




4-1 True-False Exercises

- a. A vector is any element of a vector space. ○
- b. A vector space must contain at least two vectors. ✗
- c. If \mathbf{u} is a vector and k is a scalar such that $k\mathbf{u} = \mathbf{0}$, then it must be true that $k = 0$. ✗
- d. The set of positive real numbers is a vector space if vector addition and scalar multiplication are the usual operations of addition and multiplication of real numbers. ✗
- e. In every vector space the vectors $(-1)\mathbf{u}$ and $-\mathbf{u}$ are the same. ○

- (a) True. This is a part of Definition 1.
- (b) False. Example 1 discusses a vector space containing only one vector.
- (c) False. By part (d) of Theorem 4.1.1, if $k\mathbf{u} = \mathbf{0}$ then $k = 0$ or $\mathbf{u} = \mathbf{0}$.
- (d) False. Axiom 6 fails to hold if $k < 0$. (Also, Axiom 4 fails to hold.)
- (e) True. This follows from part (c) of Theorem 4.1.1.
- (f) False. This function must have a value of zero at *every* point in $(-\infty, \infty)$.

4-2 True-False Exercises


- a. Every subspace of a vector space is itself a vector space. ○
- b. Every vector space is a subspace of itself. ○
- c. Every subset of a vector space V that contains the zero vector in V is a subspace of V . ✗
- d. The kernel of a matrix transformation $T_A : R^n \rightarrow R^m$ is a subspace of R^m . ✗
- e. The solution set of a consistent linear system $A\mathbf{x} = \mathbf{b}$ of m equations in n unknowns is a subspace of R^n . ✗


- f.** The intersection of any two subspaces of a vector space V is a subspace of V . 
- g.** The union of any two subspaces of a vector space V is a subspace of V . 
- h.** The set of upper triangular $n \times n$ matrices is a subspace of the vector space of all $n \times n$ matrices. 

- (a) True. This follows from Definition 1.
- (b) True.
- (c) False. The set of all nonnegative real numbers is a subset of the vector space R containing 0 , but it is not closed under scalar multiplication.
- (d) False. By Theorem 4.2.4, the kernel of $T_A : R^n \rightarrow R^m$ is a subspace of R^n .
- (e) False. The solution set of a nonhomogeneous system is not closed under addition: $A\mathbf{x} = \mathbf{b}$ and $A\mathbf{y} = \mathbf{b}$ do not imply $A(\mathbf{x} + \mathbf{y}) = \mathbf{b}$.
- (f) True. This follows from Theorem 4.2.2.
- (g) False. Consider $W_1 = \text{span}\{(1,0)\}$ and $W_2 = \text{span}\{(0,1)\}$. The union of these sets is not closed under vector addition, e.g. $(1,0) + (0,1) = (1,1)$ is outside the union.
- (h) True. This set contains at least one matrix (e.g., I_n). A sum of two upper triangular matrices is also upper triangular, therefore the set is closed under addition. A scalar multiple of an upper triangular matrix is also upper triangular, hence the set is closed under scalar multiplication.

4-3 True-False Exercises






- a. An expression of the form $k_1\mathbf{v}_1 + k_2\mathbf{v}_2 + \cdots + k_r\mathbf{v}_r$ is called a linear combination. ☐
- b. The span of a single vector in R^2 is a line. ☒
- c. The span of two vectors in R^3 is a plane. ☒
- d. The span of a nonempty set S of vectors in V is the smallest subspace of V that contains S . ☐
- e. The span of any finite set of vectors in a vector space is closed under addition and scalar multiplication. ☐




f. Two subsets of a vector space V that span the same subspace of V must be equal. 

g. The polynomials $x - 1$, $(x - 1)^2$, and $(x - 1)^3$ span P_3 . 

- (a) True.
- (b) False. The span of the zero vector is just the zero vector.
- (c) False. For example the vectors $(1,1,1)$ and $(2,2,2)$ span a line.
- (d) True.
- (e) True. This follows from part (a) of Theorem 4.2.1.
- (f) False. For any nonzero vector \mathbf{v} in a vector space V , both $\{\mathbf{v}\}$ and $\{2\mathbf{v}\}$ span the same subspace of V .
- (g) False. The constant polynomial $p(x)=1$ cannot be represented as a linear combination of these, since at $x=1$ all three are zero, whereas $p(1)=1$.

4-4 True-False Exercises

- a. A set containing a single vector is linearly independent. 
- b. No linearly independent set contains the zero vector. 
- c. Every linearly dependent set contains the zero vector. 
- d. If the set of vectors $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ is linearly independent, then $\{k\mathbf{v}_1, k\mathbf{v}_2, k\mathbf{v}_3\}$ is also linearly independent for every nonzero scalar k . 
- e. If $\mathbf{v}_1, \dots, \mathbf{v}_n$ are linearly dependent nonzero vectors, then at least one vector \mathbf{v}_k is a unique linear combination of $\mathbf{v}_1, \dots, \mathbf{v}_{k-1}$. 

- f. The set of 2×2 matrices that contain exactly two 1's and two 0's is a linearly independent set in M_{22} . 
- g. The three polynomials $(x - 1)(x + 2)$, $x(x + 2)$, and $x(x - 1)$ are linearly independent. 
- h. The functions f_1 and f_2 are linearly dependent if there is a real number x such that $k_1 f_1(x) + k_2 f_2(x) = 0$ for some scalars k_1 and k_2 . 

- (a) False. By part (b) of Theorem 4.4.2, a set containing a single *nonzero* vector is linearly independent.
- (b) True. This follows directly from Definition 1.
- (c) False. For instance $\{(1,1),(2,2)\}$ is a linearly dependent set that does not contain $(0,0)$.
- (d) True. If $a\mathbf{v}_1 + b\mathbf{v}_2 + c\mathbf{v}_3 = \mathbf{0}$ has only one solution $a = b = c = 0$ then $a(kv_1) + b(kv_2) + c(kv_3) = k(a\mathbf{v}_1 + b\mathbf{v}_2 + c\mathbf{v}_3)$ can only equal $\mathbf{0}$ when $a = b = c = 0$ as well.
- (e) True. Since the vectors must be nonzero, $\{\mathbf{v}_1\}$ must be linearly independent.

Let us begin adding vectors to the set until the set $\{\mathbf{v}_1, \dots, \mathbf{v}_k\}$ becomes linearly dependent, therefore, by construction, $\{\mathbf{v}_1, \dots, \mathbf{v}_{k-1}\}$ is linearly independent. The equation $c_1\mathbf{v}_1 + \dots + c_{k-1}\mathbf{v}_{k-1} + c_k\mathbf{v}_k = \mathbf{0}$ must have a solution with $c_k \neq 0$, therefore $\mathbf{v}_k = -\frac{c_1}{c_k}\mathbf{v}_1 - \dots - \frac{c_{k-1}}{c_k}\mathbf{v}_{k-1}$. Let us assume there exists another representation $\mathbf{v}_k = d_1\mathbf{v}_1 + \dots + d_{k-1}\mathbf{v}_{k-1}$. Subtracting both sides yields $0 = \left(d_1 + \frac{c_1}{c_k}\right)\mathbf{v}_1 + \dots + \left(d_{k-1} + \frac{c_{k-1}}{c_k}\right)\mathbf{v}_{k-1}$. By linear independence of $\{\mathbf{v}_1, \dots, \mathbf{v}_{k-1}\}$, we must have $d_1 = -\frac{c_1}{c_k}, \dots, d_{k-1} = -\frac{c_{k-1}}{c_k}$, which shows that \mathbf{v}_k is a *unique* linear combination of $\mathbf{v}_1, \dots, \mathbf{v}_{k-1}$.






(f) False. The set $\left\{ \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \right\}$ is linearly dependent since

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

(g) True. Requiring that for all x values $a(x-1)(x+2) + bx(x+2) + cx(x-1) = 0$ holds true implies that the equality must be true for any specific x value. Setting $x=0$ yields $a=0$. Likewise, $x=1$ implies $b=0$, and $x=-2$ implies $c=0$. Since $a=b=c=0$ is required, we conclude that the three given polynomials are linearly independent.

(h) False. The functions f_1 and f_2 are linearly dependent if there exist scalars k_1 and k_2 , not both equal 0, such that $k_1 f_1(x) + k_2 f_2(x) = 0$ for all real numbers x .

4-5 True-False Exercises

- a. If $V = \text{span}\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$, then $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ is a basis for V . 
- b. Every linearly independent subset of a vector space V is a basis for V . 
- c. If $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ is a basis for a vector space V , then every vector in V can be expressed as a linear combination of $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$. 
- d. The coordinate vector of a vector \mathbf{x} in R^n relative to the standard basis for R^n is \mathbf{x} . 
- e. Every basis of P_4 contains at least one polynomial of degree 3 or less. 

- (a) False. The set must also be linearly independent.
- (b) False. The subset must also span V .
- (c) True. This follows from Theorem 4.5.1.
- (d) True. For any vector $\mathbf{v} = (a_1, \dots, a_n)$ in R^n , we have $\mathbf{v} = a_1 \mathbf{e}_1 + \dots + a_n \mathbf{e}_n$ therefore the coordinate vector of \mathbf{v} with respect to the standard basis $S = \{\mathbf{e}_1, \dots, \mathbf{e}_n\}$ is $(\mathbf{v})_S = (a_1, \dots, a_n) = \mathbf{v}$.
- (e) False. For instance, $\{1 + t^4, t + t^4, t^2 + t^4, t^3 + t^4, t^4\}$ is a basis for P_4 .






4-6 True-False Exercises

- a. The zero vector space has dimension zero. ☐
- b. There is a set of 17 linearly independent vectors in R^{17} . ☐
- c. There is a set of 11 vectors that span R^{17} . ☒
- d. Every linearly independent set of five vectors in R^5 is a basis for R^5 . ☐
- e. Every set of five vectors that spans R^5 is a basis for R^5 . ☐
- f. Every set of vectors that spans R^n contains a basis for R^n . ☐

- g.** Every linearly independent set of vectors in R^n is contained in some basis for R^n . ○
- h.** There is a basis for M_{22} consisting of invertible matrices. ○
- i.** If A has size $n \times n$ and $I_n, A, A^2, \dots, A^{n^2}$ are distinct matrices, then $\{I_n, A, A^2, \dots, A^{n^2}\}$ is a linearly dependent set. ○
- j.** There are at least two distinct three-dimensional subspaces of P_2 . ✗
- k.** There are only three distinct two-dimensional subspaces of P_2 . ✗

- (a) True.
- (b) True. For instance, $\mathbf{e}_1, \dots, \mathbf{e}_{17}$.
- (c) False. This follows from Theorem 4.6.2(b).
- (d) True. This follows from Theorem 4.6.4.
- (e) True. This follows from Theorem 4.6.4.
- (f) True. This follows from Theorem 4.6.5(a).
- (g) True. This follows from Theorem 4.6.5(b).
- (h) True. For instance, invertible matrices $\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$ form a basis for M_{22} .
- (i) True. The set has $n^2 + 1$ matrices, which exceeds $\dim(M_m) = n^2$.
- (j) False. This follows from Theorem 4.6.6(c).
- (k) False. For instance, for any constant c , $\text{span}\{x - c, x^2 - c^2\}$ is a two-dimensional subspace of P_2 consisting of all polynomials in P_2 for which $p(c) = 0$. Clearly, there are infinitely many different subspaces of this type.






4-8 True-False Exercises

- a. The span of $\mathbf{v}_1, \dots, \mathbf{v}_n$ is the column space of the matrix whose column vectors are $\mathbf{v}_1, \dots, \mathbf{v}_n$. 
- b. The column space of a matrix A is the set of solutions of $A\mathbf{x} = \mathbf{b}$. 
- c. If R is the reduced row echelon form of A , then those column vectors of R that contain the leading 1's form a basis for the column space of A . 
- d. The set of nonzero row vectors of a matrix A is a basis for the row space of A . 
- e. If A and B are $n \times n$ matrices that have the same row space, then A and B have the same column space. 

- f. If E is an $m \times m$ elementary matrix and A is an $m \times n$ matrix, then the null space of EA is the same as the null space of A . ○
- g. If E is an $m \times m$ elementary matrix and A is an $m \times n$ matrix, then the row space of EA is the same as the row space of A . ○
- h. If E is an $m \times m$ elementary matrix and A is an $m \times n$ matrix, then the column space of EA is the same as the column space of A . ✗
- i. The system $A\mathbf{x} = \mathbf{b}$ is inconsistent if and only if \mathbf{b} is not in the column space of A . ○
- j. There is an invertible matrix A and a singular matrix B such that the row spaces of A and B are the same. ✗

- (a) True.
- (b) False. The column space of A is the space spanned by all column vectors of A .
- (c) False. Those column vectors form a basis for the column space of R .
- (d) False. This would be true if A were in row echelon form.
- (e) False. For instance $A = \begin{bmatrix} 1 & 0 \\ 2 & 0 \end{bmatrix}$ and $B = \begin{bmatrix} 1 & 0 \\ 3 & 0 \end{bmatrix}$ have the same row space, but different column spaces.
- (f) True. This follows from Theorem 4.8.3.
- (g) True. This follows from Theorem 4.8.3.
- (h) False. Elementary row operations generally can change the column space of a matrix.
- (i) True. This follows from Theorem 4.8.1.
- (j) False. Let both A and B be $n \times n$ matrices. By Theorem 4.8.3, row operations do not change the row space of a matrix. An invertible matrix can be reduced to I thus its row space is always R^n . On the other hand, a singular matrix cannot be reduced to identity matrix - at least one row in its reduced row echelon form is made up of zeros. Consequently, its row space is spanned by fewer than n vectors, therefore the dimension of this space is less than n .

4-9 True-False Exercises

- a. Either the row vectors or the column vectors of a square matrix are linearly independent. 
- b. A matrix with linearly independent row vectors and linearly independent column vectors is square. 
- c. The nullity of a nonzero $m \times n$ matrix is at most m . 
- d. Adding one additional column to a matrix increases its rank by one. 
- e. The nullity of a square matrix with linearly dependent rows is at least one. 

- f. If A is square and $A\mathbf{x} = \mathbf{b}$ is inconsistent for some vector \mathbf{b} , then the nullity of A is zero. \times
- g. If a matrix A has more rows than columns, then the dimension of the row space is greater than the dimension of the column space. \times
- h. If $\text{rank}(A^T) = \text{rank}(A)$, then A is square. \times
- i. There is no 3×3 matrix whose row space and null space are both lines in 3-space. \bigcirc
- j. If V is a subspace of R^n and W is a subspace of V , then W^\perp is a subspace of V^\perp . \times

- (a) False. For instance, in, neither row vectors nor column $\begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix}$ vectors are linearly independent.
- (b) True. In an $m \times n$ matrix, if $m < n$ then by Theorem 4.6.2(a), the n columns in R^m must be linearly dependent. If $m > n$, then by the same theorem, the m rows in R^n must be linearly dependent. We conclude that $m = n$.
- (c) False. The nullity in an $m \times n$ matrix is at most n .
- (d) False. For instance, if the column contains all zeros, adding it to a matrix does not change the rank.
- (e) True. In an $n \times n$ matrix A with linearly dependent rows, $\text{rank}(A) \leq n - 1$.
By Formula (4), $\text{nullity}(A) = n - \text{rank}(A) \geq 1$.
- (f) False. By Theorem 4.9.7, the nullity must be nonzero.
- (g) False. This follows from Theorem 4.9.1.
- (h) False. By Theorem 4.9.4, $\text{rank}(A^T) = \text{rank}(A)$ for any matrix A .
- (i) True. Since each of the two spaces has dimension 1, these dimensions would add up to 2 instead of 3 as required by Formula (4).
- (j) False. For instance, if $n = 3$, $V = \text{span}\{\mathbf{i}, \mathbf{j}\}$ (the xy -plane), and $W = \text{span}\{\mathbf{i}\}$ (the x -axis) then $W^\perp = \text{span}\{\mathbf{j}, \mathbf{k}\}$ (the yz -plane) is not a subspace of $V^\perp = \text{span}\{\mathbf{k}\}$ (the z -axis).
(Note that it is true that V^\perp is a subspace of W^\perp .)