

Indeed linear transformations can be composed. So here's a natural question: ①

Given  $l_1, l_2$  linear transformations with assoc. matrices  $A_1, A_2$

And  $l_1 \circ l_2$  with associated matrix  $B$ , can we give matrix operation

\* so that  $A_1 * A_2 = B$ ? i.e. what operation on matrices corresponds to composition?

Answer, of course, is matrix multiplication and explains our funny definition of matrix mult. in the first place.

But let's prove it carefully:  $l_2: \mathbb{R}^n \rightarrow \mathbb{R}^m$   $l_1: \mathbb{R}^m \rightarrow \mathbb{R}^k$

Part 1: If  $l_1, l_2$  linear, then  $l_1 \circ l_2$  linear

$$\begin{aligned} l_1 \circ l_2 (c_1 \vec{u} + c_2 \vec{v}) &= l_1 (c_1 l_2(\vec{u}) + c_2 l_2(\vec{v})) \text{ since } l_2 \text{ lin.} \\ &= c_1 l_1 \circ l_2(\vec{u}) + c_2 l_1 \circ l_2(\vec{v}) \text{ since } l_1 \text{ lin.} \end{aligned}$$

part 2: To find matrix of  $l_1 \circ l_2$ , evaluate standard basis vectors  $\vec{e}_i$ .

$$l_1 \circ l_2(\vec{e}_i) = l_1(\text{i}^{\text{th}} \text{ column vector of } A_2)$$

$$\begin{bmatrix} a_{1i}^{(2)} \\ \vdots \\ a_{mi}^{(2)} \end{bmatrix}$$

Now it gets a little messy: Write this as

$$(*) \quad a_{1i}^{(2)} \vec{e}_1 + \dots + a_{mi}^{(2)} \vec{e}_m$$

and note  $l_1(\vec{e}_j) = j^{\text{th}}$  column of

$$\text{so } (*) = \begin{bmatrix} a_{1i}^{(1)} a_{1i}^{(2)} + \dots + a_{1m}^{(1)} a_{mi}^{(2)} \\ \vdots \\ a_{ki}^{(1)} a_{mi}^{(2)} \end{bmatrix} = \begin{bmatrix} a_{1i}^{(1)} \\ \vdots \\ a_{ki}^{(1)} \end{bmatrix} A_1$$

Now that we know all linear transformations  $l: \mathbb{R}^n \rightarrow \mathbb{R}^m$  are associated with  $m \times n$  matrix, ask questions about matrix mult. and answer with linear trans. knowledge.

Q1: Is matrix multiplication associative?   
 Yes.   
 Proof 1: check it directly from defin

Proof 2: composition of functions is associative.

Q2: Is matrix multiplication commutative?

Not in general. Find two linear transformations that don't commute.

Do example of reflection  $\circ$  rotation.

But some matrices do commute -

think of examples - e.g. pair of rotations.

Q3: When does a linear transformation have an inverse?

Given  $l$ , seek <sup>linear</sup> transformation  $l'$  s.t.  $l \circ l' = Id_{\mathbb{R}^n \rightarrow \mathbb{R}^n}$   
 $l: \mathbb{R}^n \rightarrow \mathbb{R}^m$        $l': \mathbb{R}^m \rightarrow \mathbb{R}^n$        $l' \circ l = Id_{\mathbb{R}^m \rightarrow \mathbb{R}^m}$

Non-examples: What about projection to a line?

No, since function is not onto and not one-one.

often these conditions easier to check in practice

In fact, Hubbard and Hubbard define a linear transformation to be

invertible if one-one and onto  $\sim$  hit every point in  $\mathbb{R}^m$   
unique image for each point in  $\mathbb{R}^n$

(this is equivalent to definition above)  
as these conditions ensure well-defined inverse

You might be thinking that  $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$  can only be one-one, onto

if  $m=n$ . i.e. if  $m>n$ , can't map to all points.

e.g.  $f: \mathbb{R} \rightarrow \mathbb{R}^2$

or if  $f: \mathbb{R}^2 \rightarrow \mathbb{R}$  then "not enough room". Some points in  $\mathbb{R}^2$  must go to same place.

Sneaky point: True that only linear functions  $l: \mathbb{R}^n \rightarrow \mathbb{R}^m$  that are invertible

have  $m=n$ . However, there are onto functions  $f: \mathbb{R} \rightarrow \mathbb{R}^2$

"Peano curves", but the  $f$  is not linear (nor continuous)

Proposition: (1.3.14 in Hubbard and Hubbard) Let  $l$  be linear transformation  $\mathbb{R}^n \rightarrow \mathbb{R}^m$  with matrix  $A$

$l$  is invertible if and only if

$$\exists B \text{ s.t. } A \cdot B = B \cdot A = I$$

identity transformation as matrix:  $\begin{pmatrix} 1 & & 0 \\ & \ddots & \\ 0 & & 1 \end{pmatrix}$

better to write.

$$A \cdot B = I_m, B \cdot A = I_n$$

we say  $B$  is a "matrix inverse" for  $A$ , write  $B = A^{-1}$

pf: Suppose  $A$  invertible with inverse  $B$ . Show that linear map assoc.

to  $B$ , call it  $l'$ , is the inverse of  $l$ .

Given any  $\vec{y} \in \mathbb{R}^m$  then  $\vec{y} = I_m \cdot \vec{y} = A \cdot B \cdot \vec{y} = l'_A \cdot l_B \cdot \vec{y}$  (Shows  $l$  onto)

Given any  $\vec{x} \in \mathbb{R}^n$  then  $\vec{x} = I_n \cdot \vec{x} = B \cdot A \cdot \vec{x} = l' \cdot l \vec{x}$  (Shows  $l$  one-one)

Other direction: If  $l$  invertible with inverse  $l'$  and assoc. matrix  $B$ , (4)

then since  $l \circ l' = \text{Id}_{\mathbb{R}^n \rightarrow \mathbb{R}^n}$ , then  $A \cdot B = I_n$

Similarly  $l' \circ l = \text{Id}_{\mathbb{R}^m \rightarrow \mathbb{R}^m}$  since  $A \cdot B$  is matrix assoc. to  $l \circ l'$

$$\Rightarrow B \cdot A = I_m.$$

One problem: Haven't showed  $l'$  linear yet, so can't assume  $l'$  is assoc. to a matrix  $B$ .

i.e. show if  $l$  linear with inverse  $l'$ , then

$$\text{for any } \vec{y}_1, \vec{y}_2 \in \mathbb{R}^m, c_1, c_2 \in \mathbb{R} : \quad l'(c_1 \vec{y}_1 + c_2 \vec{y}_2) \\ = c_1 l'(\vec{y}_1) + c_2 l'(\vec{y}_2).$$

$$\begin{aligned} \text{pf: } l \circ l'(c_1 \vec{y}_1 + c_2 \vec{y}_2) &= c_1 \vec{y}_1 + c_2 \vec{y}_2 \\ &= c_1 l \circ l'(\vec{y}_1) + c_2 l \circ l'(\vec{y}_2) \\ &= l(c_1 l'(\vec{y}_1) + c_2 l'(\vec{y}_2)) \end{aligned}$$

↑  
since  $l$  linear

Now use that  $l$  one-one, so if  $l(\vec{v}) = l(\vec{u})$  then  $\vec{v} = \vec{u}$ .

i.e. in our case  $l'(c_1 \vec{y}_1 + c_2 \vec{y}_2) = c_1 l'(\vec{y}_1) + c_2 l'(\vec{y}_2)$  as desired.