A Simple Understanding of Linear Independence and Linear Correlation

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Abstract

There are many important concepts in linear algebra, such as linear correlation and linear independence, eigenvalues and eigenvectors, and so on. Among them, linear correlation and linear independence have irreplaceable importance, and have important applications in fields such as algebra, signal processing, and artificial intelligence. This article provides a graphical explanation of how to distinguish between the concepts of linear correlation and linear independence, the method provided in the paper is easy to grasp. The conclusion points out that linear independence means that there are no two (base) vectors with the same direction in a vector graph; otherwise, it is a linear correlation.

Keywords

Linear Correlation, Linear Independence, Vector, Linear Algebra

1. Introduction

Linear algebra is a compulsory course for engineering majors in universities. There are many important concepts in linear algebra. For example, linear correlation and linear independence, solution space of equations, eigenvalues and eigenvectors, similarity transformation of matrices, congruent transformation, orthogonal transformation; diagonalization of matrices, and so on.

For beginners, it is difficult to truly understand the differences between the concepts of linear correlation and linear independence without specific methods. The intuitive form provided by graphics usually helps learners better understand the differences between these two concepts. So the way to learn linear algebra well is to accurately grasp the connections and differences between the above concepts through graphics.

This article mainly introduces how to distinguish between the concepts of li-
near correlation and linear independence (Buffa, Cho, & Sangalli, 2010; Floater & Quak, 2000; Farnoosh & Haibe-Kains, 2021; Veiga et al., 2013; Bownik & Speegle, 2013; Magalhes, 2021; Ashur, Khan, & Nyberg, 2022; Esmi et al., 2023).

Linear independence refers to the fact that no quantity in a set of data can be represented by other quantities, corresponding to linear correlation. In linear algebra, if there is no vector in a set of elements of a vector space that can be represented by a finite linear combination of other vectors, it is called linear independence. On the contrary, it is called linear correlation.

Let’s start with the simplest two-dimensional plane, as shown in Figure 1.

The decomposition of forces in a two-dimensional plane is a simple operation, where force F can be decomposed along the X and Y axes.

The same is true in three-dimensional space (in Figure 2), where point p in the figure above can be represented as \( OP = 4x + 5y + 3z \).

Expand this concept to n-dimensional space and imagine each coordinate axis as a vector, resulting in Figure 3.

So, point \( x \) (actually an n-dimensional vector) in Figure 3 can be represented as:

\[
X = k_1x_1 + k_2x_2 + \ldots + k_nx_n, \text{ where } k_1, k_2, k_n \text{ is the coordinate value of the coordinate axis. This expression is called a linear space, which means decomposing any vector x in an n-dimensional space to obtain the corresponding coordinate values.}
\]

So, what do the so-called linear independence and linear correlation mean?

From Figure 4, it can be seen that any point in the two-dimensional plane can be decomposed along the \( X (1, 0) \) and \( Y (0, 1) \) axes, or by the following two column vectors:

\[
\begin{bmatrix}
a \\
b
\end{bmatrix} = a \begin{bmatrix} 1 \\ 0 \end{bmatrix} + b \begin{bmatrix} 0 \\ 1 \end{bmatrix}
\]

But if you replace the \( X \) and \( Y \) axes with the \( X (1, 0) \) and \( Y (2, 0) \) axes:

\[
\begin{bmatrix}
a \\
b
\end{bmatrix} = a \begin{bmatrix} 1 \\ 3 \end{bmatrix} + b \begin{bmatrix} 2 \\ 4 \end{bmatrix}
\]
We see that when \( b \) is not equal to 0, the above figure is solveless, which means that the vector formed by point \( P (2, 3) \) in **Figure 2** cannot be decomposed along \((1, 0)\) and \((2, 0)\). In fact, \((1, 0)\) and \((2, 0)\) are the same vector, both
on the X-axis. In this case, the vector \( OP \) in Figure 2 cannot be decomposed solely along the X-axis because its vertical component cannot be obtained.

Linear independence refers to: \( k_1x_1 + k_2x_2 \), the equation system with \( kn = 0 \) only has 0 solutions, which is \( k_1, k_2, \) when \( kn \) must be equal to 0, this equation will be equal to 0. At this point, we call vectors \( x_1, x_2, x_3, Xn \) is linearly independent.

Referring to Figure 1, the so-called linear independence actually means that there are no vectors with the same direction in the \( n \) vectors in Figure 1. If there are, then these \( n \) vectors are linearly related.

For example, assuming \( x_1 \) and \( x_2 \) are two vectors \((1, 0)\) and \((2, 0)\), the equation system can be obtained from \( k_1x_1 + k_2x_2 = 0 \):

\[
\begin{align*}
k_1 + 2k_2 &= 0 \\
0k_1 + 0k_2 &= 0
\end{align*}
\]

The above equation system has non-zero solutions, so the vectors \((1, 0)\) and \((2, 0)\) are linearly correlated.

Assuming vectors \( x_1, x_2, x_3 \). If the \( xn \) vector forms matrix \( A \), then for the equation system \( Ax = 0 \), it is obvious that when the determinant of \( A \) is not equal to 0, there is only 0 solution, which means that \( x_1, x_2, x_3 \). The \( n \) vectors \( xn \) are linearly independent; if the determinant of \( A \) is equal to 0, then it is linearly correlated.

For the equation system \( Ax = b \), when the determinant of \( A \) is not equal to 0, the equation system has a unique solution, which is the vector decomposition \( x = k_1x_1 + k_2x_2 + \ldots + kn = 0 \) will obtain a set of determined \( k_1, k_2, \) the \( kn \) value is the coordinate value of the vector \( x \). And if the determinant of \( A \) is not equal to 0, it means that the vectors \( x_1, x_2, x_3, Xn \) is linearly independent (Zhao, 2021; Guo, Li, & Yang, 2023; Ma et al., 2022; Aparkin, 2021).

We know that if a determinant is equal to 0, it means that there are equal or proportional rows or columns in the determinant, and if two rows or columns are proportional, it precisely indicates that these two row or column vectors are vectors with the same direction, that is, the same vector.

### 2. Conclusion

Linear independence means that there are no two (basis) vectors with the same direction in Figure 1; otherwise, it is a linear correlation.

This article explains the concepts of linear correlation and linear independence through graphical methods. The intuitive nature of graphics makes abstract mathematical concepts more concrete, thereby reducing learning difficulty and enabling learners to gain a deeper understanding while learning these concepts, as well as enhancing memory and understanding.

### Conflicts of Interest

The authors declare no conflicts of interest regarding the publication of this paper.
References


