

Review of Linear Independence

Definition: A set of vectors $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_k \in \mathbb{R}^n$ is called a **linear independent** set if

$$(1) \quad c_1\vec{v}_1 + \dots + c_k\vec{v}_k = \vec{0}, \quad c_1, \dots, c_k \in \mathbb{R} \\ \Rightarrow c_1 = c_2 = \dots = c_k = 0.$$

Note: **Linear independence** can be checked by **row reduction**. More precisely, if

$A = (\vec{v}_1 | \vec{v}_2 | \dots | \vec{v}_k)$ is the associated $n \times k$ matrix, then condition (1) holds if and only if

$$(2) \quad A\vec{w} = \vec{0} \Rightarrow \vec{w} = \vec{0}, \quad \text{for all } \vec{w} \in \mathbb{R}^k$$

because $A(c_1, \dots, c_k)^t = c_1\vec{v}_1 + \dots + c_k\vec{v}_k$. By **row reduction**, (2) holds if and only if

$$(3) \quad \text{rank}(A) = k,$$

where, if R is a **row echelon form** (REF) of A ,

$$\text{rank}(A) \stackrel{\text{defn}}{=} \# \text{ of nonzero rows of } R \\ = \# \text{ of leading 1's in } R.$$

Refinement: As before, let $A = (\vec{v}_1 | \vec{v}_2 | \dots | \vec{v}_k)$ be an $n \times k$ matrix, and let R be a REF of A . Put

$$I = \{i : \text{column } i \text{ of } R \text{ contains a leading } 1\}$$

so that (by definition) $\text{rank}(A) = \#I$.

Then we have:

- 1) the set $\{\vec{v}_i\}_{i \in I}$ is a linear independent set;
 - 2) the set $\{\vec{v}_i\}_{i \in I}$ spans the space $V = \langle \vec{v}_1, \dots, \vec{v}_k \rangle$.
- Thus, $\{\vec{v}_i\}_{i \in I}$ is a basis of V and hence

$$\dim V = \text{rank}(A) = \#I.$$

Example. Find a basis of $V = \langle \vec{v}_1, \dots, \vec{v}_4 \rangle$, where
 $\vec{v}_1 = (1, 2, 3, 4, 5)$, $\vec{v}_2 = (1, 2, 2, 2, 2)$,
 $\vec{v}_3 = (0, 0, 1, 2, 3)$, $\vec{v}_4 = (1, 1, 1, 1, 1) \in \mathbb{R}^5$.

Solution. Put $A = (\vec{v}_1 | \vec{v}_2 | \vec{v}_3 | \vec{v}_4)$. By row reduction we obtain

$$A = \begin{pmatrix} 1 & 1 & 0 & 1 \\ 2 & 2 & 0 & 1 \\ 3 & 2 & 1 & 1 \\ 4 & 2 & 2 & 1 \\ 5 & 2 & 3 & 1 \end{pmatrix} \leftrightarrow \begin{pmatrix} 1 & 1 & 0 & 1 \\ 0 & 1 & -1 & 2 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} = R.$$

and so the leading 1's of R (REF) are in columns 1, 2 and 4. Thus, $I = \{1, 2, 4\}$ and hence by 1), 2):

$$\{\vec{v}_1, \vec{v}_2, \vec{v}_4\} \quad \text{is a basis of } V.$$

In particular, $\dim V = \text{rank}(A) = \#I = 3$, so V is 3-dimensional.

Note: If $n = k$, then the linear independence of the vectors $\vec{v}_1, \dots, \vec{v}_n \in \mathbb{R}^n$ is equivalent to several other properties of the $n \times n$ matrix $A = (\vec{v}_1 | \dots | \vec{v}_n)$, as the following basic fact from Linear Algebra shows:

Fact: If A is an $n \times n$ matrix, then the following properties are equivalent:

- (i) A is **invertible**, i.e., there is a matrix B such that $AB = BA = I$.
- (ii) The columns of A are linearly independent.
- (iii) $A\vec{x} = \vec{0} \Rightarrow \vec{x} = \vec{0}$.
- (iv) $A\vec{x} = \vec{y}$ has a solution \vec{x} for every $\vec{y} \in \mathbb{R}^n$.
- (iv') $A\vec{x} = \vec{y}$ has a unique solution for every $\vec{y} \in \mathbb{R}^n$.
- (v) $\text{rank}(A) = n$.
- (vi) $\det(A) \neq 0$.

Proof Sketch:

- (ii) \Leftrightarrow (iii) \Leftrightarrow (v): See the above discussion.
- (i) \Rightarrow (iv)' \Rightarrow (iv): Take $\vec{x} = B\vec{y}$.
- (iv) \Rightarrow (i): Take $B = (\vec{v}_1 | \dots | \vec{v}_n)$, where $A\vec{v}_i = \vec{e}_i$.
- (i) \Rightarrow (vi): Since $\det(A) \det(B) = \det(AB) = \det(I) = 1$, we have that $\det(A) \neq 0$.
- (vi) \Rightarrow (i): Cofactor formula for $B = A^{-1}$.