

Name \_\_\_\_\_

Read each question carefully. You must SHOW ALL WORK for full credit.  
NO CALCULATORS. Good luck.

1. (20%) For the linear transformation  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  defined by

$$T \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} x_1 + x_2 \\ x_3 \end{pmatrix}$$

Find the Range and Kernel of  $T$ .

2. (25%)

$$\text{For } A = \begin{pmatrix} -1 & 1 \\ -2 & 2 \end{pmatrix}$$

- (a) Find the eigenvalues of  $A$ .  
(b) Find the corresponding eigenspaces of  $A$ .  
(c) Find an invertible matrix  $P$  and a diagonal matrix  $D$  such that  $P^{-1}AP = D$ .  
(d) Find  $A^{100}$ .  
(e) Find  $e^A$ .
3. (10%) Determine the values of  $\alpha$  for which

$$A = \begin{pmatrix} \alpha & 0 & 0 \\ 0 & 1 & 1 \\ 0 & 1 & 1 \end{pmatrix}$$

is not diagonalizable. Justify.

4. (10%) Find the point on the line  $y = 2x$  that is closest to the point  $(1, 1)$ .
5. (15%) Let  $v_1 = \frac{1}{3}(2, -2, 1, 0)^T$  and  $v_2 = \frac{1}{3}(0, 1, 2, 2)^T$
- (a) Show  $\{v_1, v_2\}$  is an orthonormal set.  
(b) Write  $v = (4, 0, 10, 8)^T$  as a linear combination of  $v_1$  and  $v_2$ . Use dot products to solve, NOT a linear system.
6. (20%) Let  $S = \text{span}\{v_1 = (1, 1, 1, 2)^T, v_2 = (1, 1, 2, 3)^T\}$ . Find a basis for  $S^\perp$ .

BONUS: For  $A = (a_{ij})_{n \times n}$ , suppose  $\alpha$  is not an eigenvalue of  $A$ .

Prove that  $X$  is an eigenvector of  $A$  if and only if  $X$  is an eigenvector of  $(A - \alpha I)^{-1}$ .

TEST 4 SOLUTIONS

1.

$$T \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} x_1 + x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$
$$\text{Range}(T) = CS \begin{pmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \text{span} \left\{ \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \end{pmatrix} \right\} = \mathbb{R}^2$$
$$\text{Kernel}(T) = N \begin{pmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \text{span} \left\{ \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix} \right\}$$

2. (a)

$$\det(\lambda I - A) = \det \begin{pmatrix} \lambda + 1 & -1 \\ 2 & \lambda - 2 \end{pmatrix} = \lambda^2 - \lambda = \lambda(\lambda - 1)$$

Eigenvalues are 0 and 1.

(b)

$$\underline{\lambda = 0} \text{ eigenspace} = N \begin{pmatrix} 1 & -1 \\ 2 & -2 \end{pmatrix} \text{ and } \begin{pmatrix} 1 & -1 \\ 2 & -2 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & -1 \\ 0 & 0 \end{pmatrix}$$
$$\Rightarrow N(0I - A) = \text{span}\{(1, 1)^T\}$$

$$\underline{\lambda = 1} \text{ eigenspace} = N \begin{pmatrix} 2 & -1 \\ 2 & -1 \end{pmatrix} \text{ and } \begin{pmatrix} 2 & -1 \\ 2 & -1 \end{pmatrix} \rightarrow \begin{pmatrix} 2 & -1 \\ 0 & 0 \end{pmatrix}$$

$$\Rightarrow N(1I - A) = \text{span}\{(1, 2)^T\}$$

(c) For example

$$D = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \text{ and } P = \begin{pmatrix} 1 & 1 \\ 1 & 2 \end{pmatrix}$$

Note: Answer not unique.

(d)  $P^{-1}AP = D$ , therefore  $P^{-1}A^kP = D^k$ .

However  $D^k = \begin{pmatrix} 0^k & 0 \\ 0 & 1^k \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = D$ . Therefore  $A^k = PDP^{-1} = A$  for all  $k$ .

$$\text{So } A^{100} = A = \begin{pmatrix} -1 & 1 \\ -2 & 2 \end{pmatrix}.$$

(e)  $A = PDP^{-1} \Rightarrow e^A = Pe^DP^{-1}$ .

$$e^D = \begin{pmatrix} e^0 & 0 \\ 0 & e^1 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & e \end{pmatrix} \Rightarrow A = P \begin{pmatrix} 1 & 0 \\ 0 & e \end{pmatrix} P^{-1} = \begin{pmatrix} 2 - e & e - 1 \\ 2 - 2e & 2e - 1 \end{pmatrix}$$

3.  $\det(\lambda I - A) = (\lambda - \alpha)(\lambda)(\lambda - 2)$ . So the eigenvalues are  $\lambda = 0, 2, \alpha$ . If  $\alpha \neq 0, 2$  then we have three distinct eigenvalues, hence  $A$  is diagonalizable. It remains to consider what happens if  $\alpha$  is 0 or 2.

Case 1  $\alpha = 0$ . The eigenvalues are  $\lambda = 0, 0, 2$ . The eigenvalue of 2 will correspond to one independent eigenvector since it's only a root of the characteristic polynomial once. It remains to consider  $\lambda = 0$  which will correspond to either one or two independent eigenvectors.

$$0I - A = \begin{pmatrix} 0 & 0 & 0 \\ 0 & -1 & -1 \\ 0 & -1 & -1 \end{pmatrix} \rightarrow \dots \rightarrow \begin{pmatrix} 0 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

which has nullity two, and hence corresponds to two independent eigenvectors. This gives a total of 3 independent eigenvectors, hence  $A$  diagonalizable.

Case 2  $\alpha = 2$ . The eigenvalues are  $\lambda = 0, 2, 2$ . The eigenvalue of 0 will correspond to one independent eigenvector since it's only a root of the characteristic polynomial once. It remains to consider  $\lambda = 2$  which will correspond to either one or two independent eigenvectors.

$$2I - A = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & -1 \\ 0 & -1 & 1 \end{pmatrix} \rightarrow \dots \rightarrow \begin{pmatrix} 0 & 1 & -1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

which has nullity two, and hence corresponds to two independent eigenvectors. This gives a total of 3 independent eigenvectors, hence  $A$  diagonalizable.

Therefore  $A$  is never defective, i.e. diagonalizable for all values of  $\alpha$ .

4. Let  $X$  denote a vector along the line  $y = 2x$ , e.g.  $X = (1, 2)^T$ . Let  $Y$  denote the vector from the origin to the point of interest, i.e.  $Y = (1, 1)^T$ . Therefore we want to find  $\text{proj}_X Y$ .

$$\text{proj}_X Y = \frac{\langle X, Y \rangle}{\langle X, X \rangle} X = \frac{1+2}{1+4} X = \frac{3}{5} X = \frac{3}{5} (1, 2)^T = \left(\frac{3}{5}, \frac{6}{5}\right)^T.$$

5. (a)  $\langle v_1, v_2 \rangle = 0 \Rightarrow$  orthogonal.  
 $\langle v_1, v_1 \rangle = 1 = \langle v_2, v_2 \rangle \Rightarrow$  orthonormal.  
 (b)  $\langle v, v_1 \rangle = \frac{8}{3} + 0 + \frac{10}{3} + 0 = 6$   
 $\langle v, v_2 \rangle = 0 + 0 + \frac{20}{3} + \frac{16}{3} = 12$

$$\text{Therefore } v = 6v_1 + 12v_2$$

6. If  $x \in S^\perp$  then  $v_1^T x = v_2^T x = 0$ . That is if  $x = (x_1, x_2, x_3, x_4)^T$  then for

$$A = \begin{pmatrix} 1 & 1 & 1 & 2 \\ 1 & 1 & 2 & 3 \end{pmatrix}, \text{ we must have } Ax = \underline{0} \text{ or } \begin{pmatrix} 1 & 1 & 1 & 2 \\ 1 & 1 & 2 & 3 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

In other words  $S^\perp = N(A)$

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

so  $S^\perp = N(A) = \text{span}\{(-1, 1, 0, 0)^T, (-1, 0, -1, 1)^T\}$ .

BONUS: Proof:

$X$  is an eigenvector for  $A \Leftrightarrow AX = \lambda X$  (for some  $\lambda$ )  $\Leftrightarrow AX - \alpha X = \lambda X - \alpha X \Leftrightarrow$   
 $AX - \alpha IX = \lambda X - \alpha X \Leftrightarrow (A - \alpha I)X = (\lambda - \alpha)X \Leftrightarrow \frac{1}{\lambda - \alpha}X = (A - \alpha I)^{-1}X \Leftrightarrow$   
 $X$  is an eigenvector of  $(A - \alpha I)^{-1}$

Note:  $A - \alpha I$  is invertible because  $\alpha$  is not an eigenvalue of  $A$ .