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What is on today

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Lay–Lay–McDonald §4.3 pp. 210 – 215

Today we study the subsets that span a vector space V or a subspace H as “efficiently” as possible. The main idea is that of linear independence.

An indexed set of vectors $\{\mathbf{v}_1, \dots, \mathbf{v}_p\}$ in V is said to be *linearly independent* if the equation

$$c_1\mathbf{v}_1 + \dots + c_p\mathbf{v}_p = \mathbf{0} \quad (1)$$

has only the trivial solution $c_1 = \dots = c_p = 0$. The set $\{\mathbf{v}_1, \dots, \mathbf{v}_p\}$ is said to be *linearly dependent* if (1) has a nontrivial solution: that is, if there are weights c_1, \dots, c_p not all zero such that (1) holds. In this case, there is said to be a *linear dependence relation* among $\mathbf{v}_1, \dots, \mathbf{v}_p$. All of this should sound familiar – we discussed the analogous definition over $V = \mathbb{R}^n$. In fact, the following theorem we saw over \mathbb{R}^n also holds true:

Theorem 1. *An indexed set $\{\mathbf{v}_1, \dots, \mathbf{v}_p\}$ of two or more vectors, with $\mathbf{v}_1 \neq \mathbf{0}$, is linearly dependent if and only if some \mathbf{v}_j (with $j > 1$) is a linear combination of the preceding vectors $\mathbf{v}_1, \dots, \mathbf{v}_j$.*

Example 2. *Let $\mathbf{p}_1(t) = 1$, $\mathbf{p}_2(t) = t$, $\mathbf{p}_3(t) = 4 - t$. Is the set $\{\mathbf{p}_1, \mathbf{p}_2, \mathbf{p}_3\}$ linearly independent in the vector space of polynomials of degree at most 1?*

Definition 3. *Let H be a subspace of a vector space V . An indexed set of vectors $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_p\}$ in V is a basis for H if*

1. \mathcal{B} is a linearly independent set, and
2. the subspace spanned by \mathcal{B} coincides with H ; that is, $H = \text{Span}\{\mathbf{b}_1, \dots, \mathbf{b}_p\}$.

Example 4. *Let A be an invertible $n \times n$ matrix. Then the columns of A form a basis for \mathbb{R}^n because they are linearly independent and they span \mathbb{R}^n , by the Invertible Matrix Theorem.*

Example 5. Let $\mathbf{e}_1, \dots, \mathbf{e}_n$ be the columns of the $n \times n$ identity matrix I_n :

$$\mathbf{e}_1 = \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \mathbf{e}_2 = \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix}, \dots, \mathbf{e}_n = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ 1 \end{bmatrix}.$$

The set $\{\mathbf{e}_1, \dots, \mathbf{e}_n\}$ is called the *standard basis* of \mathbb{R}^n .

Example 6. Let $\mathbf{v}_1 = \begin{bmatrix} 3 \\ 0 \\ -6 \end{bmatrix}$, $\mathbf{v}_2 = \begin{bmatrix} -4 \\ 1 \\ 7 \end{bmatrix}$, $\mathbf{v}_3 = \begin{bmatrix} -2 \\ 1 \\ 5 \end{bmatrix}$. Determine if $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ is a basis for \mathbb{R}^3 .

Example 7. Let $S = \{1, t, \dots, t^n\}$. Verify that S is a basis for P_n . This is called the *standard basis* for P_n .

We we'll see, a basis is an “efficient” spanning set that contains no unnecessary vectors. In fact, a basis can be constructed from a spanning set by discarding unneeded vectors.

Example 8. Let $\mathbf{v}_1 = \begin{bmatrix} 0 \\ 2 \\ -1 \end{bmatrix}$, $\mathbf{v}_2 = \begin{bmatrix} 2 \\ 2 \\ 0 \end{bmatrix}$, $\mathbf{v}_3 = \begin{bmatrix} 6 \\ 16 \\ -5 \end{bmatrix}$, and $H = \text{Span}\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$. Note that $\mathbf{v}_3 = 5\mathbf{v}_1 + 3\mathbf{v}_2$ and show that $\text{Span}\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\} = \text{Span}\{\mathbf{v}_1, \mathbf{v}_2\}$. Then find a basis for H .

The next theorem generalizes the previous example:

Theorem 9 (Spanning set theorem). Let $S = \{\mathbf{v}_1, \dots, \mathbf{v}_p\}$ be a set in V , and let $H = \text{Span}\{\mathbf{v}_1, \dots, \mathbf{v}_p\}$.

1. If one of the vectors in S – say \mathbf{v}_k – is a linear combination of the remaining vectors in S , then the set formed from S by removing \mathbf{v}_k still spans H .
2. If $H \neq \{\mathbf{0}\}$, some subset of S is a basis for H .

We know how to find vectors that span the null space of a matrix A (compute reduced echelon form, write the basic variables in terms of free variables, and decompose as a linear combination of vectors using the free variables as weights); in fact, the method produces a linearly independent set when $\text{Nul } A$ contains nonzero vectors, and in that case, a *basis* for $\text{Nul } A$. Now we describe how to find a basis for the column space, through two examples:

Example 10. Find a basis for $\text{Col } B$, where $B = [\mathbf{b}_1 \quad \mathbf{b}_2 \quad \cdots \quad \mathbf{b}_5] = \begin{bmatrix} 1 & 4 & 0 & 2 & 0 \\ 0 & 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$.

Example 11. *It can be shown that the matrix $A = [\mathbf{a}_1 \quad \mathbf{a}_2 \quad \cdots \quad \mathbf{a}_5] = \begin{bmatrix} 1 & 4 & 0 & 2 & -1 \\ 3 & 12 & 1 & 5 & 5 \\ 2 & 8 & 1 & 3 & 2 \\ 5 & 20 & 2 & 8 & 2 \end{bmatrix}$*

is row equivalent to the matrix B in the previous example. Find a basis for $\text{Col } A$.

These two examples illustrate the following useful fact:

Theorem 12. *The pivot columns of a matrix A form a basis for $\text{Col } A$.*