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# AN INTRODUCTION TO OPTIMIZATION

SOLUTIONS MANUAL

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Fourth Edition

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## 1. Methods of Proof and Some Notation

### 1.1

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A	B	not A	not B	$A \Rightarrow B$	$(\text{not } B) \Rightarrow (\text{not } A)$
F	F	T	T	T	T
F	T	T	F	T	T
T	F	F	T	F	F
T	T	F	F	T	T

### 1.2

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A	B	not A	not B	$A \Rightarrow B$	not (A and (not B))
F	F	T	T	T	T
F	T	T	F	T	T
T	F	F	T	F	F
T	T	F	F	T	T

### 1.3

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A	B	not (A and B)	not A	not B	(not A) or (not B)
F	F	T	T	T	T
F	T	T	T	F	T
T	F	T	F	T	T
T	T	F	F	F	F

### 1.4

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A	B	A and B	A and (not B)	(A and B) or (A and (not B))
F	F	F	F	F
F	T	F	F	F
T	F	F	T	T
T	T	T	F	T

### 1.5

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The cards that you should turn over are 3 and  $A$ . The remaining cards are irrelevant to ascertaining the truth or falsity of the rule. The card with  $S$  is irrelevant because  $S$  is not a vowel. The card with 8 is not relevant because the rule does not say that if a card has an even number on one side, then it has a vowel on the other side.

Turning over the  $A$  card directly verifies the rule, while turning over the 3 card verifies the contraposition.

## 2. Vector Spaces and Matrices

### 2.1

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We show this by contradiction. Suppose  $n < m$ . Then, the number of columns of  $\mathbf{A}$  is  $n$ . Since  $\text{rank } \mathbf{A}$  is the maximum number of linearly independent columns of  $\mathbf{A}$ , then  $\text{rank } \mathbf{A}$  cannot be greater than  $n < m$ , which contradicts the assumption that  $\text{rank } \mathbf{A} = m$ .

### 2.2

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$\Rightarrow$ : Since there exists a solution, then by Theorem 2.1,  $\text{rank } \mathbf{A} = \text{rank}[\mathbf{A}; \mathbf{b}]$ . So, it remains to prove that  $\text{rank } \mathbf{A} = n$ . For this, suppose that  $\text{rank } \mathbf{A} < n$  (note that it is impossible for  $\text{rank } \mathbf{A} > n$  since  $\mathbf{A}$  has only  $n$  columns). Hence, there exists  $\mathbf{y} \in \mathbb{R}^n$ ,  $\mathbf{y} \neq \mathbf{0}$ , such that  $\mathbf{A}\mathbf{y} = \mathbf{0}$  (this is because the columns of

$\mathbf{A}$  are linearly dependent, and  $\mathbf{A}\mathbf{y}$  is a linear combination of the columns of  $\mathbf{A}$ ). Let  $\mathbf{x}$  be a solution to  $\mathbf{A}\mathbf{x} = \mathbf{b}$ . Then clearly  $\mathbf{x} + \mathbf{y} \neq \mathbf{x}$  is also a solution. This contradicts the uniqueness of the solution. Hence,  $\text{rank } \mathbf{A} = n$ .

$\Leftarrow$ : By Theorem 2.1, a solution exists. It remains to prove that it is unique. For this, let  $\mathbf{x}$  and  $\mathbf{y}$  be solutions, i.e.,  $\mathbf{A}\mathbf{x} = \mathbf{b}$  and  $\mathbf{A}\mathbf{y} = \mathbf{b}$ . Subtracting, we get  $\mathbf{A}(\mathbf{x} - \mathbf{y}) = \mathbf{0}$ . Since  $\text{rank } \mathbf{A} = n$  and  $\mathbf{A}$  has  $n$  columns, then  $\mathbf{x} - \mathbf{y} = \mathbf{0}$  and hence  $\mathbf{x} = \mathbf{y}$ , which shows that the solution is unique.

### 2.3

Consider the vectors  $\bar{\mathbf{a}}_i = [1, \mathbf{a}_i^\top]^\top \in \mathbb{R}^{n+1}$ ,  $i = 1, \dots, k$ . Since  $k \geq n + 2$ , then the vectors  $\bar{\mathbf{a}}_1, \dots, \bar{\mathbf{a}}_k$  must be linearly independent in  $\mathbb{R}^{n+1}$ . Hence, there exist  $\alpha_1, \dots, \alpha_k$ , not all zero, such that

$$\sum_{i=1}^k \alpha_i \bar{\mathbf{a}}_i = \mathbf{0}.$$

The first component of the above vector equation is  $\sum_{i=1}^k \alpha_i = 0$ , while the last  $n$  components have the form  $\sum_{i=1}^k \alpha_i \mathbf{a}_i = \mathbf{0}$ , completing the proof.

### 2.4

a. We first postmultiply  $\mathbf{M}$  by the matrix

$$\begin{bmatrix} \mathbf{I}_k & \mathbf{O} \\ -\mathbf{M}_{m-k,k} & \mathbf{I}_{m-k} \end{bmatrix}$$

to obtain

$$\begin{bmatrix} \mathbf{M}_{m-k,k} & \mathbf{I}_{m-k} \\ \mathbf{M}_{k,k} & \mathbf{O} \end{bmatrix} \begin{bmatrix} \mathbf{I}_k & \mathbf{O} \\ -\mathbf{M}_{m-k,k} & \mathbf{I}_{m-k} \end{bmatrix} = \begin{bmatrix} \mathbf{O} & \mathbf{I}_{m-k} \\ \mathbf{M}_{k,k} & \mathbf{O} \end{bmatrix}.$$

Note that the determinant of the postmultiplying matrix is 1. Next we postmultiply the resulting product by

$$\begin{bmatrix} \mathbf{O} & \mathbf{I}_k \\ \mathbf{I}_{m-k} & \mathbf{O} \end{bmatrix}$$

to obtain

$$\begin{bmatrix} \mathbf{O} & \mathbf{I}_{m-k} \\ \mathbf{M}_{k,k} & \mathbf{O} \end{bmatrix} \begin{bmatrix} \mathbf{O} & \mathbf{I}_k \\ \mathbf{I}_{m-k} & \mathbf{O} \end{bmatrix} = \begin{bmatrix} \mathbf{I}_k & \mathbf{O} \\ \mathbf{O} & \mathbf{M}_{k,k} \end{bmatrix}.$$

Notice that

$$\det \mathbf{M} = \det \left( \begin{bmatrix} \mathbf{I}_k & \mathbf{O} \\ \mathbf{O} & \mathbf{M}_{k,k} \end{bmatrix} \right) \det \left( \begin{bmatrix} \mathbf{O} & \mathbf{I}_k \\ \mathbf{I}_{m-k} & \mathbf{O} \end{bmatrix} \right),$$

where

$$\det \left( \begin{bmatrix} \mathbf{O} & \mathbf{I}_k \\ \mathbf{I}_{m-k} & \mathbf{O} \end{bmatrix} \right) = \pm 1.$$

The above easily follows from the fact that the determinant changes its sign if we interchange columns, as discussed in Section 2.2. Moreover,

$$\det \left( \begin{bmatrix} \mathbf{I}_k & \mathbf{O} \\ \mathbf{O} & \mathbf{M}_{k,k} \end{bmatrix} \right) = \det(\mathbf{I}_k) \det(\mathbf{M}_{k,k}) = \det(\mathbf{M}_{k,k}).$$

Hence,

$$\det \mathbf{M} = \pm \det \mathbf{M}_{k,k}.$$

b. We can see this on the following examples. We assume, without loss of generality that  $\mathbf{M}_{m-k,k} = \mathbf{O}$  and let  $\mathbf{M}_{k,k} = 2$ . Thus  $k = 1$ . First consider the case when  $m = 2$ . Then we have

$$\mathbf{M} = \begin{bmatrix} \mathbf{O} & \mathbf{I}_{m-k} \\ \mathbf{M}_{k,k} & \mathbf{O} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 2 & 0 \end{bmatrix}.$$

Thus,

$$\det \mathbf{M} = -2 = \det(-\mathbf{M}_{k,k}).$$

Next consider the case when  $m = 3$ . Then

$$\det \begin{bmatrix} \mathbf{O} & \mathbf{I}_{m-k} \\ \mathbf{M}_{k,k} & \mathbf{O} \end{bmatrix} = \det \begin{bmatrix} 0 & \vdots & 1 & 0 \\ 0 & \vdots & 0 & 1 \\ \dots & \dots & \dots & \dots \\ 2 & \vdots & 0 & 0 \end{bmatrix} = 2 \neq \det(-\mathbf{M}_{k,k}).$$

Therefore, in general,

$$\det \mathbf{M} \neq \det(-\mathbf{M}_{k,k})$$

However, when  $k = m/2$ , that is, when all sub-matrices are square and of the same dimension, then it is true that

$$\det \mathbf{M} = \det(-\mathbf{M}_{k,k}).$$

See [121].

## 2.5

Let

$$\mathbf{M} = \begin{bmatrix} \mathbf{A} & \mathbf{B} \\ \mathbf{C} & \mathbf{D} \end{bmatrix}$$

and suppose that each block is  $k \times k$ . John R. Sylvester [121] showed that if at least one of the blocks is equal to  $\mathbf{O}$  (zero matrix), then the desired formula holds. Indeed, if a row or column block is zero, then the determinant is equal to zero as follows from the determinant's properties discussed Section 2.2. That is, if  $\mathbf{A} = \mathbf{B} = \mathbf{O}$ , or  $\mathbf{A} = \mathbf{C} = \mathbf{O}$ , and so on, then obviously  $\det \mathbf{M} = 0$ . This includes the case when any three or all four block matrices are zero matrices.

If  $\mathbf{B} = \mathbf{O}$  or  $\mathbf{C} = \mathbf{O}$  then

$$\det \mathbf{M} = \det \begin{bmatrix} \mathbf{A} & \mathbf{B} \\ \mathbf{C} & \mathbf{D} \end{bmatrix} = \det(\mathbf{AD}).$$

The only case left to analyze is when  $\mathbf{A} = \mathbf{O}$  or  $\mathbf{D} = \mathbf{O}$ . We will show that in either case,

$$\det \mathbf{M} = \det(-\mathbf{BC}).$$

Without loss of generality suppose that  $\mathbf{D} = \mathbf{O}$ . Following arguments of John R. Sylvester [121], we premultiply  $\mathbf{M}$  by the product of three matrices whose determinants are unity:

$$\begin{bmatrix} \mathbf{I}_k & -\mathbf{I}_k \\ \mathbf{O} & \mathbf{I}_k \end{bmatrix} \begin{bmatrix} \mathbf{I}_k & \mathbf{O} \\ \mathbf{I}_k & \mathbf{I}_k \end{bmatrix} \begin{bmatrix} \mathbf{I}_k & -\mathbf{I}_k \\ \mathbf{O} & \mathbf{I}_k \end{bmatrix} \begin{bmatrix} \mathbf{A} & \mathbf{B} \\ \mathbf{C} & \mathbf{O} \end{bmatrix} = \begin{bmatrix} -\mathbf{C} & \mathbf{O} \\ \mathbf{A} & \mathbf{B} \end{bmatrix}.$$

Hence,

$$\begin{aligned} \det \begin{bmatrix} \mathbf{A} & \mathbf{B} \\ \mathbf{C} & \mathbf{O} \end{bmatrix} &= \det \begin{bmatrix} -\mathbf{C} & \mathbf{O} \\ \mathbf{A} & \mathbf{B} \end{bmatrix} \\ &= \det(-\mathbf{C}) \det \mathbf{B} \\ &= \det(-\mathbf{I}_k) \det \mathbf{C} \det \mathbf{B}. \end{aligned}$$

Thus we have

$$\det \begin{bmatrix} \mathbf{A} & \mathbf{B} \\ \mathbf{C} & \mathbf{O} \end{bmatrix} = \det(-\mathbf{BC}) = \det(-\mathbf{CB}).$$

## 2.6

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We represent the given system of equations in the form  $\mathbf{Ax} = \mathbf{b}$ , where

$$\mathbf{A} = \begin{bmatrix} 1 & 1 & 2 & 1 \\ 1 & -2 & 0 & -1 \end{bmatrix}, \quad \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}, \quad \text{and} \quad \mathbf{b} = \begin{bmatrix} 1 \\ -2 \end{bmatrix}.$$

Using elementary row operations yields

$$\mathbf{A} = \begin{bmatrix} 1 & 1 & 2 & 1 \\ 1 & -2 & 0 & -1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 1 & 2 & 1 \\ 0 & -3 & -2 & -2 \end{bmatrix}, \quad \text{and}$$
$$[\mathbf{A}, \mathbf{b}] = \begin{bmatrix} 1 & 1 & 2 & 1 & 1 \\ 1 & -2 & 0 & -1 & -2 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 1 & 2 & 1 & 1 \\ 0 & -3 & -2 & -2 & -3 \end{bmatrix},$$

from which  $\text{rank } \mathbf{A} = 2$  and  $\text{rank}[\mathbf{A}, \mathbf{b}] = 2$ . Therefore, by Theorem 2.1, the system has a solution.

We next represent the system of equations as

$$\begin{bmatrix} 1 & 1 \\ 1 & -2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 - 2x_3 - x_4 \\ -2 + x_4 \end{bmatrix}$$

Assigning arbitrary values to  $x_3$  and  $x_4$  ( $x_3 = d_3$ ,  $x_4 = d_4$ ), we get

$$\begin{aligned} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} &= \begin{bmatrix} 1 & 1 \\ 1 & -2 \end{bmatrix}^{-1} \begin{bmatrix} 1 - 2x_3 - x_4 \\ -2 + x_4 \end{bmatrix} \\ &= -\frac{1}{3} \begin{bmatrix} -2 & -1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 1 - 2x_3 - x_4 \\ -2 + x_4 \end{bmatrix} \\ &= \begin{bmatrix} -\frac{4}{3}d_3 - \frac{1}{3}d_4 \\ 1 - \frac{2}{3}d_3 - \frac{2}{3}d_4 \end{bmatrix}. \end{aligned}$$

Therefore, a general solution is

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} -\frac{4}{3}d_3 - \frac{1}{3}d_4 \\ 1 - \frac{2}{3}d_3 - \frac{2}{3}d_4 \\ d_3 \\ d_4 \end{bmatrix} = \begin{bmatrix} -\frac{4}{3} \\ -\frac{2}{3} \\ 1 \\ 0 \end{bmatrix} d_3 + \begin{bmatrix} -\frac{1}{3} \\ -\frac{2}{3} \\ 0 \\ 1 \end{bmatrix} d_4 + \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix},$$

where  $d_3$  and  $d_4$  are arbitrary values.

## 2.7

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1. Apply the definition of  $|-a|$ :

$$\begin{aligned} |-a| &= \begin{cases} -a & \text{if } -a > 0 \\ 0 & \text{if } -a = 0 \\ -(-a) & \text{if } -a < 0 \end{cases} \\ &= \begin{cases} -a & \text{if } a < 0 \\ 0 & \text{if } a = 0 \\ a & \text{if } a > 0 \end{cases} \\ &= |a|. \end{aligned}$$

2. If  $a \geq 0$ , then  $|a| = a$ . If  $a < 0$ , then  $|a| = -a > 0 > a$ . Hence  $|a| \geq a$ . On the other hand,  $|-a| \geq -a$  (by the above). Hence,  $a \geq -|-a| = -|a|$  (by property 1).

3. We have four cases to consider. First, if  $a, b \geq 0$ , then  $a + b \geq 0$ . Hence,  $|a + b| = a + b = |a| + |b|$ .  
 Second, if  $a, b \leq 0$ , then  $a + b \leq 0$ . Hence  $|a + b| = -(a + b) = -a - b = |a| + |b|$ .  
 Third, if  $a \geq 0$  and  $b \leq 0$ , then we have two further subcases:

1. If  $a + b \geq 0$ , then  $|a + b| = a + b \leq |a| + |b|$ .
2. If  $a + b \leq 0$ , then  $|a + b| = -a - b \leq |a| + |b|$ .

The fourth case,  $a \leq 0$  and  $b \geq 0$ , is identical to the third case, with  $a$  and  $b$  interchanged.

4. We first show  $|a - b| \leq |a| + |b|$ . We have

$$\begin{aligned} |a - b| &= |a + (-b)| \\ &\leq |a| + |-b| \quad \text{by property 3} \\ &= |a| + |b| \quad \text{by property 1.} \end{aligned}$$

To show  $||a| - |b|| \leq |a - b|$ , we note that  $|a| = |a - b + b| \leq |a - b| + |b|$ , which implies  $|a| - |b| \leq |a - b|$ . On the other hand, from the above we have  $|b| - |a| \leq |b - a| = |a - b|$  by property 1. Therefore,  $||a| - |b|| \leq |a - b|$ .

5. We have four cases. First, if  $a, b \geq 0$ , we have  $ab \geq 0$  and hence  $|ab| = ab = |a||b|$ . Second, if  $a, b \leq 0$ , we have  $ab \geq 0$  and hence  $|ab| = ab = (-a)(-b) = |a||b|$ . Third, if  $a \leq 0, b \geq 0$ , we have  $ab \leq 0$  and hence  $|ab| = -ab = a(-b) = |a||b|$ . The fourth case,  $a \geq 0$  and  $b \leq 0$ , is identical to the third case, with  $a$  and  $b$  interchanged.

6. We have

$$\begin{aligned} |a + b| &\leq |a| + |b| \quad \text{by property 3} \\ &\leq c + d. \end{aligned}$$

7.  $\Rightarrow$ : By property 2,  $-a \leq |a|$  and  $a \leq |a|$ . Therefore,  $|a| < b$  implies  $-a \leq |a| < b$  and  $a \leq |a| < b$ .

$\Leftarrow$ : If  $a \geq 0$ , then  $|a| = a < b$ . If  $a < 0$ , then  $|a| = -a < b$ .

For the case when “ $<$ ” is replaced by “ $\leq$ ”, we simply repeat the above proof with “ $<$ ” replaced by “ $\leq$ ”.

8. This is simply the negation of property 7 (apply DeMorgan’s Law).

## 2.8

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Observe that we can represent  $\langle \mathbf{x}, \mathbf{y} \rangle_2$  as

$$\langle \mathbf{x}, \mathbf{y} \rangle_2 = \mathbf{x}^\top \begin{bmatrix} 2 & 3 \\ 3 & 5 \end{bmatrix} \mathbf{y} = (\mathbf{Q}\mathbf{x})^\top (\mathbf{Q}\mathbf{y}) = \mathbf{x}^\top \mathbf{Q}^2 \mathbf{y},$$

where

$$\mathbf{Q} = \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix}.$$

Note that the matrix  $\mathbf{Q} = \mathbf{Q}^\top$  is nonsingular.

1. Now,  $\langle \mathbf{x}, \mathbf{x} \rangle_2 = (\mathbf{Q}\mathbf{x})^\top (\mathbf{Q}\mathbf{x}) = \|\mathbf{Q}\mathbf{x}\|^2 \geq 0$ , and

$$\begin{aligned} \langle \mathbf{x}, \mathbf{x} \rangle_2 = 0 &\Leftrightarrow \|\mathbf{Q}\mathbf{x}\|^2 = 0 \\ &\Leftrightarrow \mathbf{Q}\mathbf{x} = \mathbf{0} \\ &\Leftrightarrow \mathbf{x} = \mathbf{0} \end{aligned}$$

since  $\mathbf{Q}$  is nonsingular.

2.  $\langle \mathbf{x}, \mathbf{y} \rangle_2 = (\mathbf{Q}\mathbf{x})^\top (\mathbf{Q}\mathbf{y}) = (\mathbf{Q}\mathbf{y})^\top (\mathbf{Q}\mathbf{x}) = \langle \mathbf{y}, \mathbf{x} \rangle_2$ .

3. We have

$$\begin{aligned} \langle \mathbf{x} + \mathbf{y}, \mathbf{z} \rangle_2 &= (\mathbf{x} + \mathbf{y})^\top \mathbf{Q}^2 \mathbf{z} \\ &= \mathbf{x}^\top \mathbf{Q}^2 \mathbf{z} + \mathbf{y}^\top \mathbf{Q}^2 \mathbf{z} \\ &= \langle \mathbf{x}, \mathbf{z} \rangle_2 + \langle \mathbf{y}, \mathbf{z} \rangle_2. \end{aligned}$$

$$4. \langle r\mathbf{x}, \mathbf{y} \rangle_2 = (r\mathbf{x})^\top \mathbf{Q}^2 \mathbf{y} = r\mathbf{x}^\top \mathbf{Q}^2 \mathbf{y} = r\langle \mathbf{x}, \mathbf{y} \rangle_2.$$

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**2.9**

We have  $\|\mathbf{x}\| = \|(\mathbf{x} - \mathbf{y}) + \mathbf{y}\| \leq \|\mathbf{x} - \mathbf{y}\| + \|\mathbf{y}\|$  by the Triangle Inequality. Hence,  $\|\mathbf{x}\| - \|\mathbf{y}\| \leq \|\mathbf{x} - \mathbf{y}\|$ . On the other hand, from the above we have  $\|\mathbf{y}\| - \|\mathbf{x}\| \leq \|\mathbf{y} - \mathbf{x}\| = \|\mathbf{x} - \mathbf{y}\|$ . Combining the two inequalities, we obtain  $|\|\mathbf{x}\| - \|\mathbf{y}\|| \leq \|\mathbf{x} - \mathbf{y}\|$ .

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**2.10**

Let  $\epsilon > 0$  be given. Set  $\delta = \epsilon$ . Hence, if  $\|\mathbf{x} - \mathbf{y}\| < \delta$ , then by Exercise 2.9,  $|\|\mathbf{x}\| - \|\mathbf{y}\|| \leq \|\mathbf{x} - \mathbf{y}\| < \delta = \epsilon$ .

### 3. Transformations

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**3.1**

Let  $\mathbf{v}$  be the vector such that  $\mathbf{x}$  are the coordinates of  $\mathbf{v}$  with respect to  $\{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n\}$ , and  $\mathbf{x}'$  are the coordinates of  $\mathbf{v}$  with respect to  $\{\mathbf{e}'_1, \mathbf{e}'_2, \dots, \mathbf{e}'_n\}$ . Then,

$$\mathbf{v} = x_1 \mathbf{e}_1 + \dots + x_n \mathbf{e}_n = [\mathbf{e}_1, \dots, \mathbf{e}_n] \mathbf{x},$$

and

$$\mathbf{v} = x'_1 \mathbf{e}'_1 + \dots + x'_n \mathbf{e}'_n = [\mathbf{e}'_1, \dots, \mathbf{e}'_n] \mathbf{x}'.$$

Hence,

$$[\mathbf{e}_1, \dots, \mathbf{e}_n] \mathbf{x} = [\mathbf{e}'_1, \dots, \mathbf{e}'_n] \mathbf{x}'$$

which implies

$$\mathbf{x}' = [\mathbf{e}'_1, \dots, \mathbf{e}'_n]^{-1} [\mathbf{e}_1, \dots, \mathbf{e}_n] \mathbf{x} = \mathbf{T} \mathbf{x}.$$

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**3.2**

a. We have

$$[\mathbf{e}'_1, \mathbf{e}'_2, \mathbf{e}'_3] = [\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3] \begin{bmatrix} 1 & 2 & 4 \\ 3 & -1 & 5 \\ -4 & 5 & 3 \end{bmatrix}.$$

Therefore,

$$\mathbf{T} = [\mathbf{e}'_1, \mathbf{e}'_2, \mathbf{e}'_3]^{-1} [\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3] = \begin{bmatrix} 1 & 2 & 4 \\ 3 & -1 & 5 \\ -4 & 5 & 3 \end{bmatrix}^{-1} = \frac{1}{42} \begin{bmatrix} 28 & -14 & -14 \\ 29 & -19 & -7 \\ -11 & 13 & 7 \end{bmatrix}.$$

b. We have

$$[\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3] = [\mathbf{e}'_1, \mathbf{e}'_2, \mathbf{e}'_3] \begin{bmatrix} 1 & 2 & 3 \\ 1 & -1 & 0 \\ 3 & 4 & 5 \end{bmatrix}.$$

Therefore,

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

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**3.3**

We have

$$[\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3] = [\mathbf{e}'_1, \mathbf{e}'_2, \mathbf{e}'_3] \begin{bmatrix} 2 & 2 & 3 \\ 1 & -1 & 0 \\ -1 & 2 & 1 \end{bmatrix}.$$