

## Linear (In)dependence.

Given a set of vectors  $v_1, v_2, \dots, v_k$ , we look at their **Linear Combinations**:

$c_1v_1 + c_2v_2 + \dots + c_kv_k$ . The trivial combination  $c_i = 0$  produces the zero **vector**, since  $0v_1 + 0v_2 + \dots + 0v_k = 0$ . The point is whether any other weights or **scalars** also produce it.

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If all non-trivial combinations of vectors are *non-zero*,  $c_1v_1 + c_2v_2 + \dots + c_kv_k \neq 0$ , unless  $c_1 = c_2 = \dots = c_k = 0$ , then the vectors  $v_1, v_2, \dots, v_k$  are **Linearly Independent**. Otherwise they are **linearly dependent**, and one of them is a **linear combination** of the others. e. g.

If one of the vectors, let's say,  $v_2$ , happen to be the zero vector, then we are certain that this combination is dependent. If we choose weights  $c_2 = 4$  and  $c_i = 0$ , this is certainly a nontrivial combination that yields the zero vector.

Ex 2: Let  $A$ :

$$A = \begin{pmatrix} 1 & 3 & 3 & 2 \\ 2 & 6 & 9 & 5 \\ -1 & -3 & 3 & 0 \end{pmatrix} \implies \begin{pmatrix} 1 & 3 & 3 & 2 \\ 0 & 0 & 3 & 1 \\ 0 & 0 & 6 & 2 \end{pmatrix}$$

Here we clearly see that row 3 is a combination of the other rows, so  $A$  has *linearly dependent rows*, we can also see dependent columns since column 2 is three times column 1.

The rows of the  $n \times n$  **Identity Matrix**  $I$ :

$$I = \begin{pmatrix} 1 & 0 & 0 & \cdot \\ 0 & 1 & 0 & \cdot \\ 0 & 0 & \cdot & 0 \\ \cdot & \cdot & 0 & 1 \end{pmatrix}$$

Are linearly independent. We give this vectors a special notation  $e_1, e_2, \dots, e_n$ , they are the unit vectors in the coordinate directions,  $e_1 = (1, 0, 0, \dots, 0)$ ,  $e_n = (0, 0, \dots, 1)$ .

## Procedure for Proving Independence

Assume that the linear combination gives zero, and prove that all weight  $c_i$  must equal zero, for example:

$$c_1e_1 + c_2e_2 + \cdots + c_ne_n = (c_1, c_2, \cdots, c_n)$$

If the combination is the zero vector then obviously all  $c_i = 0$ . e. g.

Suppose  $U$  is an  $n \times n$  **Upper Triangular Matrix**, with non-zero **pivots** in the diagonal. Then the rows of  $U$  are linearly independent.

*Proof:* We start by assuming that some linear combination of the rows is zero,  $c_1v_1 + c_2v_2 + \cdots + c_kv_k = 0$ . Then we head for the first non-zero entry in the diagonal  $u_{11}$ , since we know  $c_1v_1 = 0$  and  $v_1 \neq 0$ , this implies that  $c_1 = 0$ , and  $c_2 = 0$  since  $c_1v_1 + c_2v_2 = 0$  and  $v_2 \neq 0$ , this applies to all  $u_{ij}$  pivots, since the only weights that make  $c_1v_1 + c_2v_2 + \cdots + c_kv_k = 0$  are those of the trivial solution. Then  $U$  is linearly independent.

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The  $r$  nonzero rows of an **echelon** matrix  $U$  are linearly independent, and so are the  $r$  columns that contain nonzero pivots. → An important reminder is that the definition of linear independence is "coordinate free". Given  $k$  points in  $n$ -dimensional space, the vectors from the origin to those points either can or cannot be combined to give zero, regardless of where we put the coordinate axes. A rotation will change the coordinates however it won't affect the question of dependent or independent whatsoever. → Given an arbitrary **set** of vectors, their verification of dependency or independency of course requires some calculation,  $c_1v_1 + c_2v_2 + \cdots + c_kv_k$ , the natural step is to form a **matrix**  $A$ , whose columns are the given vectors. Then if we write  $c$  for the vector of weights:  $(c_1, c_2, \cdots, c_k)$ :

$$Ac = \begin{pmatrix} \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot \\ v_1 & v_2 & \cdots & v_k \\ \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \\ \vdots \\ c_k \end{pmatrix} = c_1v_1 + c_2v_2 + \cdots + c_kv_k$$

The vectors are dependent **if and only if there is a nontrivial solution for**  $Ac = 0$ . This is settled by **Gaussian Elimination**. If the **rank** of  $A = k$ , then there are no free variables and no **Nullspace**, (except for  $c = 0$ ), then the vectors are linearly independent. If the rank is less than  $k$  then there's at least one free variable that can be chosen nonzero and the columns are linearly dependent. → A really important thing is that if we let the vectors have  $m$  components, so that  $A$  is a  $m \times k$  matrix, and suppose now that  $k > m$ , it will be impossible for  $A$  to have rank  $k$ , since the number of **pivots** cannot surpass the number of rows. The rank must be less than  $k$  and a homogeneous system  $Ax = 0$  with more unknowns than equations always has nontrivial solutions  $x \neq 0$ .

| A Set of vectors  $K$  in  $\mathbb{R}^m$  with  $K > m$  is always linearly dependent.