**Example 15.11.** Let  $V = \mathbb{R}^4$  and consider the following subset of V:

$$W = \{(x_1, x_2, x_3, x_4) \in \mathbb{R}^4 \mid 2x_1 - 3x_2 + x_3 - 7x_4 = 0\}.$$

Is W a subspace of V?

Solution. The set W is the null space of the matrix  $1 \times 4$  matrix A given by

$$\mathbf{A} = \begin{bmatrix} 2 & -3 & 1 & -7 \end{bmatrix}.$$

Hence,  $W = \text{Null}(\mathbf{A})$  and consequently W is a subspace.

From our previous remarks, the null space of a matrix  $\mathbf{A} \in M_{m \times n}$  is just the solution set of the homogeneous system Ax = 0. Therefore, one way to explicitly describe the null space of A is to solve the system Ax = 0 and write the general solution in parametric vector form. From our previous work on solving linear systems, if the rref(A) has r leading 1's then the number of parameters in the solution set is d = n - r. Therefore, after performing back substitution, we will obtain vectors  $\mathbf{v}_1, \dots, \mathbf{v}_d$  such that the general solution in parametric  $\mathbf{x}=t_1\mathbf{v}_1+t_2\mathbf{v}_2+\dots+t_d\mathbf{v}_d$  where  $t_1,t_2,\dots,t_d$  are arbitrary numbers. Therefore,

$$\mathbf{x} = t_1 \mathbf{v}_1 + t_2 \mathbf{v}_2 + \dots + t_d \mathbf{v}_d$$

$$\operatorname{Nu}\left(\mathbf{A}\right)=\operatorname{span}\{\mathbf{v}_1,\mathbf{v}_2,\dots,\mathbf{v}_i\}.$$

**Example 15.12.** Find a spanning set for the null space of the matrix

$$\mathbf{A} = \begin{bmatrix} -3 & 6 & -1 & 1 & -7 \\ 1 & -2 & 2 & 3 & -1 \\ 2 & -4 & 5 & 8 & -4 \end{bmatrix}.$$

Solution. The null space of A is the solution set of the homogeneous system Ax = 0. Performing elementary row operations one obtains

$$\mathbf{A} \sim \begin{bmatrix} 1 & -2 & 0 & -1 & 3 \\ 0 & 0 & 1 & 2 & -2 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}.$$

Clearly  $r = \text{rank}(\mathbf{A})$  and since n = 5 we will have d = 3 vectors in a spanning set for  $\text{Null}(\mathbf{A})$ . Letting  $x_5 = t_1$ , and  $x_4 = t_2$ , then from the 2nd row we obtain

$$x_3 = -2t_2 + 2t_1.$$

Letting  $x_2 = t_3$ , then from the 1st row we obtain

$$x_1 = 2t_3 + t_2 - 3t_1.$$

Therefore, the only solution to Ax = 0 is the trivial solution. Therefore,  $\mathcal{B}$  is linearly independent. Moreover, for any  $\mathbf{b} \in \mathbb{R}^3$ , the augmented matrix  $[\mathbf{A} \ \mathbf{b}]$  is consistent. Therefore, the columns of **A** span all of  $\mathbb{R}^3$ :

$$Col(\mathbf{A}) = span\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\} = \mathbb{R}^3.$$

Therefore,  $\mathcal{B}$  is a basis for  $\mathbb{R}^3$ .

**Example 16.7.** In  $V = \mathbb{R}^4$ , consider the vectors

$$\mathbf{v}_1 = \begin{bmatrix} 1\\3\\0\\-2 \end{bmatrix}, \ \mathbf{v}_2 = \begin{bmatrix} 2\\-1\\-2\\1 \end{bmatrix}, \ \mathbf{v}_3 = \begin{bmatrix} -1\\4\\2\\-3 \end{bmatrix}.$$

Let  $W = \operatorname{span}\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ . Is  $\mathcal{B} = \{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  a basis for W?

Solution. By definition,  $\mathcal{B}$  is a spanning set for W, so we need only determine if  $\mathcal{B}$  is linearly

independent. Form the matrix, 
$$\mathbf{A} = [\mathbf{v}_1 \ \mathbf{v}_2 \ \mathbf{v}_3]$$
 and row reduce to obtain 
$$\mathbf{A} \sim \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix}$$
Hence, rank( $\mathbf{A} = \mathbf{V}$ ) and thus  $\mathcal{B}$  is linearly dependent. Notice  $\mathbf{v}_1 - \mathbf{v}_2 = \mathbf{v}_3$ . Therefore,  $\mathcal{B}$  is not a sail  $\mathbf{c}$ .

**Example 16.8.** Find a basis for the vector space of  $2 \times 2$  matrices.

**Example 16.9.** Recall that a  $n \times n$  is skew-symmetric **A** if  $\mathbf{A}^T = -\mathbf{A}$ . We proved that the set of  $n \times n$  matrices is a subspace. Find a basis for the set of  $3 \times 3$  skew-symmetric matrices.

## Dimension of a Vector Space 16.3

The following theorem will lead to the definition of the dimension of a vector space.

**Theorem 16.10:** Let V be a vector space. Then all bases of V have the same number of vectors.

We will prove the theorem for the case that  $V = \mathbb{R}^n$ . We already know that the standard unit vectors  $\{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n\}$  is a basis of  $\mathbb{R}^n$ . Let  $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_p\}$  be nonzero vectors in  $\mathbb{R}^n$  and suppose first that p > n. In Lecture 6, Theorem 6.7, we proved that any set of vectors in  $\mathbb{R}^n$  containing more than n vectors is automatically linearly dependent. The reason is that the RREF of  $\mathbf{A} = \begin{bmatrix} \mathbf{u}_1 & \mathbf{u}_2 & \cdots & \mathbf{u}_p \end{bmatrix}$  will contain at most r = n leading ones,

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On the other hand, the inverse matrix  $\mathbf{P}^{-1}$  maps standard coordinates in  $\mathbb{R}^3$  to  $\mathcal{B}$ -coordinates. One can verify that

$$\mathbf{P}^{-1} = \left[ \begin{array}{rrr} 4 & 3 & 6 \\ -1 & -1 & -1 \\ 0 & 0 & -1 \end{array} \right]$$

Therefore, the  $\mathcal{B}$  coordinates of  $\mathbf{v}$  are

$$[\mathbf{v}]_{\mathcal{B}} = \mathbf{P}^{-1}\mathbf{v} = \begin{bmatrix} 4 & 3 & 6 \\ -1 & -1 & -1 \\ 0 & 0 & -1 \end{bmatrix} \begin{bmatrix} 2 \\ -1 \\ 0 \end{bmatrix} = \begin{bmatrix} 5 \\ -1 \\ 0 \end{bmatrix}$$

When V is an abstract vector space, e.g.  $\mathbb{P}_n[t]$  or  $M_{n\times n}$ , the notion of a coordinate mapping is similar as the case when  $V = \mathbb{R}^n$ . If V is an n-dimensional vector space and  $\mathcal{B} = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  is a basis for V, we define the coordinate mapping  $\mathcal{P}: \mathsf{V} \to \mathbb{R}^n$  relative

$$P(\mathbf{v}) = [\mathbf{v}]_{\mathcal{B}}.$$

$$\mathcal{B} = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$$
 is a basis for  $V$ , we define the coordinate mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B}$  as the mapping  $\mathcal{P} : V \to \mathbb{R}^n$  relative to  $\mathcal{B} : \mathcal{B} : \mathcal{B}$ 

Then for any  $\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$  we have

$$\mathcal{P}\left(\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}\right) = \begin{bmatrix} a_{11} \\ a_{12} \\ a_{21} \\ a_{22} \end{bmatrix}.$$

## Matrix Representation of a Linear Map 18.3

Let V and W be vector spaces and let  $T: V \to W$  be a linear mapping. Then by definition of a linear mapping,  $T(\mathbf{v} + \mathbf{u}) = T(\mathbf{v}) + T(\mathbf{u})$  and  $T(\alpha \mathbf{v}) = \alpha T(\mathbf{v})$  for every  $\mathbf{v}, \mathbf{u} \in V$  and  $\alpha \in \mathbb{R}$ . Let  $\mathcal{B} = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  be a basis of V and let  $\gamma = \{\mathbf{w}_1, \mathbf{w}_2, \dots, \mathbf{w}_m\}$  be a basis of W. Then for any  $\mathbf{v} \in V$  there exists scalars  $c_1, c_2, \ldots, c_n$  such that

$$\mathbf{v} = c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 + \dots + c_n \mathbf{v}_n$$