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- Linear transformations and matrices. [6.1, 6.3, 6.4, 8.1]
- Let A be an $m \times n$ matrix. For any $n \times 1$ matrix \mathbf{x} , $A\mathbf{x}$ is an $m \times 1$ matrix. So A can be used to define a mapping from \mathbb{R}^n to \mathbb{R}^m .
- A mapping $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is said to be **linear**, if it preserves scalar multiplication and addition, i.e., for any $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$ and $\alpha \in \mathbb{R}$, $T(\alpha\mathbf{x}) = \alpha T(\mathbf{x})$ and $T(\mathbf{x} + \mathbf{y}) = T(\mathbf{x}) + T(\mathbf{y})$.

In other words, T is linear if it maps a linear combination $\alpha\mathbf{x} + \beta\mathbf{y}$ of vectors \mathbf{x} and \mathbf{y} to a linear combination $\alpha T(\mathbf{x}) + \beta T(\mathbf{y})$ of $T(\mathbf{x})$ and $T(\mathbf{y})$.

In linear algebra, a linear mapping is called a linear transformation usually.

- Kernel and range. Let $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a linear transformation. The **kernel** of T , $\ker(T)$, is the set $\{\mathbf{v} \in \mathbb{R}^n : T(\mathbf{v}) = \mathbf{0}\}$ and the **range** of T , $\text{ran}(T)$, is the set $\{T(\mathbf{v}) : \mathbf{v} \in \mathbb{R}^n\}$. We can show that $\ker(T)$ is a subspace of \mathbb{R}^n and $\text{ran}(T)$ is a subspace of \mathbb{R}^m .
- For an $m \times n$ matrix A we associate a linear transformation $T_A : \mathbb{R}^n \rightarrow \mathbb{R}^m$ by $T_A(\mathbf{x}) = A\mathbf{x}$. Then $\ker(T_A)$ is the solution space of $A\mathbf{x} = \mathbf{0}$, or the **null space** of A , $\text{null}(A)$, and $\text{ran}(T_A)$ is the column space of A .
- Let $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a linear transformation and let A be the $m \times n$ matrix whose i -th column is $T(\mathbf{e}_i)$ for all $i = 1, 2, \dots, n$. Then $T = T_A$.
- Questions. Let $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a linear transformation. Let $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k$ be vectors in \mathbb{R}^n .
 - If $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k$ are linearly independent, are $T(\mathbf{v}_1), T(\mathbf{v}_2), \dots, T(\mathbf{v}_k)$ linearly independent?
 - If $T(\mathbf{v}_1), T(\mathbf{v}_2), \dots, T(\mathbf{v}_k)$ are linearly independent, are $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k$ linearly independent?

- A transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is said to be onto if its range is the entire codomain \mathbb{R}^m .
- A transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is said to be one-to-one if T maps distinct vectors in \mathbb{R}^n into distinct vectors in \mathbb{R}^m .

- A linear transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is one-to-one if and only if $\ker(T) = \{\mathbf{0}\}$.

Proof. Suppose T is one-to-one. Since T is linear, we have that $T(\mathbf{0}) = \mathbf{0}$. The fact that T is one-to-one implies that $\ker(T) = \{\mathbf{0}\}$.

Suppose $\ker(T) = \{\mathbf{0}\}$. We will show that if $\mathbf{x}_1 \neq \mathbf{x}_2$, then $T(\mathbf{x}_1) \neq T(\mathbf{x}_2)$. If $\mathbf{x}_1 \neq \mathbf{x}_2$, then $\mathbf{x}_1 - \mathbf{x}_2 \neq \mathbf{0}$, so $\mathbf{x}_1 - \mathbf{x}_2 \notin \ker(T)$. Then, $T(\mathbf{x}_1 - \mathbf{x}_2) = T(\mathbf{x}_1) - T(\mathbf{x}_2) \neq \mathbf{0}$.

- Let $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a linear transformation. Let $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k$ be vectors in \mathbb{R}^n .
 - If $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k$ are linearly independent and $\ker(T) = \{\mathbf{0}\}$, then $T(\mathbf{v}_1), T(\mathbf{v}_2), \dots, T(\mathbf{v}_k)$ are linearly independent.
 - If $T(\mathbf{v}_1), T(\mathbf{v}_2), \dots, T(\mathbf{v}_k)$ are linearly independent, then $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k$ are linearly independent.

- A linear transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is one-to-one if and only if it is onto.

Proof. Suppose T is one-to-one. Its kernel is $\{\mathbf{0}\}$. The set $\{T(\mathbf{e}_1), \dots, T(\mathbf{e}_n)\}$ is linearly independent. So the range of T has dimension at least n , which implies that the range of T is \mathbb{R}^n .

Suppose T is onto. There are vectors \mathbf{v}_i such that $T(\mathbf{v}_i) = \mathbf{e}_i$. The set $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ is linearly independent and is a basis for \mathbb{R}^n . Let $\mathbf{x} = \sum_{i=1}^n a_i \mathbf{v}_i \in \ker(T)$. Then $T(\mathbf{x}) = \sum_{i=1}^n a_i T(\mathbf{v}_i) = \sum_{i=1}^n a_i \mathbf{e}_i = \mathbf{0}$, which implies that $a_i = 0$ for all i , and so $\mathbf{x} = \mathbf{0}$.

- If T is a one-to-one linear transformation, then so is T^{-1} .
- Let $T_A : \mathbb{R}^n \rightarrow \mathbb{R}^k$, $T_B : \mathbb{R}^k \rightarrow \mathbb{R}^m$ be linear transformations. Then $T_B \circ T_A = T_{BA}$ is also a linear transformation.

- Linear transformations in $\mathbb{R}, \mathbb{R}^2, \mathbb{R}^3$ (Details in the next lecture)
 - \mathbb{R} : $T : \mathbb{R} \rightarrow \mathbb{R}$, $x \mapsto ax$.
 - \mathbb{R}^2 : Reflection, orthogonal projection, rotation
 - \mathbb{R}^3 : Reflection, orthogonal projection, rotation
- Reflection in \mathbb{R}^n
- Projection in \mathbb{R}^n
- Rotation in \mathbb{R}^n
- When we say $\mathcal{B} = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ is a basis, we mean an ordered basis, i.e., the order of $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ is fixed.
- \mathbb{R}^n with an ordered basis $\mathcal{B} = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$. The coordinate matrix $[\mathbf{x}]_{\mathcal{B}}$
- \mathbb{R}^n with basis $\mathcal{S} = \{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n\}$, the standard basis. The coordinate matrix $[\mathbf{x}]_{\mathcal{S}} = [\mathbf{x}] = \mathbf{x}$
- $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$, \mathbb{R}^n with basis $\mathcal{B} = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$, $[T]_{\mathcal{B}}$, matrix for T with respect to the basis \mathcal{B} . The i -th column of $[T]_{\mathcal{B}}$ is $[T(\mathbf{v}_i)]_{\mathcal{B}}$.
- Let $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a linear transformation. Choose a basis $\mathcal{B} = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ for \mathbb{R}^n and a basis $\mathcal{B}' = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_m\}$ for \mathbb{R}^m . We define $[T]_{\mathcal{B}', \mathcal{B}}$, the matrix for T with respect to bases \mathcal{B} and \mathcal{B}' .
- $T : (\mathbb{R}^n, \mathcal{B}) \rightarrow (\mathbb{R}^m, \mathcal{B}')$, $[T]_{\mathcal{B}', \mathcal{B}}$, the matrix for T with respect to bases \mathcal{B} and \mathcal{B}' .

$$[T(\mathbf{x})]_{\mathcal{B}'} = [T]_{\mathcal{B}', \mathcal{B}} [\mathbf{x}]_{\mathcal{B}}$$

The i -th column of $[T]_{\mathcal{B}', \mathcal{B}}$ is $[T(\mathbf{v}_i)]_{\mathcal{B}'}$.

- $T_A : (\mathbb{R}^n, \mathcal{S}) \rightarrow (\mathbb{R}^m, \mathcal{S}')$, $[T_A]_{\mathcal{S}', \mathcal{S}} = A$, the matrix for T_A with respect to bases \mathcal{S} and \mathcal{S}' .
- $\text{id} : (\mathbb{R}^n, \mathcal{S}) \rightarrow (\mathbb{R}^n, \mathcal{B})$, $[\text{id}]_{\mathcal{B}, \mathcal{S}}$, the matrix for the identity mapping, id , with respect to bases \mathcal{S} and \mathcal{B} .

$$[\text{id}(\mathbf{x})]_{\mathcal{B}} = [\text{id}]_{\mathcal{B}, \mathcal{S}} [\mathbf{x}]_{\mathcal{S}} \quad \rightarrow \quad [\mathbf{x}]_{\mathcal{B}} = [\text{id}]_{\mathcal{B}, \mathcal{S}} [\mathbf{x}]_{\mathcal{S}}$$

- $\text{id} : (\mathbb{R}^n, \mathcal{B}) \rightarrow (\mathbb{R}^n, \mathcal{S})$, $[\text{id}]_{\mathcal{S}, \mathcal{B}}$, the matrix for the identity mapping, id , with respect to bases \mathcal{B} and \mathcal{S} .

$$[\text{id}(\mathbf{x})]_{\mathcal{S}} = [\text{id}]_{\mathcal{S}, \mathcal{B}} [\mathbf{x}]_{\mathcal{B}} \quad \rightarrow \quad [\mathbf{x}]_{\mathcal{S}} = [\text{id}]_{\mathcal{S}, \mathcal{B}} [\mathbf{x}]_{\mathcal{B}}$$

The i -th column of $[\text{id}]_{\mathcal{S}, \mathcal{B}}$ is $[\mathbf{v}_i]_{\mathcal{S}} = \mathbf{v}_i$.

- The relation between $[T]_{\mathcal{S}} : (\mathbb{R}^n, \mathcal{S}) \rightarrow (\mathbb{R}^n, \mathcal{S})$ and $[T]_{\mathcal{B}} : (\mathbb{R}^n, \mathcal{B}) \rightarrow (\mathbb{R}^n, \mathcal{B})$.

$$[T(\mathbf{x})]_{\mathcal{S}} = [T]_{\mathcal{S}} [\mathbf{x}]_{\mathcal{S}} = [\text{id}]_{\mathcal{S}, \mathcal{B}} [T]_{\mathcal{B}} [\text{id}]_{\mathcal{B}, \mathcal{S}} [\mathbf{x}]_{\mathcal{S}}$$

$$[T]_{\mathcal{S}} = [\text{id}]_{\mathcal{S}, \mathcal{B}} [T]_{\mathcal{B}} [\text{id}]_{\mathcal{B}, \mathcal{S}}$$

$$[T]_{\mathcal{B}} = [\text{id}]_{\mathcal{B}, \mathcal{S}} [T]_{\mathcal{S}} [\text{id}]_{\mathcal{S}, \mathcal{B}}$$