

Mathematics-for-Machine-Learning

1. Getting handle on vectors:

a. Definition:

- i. Vectors are usually viewed by computers as an ordered list of numbers which they can perform "operations" on - some operations are very natural and, as we will see, very useful!
 - ii. A vector in space-time can be described using 3 dimensions of space and 1 dimension of time according to some co-ordinate system.
 - iii. Vectors can be thought of in a variety of different ways - some geometrically, some algebraically, some numerically. In this way, there are a lot of techniques one can use to deal with vectors.
- b. Vector is just something that are based on two rules. Firstly, addition, and secondly, multiplication by a scalar number.

i. Addition:

$$1. r = \begin{bmatrix} 3 \\ 2 \end{bmatrix}, s = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$$

$$2. r + s = s + r (\text{associativity}) = \begin{bmatrix} 3 - 1 \\ 2 + 2 \end{bmatrix} = \begin{bmatrix} 2 \\ 4 \end{bmatrix}$$

ii. Multiplication:

$$1. r = \begin{bmatrix} 3 \\ 2 \end{bmatrix},$$

$$2. 2r = \begin{bmatrix} 3 \cdot 2 \\ 2 \cdot 2 \end{bmatrix} = \begin{bmatrix} 6 \\ 4 \end{bmatrix}$$

2. Finding the size of vectors, its size and projection

a. $r = \begin{bmatrix} 3 \\ 2 \end{bmatrix}, s = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$

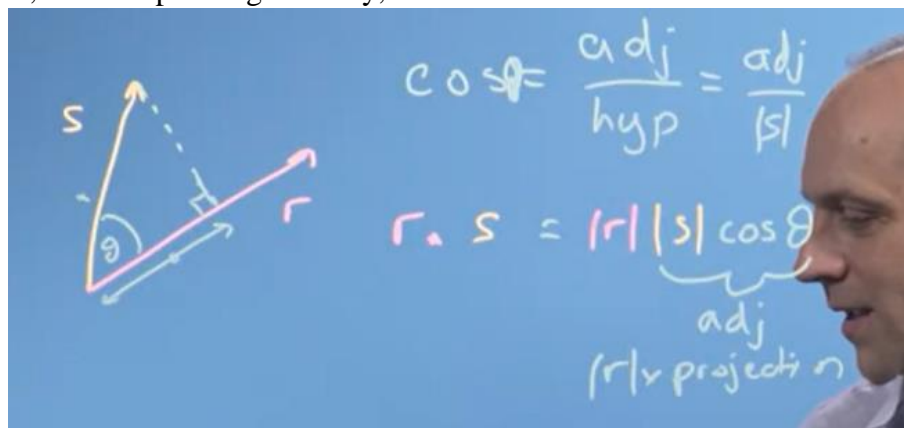
b. Size of vectors $= |r| = \sqrt{3^2 + 2^2} = \sqrt{9 + 4}$

c. Dot product of vectors:

i. $r \cdot s = s \cdot r = r_{11} \cdot s_{11} + r_{21} \cdot s_{21} = 3 \cdot (-1) + 2 \cdot 2 = 1$

d. Cosine rule $= r \cdot s = |r| \cdot |s| \cdot \cos \theta$

- e. Projection = dot product gives us, is it gives us the projection here of S on to R times the size of R. And one thing to notice here is that if S was perpendicular to R, if S was pointing this way, it would have no shadow.



i. Vector projection $= \frac{r \cdot s}{r \cdot r} \cdot r$

ii. Scalar projection $= \frac{r \cdot s}{r}$

3. Changing basis:

- a. Any vector space has multiple bases, so the question naturally arises: what are the relationships between bases of a vector space? In the first place, there must be the same number of elements in any basis of a vector space. Then, given two bases of a vector space, there is a way to translate vectors in terms of one basis into terms of the other; this is known as change of basis.

Change of basis is a technique applied to finite-dimensional vector spaces in order to rewrite vectors in terms of a different set of basis elements. It is useful for many types of matrix computations in linear algebra and can be viewed as a type of linear transformation.

b. $v = \begin{bmatrix} 10 \\ -5 \end{bmatrix}, b_1 = \begin{bmatrix} 3 \\ 4 \end{bmatrix}, b_2 = \begin{bmatrix} 4 \\ -3 \end{bmatrix}$

c. 1) $\frac{v \cdot b_1}{|b_1|^2} = \frac{2}{5}$

d. 2) $\frac{v \cdot b_2}{|b_2|^2} = \frac{11}{5}$

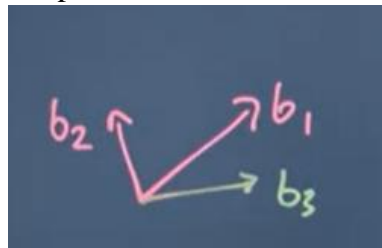
e. $v_b = \begin{bmatrix} \frac{2}{5} \\ \frac{11}{5} \end{bmatrix}$

4. Basis, vector space and linear independence

- a. Basis is a set of n vectors that:

- Are not linear combination of each other (Linear independent)
- Span the space
- The space is n -dimensional

- b. Linear independence:



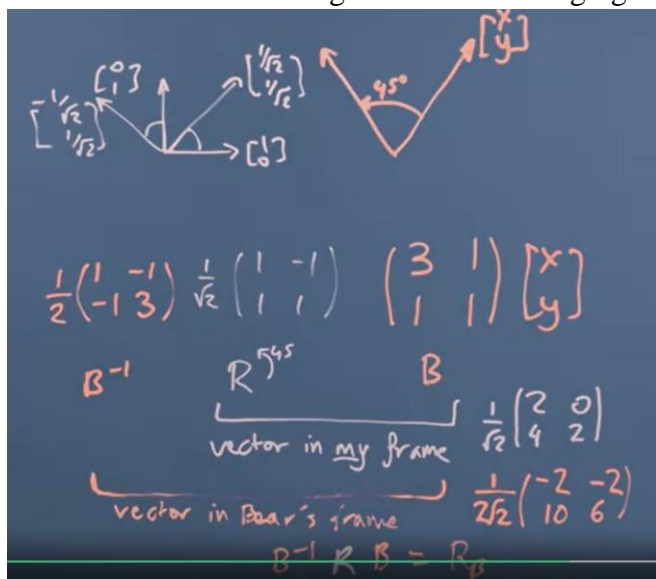
- $b_3 \neq a_1 b_1 + a_2 b_2$
- b_3 does not lie in the plane spanned by b_1 and b_2
- if $\det M \neq 0$, the given vectors are linearly independent

5. Gaussian elimination

- a. Gaussian elimination, also known as row reduction, is an algorithm in linear algebra for solving a system of linear equations. It is usually understood as a sequence of operations performed on the corresponding matrix of coefficients.

$$\left[\begin{array}{ccc|c} 1 & 3 & 1 & 9 \\ 1 & 1 & -1 & 1 \\ 3 & 11 & 5 & 35 \end{array} \right] \rightarrow \left[\begin{array}{ccc|c} 1 & 3 & 1 & 9 \\ 0 & -2 & -2 & -8 \\ 0 & 2 & 2 & 8 \end{array} \right] \rightarrow \left[\begin{array}{ccc|c} 1 & 3 & 1 & 9 \\ 0 & -2 & -2 & -8 \\ 0 & 0 & 0 & 0 \end{array} \right] \rightarrow \left[\begin{array}{ccc|c} 1 & 0 & -2 & -3 \\ 0 & 1 & 1 & 4 \\ 0 & 0 & 0 & 0 \end{array} \right]$$

6. Transformation in a changed basis and changing basis:



7. Orthogonal matrices

a. $A_{ij}^T = A_{ji}$, $A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$, $A^T = \begin{bmatrix} 1 & 3 \\ 2 & 4 \end{bmatrix}$

b. Orthogonal matrixes –

i. $Q^T \cdot Q = I$

ii. $Q^T = Q^{-1}$

c. A set of unit length basis vectors that are all perpendicular to each other are called an orthonormal basis set, and the matrix composed of them is called an orthogonal matrix. $a_i \cdot a_j = 0$ if $i \neq j$, $a_i \cdot a_j = 1$ if $i = j$

8. The Gram–Schmidt process

We define the [projection operator](#) by

$$\text{proj}_{\mathbf{u}}(\mathbf{v}) = \frac{\langle \mathbf{u}, \mathbf{v} \rangle}{\langle \mathbf{u}, \mathbf{u} \rangle} \mathbf{u},$$

where $\langle \mathbf{u}, \mathbf{v} \rangle$ denotes the [inner product](#) of the vectors \mathbf{u} and \mathbf{v} . This operator projects the vector \mathbf{v} orthogonally onto the line spanned by vector \mathbf{u} . If $\mathbf{u} = \mathbf{0}$, we define $\text{proj}_{\mathbf{0}}(\mathbf{v}) := \mathbf{0}$. i.e., the projection map $\text{proj}_{\mathbf{0}}$ is the zero map, sending every vector to the zero vector.

The Gram–Schmidt process then works as follows:

$$\begin{aligned} \mathbf{u}_1 &= \mathbf{v}_1, & \mathbf{e}_1 &= \frac{\mathbf{u}_1}{\|\mathbf{u}_1\|} \\ \mathbf{u}_2 &= \mathbf{v}_2 - \text{proj}_{\mathbf{u}_1}(\mathbf{v}_2), & \mathbf{e}_2 &= \frac{\mathbf{u}_2}{\|\mathbf{u}_2\|} \\ \mathbf{u}_3 &= \mathbf{v}_3 - \text{proj}_{\mathbf{u}_1}(\mathbf{v}_3) - \text{proj}_{\mathbf{u}_2}(\mathbf{v}_3), & \mathbf{e}_3 &= \frac{\mathbf{u}_3}{\|\mathbf{u}_3\|} \\ \mathbf{u}_4 &= \mathbf{v}_4 - \text{proj}_{\mathbf{u}_1}(\mathbf{v}_4) - \text{proj}_{\mathbf{u}_2}(\mathbf{v}_4) - \text{proj}_{\mathbf{u}_3}(\mathbf{v}_4), & \mathbf{e}_4 &= \frac{\mathbf{u}_4}{\|\mathbf{u}_4\|} \\ \vdots & & \vdots & \\ \mathbf{u}_k &= \mathbf{v}_k - \sum_{j=1}^{k-1} \text{proj}_{\mathbf{u}_j}(\mathbf{v}_k), & \mathbf{e}_k &= \frac{\mathbf{u}_k}{\|\mathbf{u}_k\|}. \end{aligned}$$

a. Gram–Schmidt process is a method for orthonormalizing a set of vectors in an inner product space, most commonly the Euclidean space

9. Transformation matrix for reflecting vectors in an arbitrarily angled mirror, construct an orthonormal basis that spans a set of input vectors $T = E T_E E^{-1}$, $r^{-1} = T \cdot r$

10. Eigenvalues and Eigenvectors:

- a. In linear algebra, an **eigenvector** or **characteristic vector** of a linear transformation is a nonzero vector that changes by a scalar factor when that linear transformation is applied to it. The corresponding eigenvalue, often denoted by λ is the factor by which the eigenvector is scaled.
- b. How to find eigenvectors and eigenvalues
 - i. $A = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$
 - ii. $Ax = \lambda x; (A - \lambda I)x = 0$
 - iii. $\det \left(\begin{bmatrix} -\lambda & -1 \\ 1 & -\lambda \end{bmatrix} \right) = \lambda^2 + 1; \lambda = 1 \text{ and } \lambda = -1$
 - iv. @ $\lambda = 1$
 1. $\begin{bmatrix} -1 & -1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$
 2. $\begin{cases} -x_1 - x_2 = 0 \\ x_1 - x_2 = 0 \end{cases}$
 3. $\begin{cases} x_2 = 0 \\ x_1 = 0 \end{cases}$
 - v. @ $\lambda = -1$
 1. $\begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$
 2. $\begin{cases} x_1 - x_2 = 0 \\ x_1 + x_2 = 0 \end{cases}$
 3. $\begin{cases} x_1 = x_2 \\ x_2 = 0 \end{cases}$
- c. To find T^n we use diagonalization for ex:
 - i. $T = \begin{bmatrix} 1 & 0 \\ 2 & -1 \end{bmatrix}$
 - ii. The find eigenvectors:
 1. $(1 - \lambda)(-1 - \lambda) = 0$
 2. $\lambda = 1 \text{ and } \lambda = -1$
 3. @ $\lambda = 1$
 - a. $\{x_1 = x_2\}$
 - b. $v_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$
 4. @ $\lambda = -1$
 - a. $\{x_1 = 0\}$
 - b. $v_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$
 - iii. $T^n = C \cdot D \cdot C^{-1}$
 - iv. $C = [v_1, v_2] = \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix}$
 - v. $C^{-1} = \begin{bmatrix} 1 & 0 \\ -1 & 1 \end{bmatrix}$
 - vi. $D = \begin{bmatrix} \lambda_1 & \\ & \lambda_2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$
 - vii. $T^n = C \cdot D^n \cdot C$