

EXAMPLES:

(1) $V \xrightarrow{T = \frac{d}{dx}} V$ had $\text{null}(T) + \text{rank}(T) = \text{dim}(V)$
 $\{ax^2+bx+c: a,b,c \in \mathbb{R}\}$ \uparrow $+$ \uparrow $=$ \uparrow
 1 $+$ 2 $=$ 3

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(2) Sums and direct sums (part of § 3.6)
 $V_1 + V_2$ $V_1 \oplus V_2$

Given two vector spaces V_1, V_2 their direct sum $V_1 \oplus V_2 = V_1 \times V_2$ with componentwise + and scaling

is another F -vector space, with $\text{dim}(V_1 \oplus V_2) = \text{dim}(V_1) + \text{dim}(V_2)$

because one can get a basis for $V_1 \oplus V_2$ by taking

~~the union of~~ a basis $\{a_1, a_2, \dots, a_{\text{dim}(V_1)}\}$ for V_1
 and a basis $\{b_1, b_2, \dots, b_{\text{dim}(V_2)}\}$ for V_2

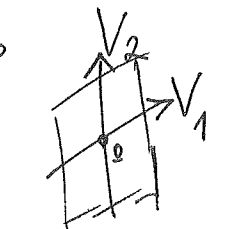
and concatenating $\{(a_1, 0), \dots, (a_{\text{dim}(V_1)}, 0), (0, b_1), \dots, (0, b_{\text{dim}(V_2)})\}$
 to get a basis for $V_1 \oplus V_2$

On the other hand, given two subspaces $V_1, V_2 \subset V$

one can form a new subspace $V_1 + V_2 := \{v_1 + v_2 : v_1 \in V_1, v_2 \in V_2\} \subset V$
 called their sum.

Q: What is $\text{dim}(V_1 + V_2)$? It depends on how V_1, V_2 intersect $\text{dim}(V_1), \text{dim}(V_2)$ and...

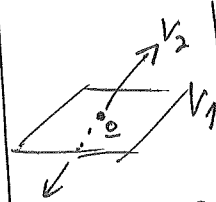
e.g. $V = \mathbb{R}^3$



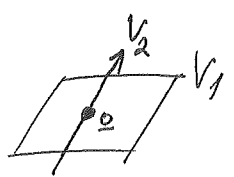
$V_1 + V_2 = \text{a plane}$
 $\text{dim}(V_1 + V_2) = 2$



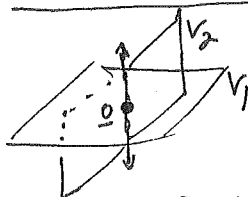
$V_1 + V_2 = V_1 = V_2$
 $\text{dim}(V_1 + V_2) = 1$



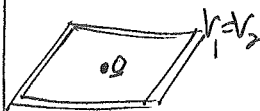
$V_1 + V_2 = V = \mathbb{R}^3$
 $\text{dim}(V_1 + V_2) = 3$



$V_1 + V_2 = V_1$
 $\text{dim}(V_1 + V_2) = 2$



$V_1 + V_2 = \mathbb{R}^3 = V$
 $\text{dim}(V_1 + V_2) = 3$



$V_1 + V_2 = V_1 = V_2$
 $\text{dim}(V_1 + V_2) = 2$

(139)

PROPOSITION: (3.6.6) Given V finite-dimensional and V_1, V_2 subspaces in V ,

$$\dim(V_1 + V_2) + \dim(V_1 \cap V_2) = \dim V_1 + \dim V_2$$

e.g. check this in all of the previous examples

proof: There is always a linear transformation

$$V_1 \oplus V_2 \xrightarrow{T} V_1 + V_2 (\subset V)$$

$$(v_1, v_2) \mapsto T((v_1, v_2)) = v_1 + v_2$$

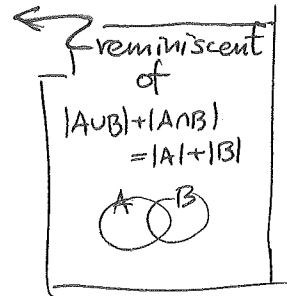
which is clearly surjective,

$$\text{having } \ker(T) = \{ (v_1, v_2) : v_1 + v_2 = 0, \text{ i.e. } v_2 = -v_1 \}$$

$$= \{ (v, -v) : v \in V_1 \cap V_2 \}, \text{ which is isomorphic to } V_1 \cap V_2$$

$$\text{via the map } \{ (v, -v) : v \in V_1 \cap V_2 \} \xrightarrow{\quad} V_1 \cap V_2$$

$$\begin{array}{ccc} (v, -v) & \mapsto & v \\ (u, -u) & \mapsto & u \end{array}$$



Hence the dimension formula gives

$$\text{rank}(T) + \text{nullity}(T) = \dim(V_1 \oplus V_2)$$

$$\begin{array}{ccc} \text{"} & \text{"} & \\ \dim \text{im}(T) & \dim \ker(T) & \parallel \\ \text{"} & \text{"} & \end{array}$$

$$\dim(V_1 + V_2) + \dim(V_1 \cap V_2) = \dim(V_1) + \dim(V_2) \quad \square$$

COROLLARY: In the setting of the Proposition,

$$\begin{array}{ccc} V_1 \oplus V_2 \rightarrow V_{\text{~~is~~} } & \text{is an isomorphism} & \Leftrightarrow \bullet V_1 + V_2 = V \\ (v_1, v_2) \mapsto v_1 + v_2 & & \bullet V_1 \cap V_2 = \{0\} \end{array}$$

(reminiscent of $H, K < G$ have $H \times K \xrightarrow{\mu} G$ an isomorphism of groups
 $(h, k) \mapsto hk$)

- \Leftrightarrow
- $HK = G$
 - $H \cap K = \{1\}$
 - H, K commute

(146)
 § 4.2 Matrices & linear transformations

This is often the right way to think about a matrix.

LEMMA (4.2.1): Every matrix $A \in F^{m \times n}$ gives a linear transformation

$$F^n \xrightarrow{T} F^m$$

$$x \mapsto T(x) = Ax$$

and conversely, every linear transformation $F^n \xrightarrow{T} F^m$

arises this way as $T(x) = Ax$ for $A = \begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{m1} & \dots & a_{mn} \end{bmatrix}$ where j th column

i.e. $A = \begin{bmatrix} | & & | \\ T(v_1) & \dots & T(v_n) \\ | & & | \end{bmatrix}$ where $\begin{bmatrix} a_{1j} \\ a_{2j} \\ \vdots \\ a_{mj} \end{bmatrix} = T(e_j)$

$$= T\left(\begin{bmatrix} 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{bmatrix}\right)$$

$$= \sum_{i=1}^m a_{ij} e_i$$

proof: Given $A \in F^{m \times n}$, certainly $T(x) = Ax$ is linear, since

$$T(cx + dy) = A(cx + dy) = cAx + dAy = cT(x) + dT(y)$$

Conversely, given $F^n \xrightarrow{T} F^m$ linear, if we define A as above

via $T(e_j) = \sum_{i=1}^m a_{ij} e_i$, then $T(x) = T\left(\sum_{j=1}^n x_j e_j\right) = \sum_{j=1}^n x_j T(e_j)$

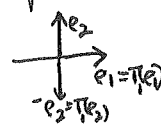
$$= \sum_{j=1}^n x_j \sum_{i=1}^m a_{ij} e_i$$

$$= \sum_{i=1}^m \left(\sum_{j=1}^n a_{ij} x_j\right) e_i = Ax$$

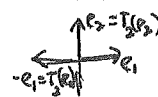
EXAMPLES:

① What are the matrices corresponding to these linear maps $\mathbb{R}^2 \xrightarrow{T_i} \mathbb{R}^2$?

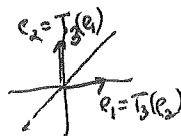
$T_1 =$ reflection across x -axis $\mapsto A_1 = \begin{matrix} e_1 & e_2 \\ e_1 & \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} \end{matrix}$



$T_2 =$ reflection across y -axis $\mapsto A_2 = \begin{matrix} e_1 & e_2 \\ e_1 & \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix} \end{matrix}$



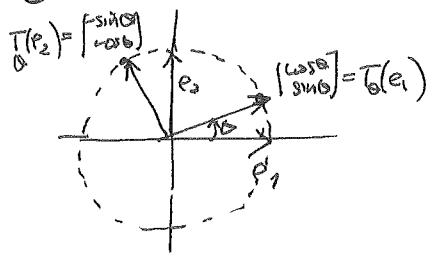
$T_3 =$ reflection across $x=y$ line $\mapsto A_3 = \begin{matrix} e_1 & e_2 \\ e_1 & \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \end{matrix}$



(141)

(2) The map $\mathbb{R}^2 \xrightarrow{T_\theta} \mathbb{R}^2$ that rotates \odot counterclockwise about \odot

has matrix $A_\theta = \begin{matrix} e_1 & e_2 \\ e_1 & \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \\ e_2 & \end{matrix}$



and since $T_{\theta_1} \circ T_{\theta_2} = T_{\theta_1 + \theta_2}$

$$\mathbb{R}^2 \xrightarrow{T_{\theta_2}} \mathbb{R}^2 \xrightarrow{T_{\theta_1}} \mathbb{R}^2 \Rightarrow \mathbb{R}^2 \xrightarrow{A_{\theta_2}} \mathbb{R}^2 \xrightarrow{A_{\theta_1}} \mathbb{R}^2$$

$T_{\theta_1} \circ T_{\theta_2} = T_{\theta_1 + \theta_2}$ $A_{\theta_1} A_{\theta_2} = A_{\theta_1 + \theta_2}$

$$\begin{bmatrix} \cos \theta_1 & -\sin \theta_1 \\ \sin \theta_1 & \cos \theta_1 \end{bmatrix} \begin{bmatrix} \cos \theta_2 & -\sin \theta_2 \\ \sin \theta_2 & \cos \theta_2 \end{bmatrix} = \begin{bmatrix} \cos(\theta_1 + \theta_2) & -\sin(\theta_1 + \theta_2) \\ \sin(\theta_1 + \theta_2) & \cos(\theta_1 + \theta_2) \end{bmatrix}$$

$$\begin{bmatrix} \cos \theta_1 \cos \theta_2 - \sin \theta_1 \sin \theta_2 & \\ \sin \theta_1 \cos \theta_2 + \cos \theta_1 \sin \theta_2 & \end{bmatrix}$$

equality gives cos, sin addition formulas!

What about $V \xrightarrow{T} W$ when V, W aren't F^n, F^m ?

If they're finite dimensional, pick bases for each of V & W and then one can write down a matrix for T .

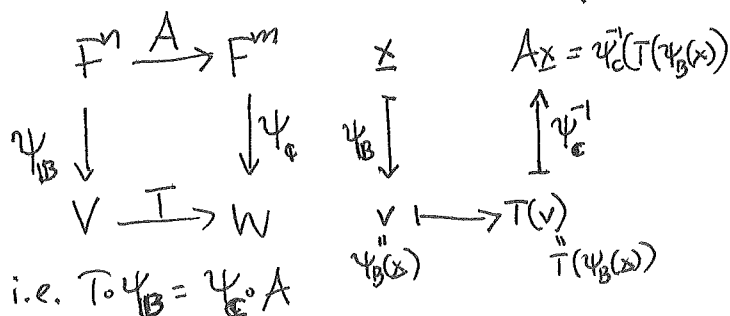
PROPOSITION: (4.2.5)

Once you've picked a basis $B = (v_1, \dots, v_n)$ for V

and a basis $C = (w_1, \dots, w_m)$ for W

then the matrix $A = (a_{ij}) \in F^{m \times n}$

defined by $T(v_j) = \sum_{i=1}^m a_{ij} w_i$ makes this diagram commute:



i.e. $T \circ \psi_B = \psi_C \circ A$

so $\psi_C^{-1} \circ T \circ \psi_B = A$

One calls A the matrix expressing T in the bases B for V and C for W