

# Common mistakes on midterm

• REF  $\rightsquigarrow$   $\begin{pmatrix} * & * & * & * & * & * \\ * & * & * & * & * & * \\ 0 & 0 & 0 & 0 & \boxed{a} & * & * & * & * & * \\ 0 & 0 & 0 & 0 & 0 & * & * & * & * & * \\ 0 & 0 & 0 & 0 & 0 & * & * & * & * & * \end{pmatrix}$

any non-zero pivot

• RREF  $\rightsquigarrow$   $\begin{pmatrix} 0 & * & * & * & * & * \\ 0 & * & * & * & * & * \\ 0 & * & * & * & * & * \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & \boxed{1} & * & * & * & * & * \\ 0 & 0 & 0 & 0 & 0 & 0 & * & * & * & * & * \end{pmatrix}$

say this is not a pivot column

$\rightsquigarrow$  when applying Gaussian elim to  $[A | b]$  perform the same row operations to the  $b$  column as to the  $A$  matrix

• Projections:  $P_V = A(A^T A)^{-1} A^T$   
always invertible as long as  
the columns of  $A$  are independent

• vector spaces can be perpendicular to each other  
(also vectors)

But it doesn't make sense to say that  
matrices are perpendicular to each other

- vectors & vector spaces are geometric objects
- matrices are just representations of geometric objects, but they do not have intrinsic geometry

$$V = \left\{ \begin{bmatrix} x \\ y \end{bmatrix} \text{ s.t. } 3x + 4y = 0 \right\}$$

write  $V = C(A)$ ; what is  $A$ ?

↪  $2 \times 1$  matrix

$$V \neq C\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) \text{ doesn't make mathematical sense}$$

$$V = C\left(\begin{bmatrix} 4 \\ -3 \end{bmatrix}\right) \text{ b/c } (x, y) = (4, -3) \text{ satisfies equation}$$

Today: linear transformations

$$\mathbb{R}^n = \left\{ \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \text{ where } x_1, \dots, x_n \in \mathbb{R} \right\}$$

Def: a linear transformation is a function

$$\phi: \mathbb{R}^n \rightarrow \mathbb{R}^m$$

such that

$$\bullet \phi(v + v') = \phi(v) + \phi(v')$$

$$\bullet \phi(c \cdot v) = c \cdot \phi(v)$$

for any  $v, v' \in \mathbb{R}^n$

for any scalar  $c$

} reminiscent  
of the defining  
properties of vector spaces

Linear transformations send objects (lines, triangles, ellipses...) to objects of the same nature

Examples:

$$\phi: \mathbb{R}^2 \rightarrow \mathbb{R}^2$$

- reflections  
(mirror images)
- dilation/contractions
- rotations
- shearing



Matrices represent linear transformations

$$\phi: \mathbb{R}^n \rightarrow \mathbb{R}^m$$

take the basis  $e_1, \dots, e_n$  of  $\mathbb{R}^n$   
 $e_1, \dots, e_m$  of  $\mathbb{R}^m$

$$e_i = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \rightarrow \text{i-th spot}$$

$$\phi(e_j) = \text{some vector in } \mathbb{R}^m = \begin{bmatrix} a_{1j} \\ a_{2j} \\ \vdots \\ a_{mj} \end{bmatrix}$$

$$= a_{1j} e_1 + a_{2j} e_2 + \dots + a_{mj} e_m$$

for all  $j \in \{1, \dots, n\}$

$\leftarrow$  j-th column of A

$\rightsquigarrow$  determines a matrix

$$A = \begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{m1} & \dots & a_{mn} \end{bmatrix}$$

linear transformation  
 what is the connection between  $\phi$  and  $A$ ?  $m \times n$  matrix

$$\phi(e_j) = a_{1j}e_1 + \dots + a_{mj}e_j$$

what about  $\phi(v)$  for any  $v \in \mathbb{R}^n$

|| b/c  $\phi$  is linear

$$\phi(v_1e_1) + \phi(v_2e_2) + \dots + \phi(v_ne_n)$$

|| b/c  $\phi$  is linear

$$v_1\phi(e_1) + v_2\phi(e_2) + \dots + v_n\phi(e_n)$$

$$\sum_{j=1}^n v_j \phi(e_j) = \sum_{j=1}^n v_j \sum_{i=1}^m a_{ij} e_i = \sum_{i=1}^m \left( \sum_{j=1}^n a_{ij} v_j \right) e_i =$$

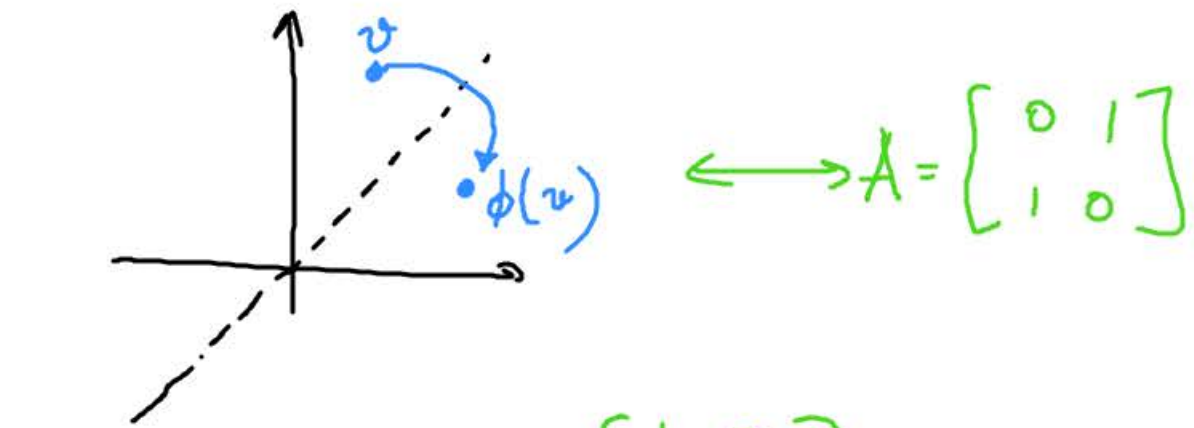
$$\begin{bmatrix} a_{11}v_1 + \dots + a_{1n}v_n \\ a_{21}v_1 + \dots + a_{2n}v_n \\ \vdots \\ a_{m1}v_1 + \dots + a_{mn}v_n \end{bmatrix} = Av$$

$$v_1e_1 + v_2e_2 + \dots + v_ne_n = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix}$$

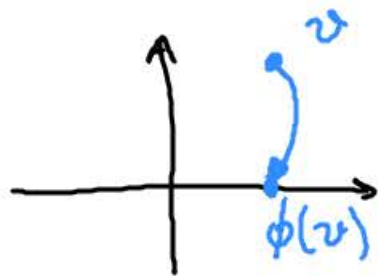
Upshot: any linear transformation  $\phi: \mathbb{R}^n \rightarrow \mathbb{R}^m$   
 is of the form  $\phi(v) = Av$   
 for a uniquely determined  $m \times n$  matrix  $A$  the matrix associated to  $\phi$

Examples:  $\phi: \mathbb{R}^2 \rightarrow \mathbb{R}^2$  is

- reflection in the line  $\{x=y\} \subset \mathbb{R}^2$



- projection onto x axis



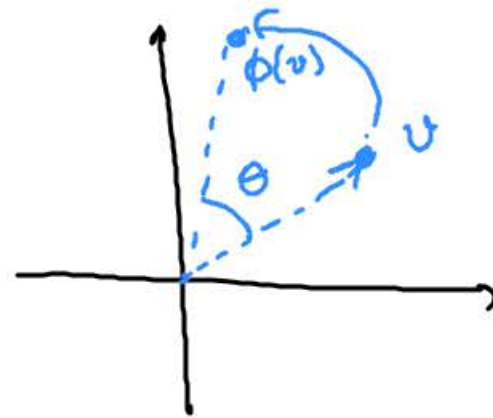
$A = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$

- dilation by  $\lambda$  in the x direction and by  $\mu$  in the y direction

$A = \begin{bmatrix} \lambda & 0 \\ 0 & \mu \end{bmatrix}$



- rotation by angle  $\theta$  around the origin of  $\mathbb{R}^2$



$$\leftrightarrow A = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

- projection onto a subspace  $V \subset \mathbb{R}^n$  is a linear transformation

$$\phi(v) = \text{proj}_V v \\ \parallel \\ A(A^T A)^{-1} A^T v$$

$$\leftrightarrow A(A^T A)^{-1} A^T$$

where  $V = C(A)$

Theorem: composition of linear transformations corresponds to matrix multiplication

$$\phi: \mathbb{R}^n \rightarrow \mathbb{R}^m \rightsquigarrow A^{m \times n}$$

$$\psi: \mathbb{R}^p \rightarrow \mathbb{R}^n \rightsquigarrow B^{n \times p}$$

$$\Rightarrow \phi \circ \psi: \mathbb{R}^p \rightarrow \mathbb{R}^m$$

$$\rightsquigarrow AB^{m \times p}$$

$$\mathbb{R}^p \xrightarrow{\psi} \mathbb{R}^n \xrightarrow{\phi} \mathbb{R}^m$$

This is why we only multiply an  $m \times n$  matrix with an  $n \times p$  matrix