

MTH 102A - Linear Algebra - 2015-16-II Semester

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P. Linear Transformations

- Let \mathbb{V}, \mathbb{W} be VS's over \mathbb{F} . A function $T : \mathbb{V} \rightarrow \mathbb{W}$ is a **linear map/ transformation** if $T(\mathbf{u} + \alpha \mathbf{v}) = T(\mathbf{u}) + \alpha T(\mathbf{v})$ for each $\mathbf{u}, \mathbf{v} \in \mathbb{V}$, $\alpha \in \mathbb{F}$. The class of all linear transformations (LT) from \mathbb{V} to \mathbb{W} is denoted by $\mathcal{L}(\mathbb{V}, \mathbb{W})$.
- Define $T(\mathbf{v}) = \mathbf{0}$ for all $\mathbf{v} \in \mathbb{V}$. Then T is a LT from \mathbb{V} to any \mathbb{W} .
- Define $T(\mathbf{v}) = \mathbf{v}$ for all $\mathbf{v} \in \mathbb{V}$. Then T is a LT from \mathbb{V} to \mathbb{V} , called the **identity map**.
- Fix $A_{m \times n}$. For $\mathbf{v} \in \mathbb{C}^n$ define $T(\mathbf{v}) = A\mathbf{v}$. Then $T \in \mathcal{L}(\mathbb{C}^n, \mathbb{C}^m)$.
- Let $\mathbb{V} = \mathbb{R}[x]$ and $T(p(x)) = p'(x)$. Then $T \in \mathcal{L}(\mathbb{V}, \mathbb{V})$.
- For $\mathbf{f} \in \mathbb{V} = \mathcal{C}(\mathbb{R}, \mathbb{R})$, define $\mathbf{g} = T(\mathbf{f})$ as $\mathbf{g}(a) = \int_0^a \mathbf{f}(t)dt$. So $[T(\sin t)](a) = \int_0^a \sin t dt = 1 - \cos a$. So $T(\sin t) = 1 - \cos t$. In general $T(\mathbf{f} + \alpha \mathbf{g}) = F + \alpha G - F(0) - \alpha G(0)$, where $F = \int \mathbf{f}$ and $G = \int \mathbf{g}$. So $T \in \mathcal{L}(\mathbb{V}, \mathbb{V})$.
- Fix $\mathbf{z} \in \mathbb{R}^n$. For $\mathbf{x} \in \mathbb{R}^n$ define $T(\mathbf{x}) = \langle \mathbf{x}, \mathbf{z} \rangle$. Then $T \in \mathcal{L}(\mathbb{R}^n, \mathbb{R})$.
- Sometimes we write **$T\mathbf{v}$** to mean $T(\mathbf{v})$.
- For $\mathbf{x} \in \mathbb{R}^2$ define $T(\mathbf{x}) = \begin{bmatrix} x_1 + x_2 \\ 3x_1 \end{bmatrix}$. Then $T \in \mathcal{L}(\mathbb{R}^2, \mathbb{R}^2)$.
- For $\mathbf{x} \in \mathbb{R}^2$ define $T(\mathbf{x}) = \begin{bmatrix} x_1 x_2 \\ 3x_1 \end{bmatrix}$. Then T is not a LT.
- Let $T \in \mathcal{L}(\mathbb{V}, \mathbb{W})$. Then $T(\mathbf{0}) = T(\mathbf{0} + \mathbf{0}) = T(\mathbf{0}) + T(\mathbf{0})$. So $T(\mathbf{0}) = \mathbf{0}$.
- Let $T \in \mathcal{L}(\mathbb{R}^2, \mathbb{R}^2)$ s.t. $T(\mathbf{e}_1) = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$, $T(\mathbf{e}_2) = \begin{bmatrix} 1 \\ 3 \end{bmatrix}$. So
$$T \begin{bmatrix} x \\ y \end{bmatrix} = xT \begin{bmatrix} 1 \\ 0 \end{bmatrix} + yT \begin{bmatrix} 0 \\ 1 \end{bmatrix} = x \begin{bmatrix} 1 \\ 1 \end{bmatrix} + y \begin{bmatrix} 1 \\ 3 \end{bmatrix} = \begin{bmatrix} x+y \\ x+3y \end{bmatrix}.$$
- Let $T \in \mathcal{L}(\mathbb{R}^2, \mathbb{R}^2)$ s.t. $T \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 3 \end{bmatrix}$, $T(\mathbf{e}_2) = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$. What is $T \begin{bmatrix} x \\ y \end{bmatrix}$?
We have $T(\mathbf{e}_1) = T \left(\begin{bmatrix} 1 \\ 1 \end{bmatrix} - \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right) = T \left(\begin{bmatrix} 1 \\ 1 \end{bmatrix} \right) - T \left(\begin{bmatrix} 0 \\ 1 \end{bmatrix} \right) = \begin{bmatrix} 1 \\ 3 \end{bmatrix} - \begin{bmatrix} 3 \\ 1 \end{bmatrix} = \begin{bmatrix} -2 \\ 2 \end{bmatrix}$.
So $T \begin{bmatrix} x \\ y \end{bmatrix} = xT(\mathbf{e}_1) + yT(\mathbf{e}_2) = x \begin{bmatrix} -2 \\ 2 \end{bmatrix} + y \begin{bmatrix} 3 \\ 1 \end{bmatrix} = \begin{bmatrix} -2x+3y \\ 2x+y \end{bmatrix}.$

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• Alternate: $T\left(\begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \alpha \\ \beta \end{bmatrix}\right) = T\left(\alpha \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \beta \begin{bmatrix} 0 \\ 1 \end{bmatrix}\right) = \alpha T\begin{bmatrix} 1 \\ 1 \end{bmatrix} + \beta T\begin{bmatrix} 0 \\ 1 \end{bmatrix} = \alpha \begin{bmatrix} 1 \\ 3 \end{bmatrix} + \beta \begin{bmatrix} 3 \\ 1 \end{bmatrix}$
 $= \begin{bmatrix} 1 & 3 \\ 3 & 1 \end{bmatrix} \begin{bmatrix} \alpha \\ \beta \end{bmatrix} = \begin{bmatrix} T\begin{bmatrix} 1 \\ 1 \end{bmatrix} & T\begin{bmatrix} 0 \\ 1 \end{bmatrix} \end{bmatrix} \begin{bmatrix} \alpha \\ \beta \end{bmatrix}.$

So, $T\begin{bmatrix} x \\ y \end{bmatrix} = T\left(\begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix} \left(\begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix}^{-1} \begin{bmatrix} x \\ y \end{bmatrix}\right)\right) = \begin{bmatrix} 1 & 3 \\ 3 & 1 \end{bmatrix} \left(\begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix}^{-1} \begin{bmatrix} x \\ y \end{bmatrix}\right)$
 $= \begin{bmatrix} 1 & 3 \\ 3 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} -2x + 3y \\ 2x + y \end{bmatrix}.$

• Let $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ be a basis of \mathbb{V} and $T \in \mathcal{L}(\mathbb{V}, \mathbb{W})$. Then T is determined by $\{T\mathbf{v}_1, \dots, T\mathbf{v}_n\}$.

If we know $\{T\mathbf{v}_1, \dots, T\mathbf{v}_n\}$, then we know $T\mathbf{v}$ for all $\mathbf{v} \in \mathbb{V}$ as $\mathbf{v} = \sum \alpha_i \mathbf{v}_i \Rightarrow T(\mathbf{v}) = \sum \alpha_i T(\mathbf{v}_i)$.

Th. Let $T \in \mathcal{L}(\mathbb{V}, \mathbb{W})$ and $S \subseteq \mathbb{V}$ be lin.dep. Then $T(S)$ is lin.dep.

Pr. As S is lin.dep, there exist $\mathbf{v}_1, \dots, \mathbf{v}_k \in \mathbb{V}$, $\alpha_i \neq 0$ s.t. $\sum \alpha_i \mathbf{v}_i = \mathbf{0}$. So $\sum \alpha_i T\mathbf{v}_i = T(\sum \alpha_i \mathbf{v}_i) = T(\mathbf{0}) = \mathbf{0}$. Hence $\{T\mathbf{v}_1, \dots, T\mathbf{v}_k\}$ is lin.dep. ■

• So $T(S)$ is linearly independent $\Rightarrow S$ is linearly independent.

P. Linear Transformations- Kernel and Range

• Let $T \in \mathcal{L}(\mathbb{V}, \mathbb{W})$. Then

the **kernel/null space** of T is $\text{KER } T = \{\mathbf{v} \in \mathbb{V} | T\mathbf{v} = \mathbf{0}\}$.

The **range/image** of T is $\text{RNG } T = \{T\mathbf{v} : \mathbf{v} \in \mathbb{V}\}$.

• If $\text{DIM } \mathbb{V}$ is finite, we define **RANK** $T := \text{DIM RNG } T$; **NULLITY** $T := \text{DIM KER } T$.

Ex. Let $T \in \mathcal{L}(\mathbb{V}, \mathbb{W})$. Is $\text{KER } T$ a subspace of \mathbb{V} ? Is $\text{RNG } T$ a subspace of \mathbb{W} ?

• Let \mathcal{B} be a basis for \mathbb{V} and $T \in \mathcal{L}(\mathbb{V}, \mathbb{W})$. Then $\text{RNG } T = \text{LS } T(\mathcal{B})$. !!

T is determined by $\{T\mathbf{v} : \mathbf{v} \in \mathcal{B}\}$

• An LT from \mathbb{V} to \mathbb{W} is also called a **linear operator** on \mathbb{V} .

P. Linear Transformations-Ordered basis

• Is $\{\mathbf{v}_1 = \mathbf{e}_1 - \mathbf{e}_2, \mathbf{v}_2 = \mathbf{e}_1 + \mathbf{e}_2\}$ a basis of \mathbb{R}^2 ? **Ans:** Yes.

• Write $\mathbf{x} = \begin{bmatrix} 4 \\ 2 \end{bmatrix}$ as lin.comb of $\mathbf{v}_1, \mathbf{v}_2$: **Ans:** $\mathbf{x} = \mathbf{v}_1 + 3\mathbf{v}_2$. Coefficient vector: $\begin{bmatrix} 1 \\ 3 \end{bmatrix}$.

• Which \mathbf{x} has coefficient vector $\begin{bmatrix} 1 \\ 2 \end{bmatrix}$? $\mathbf{x} = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$. Here we use the i th coefficient for the i th basis vector \mathbf{v}_i . That is, $\{\mathbf{v}_1, \mathbf{v}_2\}$ is ordered.

• An **ordered basis** \mathcal{B} in \mathbb{V} is an 'ordered set' of vectors that form a basis of \mathbb{V} . For finite dimension, it means calling them 1st basis vector, 2nd basis vector, etc.

• Let $\text{DIM } \mathbb{V}$ be finite, \mathcal{B} be an ordered basis and $\mathbf{x} \in \mathbb{V}$. The coefficient vector (**coordinate matrix**) of \mathbf{x} w.r.t \mathcal{B} is denoted by $[\mathbf{x}]_{\mathcal{B}}$.

• In \mathbb{R}^2 , take $\mathcal{B} = \left\{ \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$, $\mathcal{B}' = \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \end{bmatrix} \right\}$. Let $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \in \mathbb{R}^2$.

Then $[\mathbf{x}]_{\mathcal{B}} = \frac{1}{2} \begin{bmatrix} x_1 - x_2 \\ x_1 + x_2 \end{bmatrix}$ and $[\mathbf{x}]_{\mathcal{B}'} = \frac{1}{2} \begin{bmatrix} x_1 + x_2 \\ x_1 - x_2 \end{bmatrix}$. Note that

$$\begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} [\mathbf{x}]_{\mathcal{B}} = \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} \frac{1}{2} \begin{bmatrix} \mathbf{x}_1 - \mathbf{x}_2 \\ \mathbf{x}_1 + \mathbf{x}_2 \end{bmatrix} = \mathbf{x} \text{ and } \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} [\mathbf{x}]_{\mathcal{B}'} = \mathbf{x}.$$

• Let $\mathbf{v} \in \mathbb{R}^n$ and $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ be an ordered basis of \mathbb{R}^n . Form a matrix $B = [\mathbf{v}_1 \ \cdots \ \mathbf{v}_n]$. Then by definition $B[\mathbf{v}]_{\mathcal{B}} = \mathbf{v}$. We are considering the augmented matrix $[B \mid \mathbf{v}]$ to get the entries of $[\mathbf{v}]_{\mathcal{B}}$.

$$\text{P. Linear Transformations-Matrix-}[\mathbf{v}]_{\mathcal{B}} = B^{-1}\mathbf{v}$$

Th. Let $\mathbf{v} \in \mathbb{R}^n$ and $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ be an ordered basis of \mathbb{R}^n . Put $B = [\mathbf{v}_1 \ \cdots \ \mathbf{v}_n]$ (this is called the **basis matrix**). Then

$B[[\mathbf{e}_1]_{\mathcal{B}} \ \cdots \ [\mathbf{e}_n]_{\mathcal{B}}] = [B[\mathbf{e}_1]_{\mathcal{B}} \ \cdots \ B[\mathbf{e}_n]_{\mathcal{B}}] = [\mathbf{e}_1 \ \cdots \ \mathbf{e}_n] = I$. Hence, B is invertible and $[\mathbf{v}]_{\mathcal{B}} = B^{-1}\mathbf{v}$.

• Let $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ be an ordered basis of \mathbb{V} . Let $\mathcal{B}' = \{\mathbf{w}_1, \dots, \mathbf{w}_m\}$ be an ordered basis of \mathbb{W} and $T \in \mathcal{L}(\mathbb{V}, \mathbb{W})$.

• Let A be a matrix s.t. $A[\mathbf{v}]_{\mathcal{B}} = [T\mathbf{v}]_{\mathcal{B}'}$ for each $\mathbf{v} \in \mathbb{V}$. Then ‘in some sense’ A takes \mathbf{v} to $T\mathbf{v}$. So we call A a **coordinate matrix** of T .

• Note: $A(:, i) = A\mathbf{e}_i = A[\mathbf{v}_i]_{\mathcal{B}} = [T\mathbf{v}_i]_{\mathcal{B}'}$, by definition. So A is unique.

• We denote the coordinate matrix of T by $T[\mathcal{B}, \mathcal{B}']$ or simply by $[T]$.

• When there is no mention of basis, we take the standard basis.

Example. Take $T \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} x_1 + x_2 \\ x_1 - x_2 \end{bmatrix}$ on \mathbb{R}^2 .

a) Then $[T] = [Te_1 \ Te_2] = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$.

b) On the image space take the ordered basis $\mathcal{B}' = \left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$.

Then $[T] = [Te_1]_{\mathcal{B}'} \ [Te_2]_{\mathcal{B}'} = \left[\begin{bmatrix} 1 \\ 1 \end{bmatrix}_{\mathcal{B}'} \ \begin{bmatrix} 1 \\ -1 \end{bmatrix}_{\mathcal{B}'} \right] = \begin{bmatrix} 0 & 2 \\ 1 & -1 \end{bmatrix}$.

c) In b) take $\mathcal{B} = \left\{ \begin{bmatrix} -1 \\ 1 \end{bmatrix}, \begin{bmatrix} 3 \\ 1 \end{bmatrix} \right\}$ as the ordered basis of domain.

Then $[T] = \left[\left[T \begin{bmatrix} -1 \\ 1 \end{bmatrix} \right]_{\mathcal{B}'} \ \left[T \begin{bmatrix} 3 \\ 1 \end{bmatrix} \right]_{\mathcal{B}'} \right] = \left[\begin{bmatrix} 0 \\ -2 \end{bmatrix}_{\mathcal{B}'} \ \begin{bmatrix} 4 \\ 2 \end{bmatrix}_{\mathcal{B}'} \right] = \begin{bmatrix} 2 & 2 \\ -2 & 2 \end{bmatrix}$.

• Fix $A \in \mathcal{M}_n(\mathbb{C})$. Take $T \in \mathcal{L}(\mathbb{C}^n, \mathbb{C}^n)$ defined as $T\mathbf{x} = A\mathbf{x}$. Let \mathcal{B} be the standard basis in \mathbb{C}^n . Put $B = [T]$. Then $B(:, i) = [Te_i]_{\mathcal{B}} = [A\mathbf{e}_i]_{\mathcal{B}} = [A(:, i)]_{\mathcal{B}} = A(:, i)$. **Is this a bijection?**

• Fix $A \in \mathcal{M}_n(\mathbb{C})$. Define $T \in \mathcal{L}(\mathbb{C}^n, \mathbb{C}^n)$ as $T\mathbf{x} = A\mathbf{x}$. Take ordered bases $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ on the domain and $\mathcal{B}' = \{\mathbf{w}_1, \dots, \mathbf{w}_n\}$ on the range. Let B and B' be the respective basis matrices. Then

$$[T] = [T\mathbf{v}_1]_{\mathcal{B}'} \ \cdots \ [T\mathbf{v}_n]_{\mathcal{B}'} = [B'^{-1}T\mathbf{v}_1 \ \cdots \ B'^{-1}T\mathbf{v}_n] = [B'^{-1}A\mathbf{v}_1 \ \cdots \ B'^{-1}A\mathbf{v}_n] = B'^{-1}A[\mathbf{v}_1 \ \cdots \ \mathbf{v}_n] = B'^{-1}AB.$$

• In particular, if $\mathcal{B} = \mathcal{B}'$, then we have $[T] = B^{-1}AB$.

- Let $\mathbb{V} = \mathbb{R}[t; 3]$ and $\mathbb{W} = \mathbb{R}[t; 2]$. Take $\mathcal{B} = \{1, t, t^2, t^3\}$ on \mathbb{V} and $\mathcal{B}' = \{1, t, t^2\}$ on \mathbb{W} .

Take $T : \mathbb{V} \rightarrow \mathbb{W}$ defined as $T(f) = f'$, **differentiation transformation**. Then

$$[T] = \begin{bmatrix} [T1]_{\mathcal{B}'} & [Tt]_{\mathcal{B}'} & [Tt^2]_{\mathcal{B}'} & [Tt^3]_{\mathcal{B}'} \\ [0]_{\mathcal{B}'} & [1]_{\mathcal{B}'} & [2t]_{\mathcal{B}'} & [3t^2]_{\mathcal{B}'} \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 3 \end{bmatrix}.$$

- General: let $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$, $\mathcal{B}' = \{\mathbf{w}_1, \dots, \mathbf{w}_m\}$ be ordered bases in \mathbb{V} and \mathbb{W} with basis matrices B and B' , respectively. Let $T \in \mathcal{L}(\mathbb{V}, \mathbb{W})$. Then $Tv = T \sum \alpha_i v_i = \sum \alpha_i T v_i$

$$= [Tv_1 \ \dots \ Tv_n] \begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{bmatrix} = [w_1 \ \dots \ w_m] [[Tv_1]_{\mathcal{B}'} \ \dots \ [Tv_n]_{\mathcal{B}'}] \begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{bmatrix} = [w_1 \ \dots \ w_m] T[\mathcal{B}, \mathcal{B}'] [v]_{\mathcal{B}} =$$

$$B'[T][v]_{\mathcal{B}} = B'[T]B^{-1}v \quad (\text{R1})$$

- Fix $A \in \mathcal{M}_{m \times n}(\mathbb{R})$, bases \mathcal{B} in \mathbb{R}^n and \mathcal{B}' in \mathbb{R}^m . Then there exists $T \in \mathcal{L}(\mathbb{R}^n, \mathbb{R}^m)$ s.t.

$[T] = A$. Let B be the matrix of \mathcal{B} and B' be that of \mathcal{B}' . Then by (R1)

$$T\mathbf{v} = B'A[\mathbf{v}]_{\mathcal{B}} = B'AB^{-1}\mathbf{v}. \quad (\text{R2})$$

- On \mathbb{R}^2 take $\mathcal{B} = \left\{ \begin{bmatrix} -1 \\ 1 \end{bmatrix}, \begin{bmatrix} 3 \\ 1 \end{bmatrix} \right\}$ and $\mathcal{B}' = \left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$. Take $A = \begin{bmatrix} 2 & 2 \\ -2 & 2 \end{bmatrix}$.

- Then the associated linear transformation is $T\mathbf{x} = B'A[\mathbf{x}]_{\mathcal{B}} = B'AB^{-1}\mathbf{x} = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 2 & 2 \\ -2 & 2 \end{bmatrix} \begin{bmatrix} -\frac{1}{4} & \frac{3}{4} \\ \frac{1}{4} & \frac{1}{4} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} =$

$$\begin{bmatrix} 0 & 4 \\ -2 & 2 \end{bmatrix} \begin{bmatrix} -\frac{1}{4} & \frac{3}{4} \\ \frac{1}{4} & \frac{1}{4} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}.$$

P. Composition of Linear Transformations-Matrix Product

Ex. Let $\mathbb{U}, \mathbb{V}, \mathbb{W}$ be VS's over \mathbb{F} . Let $T \in \mathcal{L}(\mathbb{U}, \mathbb{V})$ and $S \in \mathcal{L}(\mathbb{V}, \mathbb{W})$. Define **composition** $ST : \mathbb{U} \rightarrow \mathbb{W}$ as $(ST)(\mathbf{x}) = S(T(\mathbf{x}))$. Then, ST is a linear transformation.

Now, let $\mathcal{B}, \mathcal{B}_1$ and \mathcal{B}_2 be bases of \mathbb{U}, \mathbb{V} and \mathbb{W} , respectively. Then, by definition $[ST(\mathbf{x})]_{\mathcal{B}_2} = ST[\mathcal{B}, \mathcal{B}_2][\mathbf{x}]_{\mathcal{B}}$.

Also,

$$[ST(\mathbf{x})]_{\mathcal{B}_2} = [S(T(\mathbf{x}))]_{\mathcal{B}_2} = S[\mathcal{B}_1, \mathcal{B}_2][T(\mathbf{x})]_{\mathcal{B}_1} = S[\mathcal{B}_1, \mathcal{B}_2]T[\mathcal{B}, \mathcal{B}_1][\mathbf{x}]_{\mathcal{B}}.$$

Hence, for all $\mathbf{x} \in \mathbb{U}$, we have

$$ST[\mathcal{B}, \mathcal{B}_2][\mathbf{x}]_{\mathcal{B}} = S[\mathcal{B}_1, \mathcal{B}_2]T[\mathcal{B}, \mathcal{B}_1][\mathbf{x}]_{\mathcal{B}}.$$

Thus,

$$ST[\mathcal{B}, \mathcal{B}_2] = S[\mathcal{B}_1, \mathcal{B}_2]T[\mathcal{B}, \mathcal{B}_1].$$