

# Exam 1

This exam covers Topics 1 - 3, Topic 4 will not be covered here.

## Part I: True/False (5 points each; 25 points)

For each of the following mark as true or false.

- a) \_\_\_\_\_ If  $A$  and  $B$  are  $n \times n$  lower triangular matrices, then  $AB$  is also lower triangular.

This is true and in fact the diagonal is the product of the diagonals.

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & \cdots & a_{1n} \\ 0 & a_{22} & a_{23} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & a_{nn} \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} & b_{13} & \cdots & b_{1n} \\ 0 & b_{22} & b_{23} & \cdots & b_{2n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & b_{nn} \end{bmatrix} = \begin{bmatrix} a_{11}b_{11} & d_{12} & d_{13} & \cdots & d_{1n} \\ 0 & a_{22}b_{22} & d_{23} & \cdots & d_{2n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & a_{nn}b_{nn} \end{bmatrix}$$

- b) \_\_\_\_\_ If  $W$  is a subspace of a vector space  $V$  and  $\mathcal{B}_W$  is a basis for  $W$ , then there is a unique subspace  $U$  so that  $V = W \oplus U$  and a basis  $\mathcal{B}_U$  for  $U$  so that  $\mathcal{B}_V = \mathcal{B}_W + \mathcal{B}_U$  is a basis for  $V$ .

This is false, there may be many, infinitely many such  $U$ . Just take  $\mathbb{R}^2$  and  $W = \text{span}\{(0, 1)\}$ , the  $y$ -axis, then  $U = \text{span}\{(a, b)\}$ , the line containing  $(a, b)$ . The only condition is that  $W$  and  $U$  are not the same line, that is  $a \neq 0$ .

- c) \_\_\_\_\_ If  $W$  is a subspace of a vector space  $V$  and  $\mathcal{B}$  is a basis for  $V$ , then  $\mathcal{B}$  can be restricted to a basis for  $W$ .

This is not true. Let  $W = \text{span}\{(1, 1)\} \subseteq \mathbb{R}^2 = V$ . The standard basis for  $\mathbb{R}^2$  can not be restricted to a basis for  $W$ .

- d) \_\_\_\_\_ Let  $A$  be an  $n \times n$  matrix over  $\mathbb{C}$ , then  $\det(\bar{A}) = \det(A)$ , where  $\bar{A}_{i,j} = \overline{A_{i,j}}$ . Here,  $\bar{z} = a - ib$  when  $z = a + ib$ , the *complex conjugate* of  $z$ .

This is true and can be seen in several ways. Induction using expansion along a row or column works. For  $n = 1$ , this is trivial,  $\det[\alpha] = \det[\bar{\alpha}]$ . If  $A$  is  $(n+1) \times (n+1)$

and we know the proposition holds for  $n \times n$  matrices, then

$$\begin{aligned}
\det(\bar{A}) &= \sum_{i=1}^n \bar{a}_{i,j} (-1)^{i+j} \det(\bar{M}_{i,j}) \\
&= \sum_{i=1}^n \bar{a}_{i,j} (-1)^{i+j} \overline{\det(M_{i,j})} && \text{(induction hypothesis)} \\
&= \overline{\sum_{i=1}^n a_{i,j} (-1)^{i+j} \det(M_{i,j})} \\
&= \overline{\det(A)}
\end{aligned}$$

- e) \_\_\_\_ For  $n \times n$  matrices  $A$  and  $B$ , define  $A \otimes B = AB - BA$ . The operator  $\otimes$  is not associative or commutative.

This is true. Failure of commutativity is trivial since  $A \otimes B = -(B \otimes A)$ , this is *anti-commutative*.

$$\begin{aligned}
A \otimes (B \otimes C) &= A(B \otimes C) - (B \otimes C)A \\
&= A(BC - CB) - (BC - CB)A \\
&= ABC - ACB - BCA + CBA
\end{aligned}$$

while

$$\begin{aligned}
(A \otimes B) \otimes C &= (A \otimes B)C - C(A \otimes B) \\
&= (AB - BA)C - C(AB - BA) \\
&= ABC - BAC - CAB + CBA
\end{aligned}$$

So

$$\begin{aligned}
A \otimes (B \otimes C) - (A \otimes B) \otimes C &= -ACB - BCA - (-BAC - CAB) \\
&= (CA - AC)B - B(CA - AC) \\
&= (C \otimes A) \otimes B
\end{aligned}$$

which in general is not  $\mathbf{0}$ .

Notice  $A \otimes (B \otimes C) = -(B \otimes C) \otimes A$  so we have

$$-(B \otimes C) \otimes A - (A \otimes B) \otimes C = (C \otimes A) \otimes B$$

or equivalently

$$(A \otimes B) \otimes C + (B \otimes C) \otimes A + (C \otimes A) \otimes B = 0$$

This is an important identity in quantum mechanics called the Jacobi identity.

## Part II: Definitions and Theorems (5 points each; 25 points)

- a) Define what it means for a set of vectors  $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  from a real vector space  $V$  to span  $V$ .

$\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  **spans**  $V$  iff for all  $\mathbf{v} \in V$ ,  $\mathbf{v}$  is a linear combination of the vectors in  $\mathcal{B}$ , that is  $\mathbf{v} = \sum_{i=1}^n \alpha_i \mathbf{v}_i$  for some coefficients  $\alpha_i \in \mathbb{R}$ .

- b) Define what it means for a set of vectors  $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  from a real vector space  $V$  to be linearly independent.

A set of vectors  $\mathcal{B}$  is **linearly independent** iff  $\sum_{i=1}^n \alpha_i \mathbf{v}_i = \mathbf{0}$ , then  $\alpha_i = 0$  for all  $i$ . Equivalently, any linear combination of the vectors that gives  $\mathbf{0}$  must be trivial.

There are equivalent definitions, but essentially, this is what you need. You can say something like, for all  $i$ ,  $\mathbf{v}_i \notin \text{span}\{\mathbf{v}_1, \dots, \mathbf{v}_{i-1}, \mathbf{v}_{i+1}, \dots, \mathbf{v}_n\}$

- c) Define what it means for a set of vectors  $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  to be a basis for a vector space  $V$ .

$\mathcal{B}$  has must be a linearly independent and span  $V$ .

- d) State the Rank-Nullity Theorem.

If  $A$  is an  $m \times n$  matrix, then  $n = \dim(\text{RS}(A)) + \dim(\text{NS}(A)) = \text{rank}(A) + \text{nullity}(A)$ .

- e) If  $B$  arises from a matrix  $A$  by elementary row operations, what is the relationship between  $\text{NS}(A)$  and  $\text{NS}(B)$ ?

$\text{NS}(A) = \text{NS}(B)$

This is because  $A\mathbf{x} = \mathbf{b} \iff B\mathbf{x} = \mathbf{b}$ , i.e., the systems of equations are equivalent. So  $A\mathbf{x} = \mathbf{0} \iff B\mathbf{x} = \mathbf{0}$  and thus  $\mathbf{x} \in \text{NS}(A) \iff \mathbf{x} \in \text{NS}(B)$ .

## Part III: Computational (15 points each; 45 point)

a) Use row ops to find an echelon form of

$$A = \begin{bmatrix} 1 & 2 & 2 & -2 & 2 \\ 2 & 4 & 1 & -2 & 5 \\ 1 & 2 & -1 & 0 & 3 \end{bmatrix}$$

Make sure to write out your steps and indicate the row ops at each step.

$$A \xrightarrow[\substack{R_3 - R_1 \rightarrow R_3}]{\substack{R_2 - 2R_1 \rightarrow R_2}} \begin{bmatrix} 1 & 2 & 2 & -2 & 2 \\ 0 & 0 & -3 & 2 & 1 \\ 0 & 0 & -3 & 2 & 1 \end{bmatrix} \xrightarrow{R_3 - R_2 \rightarrow R_3} \begin{bmatrix} 1 & 2 & 2 & -2 & 2 \\ 0 & 0 & -3 & 2 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

b) Use the echelon matrix found above to find a basis for  $\text{RS}(A)$ ,  $\text{NS}(A)$ , and  $\text{CS}(A)$ . Give a brief reason for your choice.

Without a justification, you might just have a lucky guess and I will not accept this. Your justification can be short and use facts from the text or from the notes that I have provided.

A basis for  $\text{RS}(A)$  is given by  $\{(1, 2, 2, -2, 2), (0, 0, -3, 2, 1)\}$ .

Justification: Take the non-zero rows of the echelon form.

A basis for  $\text{CS}(A)$  is given by columns 1 and 3 of  $A$ , that is,  $\{(1, 2, 1), (2, 1, -1)\}$

Justification: These correspond to the pivot columns and we know this is a basis.

For  $\text{NS}(A)$  we perform back substitution, letting  $x_2 = r$ ,  $x_4 = s$ , and  $x_5 = t$ , so

$$-3x_3 = -2s - t$$

so

$$\begin{aligned} x_3 &= (2/3)s + (1/3)t \\ x_1 &= -2r - 2x_3 + 2s - 2t \\ &= -2r - 2((2/3)s + (1/3)t) + 2s - 2t \\ &= -2r + 2/3s - 8/3t \end{aligned}$$

So a typical element of  $\text{NS}(A)$  is

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix} = \begin{bmatrix} -2r + (2/3)s - (8/3)t \\ r \\ (2/3)s + (1/3)t \\ s \\ t \end{bmatrix} = r \begin{bmatrix} -2 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} + s \begin{bmatrix} 2/3 \\ 0 \\ 2/3 \\ 1 \\ 0 \end{bmatrix} + t \begin{bmatrix} -8/3 \\ 0 \\ 1/3 \\ 0 \\ 1 \end{bmatrix}$$

A basis is

$$\{(-2, 1, 0, 0, 0), (2/3, 0, 2/3, 1, 0), (-8/3, 0, 1/3, 0, 1)\}$$

- c) Show that skew-symmetric  $3 \times 3$  matrices form a subspace of all  $3 \times 3$  matrices and find a basis for this subspace.

These matrices look like

$$\begin{bmatrix} 0 & a & b \\ -a & 0 & c \\ -b & -c & 0 \end{bmatrix}$$

So a basis is

$$\left\{ \begin{bmatrix} 0 & 1 & 0 \\ -1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \\ -1 & 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & -1 & 0 \end{bmatrix} \right\}$$

If you think about it the symmetric matrices have a similar basis of size 9, so  $M^{3 \times 3} = \text{Sym}^{3 \times 3} \oplus \text{skewSym}^{3 \times 3}$  and the basis for both are mutually orthogonal.

## Part IV: Proofs (20 points each; 60 points) - Choose three!

Provide complete arguments/proofs for three of the following. If you try more than three, I will just grade the first three, so pick three your best three! If you want to ask me about these, please do.

- a)  $A$  is invertible iff there exists a matrix  $B$  so that  $AB = BA = I$ . It is simple to show that:
- (i) If  $A$  is invertible and  $AB = I$ , then  $BA = I$  as well and  $B$  is the unique such matrix.
  - (ii) If  $A$  is invertible and  $BA = I$ , then  $AB = I$  as well and  $B$  is the unique such matrix.

This shows that if  $A$  is invertible, then there is a unique matrix  $B$  such that  $AB = I$  or  $BA = I$ . Call this unique matrix  $A^{-1}$ .

The goal here is to show that the assumption “ $A$  is invertible” is not needed in (i) or (ii).

**Prove:** Let  $A$  and  $B$  be square matrices with  $AB = I$ . Show that  $A$  is invertible and hence  $B = A^{-1}$ .

You may refer to Theorem 1.5.2 or Theorem 2.2.2, but be clear and complete in your argument.

**Proof 1:** Show that  $\text{NS}(B) = \{\mathbf{0}\}$  and hence  $B$  is invertible and from above  $A = B^{-1}$ , but then clearly  $A$  is invertible too.

Clearly,  $\mathbf{x} \in \text{NS}(B) \implies \mathbf{x} \in \text{NS}(AB) = \text{NS}(I) = \{\mathbf{0}\}$ , so

$$\{\mathbf{0}\} \subseteq \text{NS}(B) \subseteq \text{NS}(AB) = \{\mathbf{0}\}$$

so  $\text{NS}(B) = \{\mathbf{0}\}$ . So  $B$  is invertible and  $AB = I$ , so  $A = B^{-1}$ .

**Proof 2:**  $\det(AB) = \det(A)\det(B) = 1$ , so  $\det(A) \neq 0$ , hence  $A$  is invertible.

b) **Prove:**  $\text{NS}(A) = \text{NS}(A^T A)$  for any matrix  $A$ .

You have actually done this already in the homework, but you may also use the easy fact that  $\mathbf{x}^T \mathbf{x} > 0$  for  $\mathbf{x} \neq \mathbf{0}$ .

As in (a), clearly,  $\text{NS}(A) \subseteq \text{NS}(A^T A)$ , we just need the other direction.

Suppose  $\mathbf{x} \in \text{NS}(A^T A)$ , then  $A^T A \mathbf{x} = \mathbf{0}$ , so  $\mathbf{x}^T (A^T A \mathbf{x}) = 0$ . But  $\mathbf{x}^T (A^T A \mathbf{x}) = (\mathbf{x}^T A^T)(A \mathbf{x}) = (A \mathbf{x})^T (A \mathbf{x})$ . As mentioned, is  $(A \mathbf{x})^T (A \mathbf{x}) = 0$  iff  $A \mathbf{x} = \mathbf{0}$ . So  $\mathbf{x} \in \text{NS}(A)$  as needed.

c) **Prove:** If  $A$  and  $B$  are  $m \times n$  matrices such that  $A \mathbf{x} = B \mathbf{x}$  for all  $\mathbf{x} \in \mathbb{R}^n$ , then  $A = B$ .

The hypothesis is equivalent to  $(A - B) \mathbf{x} = \mathbf{0}$  for all  $\mathbf{x}$  and the conclusion is equivalent to  $A - B = \mathbf{0}$ .

It suffices to prove:

If  $A \mathbf{x} = \mathbf{0}$  for all  $\mathbf{x}$ , then  $A = \mathbf{0}$ .

This is simple, say  $A = [\mathbf{a}_1 \ \mathbf{a}_2 \ \cdots \ \mathbf{a}_n]$  (the columns of  $A$ ). Then  $A \mathbf{e}_j = \mathbf{a}_j = \mathbf{0}$ . But then  $\mathbf{a}_j(i) = A_{ij} = 0$  for all  $1 \leq i, j \leq n$ . So  $A = \mathbf{0}$  (the all 0 matrix).

d) **Prove:** If  $A$  is an  $n \times n$  matrix and  $A^k = \mathbf{0}$  for any  $k$ , then  $A^n = \mathbf{0}$ .

**Proof 1:** To do this show

- i) Show  $\text{NS}(A^{m+1}) \supseteq \text{NS}(A^m)$  for all  $m$ .
- ii) Show that if  $\text{NS}(A^{m+1}) = \text{NS}(A^m)$ , then  $\text{NS}(A^n) = \text{NS}(A^m)$  for all  $n \geq m$ .

It is clear that  $\text{NS}(A^{m+1}) \supseteq \text{NS}(A^m)$ , since  $A^m \mathbf{x} = \mathbf{0} \implies A(A^m \mathbf{x}) = \mathbf{0} \implies A^{m+1} \mathbf{x} = \mathbf{0}$ . So (i) is shown,

For (ii) suppose  $\text{NS}(A^m) = \text{NS}(A^{m+1})$ , then  $A^{m+2} \mathbf{x} = \mathbf{0} \implies A^{m+1}(A \mathbf{x}) = \mathbf{0} \implies A^m(A \mathbf{x}) = \mathbf{0} \implies A^{m+1} \mathbf{x} = \mathbf{0}$ . So  $\text{NS}(A^{m+2}) \subseteq \text{NS}(A^{m+1})$ , but then  $\text{NS}(A^{m+2}) = \text{NS}(A^{m+1}) = \text{NS}(A^m)$ . Now just keep going to get  $\text{NS}(A^k) = \text{NS}(A^m)$  for all  $k \geq m$ .

This means we have

$$\text{NS}(A^0) \subsetneq \text{NS}(A^1) \subsetneq \text{NS}(A^2) \subsetneq \cdots \subsetneq \text{NS}(A^{m-1}) \subsetneq \text{NS}(A^m) = \text{NS}(A^{m+1}) = \cdots$$

The  $m$  at which  $\text{NS}(A^k) = \text{NS}(A^m)$  for all  $m \geq k$  must itself be  $\leq n$ .

If  $A^k = \mathbf{0}$  for any  $k$ , then  $\text{NS}(A^k) = \mathbb{R}^n$  is maximal and thus  $m \leq k$  and  $\text{NS}(A^m) = \mathbb{R}^n$ . Since  $m \leq n$ ,  $\text{NS}(A^n) = \mathbb{R}^n$  and so  $A^n = \mathbf{0}$ .

**Proof 2:** You can use induction. To do this we need to prove something that sounds slightly stronger:

$P_n$  : For any  $n \times n$  matrix  $A$ , if  $A^m = \mathbf{0}$  for any  $m > n$ , then  $A^n = \mathbf{0}$ .

**base case: ( $n = 1$ )** If  $A^m = [a]^m = [a^m] = [0]$ , for  $m > 1$ , then  $a = 0$ , so  $A^1 = [a] = [0]$  as needed.

**inductive step:** Suppose  $P_{n-1}$ : For any  $m > n - 1$ ,  $A^m = \mathbf{0} \implies A^{n-1} = \mathbf{0}$  for all  $(n - 1) \times (n - 1)$  matrices. We want to prove  $P_n$ .

Assume  $A$  is an  $n \times n$  matrix and  $A^m = \mathbf{0}$  for some  $m > n$ . Notice that  $\ker(A) \neq \{\mathbf{0}\}$ , since if  $\ker(A) = \{\mathbf{0}\}$ , then  $A : \mathbb{R}^n \rightarrow \mathbb{R}^n$  is injective and thus  $A^m$  is also injective, so  $\ker(A^m) = \{\mathbf{0}\}$ . This obviously contradicts  $A^m = \mathbf{0}$ .

Let  $\mathbf{v}_1 \in \ker(A)$  and let  $\mathcal{B} = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  be a basis for  $\mathbb{R}^n$ . So letting  $B = [\mathbf{v}_1 \ \mathbf{v}_2 \ \dots \ \mathbf{v}_n]$  we have

$$A = BA'B^{-1}$$

where

$$A' = \begin{bmatrix} 0 & a'_{12} & a'_{13} & \dots & a'_{1n} \\ 0 & & & & \\ \vdots & & \hat{A} & & \\ 0 & & & & \end{bmatrix}$$

where  $\hat{A}$  is the indicated  $(n - 1) \times (n - 1)$  submatrix of  $A'$ .

$A'$  is the matrix of  $L(\mathbf{x}) = A\mathbf{x}$  with respect to the basis  $\mathcal{B}$ . Notice that  $A^m = \mathbf{0}$  means  $L^m(\mathbf{x}) = \mathbf{0}$  for all  $\mathbf{x}$  and hence  $A'\mathbf{x} = \mathbf{0}$  for all  $\mathbf{x}$ , a finally this means  $A'^m = \mathbf{0}$ .

Notice that  $A'$  has the block form

$$\begin{bmatrix} 0 & \mathbf{b}^T \\ \mathbf{0} & \hat{A} \end{bmatrix}$$

and

$$(A')^2 = \begin{bmatrix} 0 & \mathbf{b}^T \\ \mathbf{0} & \hat{A} \end{bmatrix} \begin{bmatrix} 0 & \mathbf{b}^T \\ \mathbf{0} & \hat{A} \end{bmatrix} = \begin{bmatrix} 0 & \mathbf{b}^T \hat{A} \\ \mathbf{0} & \hat{A}^2 \end{bmatrix}$$

By a second, obvious induction,

$$(A')^k = \begin{bmatrix} 0 & \mathbf{b}^T \hat{A}^{k-1} \\ \mathbf{0} & \hat{A}^k \end{bmatrix}$$

We assume  $\hat{A}'^m = \mathbf{0}$  so  $\hat{A}^m = \mathbf{0}$  and by induction  $\hat{A}^{n-1} = \mathbf{0}$  and thus

$$(A')^n = \begin{bmatrix} 0 & \mathbf{b}^T \hat{A}^{n-1} \\ \mathbf{0} & \hat{A}^n \end{bmatrix} = \mathbf{0}$$