

## Quiz #5 Solutions

(c) For the matrices with positive determinants, the transformed square has its points 1, 2, 3, and 4 in counter-clockwise order (the same order as the original square). For the matrices with negative determinants, the transformed square has its points in opposite order.

- $\mathbf{6}$ 1 3  $1 \ 0 \ 1$ 0 0 0 0 0  $\mathbf{2}$  $^{-1}$ -2 ≙  $0 \ 1$  $1 \ 0$ 0 0 -11 1  $R5 \rightarrow R5 + 2R3$ -1 $1 \ 0$  $0 \ 2$ ♠  $R5 \rightarrow R5 + R4$
- **2.** (a) Completing the reduction to echelon form, we get:

Note that we used only row replacement operations.

(b) Since we used only row replacement operations and no row interchange (or scaling) operations, the determinant of the original matrix A is given by the product of the main diagonal of the echelon form of A (the first five columns of this augmented matrix). That is,

$$\det A = (1)(1)(-1)(-3)(1) = 3$$

Since the determinant in nonzero, the matrix is invertible.

(c) Completing the reduction to reduced echelon form, we get:

As expected (since we already know that A is invertible), the left half of the reduced echelon form is the identity matrix. Therefore, the right half is  $A^{-1}$ . That is,

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

Double-checking the answer:

$$AA^{-1} = \begin{bmatrix} 1 & 2 & 0 & 6 & 1 \\ 0 & 1 & 1 & 3 & 1 \\ 1 & 1 & -2 & 0 & 1 \\ 1 & 1 & -1 & 0 & 0 \\ -2 & -2 & 4 & 3 & -1 \end{bmatrix} \begin{bmatrix} -2 & 1 & 4 & 5 & 3 \\ 1 & 0 & -3 & -2 & -2 \\ -1 & 1 & 1 & 2 & 1 \\ \frac{1}{3} & -\frac{1}{3} & 0 & -\frac{1}{3} & 0 \\ -1 & 1 & 2 & 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

- **3.** (a) No. The columns of A must be linearly dependent, since there are more vectors (n) than the number of elements in each vector (m). Therefore, T cannot be one-to-one.
  - (b) Yes. One simple example is the projection transformation  $T: \mathbb{R}^3 \to \mathbb{R}^2$  given by

$$T(\mathbf{x}) = T\left( \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right) = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

As **x** varies over all vectors in  $\mathbb{R}^3$ ,  $T(\mathbf{x})$  will take on all values in  $\mathbb{R}^2$ .

Any linear transformation T whose standard matrix has a pivot position in every row (so its columns span all of  $\mathbb{R}^m$ ) will work.

(c) Yes. One simple example is the transformation  $T \colon \mathbb{R}^2 \to \mathbb{R}^3$  given by

$$T(\mathbf{x}) = T\left( \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \right) = \begin{bmatrix} x_1 \\ x_2 \\ 0 \end{bmatrix}$$

Obviously, no two unequal vectors can map to the same image.

Any linear transformation T whose standard matrix has linearly independent columns (that is, whose standard matrix has a pivot position in every column) will work.

- (d) No. The columns of A must span  $\mathbb{R}^m$ . This can only happen if there is a pivot position in every row, but there can only be a maximum of n pivot positions (one in each column) and this is less than the number of rows m.
- 4. (a) Since  $A^T A$  is  $n \times n$ , its inverse  $(A^T A)^{-1}$  is  $n \times n$ , too. Thus, C, is a product of an  $m \times n$ , an  $n \times n$ , and an  $n \times m$  matrix. Therefore, it is an  $m \times m$  matrix.
  - (b) From the properties of inverses and transposes:

$$C^{T} = (A(A^{T}A)^{-1}A^{T})^{T}$$
  
=  $(A^{T})^{T}((A^{T}A)^{-1})^{T}(A)^{T}$   
=  $A((A^{T}A)^{T})^{-1}A^{T}$   
=  $A((A)^{T}(A^{T})^{T})^{-1}A^{T}$   
=  $A(A^{T}A)^{-1}A^{T}$   
=  $C$ 

Note that you can't use  $(A^T A)^{-1} = A^{-1} (A^T)^{-1}$ . The matrix A may not even be square, so  $A^{-1}$  and  $(A^T)^{-1}$  may not even be defined. Similarly for the second part:

$$C^{2} = (A(A^{T}A)^{-1}A^{T})(A(A^{T}A)^{-1}A^{T})$$
  
=  $A((A^{T}A)^{-1}A^{T}A)(A^{T}A)^{-1}A^{T}$   
=  $A(A^{T}A)^{-1}A^{T}$   
=  $C$ 

(c) Since  $\mathbf{v}$  is in the span of the columns of A, we can write it as  $\mathbf{v} = A\mathbf{x}$  for some vector  $\mathbf{x}$  of weights. Then

$$C\mathbf{v} = CA\mathbf{x}$$
  
=  $(A(A^TA)^{-1}A^T)A\mathbf{x}$   
=  $A((A^TA)^{-1}A^TA)\mathbf{x}$   
=  $A\mathbf{x}$   
=  $\mathbf{v}$ 

(d) For any vector  $\mathbf{w} \in \mathbb{R}^m$ , we have

$$C\mathbf{w} = \left(A(A^T A)^{-1} A^T\right)\mathbf{w} = A\left((A^T A)^{-1} A^T \mathbf{w}\right)$$

Since  $(A^T A)^{-1} A^T \mathbf{w}$  is an  $n \times 1$  vector of weights (even though it's an ugly one), this shows that  $C \mathbf{w}$  is some linear combination of the columns of A.