

Medici Solutions (incomplete)

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Chapter 3

2. a) If $m = n$, and \mathbf{AB} is invertible,

$$(\mathbf{AB})^{-1}\mathbf{AB} = \mathbf{I}$$

$$(\mathbf{B}^{-1}\mathbf{A}^{-1})\mathbf{AB} = \mathbf{I}$$

$$(\mathbf{A}^{-1}\mathbf{B}^{-1})(\mathbf{B}^{-1}\mathbf{A}^{-1})\mathbf{AB} = \mathbf{A}^{-1}\mathbf{B}^{-1}$$

$$(\mathbf{A}^{-1}\mathbf{B}^{-1}\mathbf{B}^{-1}\mathbf{A}^{-1})\mathbf{AB}(\mathbf{BA}) = (\mathbf{A}^{-1}\mathbf{B}^{-1})(\mathbf{BA})$$

Since any invertible square matrix \mathbf{X} has the property $\mathbf{X}^{-1}\mathbf{X} = \mathbf{XX}^{-1} = \mathbf{I}$, the left side simplifies to \mathbf{I} so that $\mathbf{I} = (\mathbf{A}^{-1}\mathbf{B}^{-1})\mathbf{BA}$, i.e., $\mathbf{A}^{-1}\mathbf{B}^{-1}$ is the inverse of \mathbf{BA} .

b) Note that $m < n$ because if $m > n$ then by the Corollary to Theorem V, $\mathbf{Bx} = \mathbf{0}$ has nontrivial solutions and thus $\mathbf{ABx} = \mathbf{0}$ has nontrivial solutions, i.e., \mathbf{AB} is not invertible. But if $m < n$, then $\mathbf{Ax} = \mathbf{0}$ has nontrivial solutions and thus $\mathbf{BAx} = \mathbf{0}$ also has nontrivial solutions, i.e., \mathbf{BA} is not invertible.

3. If $m < n$ then $\mathbf{Ax} = \mathbf{0}$ has nontrivial solutions by the Corollary to Theorem V. But then $\mathbf{LAx} = \mathbf{0}$ also has nontrivial solutions, so $\mathbf{LA} = \mathbf{I}$ is impossible.

4. Yes

Putting the system in matrix form ($\mathbf{Ax} = \mathbf{b}$), we have

$$\begin{bmatrix} 1 & -3 & 2 \\ 2 & 1 & -1 \\ 3 & -2 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 4 \\ 1 \\ 5 \end{bmatrix}, \text{ so } \text{rref}[\mathbf{A}|\mathbf{b}] = \begin{bmatrix} 1 & 0 & -1/7 & 1 \\ 0 & 1 & -5/7 & -1 \\ 0 & 0 & 0 & 0 \end{bmatrix}, \text{ which has a zero row and hence infinite solutions.}$$

5. Yes

\mathbf{A} is square and has a unique solution to $\mathbf{Ax} = \mathbf{b}$ for at least one $\mathbf{b} \in {}^3\mathbb{R}$, so by the Corollary to Theorem VI it is invertible, i.e., a product of elementary matrices.

6. No

\mathbf{A} cannot be invertible because it does not have a unique solution to $\mathbf{Ax} = \mathbf{b}$ for every $\mathbf{b} \in {}^3\mathbb{R}$.

7. See 2b).

Chapter 4

1. No

Axiom MIV fails, since $1(x_1, x_2) \neq (x_1, x_2)$ when $x_2 \neq 0$.

Chapter 5

1. a) No

SII fails, consider $\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$.

b) Yes

SI: $f = 0$ is the zero function.

SII: $\int_0^1 (f+g)(x) dx = 0$ for all $f, g \in U$.

SIII: If $\int_0^1 f(x) dx = 0$ then $\lambda \int_0^1 f(x) dx = 0$ for all $\lambda \in \mathbb{R}$ and $f \in U$.

2. \implies If i) is true, U is not a subspace because SI fails. Hence U must be nonempty. Further, if ii) is true, we can set $\lambda = 1$ and note that all $\mathbf{u}_1 + \mathbf{u}_2$ are in U (i.e. SII holds), and we can also set $\mathbf{x}_1 = \mathbf{0}$ and note that all $\lambda \mathbf{u}_2$ are in U (i.e. SIII holds).

\Leftarrow If U is a subspace, it contains the zero vector (SI) and hence is nonempty. Further, for all $\mathbf{u}_1, \mathbf{u}_2 \in U$, U must contain all $\mathbf{u}_1 + \lambda \mathbf{u}_2$ by SII and SIII.

4. No

SII fails, see 1a).

5. No

SII fails, consider $\begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix}$.

$$8. \mathbf{u} = \frac{1}{2}((\mathbf{u} + \mathbf{v}) + -(\mathbf{v} + \mathbf{w}) + (\mathbf{w} + \mathbf{u}))$$

$$\mathbf{v} = \frac{1}{2}((\mathbf{u} + \mathbf{v}) + (\mathbf{v} + \mathbf{w}) + -(\mathbf{w} + \mathbf{u}))$$

$$\mathbf{w} = \frac{1}{2}(-(\mathbf{u} + \mathbf{v}) + (\mathbf{v} + \mathbf{w}) + (\mathbf{w} + \mathbf{u}))$$

Thus, $\text{span}\{\mathbf{u}, \mathbf{v}, \mathbf{w}\} \supseteq \text{span}\{\mathbf{u} + \mathbf{v}, \mathbf{v} + \mathbf{w}, \mathbf{w} + \mathbf{u}\}$.

$$\mathbf{u} + \mathbf{v} = \mathbf{u} + \mathbf{v}$$

$$\mathbf{v} + \mathbf{w} = \mathbf{v} + \mathbf{w}$$

$$\mathbf{w} + \mathbf{u} = \mathbf{w} + \mathbf{u}$$

Thus, $\text{span}\{\mathbf{u}, \mathbf{v}, \mathbf{w}\} \subseteq \text{span}\{\mathbf{u} + \mathbf{v}, \mathbf{v} + \mathbf{w}, \mathbf{w} + \mathbf{u}\}$.

9. Yes

$$0(1,1) + 1(1,2) = (1,2).$$

10. No

Axiom SII fails, consider $\begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$.

11. \mathbb{R} : SI: $0 \in \mathbb{R}$

SII: For all $\mathbf{x}, \mathbf{y} \in \mathbb{R}$, $\mathbf{x} + \mathbf{y} \in \mathbb{R}$

SIII: For all $\mathbf{x} \in \mathbb{R}$ and any scalar $\lambda \in \mathbb{R}$, $\lambda \mathbf{x} \in \mathbb{R}$

$\{0\}$: SI: $0 \in \mathbb{R}$

SII: $0 + 0 \in \{0\}$

SIII: For any scalar $\lambda \in \mathbb{R}$, $\lambda 0 \in \{0\}$

For any other finite non-empty set of n elements $\{x_1, x_2, \dots, x_n\}$ in \mathbb{R} , SIII fails because some scalar $\lambda \in \mathbb{R}$ exists such that λx_1 is not in the set.

Chapter 6

1. \implies If $\{p, q, pq\}$ is linearly independent, $\lambda_1 p + \lambda_2 q + \lambda_3 pq = \mathbf{0}$ has only the trivial solution. But if $\deg p = 0$, $\lambda_1 p + \lambda_2 q + \lambda_3 pq = \mathbf{0}$ has the solution $\lambda_1 = 0, \lambda_2 = p, \lambda_3 = -1$ which is a nontrivial solution. Likewise if $\deg q = 0$, $\lambda_1 p + \lambda_2 q + \lambda_3 pq = \mathbf{0}$ has the solution $\lambda_1 = q, \lambda_2 = 0, \lambda_3 = -1$. Hence $\deg p, \deg q \neq 0$, i.e., $\deg p, \deg q \geq 1$ is required.

\Leftarrow If $\deg p, \deg q \geq 1$, then $\deg(-\lambda_3 pq) > \deg(\lambda_1 p + \lambda_2 q)$ for any choice of $\lambda_1, \lambda_2, \lambda_3$, and since it was also given that $\{p, q\}$ is linearly independent, $\lambda_1 p + \lambda_2 q + \lambda_3 pq = \mathbf{0}$ has only the trivial solution, i.e., $\{p, q, pq\}$ is linearly independent.

2. Reducing U to its basis, we have $U = \text{span}\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ and $W = \text{span}\{\mathbf{v}_1, \dots, \mathbf{v}_n, \mathbf{v}\}$, where $1 \leq n \leq k$.

If $\mathbf{v} \in \text{span}\{\mathbf{v}_1, \dots, \mathbf{v}_n\} = U$, then W can also be reduced to its basis so that $W = \text{span}\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$, i.e., $U = W$ and thus $\dim U = \dim W$.

If $\mathbf{v} \notin \text{span}\{\mathbf{v}_1, \dots, \mathbf{v}_n\} = U$, then $\{\mathbf{v}_1, \dots, \mathbf{v}_n, \mathbf{v}\}$ is a linearly independent set which spans W and is hence a basis for W . In this case, $\dim W = n + 1$, i.e., $\dim W = \dim U + 1$.

As no possibilities other than the two described above exist, we conclude that either $\dim U = \dim W$ or $\dim U + 1 = \dim W$.

3. Since $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3, \mathbf{v}_4\}$ is linearly independent, $\lambda_1 \mathbf{v}_1 + \lambda_2 \mathbf{v}_2 + \lambda_3 \mathbf{v}_3 + \lambda_4 \mathbf{v}_4 = \mathbf{0}$ has only the trivial solution.

But $\beta_1 \mathbf{v}_1 + \beta_2(\mathbf{v}_1 + \mathbf{v}_2) + \beta_3(\mathbf{v}_1 + \mathbf{v}_2 + \mathbf{v}_3) + \beta_4(\mathbf{v}_1 + \mathbf{v}_2 + \mathbf{v}_3 + \mathbf{v}_4) = \mathbf{0}$ is equivalent to

$$(\beta_1 + \beta_2 + \beta_3 + \beta_4)\mathbf{v}_1 + (\beta_2 + \beta_3 + \beta_4)\mathbf{v}_2 + (\beta_3 + \beta_4)\mathbf{v}_3 + \beta_4 \mathbf{v}_4 = \mathbf{0}, \text{ which implies}$$

$$\beta_1 + \beta_2 + \beta_3 + \beta_4 = 0$$

$$\beta_2 + \beta_3 + \beta_4 = 0$$

$$\beta_3 + \beta_4 = 0$$

$\beta_4 = 0$, i.e., $\beta_1 = \beta_2 = \beta_3 = \beta_4 = 0$. Thus, $\{\mathbf{v}_1, \mathbf{v}_1 + \mathbf{v}_2, \mathbf{v}_1 + \mathbf{v}_2 + \mathbf{v}_3, \mathbf{v}_1 + \mathbf{v}_2 + \mathbf{v}_3 + \mathbf{v}_4\}$ is linearly independent.

5. If V is an n -dimensional vector space, where n is finite, then any nonzero vector \mathbf{v} in V is linearly independent of $n - 1$ other nonzero vectors in V . Hence a basis for V can be constructed by adding $n - 1$ nonzero vectors to the set $\{\mathbf{v}\}$ such that each vector in the set cannot be produced as a linear combination of other vectors in the set.

6. If \mathbf{A} is not invertible, then its nullspace (the set of vectors such that $\mathbf{A}\mathbf{v} = \mathbf{0}$) has a dimension of at least one. But since $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ is a linearly independent set containing n vectors, and \mathbf{A} 's rowspace (the orthogonal complement to its nullspace) has a dimension of less than n , there exists at least one $\mathbf{v}_i \in \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ such that \mathbf{v}_i is in the nullspace of \mathbf{A} , so at least one of $\{\mathbf{u}_1, \dots, \mathbf{u}_n\}$ will end up being $\mathbf{0}$. Since any set containing $\mathbf{0}$ is linearly dependent, we conclude that $\{\mathbf{u}_1, \dots, \mathbf{u}_n\}$ must be linearly dependent.

7. No

The set is not a basis if $f_0 = 0$.

9. a) S is a subspace since it contains the zero vector and inherits the vector addition and scalar multiplication properties of V . Also, the maximum number of linearly independent vectors in S is k . Therefore, $\{\mathbf{v}_1, \dots, \mathbf{v}_k\}$ is a basis for S , and $S = \text{span}\{\mathbf{v}_1, \dots, \mathbf{v}_k\}$. It follows that $S \subseteq \text{span}\{\mathbf{v}_1, \dots, \mathbf{v}_k\}$.

b) $\{\mathbf{v}_1, \dots, \mathbf{v}_k, \mathbf{x}\}$ is a linearly independent set. Thus $\lambda_1 \mathbf{v}_1 + \dots + \lambda_k \mathbf{v}_k + \lambda_{k+1} \mathbf{v}_{k+1} = \mathbf{0}$ has only the trivial solution. But $\beta_1(\mathbf{v}_1 + \mathbf{x}) + \dots + \beta_k(\mathbf{v}_k + \mathbf{x}) = \mathbf{0}$ is equivalent to $\beta_1 \mathbf{v}_1 + \dots + \beta_k \mathbf{v}_k + (\beta_1 + \dots + \beta_k)\mathbf{x} = \mathbf{0}$, so the latter equation must have only the trivial solution as well and hence be linearly independent.

10. If $U \cap W \neq \{\mathbf{0}\}$, $\dim(U \cap W)$ must be greater or equal to 1 since both U and W are subspaces and must contain the zero vector. If $\dim(U \cap W) = 2$ as well, then $U = W$, since U 's basis has 2 linearly independent vectors, and W 's basis contains 2 linearly independent vectors, and the set containing these four vectors is linearly dependent. But it was given in the question that $U \neq W$, so $\dim(U \cap W) = 1$.

11. No

$$\sin^2 x + \cos^2 x = 1.$$

12. No

On the interval $\mathbf{F}[0, 1]$, $\text{span}\{x, |x|\} = \text{span}\{x\}$.

13. Since a matrix with linearly independent columns has a unique solution for every $\mathbf{A}\mathbf{x} = \mathbf{b}$, and $\mathbf{A}\mathbf{0} = \mathbf{0}$, we conclude that $\mathbf{x} = \mathbf{0}$ is the unique solution.

14. a) SI: $\mathbf{x} = \mathbf{0} \in W$

SII: $\mathbf{u}^T \mathbf{x} = \mathbf{0}$ can be rewritten as $u_1 x_1 + \dots + u_n x_n = 0$ where $\{u_1, \dots, u_n\}$ and $\{x_1, \dots, x_n\}$ are the individual entries of \mathbf{u}^T and \mathbf{x} respectively. If $\mathbf{x}, \mathbf{y} \in W$, then $u_1 x_1 + \dots + u_n x_n = 0$ and $u_1 y_1 + \dots + u_n y_n = 0$, so $u_1 x_1 + \dots + u_n x_n + u_1 y_1 + \dots + u_n y_n = u_1(x_1 + y_1) + \dots + u_n(x_n + y_n) = 0$. But this equation can be rewritten as $\mathbf{u}^T(\mathbf{x} + \mathbf{y}) = 0$, so $\mathbf{u}^T(\mathbf{x} + \mathbf{y}) = \mathbf{0}$ for all $\mathbf{x}, \mathbf{y} \in W$.

SIII: If $\mathbf{x} \in W$, then $u_1 x_1 + \dots + u_n x_n = 0$, so $\lambda(u_1 x_1 + \dots + u_n x_n) = 0$ as well. But this equation can be rewritten as $\mathbf{u}^T \lambda \mathbf{x}$, so $\lambda \mathbf{x} \in W$ for all $\mathbf{x} \in W$.

b) \mathbf{u}^T has n columns, so $n = \text{rank} \mathbf{u}^T + \text{nullity} \mathbf{u}^T$ and if $\mathbf{u} \neq \mathbf{0}$ then $\text{rank} \mathbf{u}^T$ is always equal to 1, so $\text{nullity} \mathbf{u}^T = \dim W = n - 1$.

c) If $\mathbf{u} \neq \mathbf{0}$, then $\mathbf{u}^T \mathbf{u} \neq 0$ by Medici. Hence $\{\mathbf{u}, \mathbf{w}_1, \dots, \mathbf{w}_{n-1}\}$ is a linearly independent set containing n vectors. But then this set is a basis for ${}^n \mathbb{R}$ since all of its elements are in ${}^n \mathbb{R}$.

15. Note that the equation $\lambda_1 \mathbf{u}_1 + \dots + \lambda_5 \mathbf{u}_5 + \beta_1 \mathbf{w}_1 + \dots + \beta_{10} \mathbf{w}_{10} = \mathbf{0}$ is equivalent to $\lambda_1 \mathbf{u}_1 + \dots + \lambda_5 \mathbf{u}_5 = -\beta_1 \mathbf{w}_1 - \dots - \beta_{10} \mathbf{w}_{10}$. But since $U \cap W = \{\mathbf{0}\}$, only the trivial solution exists since it is impossible to create linear combinations of vectors in U out of vectors in W and vice versa. Hence $\{\mathbf{u}_1, \dots, \mathbf{u}_5, \mathbf{w}_1, \dots, \mathbf{w}_{10}\} \in V$ is a linearly independent set containing 15 vectors, which means its span has dimension 15.

16. Yes

If $\{\mathbf{x}\} \cup S$ is linearly dependent then \mathbf{x} can be formed as a linear combination of elements in S , so $\mathbf{x} \in \text{span} S$.

18. No

Any set containing $\mathbf{0}$ is not linearly independent.

19. \implies If $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ is linearly dependent, then $\sum_{i=1}^n \lambda_i \mathbf{v}_i = 0$ has nontrivial solutions. Therefore some $\mathbf{v}_k \in \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ exists such that $\sum_{i=1}^{n-1} \lambda_i \mathbf{v}_i$, where $i \neq k$, $= -\lambda_k \mathbf{v}_k$, i.e., $\mathbf{v}_k = \sum_{i=1}^{n-1} (\lambda_i / -\lambda_k) \mathbf{v}_i$.

\Leftarrow If some \mathbf{v}_k is a linear combination of the other members in the set, then $-\lambda_k \mathbf{v}_k = \sum_{i=1}^{n-1} \lambda_i \mathbf{v}_i$ where $i \neq k$ and λ_i, λ_k are not all 0, so $\sum_{i=1}^{n-1} \lambda_i \mathbf{v}_i$ (where $i \neq k$) $+ \lambda_k \mathbf{v}_k = 0$ has non-trivial solutions, i.e., the set $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ is linearly dependent.

22. Since $\dim^n \mathbb{R}^n = n^2$, and there are $n^2 + 1$ terms in this equation, nontrivial solutions exist by FTOLA.

23. For $\dim(\text{span}S) = 3$, S must be a linearly independent set, i.e., $\lambda_1(1+x) + \lambda_2(1+kx+x^2) + \lambda_3(1+2x^2) = 0$ must only have the trivial solution. This equation is equivalent to $(\lambda_1 + \lambda_2 + \lambda_3)1 + (\lambda_1 + k\lambda_2)x + (\lambda_2 + 2\lambda_3)x^2 = 0$, and since $\{1, x, x^2\}$ is a linearly independent set,

$$\lambda_1 + \lambda_2 + \lambda_3 = 0,$$

$$\lambda_1 + k\lambda_2 = 0, \text{ and}$$

$\lambda_2 + 2\lambda_3 = 0$ are required. Solving the system of equations, we see that $k = \frac{1}{2}$ is the only value which causes this set of equations to be linearly dependent. As nontrivial solutions for $\lambda_1, \lambda_2, \lambda_3$ will exist iff the set of equations is linearly dependent, we conclude that $k \neq \frac{1}{2}$ is the only condition for $\dim(\text{span}S) = 3$.

24. \implies If $\{\mathbf{e}_1 - \mathbf{v}, \dots, \mathbf{e}_n - \mathbf{v}\}$ is a basis for V , then $\lambda_1(\mathbf{e}_1 - \mathbf{v}) + \dots + \lambda_n(\mathbf{e}_n - \mathbf{v}) = 0$ has only the trivial solution. But this equation is equivalent to $\lambda_1 \mathbf{e}_1 + \dots + \lambda_n \mathbf{e}_n = (\lambda_1 + \dots + \lambda_n) \mathbf{v}$ which in turn is equivalent to $\frac{\lambda_1 \mathbf{e}_1 + \dots + \lambda_n \mathbf{e}_n}{(\lambda_1 + \dots + \lambda_n)} = \mathbf{v}$. If $\mathbf{v} = \alpha_1 \mathbf{e}_1 + \dots + \alpha_n \mathbf{e}_n$ where $\alpha_1 + \dots + \alpha_n = 1$, then infinite solutions exist for $\{\lambda_1 \dots \lambda_n\}$ which means that $\{\mathbf{e}_1 - \mathbf{v}, \dots, \mathbf{e}_n - \mathbf{v}\}$ cannot be a basis. Hence $\mathbf{v} \neq \alpha_1 \mathbf{e}_1 + \dots + \alpha_n \mathbf{e}_n$ is required.

\Leftarrow If $\mathbf{v} \neq \alpha_1 \mathbf{e}_1 + \dots + \alpha_n \mathbf{e}_n$ (where $\alpha_1 + \dots + \alpha_n = 1$), then $\frac{\lambda_1 \mathbf{e}_1 + \dots + \lambda_n \mathbf{e}_n}{(\lambda_1 + \dots + \lambda_n)} = \mathbf{v}$ has no nontrivial solutions, i.e., only the trivial solution exists. It follows that $\{\mathbf{e}_1 - \mathbf{v}, \dots, \mathbf{e}_n - \mathbf{v}\}$ is a basis for V .

Chapter 7

1. No
rank ≤ 4 .

2. No
rank ≤ 7 , so nullity ≥ 5 .

3. No
Consider $\mathbf{A} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \mathbf{B} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$.

4. Yes
 $n = \text{rank} \mathbf{A}$, so nullity $\mathbf{A} = 0$.

5. No
Any matrix whose columns are linearly independent has a nullity of 0.

6. Yes
It must: rank ≤ 34 , so nullity ≥ 17 .

7. \implies If $\text{null} \mathbf{A} = \text{null} \mathbf{B}$ then $\text{rref} \mathbf{A} = \text{rref} \mathbf{B}$ and so $\mathbf{E} \dots \mathbf{E}_n \mathbf{A} = \mathbf{G} \dots \mathbf{G}_n \mathbf{B}$ for some combinations of elementary matrices $\mathbf{e} = \mathbf{E} \dots \mathbf{E}_n$ and $\mathbf{g} = \mathbf{G} \dots \mathbf{G}_n$. But $\mathbf{A} = e^{-1} \mathbf{g} \mathbf{B}$, and since $e^{-1} \mathbf{g}$ is invertible, we take $\mathbf{U} = e^{-1} \mathbf{g}$ and thereby conclude \mathbf{U} must exist.

\Leftarrow If $\mathbf{A} = \mathbf{U} \mathbf{B}$, then $\text{null} \mathbf{A} = \text{null} \mathbf{U} \mathbf{B}$. But since \mathbf{U} is invertible, $\text{null} \mathbf{U} \mathbf{B} = \text{null} \mathbf{B}$ since the only solution to any $\mathbf{U} \mathbf{B} \mathbf{x} = \mathbf{0}$ is $\mathbf{B} \mathbf{x} = \mathbf{0}$, and $\text{null} \mathbf{B}$ contains all \mathbf{x} such that $\mathbf{B} \mathbf{x} = \mathbf{0}$. Thus, $\text{null} \mathbf{A} = \text{null} \mathbf{B}$.

8. $\text{col} \mathbf{A} \mathbf{V} = \text{col} \mathbf{A}$ [Prop I].

Thus, $\text{rank} \mathbf{A} \mathbf{V} = \text{rank} \mathbf{A}$, which implies $\text{nullity} \mathbf{A} \mathbf{V} = \text{nullity} \mathbf{A}$ since \mathbf{A} and \mathbf{V} both have n columns.

9. Let $\mathbf{A} = \begin{bmatrix} 1 & 2 & 4 \\ -1 & 1 & -2 \\ 0 & 5 & 5 \\ 3 & 1 & 7 \end{bmatrix}$. Then $\text{rref} \mathbf{A} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}$.

Taking the corresponding columns with leading ones from $\text{rref}\mathbf{A}$, we conclude that a basis for $\text{col}\mathbf{A}$ is $\left\{ \begin{bmatrix} 1 \\ -1 \\ 0 \\ 3 \end{bmatrix}, \begin{bmatrix} 2 \\ 1 \\ 5 \\ 1 \end{bmatrix}, \begin{bmatrix} 4 \\ -2 \\ 5 \\ 7 \end{bmatrix} \right\}$.

10. a) $\text{rref}\mathbf{A} = \begin{bmatrix} 1 & 0 & 0 & -4 & -3 \\ 0 & 1 & 0 & 1 & 2 \\ 0 & 0 & 1 & 0 & -1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$.

Thus a basis for $\text{row}\mathbf{A}$ is $\{[1 \ 4 \ 5 \ 0 \ 0], [0 \ 2 \ 4 \ 2 \ 0], [0 \ 6 \ 7 \ 6 \ 5]\}$.

b) Taking the corresponding columns with leading ones from $\text{rref}\mathbf{A}$, we conclude that a basis for $\text{col}\mathbf{A}$ is $\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 4 \\ 2 \\ 6 \\ 0 \end{bmatrix}, \begin{bmatrix} 5 \\ 4 \\ 7 \\ 0 \end{bmatrix} \right\}$.

c) By looking at $\text{rref}\mathbf{A}$, we see that the system of equations needed to obtain solutions to $\mathbf{Ax} = \mathbf{0}$ is:

$$x_1 - 4x_4 - 3x_5 = 0$$

$$x_2 + x_4 + 2x_5 = 0$$

$$x_3 - x_5 = 0$$

Taking x_4 and x_5 as free variables, we obtain

$$\text{null}\mathbf{A} = x_4 \begin{bmatrix} 4 \\ -1 \\ 0 \\ 1 \\ 0 \end{bmatrix} + x_5 \begin{bmatrix} 3 \\ -2 \\ 1 \\ 0 \\ 1 \end{bmatrix}$$

and thus a basis for $\text{null}\mathbf{A}$ is $\left\{ \begin{bmatrix} 4 \\ -1 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 3 \\ -2 \\ 1 \\ 0 \\ 1 \end{bmatrix} \right\}$.

14. a) Since $\mathbf{Ax} = \mathbf{0}$, $\mathbf{A}^T \mathbf{Ax} = \mathbf{0}$.

Since $\mathbf{A}^T \mathbf{Ax} = \mathbf{0}$, $\mathbf{x}^T \mathbf{A}^T \mathbf{Ax} = (\mathbf{Ax})^T \mathbf{Ax} = \mathbf{0}$, which implies $\mathbf{Ax} = \mathbf{0}$ [Lemma III].

Thus, $\text{null}\mathbf{A} = \text{null}\mathbf{A}^T$.

b) By the above result, we have $n - \text{nullity}\mathbf{A} = \text{rank}\mathbf{A} = n - \text{nullity}\mathbf{A}^T = \text{rank}\mathbf{A}^T$.

15. $\mathbf{LA} = \mathbf{I}$ implies $\mathbf{L}\mathbf{Ax} = \mathbf{0}$ has only $\mathbf{x} = \mathbf{0}$ as a solution.

But \mathbf{A} has linearly dependent columns ($m < n$), which means its nullspace contains nontrivial solutions. It follows that $\mathbf{L}\mathbf{Ax} = \mathbf{0}$ must also have nontrivial solutions; hence, $\mathbf{LA} = \mathbf{I}$ is impossible.

18. \implies If $\mathbf{AB} = \mathbf{O}$, then $\mathbf{ABx} = \mathbf{0}$ for all \mathbf{Bx} , which implies that $\text{col}\mathbf{B} \subseteq \text{null}\mathbf{A}$.

\Leftarrow If $\text{col}\mathbf{B} \subseteq \text{null}\mathbf{A}$, then partitioning \mathbf{B} into its columns we get:

$$\mathbf{AB} = \mathbf{A} \begin{bmatrix} b_1 & \dots & b_n \end{bmatrix} = \begin{bmatrix} \mathbf{A}b_1 & \dots & \mathbf{A}b_n \end{bmatrix}$$

and $\{b_1, \dots, b_n\} \in \text{col}\mathbf{B} \subseteq \text{null}\mathbf{A}$ so this equation becomes $\begin{bmatrix} \mathbf{0} & \dots & \mathbf{0} \end{bmatrix} = \mathbf{O}$ implying that $\mathbf{AB} = \mathbf{O}$.

21. \implies If $\text{null}\mathbf{A} = \text{col}\mathbf{A}$, then $\text{rank}\mathbf{A} = n/2$ because $\dim\text{col}\mathbf{A} = \text{rank}\mathbf{A}$ and $n = \text{rank}\mathbf{A} + \text{nullity}\mathbf{A}$, so the only value which satisfies $\text{rank}\mathbf{A} = \text{nullity}\mathbf{A}$ is $n/2$.

If $\text{null}\mathbf{A} = \text{col}\mathbf{A}$, $\mathbf{A}^2 = \mathbf{O}$ because, partitioning \mathbf{A} into its columns $\{a_1, \dots, a_n\}$, $\mathbf{A}^2 = \mathbf{A} \begin{bmatrix} a_1 & \dots & a_n \end{bmatrix} = \begin{bmatrix} \mathbf{A}a_1 & \dots & \mathbf{A}a_n \end{bmatrix}$ which becomes $\begin{bmatrix} \mathbf{0} & \dots & \mathbf{0} \end{bmatrix} = \mathbf{O}$ since $\{a_1, \dots, a_n\} \in \text{col}\mathbf{A} \subseteq \text{null}\mathbf{A}$.

\Leftarrow If $\mathbf{A}^2 = \mathbf{O}$, then $\mathbf{A}^2\mathbf{x} = \mathbf{0}$ for all \mathbf{x} , so $\mathbf{A}(\mathbf{Ax}) = \mathbf{0}$ for all \mathbf{x} . But this means that $\text{col}\mathbf{A} \subseteq \text{null}\mathbf{A}$, since all \mathbf{b} such that $\mathbf{Ax} = \mathbf{b}$ are solutions to $\mathbf{Ab} = \mathbf{0}$. As $\text{col}\mathbf{A}$ and $\text{null}\mathbf{A}$ are both subspaces of ${}^n\mathbb{R}$, if $\text{col}\mathbf{A} \subseteq \text{null}\mathbf{A}$, we can only conclude that $\text{col}\mathbf{A} = \text{null}\mathbf{A}$ if $\dim\text{col}\mathbf{A} = \text{nullity}\mathbf{A}$ [Theorem VI, Chapter 6]. This is only possible if $\text{rank}\mathbf{A} = \text{nullity}\mathbf{A}$, i.e., $\text{rank}\mathbf{A} = n/2$.

Chapter 8

$$2. \ 1 + x = 1(1) + 1x$$

$$2 + 3x = 2(1) + 3x$$

Taking the transpose of the coefficients, we get $\mathbf{T} = \begin{bmatrix} 1 & 2 \\ 1 & 3 \end{bmatrix}$.

3. a) The standard basis of \mathbb{P}^2 is $\{1, x, x^2\}$.
 $(1/2)x^2 - (1/2)x = 0(1) - (1/2)x + (1/2)x^2$
 $1 - x^2 = 1(1) + 0x + (-1)x^2$
 $(1/2)x + (1/2)x^2 = 0 + (1/2)x + (1/2)x^2$

Taking the transpose of the coefficients, we get $\mathbf{T} = \begin{bmatrix} 0 & 1 & 0 \\ -1/2 & 0 & 1/2 \\ 1/2 & -1 & 1/2 \end{bmatrix}$. Since $\text{rref } \mathbf{T} = \mathbf{I}$, E is a linearly independent set in \mathbb{P}^2 containing three vectors, and hence a basis.

$$5. 2 \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + 3 \begin{bmatrix} 0 & 1 \\ 0 & -1 \end{bmatrix} = \begin{bmatrix} 2 & 3 \\ -3 & 2 \end{bmatrix} = c \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} + d \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}$$

where $\begin{bmatrix} c \\ d \end{bmatrix}$ are the coordinates in F . Solving for c and d , we get coordinates of $\begin{bmatrix} 5/2 \\ -1/2 \end{bmatrix}$.

Chapter 9

1. No

$$\det \mathbf{B} = 6 \det \mathbf{A}.$$

2. No

$$\text{Consider } \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}.$$

$$3. \det(2\mathbf{A}^{-1}) = 2^n \det \mathbf{A}^{-1} = 2^n / \det \mathbf{A} = -4$$

$$\det \mathbf{A} = -(2^{n-2})$$

$$-4 = \det \mathbf{A}^3 \det \mathbf{B}^{-1} = \det \mathbf{A} \det \mathbf{A} \det \mathbf{A} / \det \mathbf{B} = (2^n / (-4))^3 / \det \mathbf{B}$$

$$\det \mathbf{B} = 2^{3n-8}$$

4. No

$$\text{Consider } \mathbf{A} = \begin{bmatrix} 1 & 1 \end{bmatrix}, \mathbf{B} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}.$$

5. Since $\mathbf{A}^2 = -\mathbf{I}$ and \mathbf{A} is square,

$$\det \mathbf{A}^2 = \det \mathbf{A} \det \mathbf{A} = \det(-\mathbf{I}) = (-1)^n.$$

$$\text{Thus, } \det \mathbf{A} = (-1)^{n/2}.$$

$$7. \implies \text{Let } \mathbf{A}\mathbf{g} = \begin{bmatrix} a_1 & b_1 & g_1 \\ a_2 & b_2 & g_2 \\ a_3 & b_3 & g_3 \end{bmatrix}, \text{ where } g_i = -c_i. \text{ Note that this represents matrices } \mathbf{A} = \begin{bmatrix} a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \end{bmatrix} \text{ and } \mathbf{g} = \begin{bmatrix} g_1 \\ g_2 \\ g_3 \end{bmatrix} \text{ in}$$

augmented form. Also let $\mathbf{v} = \begin{bmatrix} x \\ y \end{bmatrix}$. For one unique solution to exist for any \mathbf{g} in $\mathbf{A}\mathbf{v} = \mathbf{g}$, it is required that $\text{rank } \mathbf{A} = \text{rank } \mathbf{A}\mathbf{g} = n$, where n is the number of columns in \mathbf{A} . This means that $\text{rank } \mathbf{A}\mathbf{g}$ must be 2, i.e. that $\mathbf{A}\mathbf{g}$ is singular and thus must have determinant 0. If $\mathbf{A}\mathbf{g}$ has determinant 0, then multiplying the third column of $\mathbf{A}\mathbf{g}$ by -1 will also yield a matrix (which is the matrix considered in the question) with determinant 0.

\Leftarrow If the matrix considered in the question has determinant 0, then $\det \mathbf{A}\mathbf{g} = 0$, so $\text{rank } \mathbf{A}\mathbf{g} = 2$ because if $\text{rank } \mathbf{A}\mathbf{g} \neq 2$ then $\text{rank } \mathbf{A} \neq 2$ by necessity which implies nonuniqueness of any solution (if it exists) to $\mathbf{A}\mathbf{v} = \mathbf{g}$. If $\text{rank } \mathbf{A}\mathbf{g} = 2$, then $\text{rank } \mathbf{A} = \text{rank } \mathbf{A}\mathbf{g} = n$ and we can thereby conclude that a unique solution (i.e. one single point) exists as the intersection between the three given lines for any g_1, g_2, g_3 . Since this conclusion holds for any g_1, g_2, g_3 , it also holds for any c_1, c_2, c_3 .

9. No

Any transformation matrix \mathbf{T} must be invertible, so $\det \mathbf{T} \neq 0$ is required.

10. No

$$\text{Consider } \mathbf{x} = \begin{bmatrix} 1 \end{bmatrix}, \mathbf{y} = \begin{bmatrix} 1 \end{bmatrix}.$$

11. Let $\{a_1, \dots, a_n\}$ be the rows of \mathbf{A} .

Since the determinant is multilinear along the rows of \mathbf{A} , $\mu^n \det \mathbf{A} = \det \begin{bmatrix} \mu a_1 \\ \vdots \\ \mu a_n \end{bmatrix} = \det \mu \mathbf{A}$.

13. Yes

If \mathbf{A} is symmetric, then $\mathbf{C}^T = \mathbf{C} = \text{adj} \mathbf{A}$.

14. Yes

Let $\mathbf{A} = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$. Then $(\text{tr} \mathbf{A})\mathbf{I} - \mathbf{A} = \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} = \text{adj} \mathbf{A}$.

18. \implies If \mathbf{A} is invertible, $\mathbf{A}^{-1} \mathbf{A} \text{adj} \mathbf{A} = \mathbf{A}^{-1} (\det \mathbf{A}) \mathbf{I}$, so $\text{adj} \mathbf{A} = \mathbf{A}^{-1} \det \mathbf{A}$. Since $\mathbf{A}^{-1} \det \mathbf{A}$ is always invertible if $\det \mathbf{A} \neq 0$, $\text{adj} \mathbf{A}$ must also be invertible.

\Leftarrow If $\text{adj} \mathbf{A}$ is invertible, $\mathbf{A} \text{adj} \mathbf{A} (\text{adj} \mathbf{A})^{-1} = (\det \mathbf{A}) \mathbf{I} (\text{adj} \mathbf{A})^{-1}$, so $\mathbf{A} = (\det \mathbf{A}) (\text{adj} \mathbf{A})^{-1}$. If \mathbf{A} is not invertible, $\det \mathbf{A} = 0$ and so the only solution to this equation is $\mathbf{A} = \mathbf{O}$. But it was given that $\mathbf{A} \neq \mathbf{O}$, so \mathbf{A} must also be invertible.

Chapter 10

1. Yes

Since all of \mathbf{A} 's eigenvalues are 0, $\mathbf{S}^{-1} \mathbf{A} \mathbf{S} = \mathbf{O}$. Multiplying both sides by \mathbf{S} on the left and then \mathbf{S}^{-1} on the right, we get $\mathbf{A} = \mathbf{O}$.

2. Yes

$\mathbf{A}^k = \mathbf{O}$ implies that $\det \mathbf{A} = 0$, since $\det(\mathbf{A}^k) = (\det \mathbf{A})^k = 0$. But then \mathbf{A} is not invertible, so $\text{null} \mathbf{A} = \text{null}(\mathbf{A} - 0\mathbf{I})$ contains nonzero vectors. Since $\lambda = 0$ has nonzero eigenvectors, it is an eigenvalue of \mathbf{A} .

3. No

Since $c_{\mathbf{A}}(\lambda) = \lambda^2 - 3\lambda + 2 = \lambda^2 - (\text{tr} \mathbf{A})\lambda + \det \mathbf{A}$, $\text{tr} \mathbf{A} = 3$.

4. Yes

If \mathbf{x} is a solution to $\mathbf{A}\mathbf{x} = \lambda\mathbf{x}$, then $(\mathbf{A} + \mathbf{I})\mathbf{x} = \lambda\mathbf{x} + \mathbf{x}$ and so the eigenvectors of \mathbf{A} and $\mathbf{A} + \mathbf{I}$ are the same.

5. Yes

If \mathbf{x} is a solution to $\mathbf{A}\mathbf{x} = \lambda\mathbf{x}$, then $\mathbf{A}^2\mathbf{x} = \lambda^2\mathbf{x}$ and so the eigenvectors of \mathbf{A} and \mathbf{A}^2 are the same.

7. Yes

Since \mathbf{A} is diagonalizable, the sum of the dimensions of all its eigenspaces must be equal to n . Thus E_d , the eigenspace corresponding to $\lambda = d$, must be equal to n , and since $E_d = \text{null}(d\mathbf{I} - \mathbf{A})$, we have by the Rank-Nullity theorem that $\text{rank}(d\mathbf{I} - \mathbf{A}) = 0$.

10. Yes

If \mathbf{A} does not have 0 as an eigenvalue, then $E_0 = \text{null}(\mathbf{A} - 0\mathbf{I}) = \text{null} \mathbf{A}$ must have a dimension of 0 since the eigenspace associated with $\lambda = 0$ contains no eigenvectors. $\text{nullity} \mathbf{A} = 0$ is equivalent to \mathbf{A} being invertible, i.e., a product of elementary matrices.

12. If $\det \mathbf{A} = 0$, $\det(\mathbf{A} \text{adj} \mathbf{A}) = 0$.

If $\det \mathbf{A} \neq 0$, \mathbf{A} is invertible, so $\text{adj} \mathbf{A} = \mathbf{A}^{-1} (\det \mathbf{A}) \mathbf{I}$

$\det(\text{adj} \mathbf{A}) = \det(\mathbf{A}^{-1} (\det \mathbf{A}) \mathbf{I}) = (\det \mathbf{A})^n / \det \mathbf{A} = (\det \mathbf{A})^{n-1}$

13. a) $\text{rank} \mathbf{A} = 1$, $\text{nullity} \mathbf{A} = 3$

b) Since $E_0 = \text{null} \mathbf{A}$, $\lambda = 0$ exists and its eigenspace has dimension 3.

c) Since $c_{\mathbf{A}}(\lambda) = \lambda^3(\lambda - 4)$, another eigenvalue of \mathbf{A} is $\lambda = 4$.

17. a) $\mathbf{A} \text{adj} \mathbf{A} = (\det \mathbf{A}) \mathbf{I} = (\det \mathbf{A}^T) \mathbf{I} = \mathbf{A}^T \text{adj} \mathbf{A}^T$

b) \implies If \mathbf{A} is invertible, $\det \mathbf{A} \neq 0$, so $\mathbf{A}^T \text{adj} \mathbf{A}^T = (\det \mathbf{A}) \mathbf{I}$

$\mathbf{A}^T (\text{adj} \mathbf{A}^T / \det \mathbf{A}) = \mathbf{I}$ which means \mathbf{A}^T has an inverse, $\text{adj} \mathbf{A}^T / \det \mathbf{A}$.

\Leftarrow If \mathbf{A}^T is invertible, $\det \mathbf{A}^T \neq 0$, so $\mathbf{A} \text{adj} \mathbf{A} = (\det \mathbf{A}^T) \mathbf{I}$

$\mathbf{A} (\text{adj} \mathbf{A} / \det \mathbf{A}^T) = \mathbf{I}$ which means \mathbf{A} has an inverse, $\text{adj} \mathbf{A} / \det \mathbf{A}^T$.

20. Note that at least two of the λ_i 's must be equal or else there are more distinct eigenvalues than columns of \mathbf{A} . Since 2 linearly independent vectors are in the repeated eigenvalue's eigenspace, this eigenspace has dimension 2. But then the other eigenvalue is not an eigenvalue if it is different from the two λ_i 's, since then its eigenspace would have dimension 0. Thus, all three λ_i 's are equal (call it λ), and since λ 's eigenspace has dimension 2, all $\mathbf{x} \in {}^2\mathbb{R}$ are solutions to the equation $\mathbf{Ax} = \lambda\mathbf{x}$.