

Lecture 6: Coordinates

Today we discuss coordinates with respect to a basis: this turns abstract vectors into column vectors (in \mathbb{R}^n) and abstract transformations into matrices.

Let V be a vector space (of dim. n) and $B = \{\vec{b}_1, \dots, \vec{b}_n\}$ a basis of V . Given a vector $\vec{v} \in V$, we can write $\vec{v} = \sum_{i=1}^n v_i \vec{b}_i$ in exactly one way. Define the coordinate vector (with respect to B)

$$[\vec{v}]_B := \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix} \in \mathbb{R}^n.$$

Proposition 1: $[\cdot]_B : V \rightarrow \mathbb{R}^n$ is a (linear) isomorphism.

Proof: If $\vec{w} = \sum w_i \vec{b}_i$ and $\vec{v} = \sum v_i \vec{b}_i$, then

$$\alpha \vec{w} + \beta \vec{v} = \sum (\alpha w_i + \beta v_i) \vec{b}_i \text{ and so}$$

$$[\alpha \vec{w} + \beta \vec{v}]_B = \begin{pmatrix} \alpha w_1 + \beta v_1 \\ \vdots \\ \alpha w_n + \beta v_n \end{pmatrix} = \alpha \begin{pmatrix} w_1 \\ \vdots \\ w_n \end{pmatrix} + \beta \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix} = \alpha [\vec{w}]_B + \beta [\vec{v}]_B.$$

That is,

$[\cdot]_{\mathcal{B}}$ is linear. $\mathbb{I}d$ is 1-1 b/c $[\vec{v}]_{\mathcal{B}} = \begin{pmatrix} 0 \\ \vdots \\ 0 \end{pmatrix} \Rightarrow \vec{v} = 0\vec{b}_1 + \dots + 0\vec{b}_n = \vec{0}$, and onto b/c any given $\begin{pmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{pmatrix}$ is the coord. vector of $\alpha_1\vec{b}_1 + \dots + \alpha_n\vec{b}_n$. \square

Ex / Let $\mathcal{B} = \{1, t, \dots, t^k\}$ be the basis of P_k .

$$\text{Then } [a_0 + a_1t + \dots + a_k t^k]_{\mathcal{B}} = \begin{pmatrix} a_0 \\ a_1 \\ \vdots \\ a_k \end{pmatrix}. //$$

Now let W be another vector space, of dimension m , with basis $\mathcal{C} = \{\vec{c}_1, \dots, \vec{c}_m\}$; and $T: V \rightarrow W$ a linear transformation ($T \in \mathcal{L}(V, W)$).

The (coordinate) matrix of T (w.r.t. \mathcal{B} & \mathcal{C}) is

$${}_{\mathcal{C}}[T]_{\mathcal{B}} := \begin{pmatrix} \uparrow & & \uparrow \\ [T\vec{b}_1]_{\mathcal{C}} & \dots & [T\vec{b}_n]_{\mathcal{C}} \\ \downarrow & & \downarrow \end{pmatrix} \in M_{m,n}$$

That is, if $T\vec{b}_j = \sum_{i=1}^m t_{ij} \vec{c}_i$ ($\forall j$) then t_{ij} is the (i,j) th matrix entry ($*i,j$).

Proposition 2: ${}_{\mathcal{C}}[\cdot]_{\mathcal{B}}: \mathcal{L}(V, W) \rightarrow M_{m,n}$

is an isomorphism of vector spaces.

Proof: (Linearity is left to you.) It is 1-1 b/c
 If $e [T]_{\mathcal{B}}$ is the zero matrix, $T \vec{b}_j = \sum 0 \vec{c}_i = \vec{0} (\forall j)$
 $\Rightarrow T$ is the zero transformation. It is onto b/c
 given the matrix (t_{ij}) , you can define a linear transfor-
 mation by $T \vec{b}_j := \sum_i t_{ij} \vec{c}_i$ (telling each basis vector where
 to go). \square

Ex / $V = \text{Span} \{ \overset{\vec{b}_1}{\cos(t)}, \overset{\vec{b}_2}{\sin(t)}, \overset{\vec{b}_3}{t \cos(t)}, \overset{\vec{b}_4}{t \sin(t)} \} \subset C^0(\mathbb{R})$
 $W = \mathbb{P}_4$ with basis $\mathcal{C} = \{ \overset{\vec{c}_1}{1}, \overset{\vec{c}_2}{t}, \overset{\vec{c}_3}{t^2}, \overset{\vec{c}_4}{t^3}, \overset{\vec{c}_5}{t^4} \}$.

$T: V \rightarrow W$ sends $f \mapsto \{ \text{degree-4 Taylor polynomial} \}$
 at 0 for f .

$$e [T]_{\mathcal{B}} = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ -1/2 & 0 & 0 & 0 & 0 \\ 0 & -1/6 & -1/2 & 0 & 0 \\ 1/24 & 0 & 0 & -1/6 & 0 \end{pmatrix}$$

Ex / $V = \mathbb{P}_3$, $T = \frac{d}{dt}: V \rightarrow V$, $\mathcal{B} = \{1, t, t^2, t^3\}$

$$[T]_{\mathcal{B}} := [T]_{\mathcal{B}} = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 3 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

\uparrow
 notation

Ex / When $V = \mathbb{R}^n$ & $W = \mathbb{R}^m$
 $B = \{\vec{e}_1, \dots, \vec{e}_n\}$ $C = \{\vec{e}_1, \dots, \vec{e}_m\}$ ← the standard basis is usually written \vec{e} in what follows.

and $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is defined by
 $\vec{x} \mapsto A\vec{x}$ for $A \in M_{m,n}$, we have

simply ${}_C[T]_B = A$. This is because the columns of A are precisely $A\vec{e}_1, \dots, A\vec{e}_n$ (written w.r.t. the standard basis).

Some basic examples: $A = \begin{pmatrix} 2 & -3 & 1 \\ 1 & 0 & -1 \end{pmatrix}$ gives

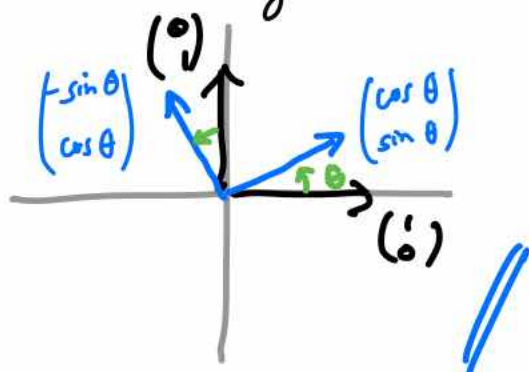
a transformation from $\mathbb{R}^3 \rightarrow \mathbb{R}^2$:

$$\begin{pmatrix} x \\ y \\ z \end{pmatrix} \mapsto \begin{pmatrix} 2 & -3 & 1 \\ 1 & 0 & -1 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 2x - 3y + z \\ x - z \end{pmatrix}.$$

$\begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ reflects about the x -axis
 $\begin{pmatrix} x \\ y \end{pmatrix} \mapsto \begin{pmatrix} x \\ -y \end{pmatrix}$

$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}: \mathbb{R}^3 \rightarrow \mathbb{R}^3$ projects to the xy -plane
 $\begin{pmatrix} x \\ y \\ z \end{pmatrix} \mapsto \begin{pmatrix} x \\ y \\ 0 \end{pmatrix}$

$R_\theta := \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ rotates counterclockwise by θ
 $\begin{pmatrix} 1 \\ 0 \end{pmatrix} \mapsto \begin{pmatrix} \cos \theta \\ \sin \theta \end{pmatrix}$
 $\begin{pmatrix} 0 \\ 1 \end{pmatrix} \mapsto \begin{pmatrix} -\sin \theta \\ \cos \theta \end{pmatrix}$



Properties:

$$\textcircled{1} \quad \underset{m \times n}{C} \begin{bmatrix} T \end{bmatrix} \underset{n \times 1}{B} \begin{bmatrix} \vec{v} \end{bmatrix} = \begin{bmatrix} T \vec{v} \end{bmatrix} \underset{m \times 1}{C} \quad (\text{in } \mathbb{R}^m)$$

Proof: If $\vec{v} = \sum_{j=1}^n v_j \vec{b}_j$, $T \vec{b}_j = \sum_{i=1}^m t_{ij} \vec{c}_i$, then

$$T \vec{v} = T \sum_j v_j \vec{b}_j = \sum_j v_j \underset{\substack{\uparrow \\ \text{linearity}}}{T \vec{b}_j} = \sum_j v_j \sum_i t_{ij} \vec{c}_i = \sum_{i=1}^m \left(\sum_{j=1}^n t_{ij} v_j \right) \vec{c}_i$$

$$\Rightarrow \text{ith entry of RHS} = \sum_i t_{ij} v_j = \text{ith entry of LHS. } \square$$

$\textcircled{2}$ Writing $A = \underset{B}{C} \begin{bmatrix} T \end{bmatrix}$, $\textcircled{1}$ has the 2 consequences:

\textcircled{a} The span of A's columns is $[\text{Im}(T)]_C$ \Rightarrow

TFAE —

- T is onto
- columns of A span \mathbb{R}^m
- $\text{rref}(A)$ has no rows of all 0's

\textcircled{b} Null space of A (solns of $A\vec{x} = \vec{0}$) is $[\text{Ker}(T)]_B$ \Rightarrow

TFAE —

- T is 1-1
- columns of A are independent
- $\text{rref}(A)$ has a leading 1 in every column.

If $\dim V = \dim W$, then: the 2 statements about $\text{ref}(A)$ are equivalent to $\text{ref}(A) = \mathbb{I}_n$, hence to invertibility of A (Lecture 5); and the 2 statements about T are equivalent (to each other and) to T being an isomorphism/invertible. So we get ...

- ③ $TFAE$ —
- T is an isomorphism/invertible
 - $\text{ref}(A) = \mathbb{I}_n$
 - A is invertible.

④ $\text{rank}(T) = \text{rank}(A)$

$\nwarrow \text{dim}(\text{Im}(T))$ $\nwarrow \# \text{ of pivot columns}$

Pf: The null space of A is the same as the null space of $\text{ref}(A)$, which is spanned by the solutions of $\text{ref}(A)\vec{x} = \vec{0}$ obtained by setting each free variable to 1 (all others to 0)

$$\Rightarrow \dim(\text{null space}(A)) = n - \# \text{ pivots} = n - \text{rank}(A).$$

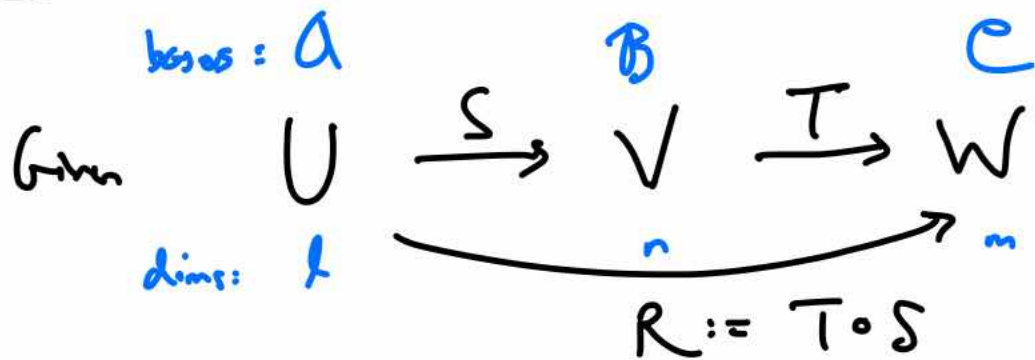
$$\overset{\parallel}{\dim(\ker(T))} \stackrel{R+N}{=} n - \text{rank}(T). \quad \square$$

Ex/ Suppose $e^{(T)}_{\mathcal{B}} = \begin{pmatrix} 1 & -4 & 8 & 1 \\ 0 & 2 & -1 & 3 \\ -1 & 0 & -6 & -2 \end{pmatrix}$. Is T 1-1? onto? //

Ex/ Suppose $e^{(T)}_{\mathcal{B}} = \begin{pmatrix} 3 & 4 & 7 & 4 \\ 0 & 1 & 4 & 6 \\ 0 & 0 & 2 & 8 \\ 0 & 0 & 0 & 1 \end{pmatrix}$. Is T invertible? //

How about $\begin{pmatrix} 3 & 3 & 1 \\ 3 & 3 & 1 \\ 1 & 1 & 1 \end{pmatrix}$? (If not, what is its rank?) //

⑤ Compositors of linear transformations



I claim that

matrix multiplication

$$(*) \quad \underset{m \times n}{[T]_C} \cdot \underset{n \times l}{[S]_A} = \underset{m \times l}{[R]_A} a.$$

Proof: Start with $S \vec{a}_k = \sum_j s_{jk} \vec{b}_j$, $T \vec{b}_j = \sum_i t_{ij} \vec{c}_i$,

$$\begin{aligned}
 \circ \sum_i r_{ik} \vec{c}_i &= R \vec{a}_k = T(S \vec{a}_k) = T\left(\sum_j s_{jk} \vec{b}_j\right) \\
 &= \sum_j s_{jk} T \vec{b}_j = \sum_j s_{jk} \sum_i t_{ij} \vec{c}_i \\
 &= \sum_i \left(\sum_j t_{ij} s_{jk}\right) \vec{c}_i.
 \end{aligned}$$

Since the \vec{c}_i are l.i., the coeffs. must agree:

$$\sum_j t_{ij} s_{jk} = r_{ik} \quad \text{for all } i, k.$$

But this says the LHS & RHS of (*) matrix entries agree. □

Here are two consequences:

Ex / $R_\theta = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix} = [T_\theta]_{\mathcal{E}}$ where

$T_\theta: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ is rotation by θ . Clearly

$T_{\theta+\phi} = T_\theta \circ T_\phi$. Taking matrices, we get

$$R_{\theta+\phi} = [T_{\theta+\phi}]_{\mathcal{E}} = [T_\theta]_{\mathcal{E}} \cdot [T_\phi]_{\mathcal{E}} = R_\theta R_\phi.$$

This recovers the angle addition formulas! —

$$\begin{pmatrix} \cos(\theta+\phi) & -\sin(\theta+\phi) \\ \sin(\theta+\phi) & \cos(\theta+\phi) \end{pmatrix} = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix} \begin{pmatrix} \cos \phi & -\sin \phi \\ \sin \phi & \cos \phi \end{pmatrix} \\ = \left(\begin{array}{cc|cc} \cos \theta \cos \phi - \sin \theta \sin \phi & -\sin \theta \cos \phi - \cos \theta \sin \phi \\ \sin \theta \cos \phi + \cos \theta \sin \phi & \cos \theta \cos \phi - \sin \theta \sin \phi \end{array} \right)$$

We also get $R_\theta^n = R_{n\theta}$, with no effort! //

Ex / If T is invertible and $A = {}_C [T]_B$,

$$\text{then } {}_B [T^{-1}]_C [T]_B = {}_B [T^{-1} \circ T]_B = [Id_V]_B = I_n$$

$\Rightarrow {}_B [T^{-1}]_C = A^{-1}$; that is, the inverse matrix

is the matrix of the inverse transformation. //