

**Final**  
**Linear Algebra Practice Midterm 2**  
**Spring 2019**

1. (10 points) Let

$$A = \begin{pmatrix} 1 & 3 & 0 \\ 0 & -1 & 2 \end{pmatrix}$$

Compute  $AA^T$  and determine if it is diagonalizable. Justify your answer either with a clear demonstration or with reference to a relevant theorem.

$$\begin{aligned} AA^T &= \begin{pmatrix} 1 & 3 & 0 \\ 0 & -1 & 2 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 3 & -1 \\ 0 & 2 \end{pmatrix} \\ &= \begin{pmatrix} 10 & -3 \\ -3 & 5 \end{pmatrix} \end{aligned}$$

Since  $AA^T$  is symmetric, we already know it is diagonalizable with real eigenvalues, but to see it directly, we compute:

$$\begin{aligned} c(\lambda) &= \det(AA^T - \lambda I) \\ &= \det \begin{pmatrix} 10 - \lambda & -3 \\ -3 & 5 - \lambda \end{pmatrix} \\ &= (10 - \lambda)(5 - \lambda) - 9 \\ &= \lambda^2 - 15\lambda + 41 \\ &= 0 \end{aligned}$$

distinct  
eigenvalues  
 $\Rightarrow$   
diagonalizable

$$\begin{aligned} \Rightarrow \lambda &= \frac{15 \pm \sqrt{15^2 - 4 \cdot 1 \cdot 41}}{2} = \frac{15 \pm \sqrt{225 - 164}}{2} \\ &= \frac{15 \pm \sqrt{61}}{2} \end{aligned}$$

2. Using the same matrix  $A$  as in problem (1),

$$A = \begin{pmatrix} 1 & 3 & 0 \\ 0 & -1 & 2 \end{pmatrix}$$

answer the following questions:

(a) (5 points) What is  $\det(A^T A)$ ?

$$\begin{aligned} \det(A^T A) &= \det \left( \begin{pmatrix} 1 & 0 \\ 3 & -1 \\ 0 & 2 \end{pmatrix} \begin{pmatrix} 1 & 3 & 0 \\ 0 & -1 & 2 \end{pmatrix} \right) \\ &= \det \begin{pmatrix} 1 & 3 & 0 \\ 3 & 10 & -2 \\ 0 & -2 & 4 \end{pmatrix} \quad \left. \begin{array}{l} \\ \end{array} \right\} \text{add } -3I \text{ to II} \\ &= \det \begin{pmatrix} 1 & 3 & 0 \\ 0 & 1 & -2 \\ 0 & -2 & 4 \end{pmatrix} \quad \left. \begin{array}{l} \\ \end{array} \right\} \text{add } 2II \text{ to III} \\ &= \det \begin{pmatrix} 1 & 3 & 0 \\ 0 & 1 & -2 \\ 0 & 0 & 0 \end{pmatrix} \\ &= 1 \cdot 1 \cdot 0 = \boxed{0} \end{aligned}$$

(b) (5 points) Use the above result to determine if the columns of  $A^T A$  are linearly independent.

No, not linearly independent,  
because  $A \in GL(n, \mathbb{R}) \iff A$ 's  
columns are lin. indep  $\iff \det A \neq 0$ .  
(This is a Theorem.)

3. Consider the matrices  $A = \begin{pmatrix} 2 & 1 \\ 0 & 2 \end{pmatrix}$  and  $B = \begin{pmatrix} 2 & 0 \\ 0 & 2 \end{pmatrix}$ .

(a) (5 points) Find the eigenvalues of each matrix.

$\lambda = 2$  for both:  $A$  &  $B$  are upper triangular, so the  $\lambda$ 's sit along their diagonal

(b) (5 points) Determine the geometric multiplicities of each eigenvalue, for each matrix. Which of  $A$  and  $B$  is diagonalizable?

$A$  has geometric mult. of  $\lambda = 1 < 2$   $\swarrow$   
 alg. mult.  $\left\{ \begin{array}{l} A - 2I = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \text{ has 1 pivot col., therefore} \\ \dim E_\lambda = \dim N(A - 2I) = 1 \end{array} \right.$

$B$  has geom. mult.  $= 2$   $\left\{ \begin{array}{l} B - 2I = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$ ; the zero matrix, whose null space is all of  $\mathbb{R}^2$ , therefore  $\dim E_\lambda = 2$

Therefore,  $A$  is not,  $B$  is diagonalizable.

- (c) (5 points) Determine whether the two matrices are similar, i.e. whether  $B = PAP^{-1}$  for some  $P \in GL(2, \mathbb{R})$ .

No, because:

$B$  is diagonalizable (it's already diagonal!), whereas  $A$  is not.

That is,  $B \sim D$  (similar to)  
 $\uparrow$  diag. (in fact  $B=D$ )

but  $A \not\sim D$ ,

whereas if  $A \sim B$ , then

$$A \sim B \sim D \Rightarrow A \sim D$$

if we know  
this isn't poss.

4. Consider the vector  $\mathbf{v} = \begin{pmatrix} 5 \\ 2 \end{pmatrix}$  in  $\mathbb{R}^2$ .

(a) (5 points) Determine  $[\mathbf{v}]_\beta$  if  $\beta = \left( \begin{pmatrix} 1 \\ 2 \end{pmatrix}, \begin{pmatrix} 3 \\ 4 \end{pmatrix} \right)$ .

$$M_{\beta\sigma} = \begin{pmatrix} 1 & 3 \\ 2 & 4 \end{pmatrix}, \quad M_{\sigma\beta} = M_{\beta\sigma}^{-1} = \begin{pmatrix} 1 & 3 \\ 2 & 4 \end{pmatrix}^{-1}$$

$$= -\frac{1}{2} \begin{pmatrix} 4 & -3 \\ -2 & 1 \end{pmatrix}$$

$$= \frac{1}{2} \begin{pmatrix} -4 & 3 \\ 2 & -1 \end{pmatrix}$$

$$\Rightarrow [\vec{v}]_\beta = M_{\sigma\beta} \vec{v} = \frac{1}{2} \begin{pmatrix} -4 & 3 \\ 2 & -1 \end{pmatrix} \begin{pmatrix} 5 \\ 2 \end{pmatrix}$$

$$= \frac{1}{2} \begin{pmatrix} -20 + 6 \\ 10 - 2 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} -14 \\ 8 \end{pmatrix}$$

$$= \boxed{\begin{pmatrix} -7 \\ 4 \end{pmatrix}}$$

(b) (5 points) Find the standard representation of the projection  $\pi_{\mathbf{w}}$  onto the vector  $\mathbf{w} = R_{\pi/2} \mathbf{v}$  orthogonal to  $\mathbf{v}$ .

$$R_{\pi/2} \vec{v} = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 5 \\ 2 \end{pmatrix} = \begin{pmatrix} -2 \\ 5 \end{pmatrix} = \vec{w}$$

$$\Rightarrow \pi_{\vec{w}}(\vec{e}_1) = \frac{\vec{w} \cdot \vec{e}_1}{\|\vec{w}\|^2} \vec{w} = \frac{\begin{pmatrix} -2 \\ 5 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 0 \end{pmatrix}}{(-2)^2 + 5^2} \begin{pmatrix} -2 \\ 5 \end{pmatrix}$$

$$= \frac{-2}{29} \begin{pmatrix} -2 \\ 5 \end{pmatrix}$$

$$\pi_{\vec{w}}(\vec{e}_2) = \frac{\vec{w} \cdot \vec{e}_2}{\|\vec{w}\|^2} \vec{w} = \frac{\begin{pmatrix} -2 \\ 5 \end{pmatrix} \cdot \begin{pmatrix} 0 \\ 1 \end{pmatrix}}{(-2)^2 + 5^2} \begin{pmatrix} -2 \\ 5 \end{pmatrix}$$

$$= \frac{5}{29} \begin{pmatrix} -2 \\ 5 \end{pmatrix}$$

$$\Rightarrow [\pi_{\vec{w}}]_\sigma = \frac{1}{29} \begin{pmatrix} -2 \begin{pmatrix} -2 \\ 5 \end{pmatrix} & 5 \begin{pmatrix} -2 \\ 5 \end{pmatrix} \end{pmatrix} = \boxed{\frac{1}{29} \begin{pmatrix} 4 & -10 \\ -10 & 25 \end{pmatrix}}$$

- (c) (5 points) Find the standard representation of the reflection  $R_\ell$  across  $\ell = \text{span}(\mathbf{v})$ .

Since  $R_\ell(\vec{v}) = \vec{v}$ , where  $\vec{v} = \begin{pmatrix} 5 \\ 2 \end{pmatrix}$

and  $R_\ell(\vec{w}) = -\vec{w}$ ,  $\vec{w} = \begin{pmatrix} -2 \\ 5 \end{pmatrix}$ ,

letting  $\beta = (\vec{v}, \vec{w}) = \left( \begin{pmatrix} 5 \\ 2 \end{pmatrix}, \begin{pmatrix} -2 \\ 5 \end{pmatrix} \right)$ ,

$$[R_\ell]_\beta = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

and therefore

$$\begin{aligned} [R_\ell]_\sigma &= P_{\beta\sigma} [R_\ell]_\beta P_{\beta\sigma}^{-1} \\ &= \begin{pmatrix} 5 & -2 \\ 2 & 5 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} 5 & -2 \\ 2 & 5 \end{pmatrix}^{-1} \\ &= \begin{pmatrix} 5 & 2 \\ 2 & -5 \end{pmatrix} \left( \frac{1}{29} \begin{pmatrix} 5 & 2 \\ -2 & 5 \end{pmatrix} \right) \end{aligned}$$

$$= \frac{1}{29} \begin{pmatrix} 21 & 20 \\ 20 & -21 \end{pmatrix}$$

5. Let

$$A = \begin{pmatrix} 2 & 3 \\ 3 & -6 \end{pmatrix}$$

(a) (5 points) Find the eigenvalues of  $A$ .

$$\begin{aligned} \det(A - \lambda I) &= \det \begin{pmatrix} 2-\lambda & 3 \\ 3 & -6-\lambda \end{pmatrix} \\ &= (2-\lambda)(-6-\lambda) - 9 \\ &= \lambda^2 + 4\lambda - 21 \\ &= (\lambda + 7)(\lambda - 3) \\ &= 0 \end{aligned}$$

$$\Rightarrow \boxed{\lambda = -7, 3}$$

(b) (5 points) Find corresponding eigenvectors of  $A$ .

$$\begin{aligned} \textcircled{1} \quad \underline{\lambda = -7}: \quad A + 7I &= \begin{pmatrix} 9 & 3 \\ 3 & 1 \end{pmatrix} \xrightarrow{\frac{1}{3}I} \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} \\ &\xrightarrow{-I+II} \begin{pmatrix} 3 & 1 \\ 0 & 0 \end{pmatrix} \end{aligned}$$

$$\begin{aligned} \text{Let } y = t, \text{ then } x &= -\frac{1}{3}t \Rightarrow \begin{pmatrix} x \\ y \end{pmatrix} = t \begin{pmatrix} -1/3 \\ 1 \end{pmatrix} \\ \text{so choosing } t &= 3, \quad \boxed{\vec{v}_1 = \begin{pmatrix} -1 \\ 3 \end{pmatrix}} \end{aligned}$$

$$\textcircled{2} \quad \underline{\lambda = 3}: \quad A - 3I = \begin{pmatrix} -1 & 3 \\ 3 & -9 \end{pmatrix} \xrightarrow{(-1)I} \begin{pmatrix} 1 & -3 \\ 3 & -9 \end{pmatrix} \xrightarrow{-3I+II} \begin{pmatrix} 1 & -3 \\ 0 & 0 \end{pmatrix}$$

$$\begin{aligned} \text{Let } y = t, \text{ then } x &= 3t \Rightarrow \begin{pmatrix} x \\ y \end{pmatrix} = t \begin{pmatrix} 3 \\ 1 \end{pmatrix}, \\ \text{so } \quad \boxed{\vec{v}_2 = \begin{pmatrix} 3 \\ 1 \end{pmatrix}} \end{aligned}$$

- (c) (5 points) Is  $A$  orthogonally diagonalizable? Justify your answer with a clear demonstration or with reference to a relevant theorem.

Correction: Yes because  $A$  is symmetric iff it is orthogonally diagonalizable! This is a theorem (Thm. 7, p. 19 of 'Symmetric...' notes). I didn't heed my own warning at the bottom of p. 17 of those notes, and therefore I lose one point!

Yes, because  $A \in \text{Sym}_2(\mathbb{R})$ ,  $\nexists$   
 therefore normal,  $A^T A = A A^T$ .  
 Normal  $\Leftrightarrow$  orthogonally diagonalizable  
 is a theorem.

- (d) (5 points) Let

$$D = [A]_{\beta} = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}$$

be the diagonal matrix with the eigenvalues of  $A$ ,  $\lambda_1$  and  $\lambda_2$ , you got in part (b), whose corresponding eigenvectors  $\mathbf{v}_1$  and  $\mathbf{v}_2$  you got in part (c). Use the fact that  $A = P D P^{-1}$ , and therefore  $e^A = e^{P D P^{-1}} = P e^D P^{-1}$ , to compute  $e^A$ .

Here, of course,  $e^A \stackrel{\text{def}}{=} \sum_{n=0}^{\infty} \frac{1}{n!} A^n$ .

$$e^A = e^{P D P^{-1}}$$

$$= P e^D P^{-1}$$

$$= P e^{\begin{pmatrix} -7 & 0 \\ 0 & 3 \end{pmatrix}} P^{-1}$$

$$= P \begin{pmatrix} e^{-7} & 0 \\ 0 & e^3 \end{pmatrix} P^{-1}$$

$$= \begin{pmatrix} -1 & 3 \\ 3 & 1 \end{pmatrix} \begin{pmatrix} e^{-7} & 0 \\ 0 & e^3 \end{pmatrix} \begin{pmatrix} \frac{1}{10} & -\frac{1}{10} \\ \frac{3}{10} & \frac{1}{10} \end{pmatrix}$$

$$= \frac{1}{10} \begin{pmatrix} -e^{-7} & 3e^3 \\ 3e^{-7} & e^3 \end{pmatrix} \begin{pmatrix} -1 & 3 \\ 3 & 1 \end{pmatrix}$$

$$= \frac{1}{10} \begin{pmatrix} e^{-7} + 9e^3 & -3e^{-7} + 3e^3 \\ -3e^{-7} + 3e^3 & 9e^{-7} + e^3 \end{pmatrix}$$

$$\begin{pmatrix} P = (\vec{v}_1 \ \vec{v}_2) \\ = \begin{pmatrix} -1 & 3 \\ 3 & 1 \end{pmatrix} \\ P^{-1} = \frac{1}{10} \begin{pmatrix} 1 & -3 \\ -3 & 1 \end{pmatrix} \end{pmatrix}$$

$$= \frac{1}{10} \begin{pmatrix} -1 & 3 \\ 3 & 1 \end{pmatrix}$$

6. (5 points) Show that  $A = \begin{pmatrix} 1 & 2-i \\ 2+i & 3 \end{pmatrix}$  is Hermitian.

$$A^* = \overline{A}^T = \begin{pmatrix} \overline{1} & \overline{2+i} \\ \overline{2-i} & \overline{3} \end{pmatrix}$$

$$= \begin{pmatrix} 1 & 2-i \\ 2+i & 3 \end{pmatrix}$$

$$= A$$

7. (5 points) Let  $A \in O(n)$  be an  $n \times n$  orthogonal matrix. Show that

$$(Ax) \cdot (Ay) = x \cdot y$$

for all  $x, y \in \mathbb{R}^n$ . *Hint: the dot product  $x \cdot y$  can be computed in terms of matrices,*

$$x \cdot y = (x_1 \ \cdots \ x_n) \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix} = x^T y$$

$$A \in O(n) \stackrel{\text{def}}{\iff} A^T A = I$$

$$\iff (A\vec{x}) \cdot (A\vec{y})$$

$$= (A\vec{x})^T (A\vec{y})$$

$$= (\vec{x}^T A^T) (A\vec{y})$$

$$= \vec{x}^T (A^T A) \vec{y}$$

$$= \vec{x}^T I \vec{y}$$

$$= \vec{x}^T \vec{y}$$

$$= \vec{x} \cdot \vec{y}$$

8. (5 points) Show that for all  $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$ , we have

$$\mathbf{x} \cdot \mathbf{y} = \frac{1}{4} \|\mathbf{x} + \mathbf{y}\|^2 - \frac{1}{4} \|\mathbf{x} - \mathbf{y}\|^2$$

This is the **polar identity**.<sup>1</sup>

Start with the RHS:

$$\begin{aligned} & \frac{1}{4} \|\bar{\mathbf{x}} + \bar{\mathbf{y}}\|^2 - \frac{1}{4} \|\bar{\mathbf{x}} - \bar{\mathbf{y}}\|^2 \\ &= \frac{1}{4} ((\bar{\mathbf{x}} + \bar{\mathbf{y}}) \cdot (\bar{\mathbf{x}} + \bar{\mathbf{y}})) - \frac{1}{4} ((\bar{\mathbf{x}} - \bar{\mathbf{y}}) \cdot (\bar{\mathbf{x}} - \bar{\mathbf{y}})) \\ &= \frac{1}{4} (\cancel{\bar{\mathbf{x}} \cdot \bar{\mathbf{x}}} + 2\bar{\mathbf{x}} \cdot \bar{\mathbf{y}} + \bar{\mathbf{y}} \cdot \bar{\mathbf{y}}) \\ &\quad - \frac{1}{4} (\cancel{\bar{\mathbf{x}} \cdot \bar{\mathbf{x}}} - 2\bar{\mathbf{x}} \cdot \bar{\mathbf{y}} + \bar{\mathbf{y}} \cdot \bar{\mathbf{y}}) \\ &= \frac{1}{4} (\cancel{\bar{\mathbf{x}} \cdot \bar{\mathbf{x}} - \bar{\mathbf{x}} \cdot \bar{\mathbf{x}}}) + \frac{1}{4} (2\bar{\mathbf{x}} \cdot \bar{\mathbf{y}} - (-2\bar{\mathbf{x}} \cdot \bar{\mathbf{y}})) \\ &\quad + \frac{1}{4} (\cancel{\bar{\mathbf{y}} \cdot \bar{\mathbf{y}} - \bar{\mathbf{y}} \cdot \bar{\mathbf{y}}}) \\ &= \frac{1}{4} 4 \bar{\mathbf{x}} \cdot \bar{\mathbf{y}} \\ &= \bar{\mathbf{x}} \cdot \bar{\mathbf{y}} \end{aligned}$$

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<sup>1</sup>There is also a complex polar identity in  $\mathbb{C}^n$ , in terms of the Hermitian inner product,

$$\mathbf{z} \cdot \mathbf{w} = \frac{1}{4} \sum_{k=1}^4 \|\mathbf{z} + i^k \mathbf{w}\|^2$$

which it would be a worthwhile exercise to try to prove.

9. (10 points) Determine if  $\mathbf{w}$  lies in the span of  $\mathbf{u}$  and  $\mathbf{v}$ . If so, write  $\mathbf{w}$  as a linear combination of  $\mathbf{u}$  and  $\mathbf{v}$ .

$$\mathbf{u} = \begin{pmatrix} 0 \\ 2 \\ -1 \end{pmatrix}, \quad \mathbf{v} = \begin{pmatrix} 2 \\ 2 \\ 0 \end{pmatrix}, \quad \mathbf{w} = \begin{pmatrix} 6 \\ 16 \\ -5 \end{pmatrix}$$

$$\det \begin{pmatrix} \vec{u} & \vec{v} & \vec{w} \end{pmatrix} = \det \begin{pmatrix} 0 & 2 & 6 \\ 2 & 2 & 16 \\ -1 & 0 & -5 \end{pmatrix}$$

$$= 0 \cdot \det \begin{pmatrix} 2 & 16 \\ 0 & -5 \end{pmatrix} - 2 \det \begin{pmatrix} 2 & 16 \\ -1 & -5 \end{pmatrix} + 6 \det \begin{pmatrix} 2 & 2 \\ -1 & 0 \end{pmatrix}$$

$$= -2(-10+16) + 6(0+2)$$

$$= ~~-12~~ -12 + 12$$

$$= 0$$

so, no, not linearly indep. let's find  $a, b \in \mathbb{R}$  such that  $a\vec{u} + b\vec{v} = \vec{w}$ :

$$a\vec{u} + b\vec{v} = \vec{w} \iff \begin{pmatrix} \vec{u} & \vec{v} \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = \vec{w}$$

$$\iff \begin{pmatrix} 0 & 2 \\ 2 & 2 \\ -1 & 0 \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} 6 \\ 16 \\ -5 \end{pmatrix}$$

$$\iff \left( \begin{array}{cc|c} 0 & 2 & 6 \\ 2 & 2 & 16 \\ -1 & 0 & -5 \end{array} \right) \rightarrow \left( \begin{array}{cc|c} 1 & 0 & 5 \\ 2 & 2 & 16 \\ 0 & 2 & 6 \end{array} \right) \rightarrow \left( \begin{array}{cc|c} 1 & 0 & 5 \\ 0 & 2 & 6 \\ 0 & 2 & 6 \end{array} \right)$$

$$\rightarrow \left( \begin{array}{cc|c} 1 & 0 & 5 \\ 0 & 1 & 3 \\ 0 & 0 & 0 \end{array} \right) \Rightarrow \boxed{a=5, b=3}$$

Verify:  $5\vec{u} + 3\vec{v} = 5 \begin{pmatrix} 0 \\ 2 \\ -1 \end{pmatrix} + 3 \begin{pmatrix} 2 \\ 2 \\ 0 \end{pmatrix} = \begin{pmatrix} 6 \\ 16 \\ -5 \end{pmatrix} = \vec{w} \checkmark$

10. (10 points) Use the following properties of the determinant,

- $\det(AB) = (\det A)(\det B)$
- $\det(A^T) = \det A$

to prove that for any square matrix  $A \in M_n(\mathbb{R})$  we have

$$|\det A| = \sqrt{\det(A^T A)}$$

The RHS of this equation is sometimes taken for the definition of the volume of the parallelepiped spanned by the columns of  $A$ . What's even nicer, this also works for non-square  $m \times n$  matrices, with  $m > n$ . For example, a  $3 \times 2$  matrix with two columns of size 3, would give the area of the parallelogram spanned by its columns in  $\mathbb{R}^3$ .

$$\begin{aligned}\det(A^T A) &= (\det(A^T))(\det A) \\ &= (\det A)^2\end{aligned}$$

$$\Rightarrow |\det A| = \sqrt{\det(A^T A)}$$

11. (10 points) The **trace** of a square matrix  $A \in M_n(\mathbb{R})$  is the sum of its diagonal entries,

$$\operatorname{tr} A = \sum_{i=1}^n a_{ii}$$

and we know that  $\operatorname{tr}(AB) = \operatorname{tr}(BA)$ . It follows from this that if  $A$  is diagonalizable, then  $\operatorname{tr} A = \operatorname{tr} D = \sum_{i=1}^n \lambda_i$ . Use this and the fact that any real or complex matrix is diagonalizable over  $\mathbb{C}$  to show that

$$\det e^A = e^{\operatorname{tr} A}$$

[Hint: use the diagonalizability of  $A$  in combination with the exponential,  $e^A = e^{PDP^{-1}} = Pe^D P^{-1}$ .]

First notice that

$$\begin{aligned} e^D &= \sum_{n=0}^{\infty} \frac{1}{n!} D^n = \sum_{n=0}^{\infty} \frac{1}{n!} \begin{pmatrix} \lambda_1^n & & 0 \\ & \ddots & \\ 0 & & \lambda_n^n \end{pmatrix} \\ &= \begin{pmatrix} \sum_{n=0}^{\infty} \frac{1}{n!} \lambda_1^n & & 0 \\ & \ddots & \\ 0 & & \sum_{n=0}^{\infty} \frac{1}{n!} \lambda_n^n \end{pmatrix} \\ &= \begin{pmatrix} e^{\lambda_1} & & 0 \\ & \ddots & \\ 0 & & e^{\lambda_n} \end{pmatrix} \end{aligned}$$

$$\begin{aligned} \Rightarrow \det e^A &= \det (Pe^D P^{-1}) = \det (P) \det (e^D) \det (P^{-1}) \\ &= \cancel{\det(P)} (\det e^D) \cancel{(\det P)^{-1}} \\ &= \det e^D \\ &= \det \begin{pmatrix} e^{\lambda_1} & & 0 \\ & \ddots & \\ 0 & & e^{\lambda_n} \end{pmatrix} \\ &= e^{\lambda_1} e^{\lambda_2} \dots e^{\lambda_n} \\ &= e^{\sum_{i=1}^n \lambda_i} \\ &= e^{\operatorname{tr} D} = e^{\operatorname{tr} A} \end{aligned}$$

12. (5 points) **True or False:** If  $A \in \text{Skew}_n(\mathbb{R})$ , then  $\text{tr } A = 0$ .

Yes, true:  $A^T = -A \implies a_{ii} = -a_{ii}$   
for all  $i$

$$\implies a_{ii} = 0$$

$$\implies \text{tr } A = \sum_{i=1}^n a_{ii} = 0$$

13. (5 points) **True or False:** If  $A$  is diagonalizable with characteristic polynomial factoring as

$$c_A(\lambda) = \det(A - \lambda I) = (-1)^n (\lambda - \lambda_1)^{n_1} (\lambda - \lambda_2)^{n_2} \cdots (\lambda - \lambda_k)^{n_k}$$

where  $\lambda_1, \dots, \lambda_k$  are  $A$ 's distinct eigenvalues, then

$$\det A = \lambda_1^{n_1} \lambda_2^{n_2} \cdots \lambda_k^{n_k}$$

True:  $A = PDP^{-1} = P \begin{pmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_k \end{pmatrix} P^{-1}$   
 $\implies \det A = \det D$   
 $= \lambda_1^{n_1} \cdots \lambda_k^{n_k}$   
 (Annotations:  $\lambda_1$  occurs  $n_1$  times,  $\lambda_k$  occurs  $n_k$  times)

14. (5 points) **True or False:** The standard representation of a reflection is always an orthogonal matrix.

True: Let  $\vec{u}$  be a unit vector in  $\ell$ , so  
 $\ell = \text{span}(\vec{u})$ , & let  $\vec{v} = R_{\pi/2} \vec{u}$ ,  
 also unit length. Thus,  $\beta = (\vec{u}, \vec{v})$   
 is an orthonormal basis, &  $P = \pi_{\beta \circ}$   
 $\in O(2)$ , &  $P[R_\ell]_\beta P^{-1} = [R_\ell]_\sigma$   
 &  $[R_\ell]_\beta = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \sim \text{diag. orth.}$

15. (5 points) **True or False:** If  $A$  is diagonalizable with distinct eigenvalues  $\lambda_1, \dots, \lambda_k$ , then we may write  $A$  as a sum of projections onto the corresponding eigenspaces,

$$A = \lambda_1 \pi_{E_{\lambda_1}} + \lambda_2 \pi_{E_{\lambda_2}} + \dots + \lambda_k \pi_{E_{\lambda_k}}$$

Here, of course, and only to round things off, we mention that

$$\begin{aligned} \pi_{E_{\lambda_i}}(\mathbf{x}) &= \pi_{\mathbf{v}_1}(\mathbf{x}) + \dots + \pi_{\mathbf{v}_k}(\mathbf{x}) \\ &= \frac{\mathbf{v}_1 \cdot \mathbf{x}}{\|\mathbf{v}_1\|^2} \mathbf{v}_1 + \dots + \frac{\mathbf{v}_{n_i} \cdot \mathbf{x}}{\|\mathbf{v}_{n_i}\|^2} \mathbf{v}_{n_i} \end{aligned}$$

where  $\mathbf{v}_1, \dots, \mathbf{v}_{n_i}$  are the basis eigenvectors for  $E_{\lambda_i}$ .

True:  $A = PDP^{-1}$ ,  $D = \begin{pmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_k \end{pmatrix}$

not counting repetitions

$= \begin{pmatrix} \underbrace{\lambda_1 \dots \lambda_1}_{n_1} & & 0 \\ & \ddots & \\ 0 & & \underbrace{\lambda_k \dots \lambda_k}_{n_k} \end{pmatrix}$

$\Rightarrow D = P^{-1}AP$  acts on

$\mathbb{R}^n = E_{\lambda_1} \oplus \dots \oplus E_{\lambda_n}$

by multiplication by  $\lambda_i$  on each  $E_{\lambda_i}$

$\Leftrightarrow A\vec{v}_i = \lambda_i v_i = \lambda_i \frac{\vec{v}_i \cdot \vec{v}_i}{\vec{v}_i \cdot \vec{v}_i} v_i \Leftrightarrow \vec{x} = \vec{x}_1 + \dots + \vec{x}_k$

writing:

$\vec{x} = \vec{x}_1 + \dots + \vec{x}_k$

$\vec{x}_1 \in E_{\lambda_1}, \vec{x}_k \in E_{\lambda_k}$

$A\vec{x} = \sum_i A\vec{x}_i = \sum_i \lambda_i \vec{x}_i$

16. (5 points) Show that if  $U$  and  $V$  are two subspaces of  $\mathbb{C}^n$ , then so is their sum

$$U + V \stackrel{\text{def}}{=} \{\mathbf{u} + \mathbf{v} \mid \mathbf{u} \in U \text{ and } \mathbf{v} \in V\}$$

(1)  $\vec{0} \in U, \vec{0} \in V$  b.c.  $U$  &  $V$  are subspaces

$$\Rightarrow \vec{0} = \vec{0} + \vec{0} \in U + V$$

(2)  $\vec{x} = \vec{u}_1 + \vec{v}_1, \vec{y} = \vec{u}_2 + \vec{v}_2 \in U + V$

$$\Rightarrow \vec{x} + \vec{y} = (\vec{u}_1 + \vec{v}_1) + (\vec{u}_2 + \vec{v}_2)$$

$$= (\underbrace{\vec{u}_1 + \vec{u}_2}_U) + (\underbrace{\vec{v}_1 + \vec{v}_2}_V) \quad (\text{associativity})$$

b.c.  $U$  &  $V$  are subspaces

$\in U + V$  by the def. of  $U + V$ .

(3)  $\vec{x} = \vec{u} + \vec{v} \in U + V, c \in \mathbb{C} \Rightarrow$

$$c\vec{x} = \underbrace{c\vec{u}}_U + \underbrace{c\vec{v}}_V \in U + V$$

