

(1) Vectors $v_1, \dots, v_d \in \mathbb{R}^n$ are linearly independent if $\forall \alpha_1, \dots, \alpha_d \in \mathbb{R}$, we have $\sum_{i=1}^d \alpha_i v_i = 0$ if and only if

$\alpha_1 = \alpha_2 = \dots = \alpha_d = 0$. This implies, in particular, that if

v_1, \dots, v_d are linearly independent, we cannot write $v_i = \sum_{j \neq i} \beta_j v_j$

for coefficients $\beta_j \in \mathbb{R}$, as otherwise $-v_i + \sum_{j \neq i} \beta_j v_j = 0$.

~~(2) The span of a set of vectors v_1, \dots, v_d is the set $\{y \in \mathbb{R}^n \text{ such that } y = \sum_{i=1}^d \alpha_i v_i \text{ for some coefficients } \alpha_1, \dots, \alpha_d \in \mathbb{R}\}$.~~

(2) A subspace S of \mathbb{R}^n is a set of vectors such that
(1) if $y \in S$ then $\alpha y \in S$ for all $\alpha \in \mathbb{R}$, and
(2) if $x, y \in S$ then $x + y \in S$.

(3) The span of a set of vectors v_1, \dots, v_d is the set $\{y \in \mathbb{R}^n \text{ such that } y = \sum_{i=1}^d \alpha_i v_i \text{ for some coefficients } \alpha_1, \dots, \alpha_d \in \mathbb{R}\}$. Note that the span of any set of vectors is a subspace. If v_1, \dots, v_d are linearly independent, then d is called the dimension of the subspace, and v_1, \dots, v_d is called a basis.

(4) ~~A dxd matrix R~~ A $d \times d$ matrix R is said to be invertible or non-singular if there exists a matrix R^{-1} so that $R \cdot R^{-1} = R^{-1} \cdot R = I_d$, where I_d is the identity, so has d ones on the diagonal and is zero otherwise.

(5) If v_1, \dots, v_d is a basis of a d -dimensional subspace S , then $(VR)_1, (VR)_2, \dots, (VR)_d$ is also a basis for the same subspace S , where R is a $d \times d$ invertible matrix. Indeed $S = \{ \sum_{i=1}^d \alpha_i v_i \mid \alpha_i \in \mathbb{R} \}$. Let V be an $n \times d$ matrix whose columns are v_1, \dots, v_d . Then $S = \{ V\alpha \mid \alpha \in \mathbb{R}^d \}$. On the other hand the span of Rv_1, \dots, Rv_d is the set T of vectors $T = \{ VR\beta \mid \beta \in \mathbb{R}^d \}$. But for any y of the form $y = V\alpha$, if we set $\beta = R^{-1}\alpha$ we have that $y \in T$, and on the other hand if $y = VR\beta$, if we set $\alpha = R\beta$, we have $y \in S$.

(6) Let A be an $n \times d$ matrix, $n \geq d$, such that the columns of A are linearly independent. Then $A^T A$ is a $d \times d$ matrix and for all $x \neq 0$, $A^T A x \neq 0$. Indeed, suppose $A^T A x = 0$. Then $x^T (A^T A x) = 0$. But $x^T A^T A x = \|Ax\|_2^2 = \sum_{i=1}^n (Ax)_i^2$, and so this is equal to 0 if and only if $x = 0$.

(7) If a ^{symmetric} $d \times d$ matrix B is such that $Bx = 0$ if and only if $x = 0$, then there is a matrix B^{-1} for which ~~the~~ $B^{-1}B = I$.

It suffices to show for any vector y , there is a solution x to the equation $xB = y$. Then we can apply this to each row of I separately.

~~Need simple fact: (1) if you add a multiple of one row to another,~~

Need a simple fact: suppose you start with d linearly independent rows in \mathbb{R}^d . Then the following operations preserve linear independence

- (1) scaling a row by a non-zero real number λ
- (2) swapping two rows
- (3) adding a multiple of one row to another

Note that if there is no $x \neq 0$ for which $Bx = 0$, then there is also no $x \neq 0$ for which $x^T B = 0$, since B is symmetric.

The proof is an algorithm, called Gauss-Jordan algorithm.

Let's do an example.

$$\left[\begin{array}{ccc|ccc} 3 & 0 & 2 & 1 & 0 & 0 \\ 2 & 0 & -2 & 0 & 1 & 0 \\ 0 & 1 & 1 & 0 & 0 & 1 \end{array} \right]$$

$$\left[\begin{array}{ccc|ccc} 5 & 0 & 0 & 1 & 1 & 0 \\ 2 & 0 & -2 & 0 & 1 & 0 \\ 0 & 1 & 1 & 0 & 0 & 1 \end{array} \right] \leftarrow \text{ADD}$$

$$\left[\begin{array}{ccc|ccc} 1 & 0 & 0 & .2 & .2 & 0 \\ 2 & 0 & -2 & 0 & 1 & 0 \\ 0 & 1 & 1 & 0 & 0 & 1 \end{array} \right] \text{ Divide by 5}$$

$$\left[\begin{array}{ccc|ccc} 1 & 0 & 0 & .2 & .2 & 0 \\ 0 & 0 & -2 & -.4 & .6 & 0 \\ 0 & 1 & 1 & 0 & 0 & 1 \end{array} \right] \leftarrow \text{Subtract } \times 2$$

~~$$\left[\begin{array}{ccc|ccc} 1 & 0 & 0 & .2 & .2 & 0 \\ 0 & 0 & -2 & -.4 & .6 & 0 \\ 0 & 1 & 1 & 0 & 0 & 1 \end{array} \right]$$~~

$$\left[\begin{array}{ccc|ccc} 1 & 0 & 0 & .2 & .2 & 0 \\ 0 & 0 & 1 & .2 & -.3 & 0 \\ 0 & 1 & 1 & 0 & 0 & 1 \end{array} \right] \text{ Multiply by } -1/2$$

$$\left[\begin{array}{ccc|ccc} 1 & 0 & 0 & .2 & .2 & 0 \\ 0 & 1 & 1 & 0 & 0 & 1 \\ 0 & 0 & 1 & .2 & -.3 & 0 \end{array} \right] \leftarrow \text{Swap}$$

$$\left[\begin{array}{ccc|ccc} 1 & 0 & 0 & .2 & .2 & 0 \\ 0 & 1 & 0 & -.2 & .3 & 1 \\ 0 & 0 & 1 & .2 & -.3 & 0 \end{array} \right] \leftarrow \text{Subtract}$$

Why can you
always do
this?