

Shifrin 1.4 Linear Transformations.

①

Definition. A function $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is called a linear transformation or linear map if

$$1) T(\vec{u} + \vec{v}) = T(\vec{u}) + T(\vec{v})$$

$$2) T(c\vec{v}) = cT(\vec{v}).$$

These properties are so natural that one tends to assume they always hold.

Self-check. Which of the following are linear?

$$x \mapsto ax + b$$

$$f(x) \mapsto f'(x)$$

$$x \mapsto ax$$

$$f(x) \mapsto \int_0^1 f(x) dx$$

$$x \mapsto x^2$$

$$f(x) \mapsto \int f(x) dx$$

$$\vec{x} \mapsto \vec{v} \cdot \vec{x}$$

②

Notice that if $\vec{X} = x_1 \vec{e}_1 + \dots + x_n \vec{e}_n$
~~then~~ and T is linear, then

$$\begin{aligned} T(\vec{X}) &= T(x_1 \vec{e}_1 + \dots + x_n \vec{e}_n) \\ &= T(x_1 \vec{e}_1) + \dots + T(x_n \vec{e}_n) \\ &= x_1 T(\vec{e}_1) + \dots + x_n T(\vec{e}_n). \end{aligned}$$

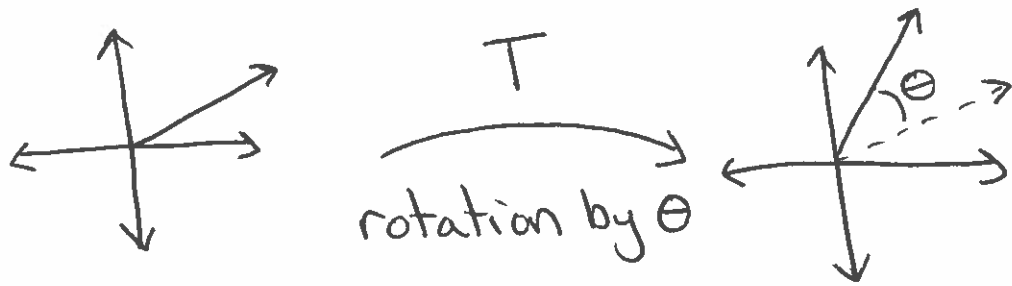
This means that T is really determined
by the vectors $T(\vec{e}_1), \dots, T(\vec{e}_n)$. We

encode $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ by the $m \times n$ array

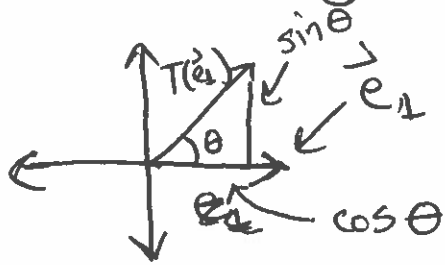
$$[T] = \underbrace{\left[\begin{array}{ccc} \uparrow & & \uparrow \\ T(\vec{e}_1) & \dots & T(\vec{e}_n) \\ \downarrow & & \downarrow \end{array} \right]}_{n \text{ columns}} \left. \vphantom{\left[\begin{array}{ccc} \uparrow & & \uparrow \\ T(\vec{e}_1) & \dots & T(\vec{e}_n) \\ \downarrow & & \downarrow \end{array} \right]} \right\} m \text{ rows}$$

which is called the standard matrix for T .

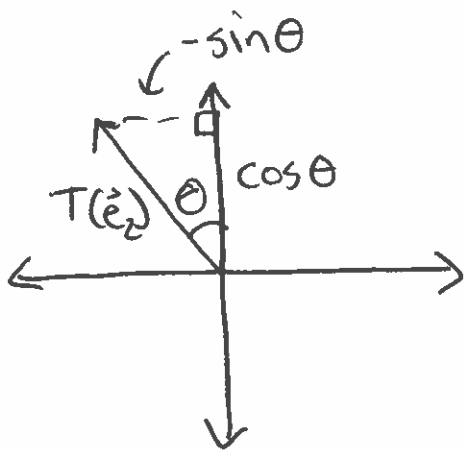
We now give some useful examples. ③



We can find $T(\vec{e}_1)$ and $T(\vec{e}_2)$ with some trigonometry



$$T(\vec{e}_1) = \begin{bmatrix} \cos \theta \\ \sin \theta \end{bmatrix}$$

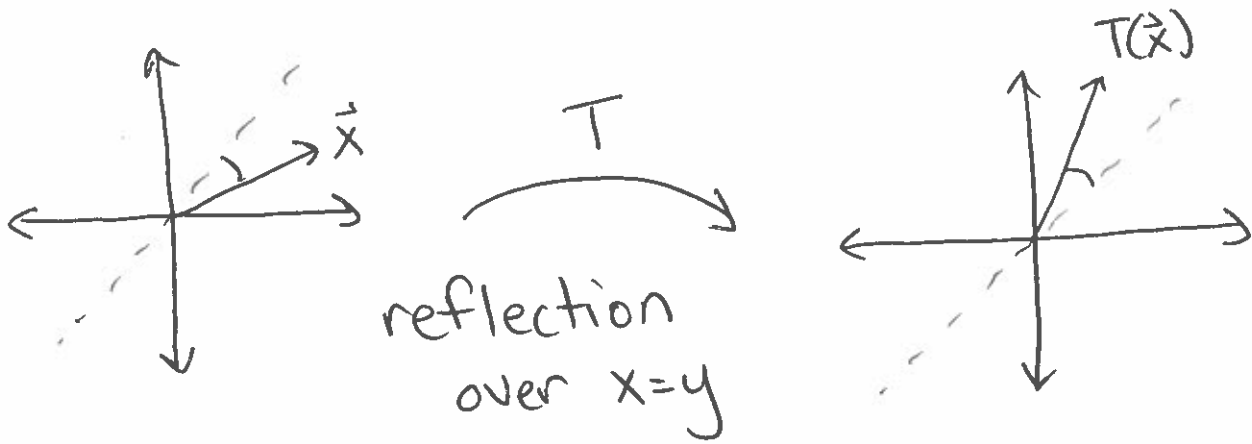


$$T(\vec{e}_2) = \begin{bmatrix} -\sin \theta \\ \cos \theta \end{bmatrix}$$

so

$$[T] = \begin{bmatrix} \uparrow & \uparrow \\ T(\vec{e}_1) & T(\vec{e}_2) \\ \downarrow & \downarrow \end{bmatrix} = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

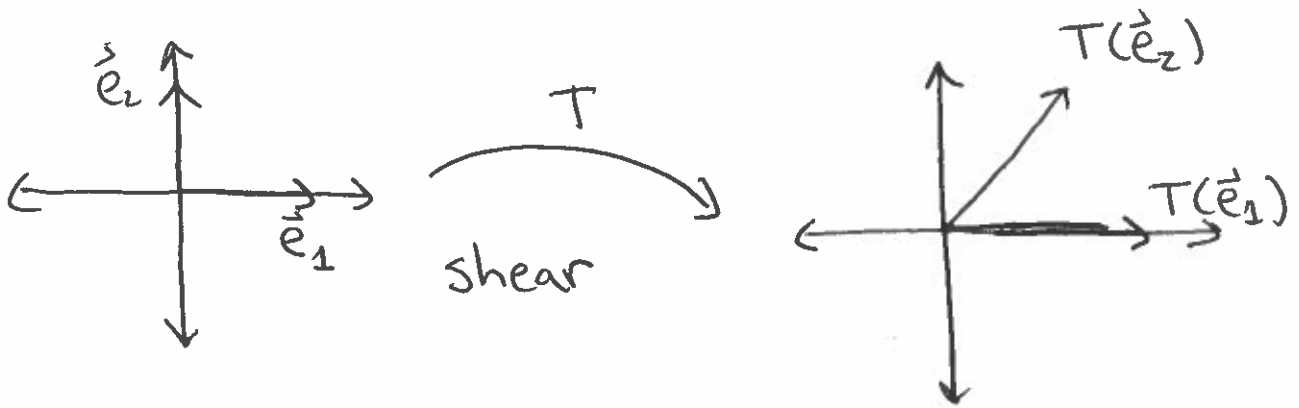
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Here $T\left(\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}\right) = \begin{bmatrix} x_2 \\ x_1 \end{bmatrix}$. (swaps coords)

so

$$[T] = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$



$$[T] = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$$

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Definition. We call an $m \times n$ array of real numbers an $m \times n$ matrix, A . We define (for a vector $\vec{x} \in \mathbb{R}^n$)

$$A\vec{x} = \begin{matrix} \text{[Scribbled matrix]} \\ \left[\begin{array}{c} \leftarrow A_1 \rightarrow \\ \leftarrow A_m \rightarrow \end{array} \right] \end{matrix} \vec{x} = \begin{bmatrix} A_1 \cdot \vec{x} \\ \vdots \\ A_m \cdot \vec{x} \end{bmatrix} \in \mathbb{R}^m$$

or

$$(A\vec{x})_j = \sum_{i=1}^n A_{ji} x_i$$

Proposition. If $A = [T]$ is the standard matrix for a linear transformation T , then $A\vec{x} = T(\vec{x})$.

Proof. If $\vec{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = x_1 \vec{e}_1 + \dots + x_n \vec{e}_n$, then

$$T(\vec{x}) = x_1 T(\vec{e}_1) + \dots + x_n T(\vec{e}_n).$$

⑥

Now each $T(\vec{e}_j)$ can be written in terms of the standard basis for \mathbb{R}^m , which we call $\vec{d}_1, \dots, \vec{d}_m$ just this once. So

$$T(\vec{x}) = x_1 (T(\vec{e}_1)_1 \vec{d}_1 + \dots + T(\vec{e}_1)_m \vec{d}_m) + \dots + x_n (T(\vec{e}_n)_1 \vec{d}_1 + \dots + T(\vec{e}_n)_m \vec{d}_m)$$

and regrouping, we get

$$\begin{aligned} &= (x_1 T(\vec{e}_1)_1 + \dots + x_n T(\vec{e}_n)_1) \vec{d}_1 \\ &\quad + \dots + (x_1 T(\vec{e}_1)_m + \dots + x_n T(\vec{e}_n)_m) \vec{d}_m \\ &= (x_1 A_{11} + \dots + x_n A_{1n}) \vec{d}_1 \\ &\quad + \dots + (x_1 A_{m1} + \dots + x_n A_{mn}) \vec{d}_m \end{aligned}$$

⑦

$$= \begin{bmatrix} A_1 \cdot \vec{x} \\ \vdots \\ A_m \cdot \vec{x} \end{bmatrix} \in \mathbb{R}^m, \text{ as claimed. } \square$$

Motto of the day: The linear transformation T and the standard matrix $[T]$ are not the same thing.

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We will call the set of $m \times n$ matrices $\text{Mat}_{m \times n}$ or $M_{m \times n}$.

Look up: square matrix, diagonal entries, diagonal matrix, upper triangular, lower triangular.

Definition. If $S, T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ are linear maps, and $c \in \mathbb{R}$, we define

$$(cT)(\vec{x}) = c(T(\vec{x}))$$

$$(S+T)(\vec{x}) = S(\vec{x}) + T(\vec{x}).$$

There are corresponding operations on the standard matrices for these linear maps.

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For $A \in \text{Mat}_{m \times n}$, $c \in \mathbb{R}$, we have

$$cA = c \begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \dots & a_{mn} \end{bmatrix} = \begin{bmatrix} ca_{11} & \dots & ca_{1n} \\ \vdots & \ddots & \vdots \\ ca_{m1} & \dots & ca_{mn} \end{bmatrix}$$

And for $A, B \in \text{Mat}_{m \times n}$,

$$\begin{aligned} A+B &= \begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \dots & a_{mn} \end{bmatrix} + \begin{bmatrix} b_{11} & \dots & b_{1n} \\ \vdots & \ddots & \vdots \\ b_{m1} & \dots & b_{mn} \end{bmatrix} \\ &= \begin{bmatrix} a_{11}+b_{11} & \dots & a_{1n}+b_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1}+b_{m1} & \dots & a_{mn}+b_{mn} \end{bmatrix} \end{aligned}$$

Proposition. $c[T] = [cT]$, $[S]+[T] = [S+T]$.

Proof. To be done in homework.

Example.

$$A = \begin{bmatrix} 1 & 2 \\ 4 & 6 \end{bmatrix} \quad B = \begin{bmatrix} -2 & 3 \\ 5 & -1 \end{bmatrix} \quad C = \begin{bmatrix} 1 & 2 & 1 \\ 3 & -4 & 7 \end{bmatrix}$$

Find $A+B$, $2C$. Can ~~you~~ you
add $A+C$? (No!)

There is an algebra of matrices like
the algebra of vectors.

Proposition. Let A, B, C be matrices in $\text{Mat}_{m \times n}$
and $c, d \in \mathbb{R}$, and O be the $m \times n$ matrix of 0 's

1. $A + B = B + A$

5. $c(dA) = (cd)A$

2. $(A+B)+C = A+(B+C)$

6. $c(A+B) = cA + cB$

3. $O + A = A$

7. $(c+d)A = cA + dA$

4. There is a matrix

8. $1A = A$

$-A$ so that $A + (-A) = O$

(so matrices are a real vector space).

Now let's go back to our basic idea.

(11)

$T: \mathbb{R}^n \rightarrow \mathbb{R}^m$
a linear transformation $\longleftrightarrow [T] \in \text{Mat}_{m \times n}$
an $m \times n$ matrix

If we have $S: \mathbb{R}^p \rightarrow \mathbb{R}^n$ and $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$,
we ~~must~~ can define the composition.

$T \circ S: \mathbb{R}^p \rightarrow \mathbb{R}^m$ by $(T \circ S)(\vec{x}) = T(S(\vec{x}))$.

Lemma. If $S: \mathbb{R}^p \rightarrow \mathbb{R}^n$ and $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ are
linear transformations, then $T \circ S: \mathbb{R}^p \rightarrow \mathbb{R}^m$
is also a linear transformation.

Proof. $(T \circ S)(\vec{x} + \vec{y}) = T(S(\vec{x} + \vec{y}))$
 $= T(S(\vec{x}) + S(\vec{y}))$
 $= T(S(\vec{x})) + T(S(\vec{y}))$
 $= (T \circ S)(\vec{x}) + (T \circ S)(\vec{y}).$

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$$\begin{aligned}
 (T \circ S)(c\vec{x}) &= T(S(c\vec{x})) \\
 &= T(cS(\vec{x})) \\
 &= cT(S(\vec{x})). \quad \square
 \end{aligned}$$

This raises a natural question: how are the standard matrices $[S] \in \text{Mat}_{p \times n}$, $[T] \in \text{Mat}_{m \times n}$ and $[T \circ S] \in \text{Mat}_{m \times p}$ related?

Definition. Let A be an $m \times n$ matrix and B be an $n \times p$ matrix. The matrix product AB is defined by

$$\begin{aligned}
 (AB)_{ij} &= \sum_{k=1}^n a_{ik} b_{kj} \\
 &= (\text{ith row of } A) \cdot (\text{jth column of } B)
 \end{aligned}$$

Example.

$$\begin{bmatrix} 1 & 3 \\ 2 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 4 & 1 \\ -1 & 1 \end{bmatrix} = \begin{bmatrix} 4-3 & 1+3 \\ 8+1 & 2-1 \\ 4-1 & 1+1 \end{bmatrix} = \begin{bmatrix} 1 & 4 \\ 9 & 1 \\ 3 & 2 \end{bmatrix}$$

3×2 2×2 3×2

Definition. If the matrix product AB exists, we say A, B are conformable for multiplication.

Lemma. A, B are conformable \Leftrightarrow the # of columns of $A =$ # of rows of B .

Proposition. Suppose $S: \mathbb{R}^p \rightarrow \mathbb{R}^n$ and $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ are linear transformations. Then

$$[T \circ S] = [T][S]$$

\uparrow
 matrix product

Proof. Suppose that $A = [T]$ and $B = [S]$. For any $\vec{x} \in \mathbb{R}^p$, we know

$$\begin{aligned}
 [T \circ S] \vec{x} &= T(S(\vec{x})) \\
 &= T(B\vec{x}) \\
 &= A(B\vec{x})
 \end{aligned}$$

matrix vector mult

We must therefore show that

$$A(B\vec{x}) = (AB)\vec{x}$$

matrix vector mult matrix matrix mult matrix vector mult

Now

$$\begin{aligned}
 (A(B\vec{x}))_{j_1} &= \sum_{i=1}^n A_{ji} (B\vec{x})_{i_1} \\
 &= \sum_{i=1}^n A_{ji} \left(\sum_{k=1}^p B_{ik} \vec{x}_{k_1} \right)
 \end{aligned}$$

$$= \sum_{i=1}^n \sum_{k=1}^p A_{ji} B_{ik} \vec{x}_{k1}$$

$$= \sum_{k=1}^p \left(\sum_{i=1}^n A_{ji} B_{ik} \right) \vec{x}_{k1}$$

$$= \sum_{k=1}^p (AB)_{jk} \vec{x}_{k1}$$

$$= ((AB) \vec{x})_{j1} \quad \square$$