Section 4.5

- 29. a. True, dim V is the number of pivots that a matrix with column vectors given by the set given in the problem would have. There can be at least p pivots, because the vectors must be V dimensional. dim V must be less than or equal to p because the set of vectors must span a subset of V.
 - b. True, if there are p V dimensional vectors in a linearly independent set, those vectors must be spanning a subset of V. Therefore there must be at least p basis vectors for V, and potentially more.
 - c. True, if a set of vectors already spans V, adding another vector to the set isn't going to prevent those vectors from spanning V.

30. a. False, if V is
$$\mathbb{R}^3$$
 and the set is $\left\{ \begin{bmatrix} 1\\0\\0 \end{bmatrix}, \begin{bmatrix} 1\\0\\0 \end{bmatrix} \right\}$, dim $V = 3$ but $p = 2$.

- b. True, since every set of p vectors fails to span all of V, the set fails to be a basis for V, dim V > p.
- c. False, the set of vectors doesn't have to be a basis for V, so it could be made out of vectors that are multiples of one another, which would make the set linearly dependent.

Section 4.6

- 12. If the null space of a 5×6 matrix is 4 dimensional, it means that that matrix must have 4 pivotless columns. If the matrix has 4 pivotless columns, then two columns must have pivots. If two columns have pivots, then the matrix must have 2 pivots, which means that the matrix has 2 row pivots. Therefore, dim Row A = 2.
- 19. 5 linear equations in 6 variables would produce a 5×6 coefficient matrix. We are told that there the non-trivial solutions to the system span a line. Therefore, we know the system has exactly one free variable and the coefficient matrix has exactly one pivotless column. Therefore, the matrix will have 5 pivots. Since there are 5 rows and 5 pivots, there is one pivot in every row, which means the system will always be consistent for any vector in \mathbb{R}^5 .

Section 4.7

- 3. Only the equation in part (ii) is satisfied. Matrix P is $\begin{bmatrix} \vec{u_1} & \vec{u_2} \end{bmatrix}_{\mathcal{W}}$. This makes P the $\mathcal{U} \to \mathcal{W}$ conversion matrix.
- 4. P is the $\mathcal{D} \to \mathcal{A}$ conversion matrix. Only the equation in (i) is satisfied.

$$\begin{array}{l} 6. \ \mathcal{D} = \left\{ d_{1}^{-} \ d_{2}^{-} \ d_{3}^{-} \right\}, \mathcal{F} = \left\{ f_{1}^{-} \ f_{2}^{-} \ f_{3}^{-} \right\}, \left[f_{1}^{-} \right]_{\mathcal{D}} = \begin{bmatrix} 2 \\ -1 \\ 1 \end{bmatrix}_{\mathcal{D}}^{-}, \left[f_{2}^{-} \right]_{\mathcal{D}} = \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}_{\mathcal{D}}^{-}, \left[f_{3}^{-} \right]_{\mathcal{D}} = \begin{bmatrix} -3 \\ 0 \\ 2 \end{bmatrix}_{\mathcal{D}}^{-}, \\ \text{For any } \begin{bmatrix} \vec{x} \end{bmatrix}_{\mathcal{F}} \Rightarrow \begin{bmatrix} \vec{x} \end{bmatrix}_{\mathcal{D}} = \begin{bmatrix} 2 & 0 & -3 \\ -1 & 3 & 0 \\ 1 & 1 & 2 \end{bmatrix} \begin{bmatrix} \vec{x} \end{bmatrix}_{\mathcal{F}} \Rightarrow \mathcal{F}_{\mathcal{F} \mathcal{D} \mathcal{D}}^{-} = \begin{bmatrix} 2 & 0 & -3 \\ -1 & 3 & 0 \\ 1 & 1 & 2 \end{bmatrix} \begin{bmatrix} \vec{x} \end{bmatrix}_{\mathcal{F}} \Rightarrow \mathcal{F}_{\mathcal{F} \mathcal{D} \mathcal{D}}^{-} = \begin{bmatrix} 2 & 0 & -3 \\ -1 & 3 & 0 \\ 1 & 1 & 2 \end{bmatrix} \\ \text{b. } \begin{bmatrix} \vec{x} \end{bmatrix}_{\mathcal{F}} = \begin{bmatrix} 1 \\ -2 \\ 2 \end{bmatrix} \Rightarrow \begin{bmatrix} 2 & 0 & -3 \\ -1 & 3 & 0 \\ 1 & 1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ -2 \\ 2 \end{bmatrix}_{\mathcal{F}} = \begin{bmatrix} -4 \\ -7 \\ 3 \end{bmatrix}_{\mathcal{D}}^{-} = \begin{bmatrix} \vec{x} \end{bmatrix}_{\mathcal{D}}^{-} \\ \text{b. } \begin{bmatrix} \vec{x} \end{bmatrix}_{\mathcal{D}}^{-} = \begin{bmatrix} 2 \\ 0 \end{bmatrix}, \vec{c} = \begin{bmatrix} 2 \\ -1 \end{bmatrix}, \vec{c} \neq \begin{bmatrix} 6 \\ -2 \end{bmatrix} \\ \begin{bmatrix} -4 \\ -7 \\ 3 \end{bmatrix}_{\mathcal{D}}^{-} = \begin{bmatrix} \vec{x} \end{bmatrix}_{\mathcal{D}}^{-} \\ \text{b. } \begin{bmatrix} \vec{x} \end{bmatrix}_{\mathcal{F}}^{-} \Rightarrow \vec{x} \begin{bmatrix} 2 \\ -1 \end{bmatrix}, \vec{c} \neq \begin{bmatrix} 6 \\ -2 \end{bmatrix} \\ \begin{bmatrix} -4 \\ -7 \\ 3 \end{bmatrix}_{\mathcal{D}}^{-} = \begin{bmatrix} \vec{x} \end{bmatrix}_{\mathcal{D}}^{-} \\ \text{b. } \begin{bmatrix} \vec{x} \end{bmatrix}_{\mathcal{D}^{-}}^{-} \Rightarrow \vec{x} \end{bmatrix} \begin{bmatrix} 2 & 0 & -3 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} \vec{x} \end{bmatrix}_{\mathcal{D}^{-} = \begin{bmatrix} \vec{x} \end{bmatrix}_{\mathcal{D}}^{-} \\ \text{b. } \begin{bmatrix} \vec{x} \end{bmatrix}_{\mathcal{D}^{-} \Rightarrow \vec{x} \end{bmatrix} \begin{bmatrix} 2 & 0 & -3 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} \vec{x} \end{bmatrix}_{\mathcal{D}^{-} = \begin{bmatrix} -3 \\ -1 \end{bmatrix} \\ \text{b. } \begin{bmatrix} \vec{x} \end{bmatrix}_{\mathcal{D}^{-} \Rightarrow \vec{x} \end{bmatrix} \begin{bmatrix} 2 & -1 \\ -1 \end{bmatrix} + x_{2} \begin{bmatrix} 6 \\ -2 \end{bmatrix} \\ \text{b. } \begin{bmatrix} \vec{x} \end{bmatrix}_{\mathcal{D}^{-} = \begin{bmatrix} -6 \\ -1 \end{bmatrix} \\ \text{c. } \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix}_{\mathcal{D}^{-} \Rightarrow \vec{x} \end{bmatrix} \begin{bmatrix} -1 \\ -1 \end{bmatrix} + x_{2} \begin{bmatrix} 6 \\ -2 \end{bmatrix} = \vec{b}_{1}^{-} \Rightarrow \begin{bmatrix} 2 & -6 \\ -1 & -2 \end{bmatrix} \begin{bmatrix} \vec{x} \end{bmatrix}_{\mathcal{D}^{-} \end{bmatrix} \\ \begin{bmatrix} \vec{x} \end{bmatrix}_{\mathcal{D}^{-} = \begin{bmatrix} -6 \\ -1 \end{bmatrix} \\ \text{b. } \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \end{bmatrix} \\ \text{c. } \begin{bmatrix} x_{1} \\ y_{2} \\ y_{2} \end{bmatrix} = \begin{bmatrix} -6 \\ -2 \end{bmatrix} \\ \text{c. } \begin{bmatrix} x_{1} \\ y_{2} \\ y_{2} \end{bmatrix} \end{bmatrix} = \begin{bmatrix} -6 \\ -2 \end{bmatrix} \\ \text{c. } \begin{bmatrix} x_{1} \\ y_{2} \\ y_{2} \end{bmatrix} \end{bmatrix} \\ \text{c. } \begin{bmatrix} x_{1} \\ y_{1} \\ y_{2} \end{bmatrix} \end{bmatrix} \\ \text{c. } \begin{bmatrix} x_{1} \\ y_{1} \\ y_{2} \end{bmatrix} \end{bmatrix} \\ \text{c. } \begin{bmatrix} x_{1} \\ y_{1} \\ y_{2} \end{bmatrix} \end{bmatrix} \\ \text{c. } \begin{bmatrix} x_{1} \\ y_{1} \\ y_{2} \end{bmatrix} \end{bmatrix} \\ \text{c. } \begin{bmatrix} x_{1} \\ y_{2} \\ y_{2} \end{bmatrix} \end{bmatrix} \\ \text{c. } \begin{bmatrix} x_{1} \\ y_{2} \\ y_{2} \end{bmatrix} \end{bmatrix} \\ \text{c. } \begin{bmatrix} x_{1} \\ y_{2} \\ y_{2} \end{bmatrix} \end{bmatrix} \\ \text{c. } \begin{bmatrix} x_{1} \\ y_{2} \\ y_{2} \end{bmatrix} \end{bmatrix} \\ \text{c. } \begin{bmatrix} x_{1} \\ y_{2} \\ y_{2} \end{bmatrix} \end{bmatrix} \\ \text{c. } \begin{bmatrix} x_{1} \\ y_{2} \\ y_{2} \end{bmatrix} \end{bmatrix} \\ \text{c. } \begin{bmatrix} x_{1} \\ y_{2} \\ y_{2} \end{bmatrix} \end{bmatrix} \\ \text{c. } \begin{bmatrix} x_{1} \\ y_{2} \\ y_{2} \end{bmatrix} \end{bmatrix} \\ \text{c. } \begin{bmatrix} x_{1} \\ y_$$

Chapter 4 Supplementary Problems

1. a. True, the set of all linear combinations of a set of vectors is the same as the span of those vectors which is the definition for a subspace.

- b. True, if a set of vectors already spans the entirety of a space, adding another vector to that set will still allow the vectors to span the entirety of a space.
- c. False, if you add a vector to a set of linearly independent vectors it is not necessarily true that the new set will be linearly independent. It is possible for the new vector to be a multiple of a vector already in the set, which would make the new set linearly dependent.
- d. False, this is only one of the conditions for being a subspace. If set S spans all of V and is linearly independent then it is a basis for V.
- e. True, if a set S spans all of a space, it is not necessarily true that the set is linearly independent. However, it would be possible to find a subset of S that is linearly independent and spans all of V, which would satisfy the conditions for a basis for space V.
- f. True, if the set of vectors S is a basis for a space V and dim S = the number of vectors in S, then S must be a basis for V, and the vectors in S must be linearly independent.
- g. False, if the plane does not contain the origin, then it can't be a subspace.
- h. False, $\begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}$. The non pivot columns in this matrix form the set $\left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right\}$, which is a linearly independent set.

i. True,
$$R_1 + R_3 = R_2 \begin{bmatrix} 1 & 2 & 0 \\ 1 & 2 & 1 \\ 0 & 0 & 1 \end{bmatrix} \xrightarrow{rref} \begin{bmatrix} 1 & 2 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix} R_1 + R_3 \neq R_2$$
. The dependency relationship between the three rows has changed by row reduction

tionship between the three rows has changed by row reduction.

- j. False, row operations do not change the location of pivot columns. As a result row operations do not change the dependency relations of columns. Therefore row reduction does not change the null space of a matrix.
- k. False, the rank of a matrix is equal to the number of pivot columns in the matrix.
- 1. False, nonzero rows don't tell us anything about the null space, only the number of pivotless columns tell us anything about the null space.
- m. True, if B is obtained by multiplying A by elementary matrices, then B is row equivalent to A. Therefore the two matrices will have the same pivots, and their rank will be equal.
- n. False, the rows with pivots in A form the basis for Row A.
- o. True, if A and B share the same rref form, then they are row equivalent. Therefore the two matrices will have the same pivots and the same bases for their row subspaces.
- q. False, if rank A = m there must be m pivots. If A is 2×35 there are 2 columns with pivots and 33 columns that are pivotless in A. For the transformation to be one-to-one, there must be a pivot in every column of A.
- r. True, if a linear transformation is onto, there must be a pivot in every row. It there is a pivot in every row $n \ge m$. Therefore there will be m pivots, and m columns will have pivots making Rank A = m.

t. False, the
$$j^{\text{th}}$$
 column of the $\underset{\mathcal{B}\leftarrow\mathcal{C}}{P}$ would be $\begin{bmatrix} \vec{b_j} \end{bmatrix}_{\mathcal{C}}$
2. $a \begin{bmatrix} 1\\2\\-1\\3 \end{bmatrix} + b \begin{bmatrix} -2\\5\\-4\\1 \end{bmatrix} + c \begin{bmatrix} 5\\-8\\7\\1 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & -2 & 5\\2 & 5 & -8\\-1 & -4 & 7\\3 & 1 & 1 \end{bmatrix} \begin{bmatrix} a\\b\\c \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & -2 & 5\\2 & 5 & -8\\-1 & -4 & 7\\3 & 1 & 1 \end{bmatrix} \xrightarrow{\text{rref}} \begin{bmatrix} 1 & 0 & 1\\0 & 1 & -2\\0 & 0 & 0\\0 & 0 & 0 \end{bmatrix}$
Basis = $\left\{ \begin{bmatrix} 1\\2\\-1\\3 \end{bmatrix}, \begin{bmatrix} -2\\5\\-4\\1 \end{bmatrix} \right\}$