



- New concepts:
 - Span of a set of vectors, matrix column space
 - Matrix null space
 - Matrix row space
 - Matrix left null space
 - Vector space basis
 - Vector space sums

Definition. The *span* of vectors $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n \in \mathcal{V}$, is the set of vectors reachable by linear combination

$$\text{span}\{\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n\} = \{\mathbf{b} \in \mathcal{V} \mid \exists x_1, \dots, x_n \in \mathcal{S} \text{ such that } \mathbf{b} = x_1\mathbf{a}_1 + \dots x_n\mathbf{a}_n\}.$$

The notation used for set on the right hand side is read: “those vectors $\mathbf{b}$ in $\mathcal{V}$ with the property that there exist n scalars $x_1, \dots, x_n$ to obtain $\mathbf{b}$ by linear combination of $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n$.”

- A linear combination is conveniently expressed as a matrix-vector product leading to a different formulation of the same concept

Definition. The *column space* (or *range*) of matrix $\mathbf{A} \in \mathbb{R}^{m \times n}$ is the set of vectors reachable by linear combination of the matrix column vectors

$$C(\mathbf{A}) = \text{range}(\mathbf{A}) = \{\mathbf{b} \in \mathbb{R}^m \mid \exists \mathbf{x} \in \mathbb{R}^n \text{ such that } \mathbf{b} = \mathbf{A}\mathbf{x}\} \subseteq \mathbb{R}^m$$



Introduce a characterization of the column vectors of a matrix related to linear dependence

Definition. The *null space* of a matrix $\mathbf{A} \in \mathbb{R}^{m \times n}$ is the set

$$N(\mathbf{A}) = \text{null}(\mathbf{A}) = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{A}\mathbf{x} = \mathbf{0}\} \subseteq \mathbb{R}^n$$

- If $\text{null}(\mathbf{A}) = \{\mathbf{0}\}$ then the column vectors of $\mathbf{A}$ are linearly independent, since the only way to satisfy (?) is by the trivial solution $\mathbf{x} = \mathbf{0}$

For example $\mathbf{A} = [\mathbf{a}_1 \ \mathbf{a}_2 \ \mathbf{a}_3]$ below, $c(\mathbf{a}_1 + \mathbf{a}_2 - \mathbf{a}_3) = \mathbf{0}$ for any scalar c , hence

$$\mathbf{A} = \begin{bmatrix} 1 & 2 & 3 \\ 0 & 1 & 1 \\ 1 & 2 & 3 \end{bmatrix} \Rightarrow C(\mathbf{A}) = \text{span}\{\mathbf{a}_1, \mathbf{a}_2\}, N(\mathbf{A}) = \text{span}\left\{\begin{pmatrix} 1 \\ 1 \\ -1 \end{pmatrix}\right\}$$

Recall definitions of column space, null space of $\mathbf{A} \in \mathbb{R}^{m \times n}$

$$C(\mathbf{A}) = \{\mathbf{b} \in \mathbb{R}^m \mid \exists \mathbf{x} \in \mathbb{R}^n \text{ such that } \mathbf{b} = \mathbf{A}\mathbf{x}\} \subseteq \mathbb{R}^m$$

$$N(\mathbf{A}) = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{A}\mathbf{x} = \mathbf{0}\} \subseteq \mathbb{R}^n$$

Note that $C(\mathbf{A}) \subseteq \mathbb{R}^m$, $N(\mathbf{A}) \subseteq \mathbb{R}^n$ means that $C(\mathbf{A})$, $N(\mathbf{A})$ are subsets of $\mathbb{R}^m$, $\mathbb{R}^n$ respectively. In fact, we can make a stronger statement, that they are vector subspaces

$$C(\mathbf{A}) \leq \mathbb{R}^m, N(\mathbf{A}) \leq \mathbb{R}^n$$

Proof. Let $\mathbf{u}, \mathbf{v} \in C(\mathbf{A})$, $\alpha, \beta \in \mathcal{S}$. By definition of $C(\mathbf{A})$ there exist $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$ such that $\mathbf{u} = \mathbf{A}\mathbf{x}$ and $\mathbf{v} = \mathbf{A}\mathbf{y}$. Using vector space properties

$$\alpha\mathbf{u} + \beta\mathbf{v} = \alpha\mathbf{A}\mathbf{x} + \beta\mathbf{A}\mathbf{y} = \mathbf{A}(\alpha\mathbf{x} + \beta\mathbf{y}),$$

hence $\alpha\mathbf{u} + \beta\mathbf{v} \in C(\mathbf{A})$ (it is obtained as the image through the linear mapping $\mathbf{A}$ of $\alpha\mathbf{x} + \beta\mathbf{y}$)

- Recall that if $\mathbf{u}^T \mathbf{v} = 0$, with $\mathbf{u}, \mathbf{v} \in \mathbb{R}^m$ then $\mathbf{u} \perp \mathbf{v}$ (orthogonal)

Proposition. If $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n \in \mathbb{R}^m$ are non-zero ($\mathbf{u}_i \neq \mathbf{0}$) and pairwise orthogonal, $\mathbf{u}_i^T \mathbf{u}_j = 0$ for $i \neq j$ then they form a linearly independent set of vectors.

Proof. Consider the equation equating the linear combination $c_1 \mathbf{u}_1 + \dots + c_n \mathbf{u}_n$ to the zero vector

$$c_1 \mathbf{u}_1 + \dots + c_n \mathbf{u}_n = \mathbf{0} \tag{1}$$

Multiply on the left by $\mathbf{u}_i^T$ and use orthogonality to obtain $c_i = 0$ for $i = 1, \dots, n$. The only solution to (1) is $c_1 = c_2 = \dots = c_n = 0$, hence $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n\}$ is a linearly independent set.



- For $\mathbf{A} \in \mathbb{R}^{m \times n}$, seen as a linear mapping $\mathbf{A}: \mathbb{R}^n \mapsto \mathbb{R}^m$, that given input vector $\mathbf{x} \in \mathbb{R}^n$ returns output vector $\mathbf{b} \in \mathbb{R}^m$, $\mathbf{b} = \mathbf{A}\mathbf{x}$, we have defined the vector space of possible outputs, the column space of $\mathbf{A}$

$$C(\mathbf{A}) = \{\mathbf{b} \in \mathbb{R}^m \mid \exists \mathbf{x} \in \mathbb{R}^n \text{ such that } \mathbf{b} = \mathbf{A}\mathbf{x}\} \subseteq \mathbb{R}^m$$

- The transpose $\mathbf{A}^T \in \mathbb{R}^{n \times m}$ can also be seen as a linear mapping. Given some input vector $\mathbf{y} \in \mathbb{R}^m$ the mapping returns the output vector $\mathbf{c} \in \mathbb{R}^n$, $\mathbf{c} = \mathbf{A}^T\mathbf{y}$. The set of possible outputs is the column space of $\mathbf{A}^T$. Since columns of $\mathbf{A}^T$ are rows of $\mathbf{A}$, we can define the *row space* of $\mathbf{A}$ as

$$R(\mathbf{A}) = C(\mathbf{A}^T) = \{\mathbf{c} \in \mathbb{R}^n \mid \exists \mathbf{y} \in \mathbb{R}^m \text{ such that } \mathbf{c} = \mathbf{A}^T\mathbf{y}\} \subseteq \mathbb{R}^n$$

- Left null space*, $N(\mathbf{A}^T) = \{\mathbf{y} \in \mathbb{R}^m \mid \mathbf{A}^T\mathbf{y} = \mathbf{0}\} \subseteq \mathbb{R}^m$, the part of $\mathbb{R}^m$ *not reachable* by linear combination of columns of $\mathbf{A}$



Definition. A set of vectors $\mathbf{u}_1, \dots, \mathbf{u}_n \in \mathcal{V}$ is a **basis** for vector space $\mathcal{V}$ if:

1. $\mathbf{u}_1, \dots, \mathbf{u}_n$ are linearly independent;
2. $\text{span}\{\mathbf{u}_1, \dots, \mathbf{u}_n\} = \mathcal{V}$.

Definition. The number of vectors $\mathbf{u}_1, \dots, \mathbf{u}_n \in \mathcal{V}$ within a basis is the **dimension** of the vector space $\mathcal{V}$.



Definition. Given two vector subspaces $(\mathcal{U}, \mathcal{S}, +)$, $(\mathcal{V}, \mathcal{S}, +)$ of the space $(\mathcal{W}, \mathcal{S}, +)$, the **sum** is the set $\mathcal{U} + \mathcal{V} = \{\mathbf{u} + \mathbf{v} \mid \mathbf{u} \in \mathcal{U}, \mathbf{v} \in \mathcal{V}\}$.

Definition. Given two vector subspaces $(\mathcal{U}, \mathcal{S}, +)$, $(\mathcal{V}, \mathcal{S}, +)$ of the space $(\mathcal{W}, \mathcal{S}, +)$, the **direct sum** is the set $\mathcal{U} \oplus \mathcal{V} = \{\mathbf{u} + \mathbf{v} \mid \exists! \mathbf{u} \in \mathcal{U}, \exists! \mathbf{v} \in \mathcal{V}\}$. (unique decomposition)

Definition. Given two vector subspaces $(\mathcal{U}, \mathcal{S}, +)$, $(\mathcal{V}, \mathcal{S}, +)$ of the space $(\mathcal{W}, \mathcal{S}, +)$, the **intersection** is the set

$$\mathcal{U} \cap \mathcal{V} = \{\mathbf{x} \mid \mathbf{x} \in \mathcal{U}, \mathbf{x} \in \mathcal{V}\}.$$

Definition. Two vector subspaces $(\mathcal{U}, \mathcal{S}, +)$, $(\mathcal{V}, \mathcal{S}, +)$ of the space $(\mathcal{W}, \mathcal{S}, +)$ are **orthogonal subspaces**, denoted $\mathcal{U} \perp \mathcal{V}$ if $\mathbf{u}^T \mathbf{v} = 0$ for any $\mathbf{u} \in \mathcal{U}, \mathbf{v} \in \mathcal{V}$.

Definition. Two vector subspaces $(\mathcal{U}, \mathcal{S}, +)$, $(\mathcal{V}, \mathcal{S}, +)$ of the space $(\mathcal{W}, \mathcal{S}, +)$ are **orthogonal complements**, denoted $\mathcal{U} = \mathcal{V}^\perp$, $\mathcal{V} = \mathcal{U}^\perp$ if they are orthogonal subspaces and $\mathcal{U} \cap \mathcal{V} = \{\mathbf{0}\}$, i.e., the null vector is the only common element of both subspaces.