogonal projection onto $\text{im}(T)$.

- (d) TRUE. For every vector space W and projection $T : W \rightarrow W$, if T is surjective then T is the identity map on W .

Proof: Let $\vec{x} \in W$. Since T is surjective, there exists $\vec{y} \in W$ such that $T(\vec{y}) = \vec{x}$. Applying T to both sides of the equation, and using that $T(T(\vec{y})) = T(\vec{y}) = \vec{x}$, we get that $\vec{x} = T(\vec{x})$. Hence, T is the identity transformation.