

EE5120 Linear Algebra: Tutorial 4, July-Dec 2017-18

1. State True or False for each of the following with proper justification:

- (a) Let matrix A be a transformation from \mathbb{R}^m to \mathbb{R}^n , then dimension of left nullspace of A , i.e. $N(A^T)$ is $m - r$.
- (b) The pseudoinverse $(A^t A)^{-1} A$ of any linear operator A exists even if the operator is not invertible.
- (c) Let V and W be vector spaces, and let $T : V \rightarrow W$ be linear. The T is one-to-one iff $N(T) = \{0\}$.
- (d) Let $v \in \mathbb{R}^n$. The nullity of matrix vv^t is n .
- (e) $A = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$ is a rotation matrix and $B = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$ is a reflection matrix.

Solution:

- (a) False. Dimension will be $n - r$.
- (b) False. $A^t A$ needs to be invertible.
- (c) True. Suppose that T is one-to-one and $x \in N(T)$. Then $T(x) = 0 = T(0)$. Hence, $N(T) = \{0\}$.
- (d) False. vv^t is a rank 1 matrix. By rank-nullity theorem, nullity = $n - 1$.
- (e) True.

2. In \mathbb{R}^2 , let L be the line $y = 2x$. Find an expression for $T(x, y)$, where T is the reflection of \mathbb{R}^2 about L .

Solution: Refer section 2.6 of Gilbert Strang. The matrix for reflection about θ line is

$$H = \begin{bmatrix} 2c^2 - 1 & 2cs \\ 2cs & 2s^2 - 1 \end{bmatrix}, \text{ where } c = \cos\theta \text{ and } s = \sin\theta. \text{ For line } y = 2x, \theta = \tan^{-1}(2).$$

Substitute these values in H and you get, $T(x, y) = (1/5) \begin{bmatrix} -3x + 4y \\ 4x + 3y \end{bmatrix}$.

3. Prove that for two matrices, A, B , the following holds: $\text{rank}(AB) \leq \min(\text{rank}(A), \text{rank}(B))$.

Solution: Step 1 (col picture): Consider the product AB in the following way: $AB = A[b_1 \dots b_n] = [Ab_1 \dots Ab_n]$, where b_i is a column of B . Each of these Ab_i is a linear combination of the columns of A , hence, $Ab_i \in C(A)$ (col space of A), thus $C(AB) \in C(A)$. This implies that the rank of AB can not exceed that of A .

Step 2 (row picture): In the product AB , every row is a linear combination* of the rows of B . We know that linear combinations of rows don't change the rank of a matrix, thus, the rank of AB can not exceed that of B . Putting the two steps together, we get the desired result.

*: To see this, $(AB)_{ij} = \sum_k a_{ik} b_{kj}$. So the i^{th} row of (AB) is:

$$[\sum_k a_{ik} b_{k1} \quad \sum_k a_{ik} b_{k2} \quad \dots \quad \sum_k a_{ik} b_{kn}] = \sum_k a_{ik} [b_{k1} \quad b_{k2} \quad \dots \quad b_{kn}] = \sum_k a_{ik} [b_k] \text{ where } [b_k] \text{ is the } k^{\text{th}} \text{ row of } B, \text{ i.e. a linear combination of the rows.}$$

4. Let T be a linear transformation from R^3 into R^2 and U be a linear transformation from R^2 into R^3 . Prove that the transformation UT is not invertible. Generalize the theorem. (Can you relate this to question no.7 of the previous tutorial?)

Solution: Since U, T are linear transformations, they must have matrix representations. T is 2×3 , and U is 2×3 , thus both their ranks can be at most 2. Using the result above, even though the size of UT is 3×3 , it can have rank at most 2. Thus, UT is not invertible.

Alt proof: For a transformation to be invertible, it should be both one-to-one and onto. In this case, T is not one-to-one and therefore UT is not one-to-one. Hence it is not invertible. Any transformation from a higher dimensional space to a lower dimensional space leads to loss of information in one or more dimensions and hence is not invertible.

5. What 3 by 3 matrices represent the transformations that,
- project every vector onto the x-y plane?
 - reflect every vector through the x-y plane?
 - rotate the x-y plane through 90° , leaving the z-axis alone?
 - rotate the x-y plane, then x-z, then y-z through 90° ?
 - carry out the same three rotations, but each one through 180° ?

Solution:

$$(a) \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$(b) \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{bmatrix}$$

$$(c) \begin{bmatrix} \cos(\pi/2) & -\sin(\pi/2) & 0 \\ \sin(\pi/2) & \cos(\pi/2) & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$(d) \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & -1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 0 & 0 & -1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 0 & -1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}$$

$$(e) \begin{bmatrix} 1 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & -1 \end{bmatrix} \begin{bmatrix} -1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{bmatrix} \begin{bmatrix} -1 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

6. Let $\alpha_1 = \left\{ \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} \right\}$ be the (ordered) basis for the vector space $M_{2 \times 2}$, which is the set of all real valued 2×2 matrices. Also, let $\alpha_2 = \{x^2, x, 1\}$ be the basis for the vector space P_2 , which is the set of all real polynomials (with real co-efficients) with minimum degree 2. Compute the matrix representations for the following linear transformations:

(a) $T_1 : M_{2 \times 2} \rightarrow M_{2 \times 2}$ with $T_1(\mathbf{A}) = \mathbf{A}^T$, for every $\mathbf{A} \in M_{2 \times 2}$.

(b) $T_2 : P_2 \rightarrow M_{2 \times 2}$ with $T_2(f(x)) = \begin{bmatrix} f'(0) & 2f(1) \\ 0 & f''(3) \end{bmatrix}$. Here, $f'(x)$ and $f''(x)$ are the 1st and 2nd derivatives of $f(x) \in P_2$.

Solution:

(a) We have the following:

$$T_1 \left(\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \right) = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} = (1) \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} + (0) \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} + (0) \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} + (0) \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}.$$

We take the co-efficients present in the linear combination shown above to construct the first column in the matrix representation of T_1 will be $[1000]^T$. Similarly,

$$T_1 \left(\begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right) = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} = (0) \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} + (1) \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} + (0) \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} + (0) \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix},$$

$$T_1 \left(\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \right) = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} = (0) \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} + (0) \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} + (0) \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} + (1) \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}$$

$$T_1 \left(\begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} \right) = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} = (0) \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} + (0) \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} + (1) \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} + (0) \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}.$$

Thus, matrix representation of T_1 is given by
$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \end{bmatrix}.$$

(b) Following the same procedure as discussed above, we get,

$$T_2(x^2) = \begin{bmatrix} 2(0) & 2(1^2) \\ 0 & 2 \end{bmatrix} = \begin{bmatrix} 0 & 2 \\ 0 & 2 \end{bmatrix} = (0) \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} + (2) \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} + (2) \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} + (0) \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix},$$

$$T_2(x) = \begin{bmatrix} 1 & 2(1) \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 0 & 0 \end{bmatrix} = (1) \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} + (0) \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} + (2) \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} + (0) \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix},$$

$$T_2(1) = \begin{bmatrix} 0 & 2(1) \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 2 \\ 0 & 0 \end{bmatrix} = (0) \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} + (0) \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} + (2) \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} + (0) \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}.$$

The matrix representation for T_2 is given by
$$\begin{bmatrix} 0 & 1 & 0 \\ 2 & 0 & 0 \\ 2 & 2 & 2 \\ 0 & 0 & 0 \end{bmatrix}.$$

7. Let V be a vector space and $T : V \rightarrow V$ be a linear transformation. Suppose $\mathbf{x} \in V$ is such that $T^k(\mathbf{x}) = \mathbf{0}$, $T^m(\mathbf{x}) \neq \mathbf{0}$, $\forall 1 \leq m < k$ and $k > 1$, then prove that the set of vectors $\{\mathbf{x}, T(\mathbf{x}), T^2(\mathbf{x}), \dots, T^{k-1}(\mathbf{x})\}$ is linearly independent.

Solution: Given $k > 1$. Thus, $T(\mathbf{x}) \neq \mathbf{0} \Rightarrow \mathbf{x} \neq \mathbf{0}$. Since $T^k(\mathbf{x}) = \mathbf{0}$, for all $p \geq 1$,

$$T^{k+p}(\mathbf{x}) = T^p(T^k(\mathbf{x})) = T^p(\mathbf{0}) = \mathbf{0}. \quad (1)$$

Assume that $\{\mathbf{x}, T(\mathbf{x}), T^2(\mathbf{x}), \dots, T^{k-1}(\mathbf{x})\}$ is linearly dependent. Then,

$$a_1\mathbf{x} + a_2T(\mathbf{x}) + \dots + a_kT^{k-1}(\mathbf{x}) = \mathbf{0},$$

with not all a_i s being zero, i.e., some a_i s are not equal to zero. Now, consider the following:

$$\begin{aligned} T^{k-1}(a_1\mathbf{x} + a_2T(\mathbf{x}) + \dots + a_kT^{k-1}(\mathbf{x})) &= T^{k-1}(\mathbf{0}) \\ \Rightarrow a_1T^{k-1}(\mathbf{x}) + a_2T^k(\mathbf{x}) + a_3T^{k+1}(\mathbf{x}) + \dots + a_kT^{2(k-1)}(\mathbf{x}) &= \mathbf{0} \\ \Rightarrow a_1T^{k-1}(\mathbf{x}) + \mathbf{0} + \mathbf{0} + \dots + \mathbf{0} &= \mathbf{0}. \end{aligned}$$

The above result is a consequence of equation (1) and other given information. Since $T^{k-1}(\mathbf{x}) \neq \mathbf{0}$, $a_1 = 0$. Now,

$$\begin{aligned} T^{k-2}(a_1\mathbf{x} + a_2T(\mathbf{x}) + \dots + a_kT^{k-1}(\mathbf{x})) &= T^{k-2}(\mathbf{0}) \\ \Rightarrow a_1T^{k-2}(\mathbf{x}) + a_2T^{k-1}(\mathbf{x}) + a_3T^k(\mathbf{x}) + \dots + a_kT^{2k-3}(\mathbf{x}) &= \mathbf{0} \\ \Rightarrow \mathbf{0} + a_2T^{k-1}(\mathbf{x}) + \mathbf{0} + \dots + \mathbf{0} &= \mathbf{0}. \end{aligned}$$

Again, since $T^{k-1}(\mathbf{x}) \neq \mathbf{0}$, we get $a_2 = 0$. On repeating this procedure, we get $a_i = 0, \forall i = 1, 2, \dots, k$, which is contradicting to the initial assumption. Hence, the initial assumption of the set $\{\mathbf{x}, T(\mathbf{x}), T^2(\mathbf{x}), \dots, T^{k-1}(\mathbf{x})\}$ being linearly dependent is incorrect. Thus, the above set is linearly independent.

8. BONUS question: **Definition:** Let V be a vector space and $T : V \rightarrow V$ be a linear transformation on V . A subspace $W \subset V$ is said to be T -invariant if for every $\mathbf{w} \in W$, $T(\mathbf{w}) \in W$. Further, if W is T -invariant, define *restriction of T on W* as, $T_W : W \rightarrow W$ such that $T_W(\mathbf{w}) = T(\mathbf{w}), \forall \mathbf{w} \in W$. Then, prove the following results:

- Subspaces $\{\mathbf{0}\}, V, N(T)$ and $R(T)$ are T -invariant. Here, $N(T) = \{\mathbf{v} \in V | T(\mathbf{v}) = \mathbf{0}\}$, and $R(T) = \{\mathbf{u} \in V | \exists \mathbf{x}_u \in V \text{ s.t } T(\mathbf{x}_u) = \mathbf{u}\}$ (The choice of \mathbf{x}_u depends on \mathbf{u}).
- For a T -invariant subspace W , the transformation T_W is linear, and $N(T_W) = N(T) \cap W$.

Solution:

- Let $U = \{\mathbf{0}\}$. Its a singleton set. Since T is linear $T(\mathbf{0}) = \mathbf{0} \Rightarrow T(\mathbf{0}) \in U$. Hence, $U = \{\mathbf{0}\}$ is T -invariant.
 - Since T is defined from V to V , for every $\mathbf{v} \in V$, $T(\mathbf{v}) \in V \Rightarrow V$ is T -invariant.
 - Since T is linear $T(\mathbf{0}) = \mathbf{0}$. Thus, $\mathbf{0} \in N(T)$ by the definition of $N(T)$. Further, $T(\mathbf{v}) = \mathbf{0} \in N(T)$, for all $\mathbf{v} \in N(T)$. Hence, $N(T)$ is T -invariant.

(iv) Let $\mathbf{u} \in R(T)$. Since, $R(T) \subset V$ (by definition), $\mathbf{u} \in V$. Thus, $T(\mathbf{u}) \in R(T) \Rightarrow R(T)$ is T -invariant.

(b) (i) Let $\mathbf{x}, \mathbf{y} \in W$. Then, $c_1\mathbf{x} + c_2\mathbf{y} \in W$, for some scalars c_1, c_2 as W is a subspace. W is T -invariant $\Rightarrow T(\mathbf{x}) \in W, T(\mathbf{y}) \in W, T(c_1\mathbf{x} + c_2\mathbf{y}) \in W$. Thus, $T_W(\mathbf{x}) = T(\mathbf{x}), T_W(\mathbf{y}) = T(\mathbf{y})$ and $T_W(c_1\mathbf{x} + c_2\mathbf{y}) = T(c_1\mathbf{x} + c_2\mathbf{y})$. Now, we have the following:

$$T_W(c_1\mathbf{x} + c_2\mathbf{y}) = T(c_1\mathbf{x} + c_2\mathbf{y}) = c_1T(\mathbf{x}) + c_2T(\mathbf{y}) = c_1T_W(\mathbf{x}) + c_2T_W(\mathbf{y}).$$

The above is true for any scalars c_1, c_2 and for any $\mathbf{x}, \mathbf{y} \in W$. Thus, T_W is linear.

(ii) Let $\mathbf{w} \in N(T_W)$. Since $N(T_W) \subset W, \mathbf{w} \in W$. Further, $T_W(\mathbf{w}) = \mathbf{0}$. But by definition, $T_W(\mathbf{w}) = T(\mathbf{w}) \Rightarrow T(\mathbf{w}) = \mathbf{0} \Rightarrow \mathbf{w} \in N(T)$. Thus, $\mathbf{w} \in N(T) \cap W \Rightarrow N(T_W) \subset N(T) \cap W$.

Let $\mathbf{u} \in N(T) \cap W$. Then, $\mathbf{u} \in N(T) \Rightarrow T(\mathbf{u}) = \mathbf{0}$. Since $\mathbf{u} \in W$ and W is T -invariant, $T_W(\mathbf{u}) = T(\mathbf{u}) = \mathbf{0} \in W \Rightarrow \mathbf{u} \in N(T_W)$. Hence, $N(T) \cap W \subset N(T_W)$.

Therefore, $N(T_W) = N(T) \cap W$. Hence, proved.