Definitions

Vector Space

Assume that u, v, w are vectors in V, and a, b, c are scalars in $\mathbb{R}$. A **vector space** is a set V with the following properties:

Commutativity:

- $u + v = v + u$

Associativity:

- $(u + v) + w = u + (v + w)$
- $(ab)v = a(bv)$

Additive Identity:

- there exists $0 \in V$ such that $v + 0 = v$ for all $v \in V$

Multiplicative Identity:

- for all $v \in V$, there exists $w \in V$ such that $v + w = 0$

Distributive Properties:

- $a(u + v) = au + av$
- $(a + b)v = av + bv$

Linear Combination

A linear combination of a list of vectors $v_1, \ldots, v_n$ is itself a vector, taking the form:

$$a_1 v_1 + \ldots + a_m v_m$$

where each $a_1, \ldots a_n \in \mathbb{R}$

Span

The set of all linear combinations of a list of vectors $v_1, \ldots, v_n$ is called the **span** of $v_1, \ldots, v_n$, or $\text{Span}(v_1, \ldots, v_n)$. Defined as:

$$\text{span}(v_1, \ldots, v_n) = \{a_1 v_1 + \cdots + a_n v_n \ : \ a_1, \ldots, a_m \in \mathbb{R}\}$$

If the span is equal to some space $\text{span}(v_1, \ldots, v_n) = V$, then you could say that $v_1, \ldots, v_n$ **spans** V.

Linearly Independent

For $v_1, \ldots, v_n \in V$ and $a_1, \ldots, a_n \in \mathbb{R}$ such that:

$$a_1 v_1 + \cdots + a_n v_n = 0$$

The list of vectors $v_1, \ldots, v_n$ is called **linearly independent** when

$$a_1 = \cdots = a_n = 0$$

for all possible values of $v_1, \ldots, v_n$.

Basis

A **basis** of V is a list of vectors in V that is both linearly independent and spans V.

The **Standard Basis** of the vector space $\mathbb{R}^{\ltimes}$ is

$$(1, 0, \ldots, 0), (0, 1, \ldots, 0), \ldots, (0, 0, \ldots, 1)$$

which could also be written, using matrix bracket notation, as:

$$\begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix}, \ldots, \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$$

Dimension

The dimension of a vector space is the length of any basis of the vector space. For example,

$$\dim \mathbb{R}^n = n$$

Inner Product

For a pair of vectors $u, v \in V$ in the same vector space (they are both in $\mathbb{R}^n$ for example), the Inner Product is defined as:

$$u \cdot v = u_1 v_1 + ... + u_n v_n$$

which is also sometimes written using angular brackets:

$$\langle u, v \rangle$$

Keep in mind that the dimension of u and v must be the same. Using matrix dimension notation:

$$u_{\{n \times 1\}} \cdot v_{\{n \times 1\}}$$

The **Inner Product** is also a function $f : (\mathbb{R}^n, \mathbb{R}^n) \to \mathbb{R}$. The input is an ordered pair of vectors, and the output is a number. Inner products have the following properties:

Positivity:

- $\langle v, v \rangle \geq 0$ for all $v \in V$

Definiteness:

- $\langle v, v \rangle = 0$ if and only if $v = 0$

Additivity in First Slot:

- $\langle u + v, w \rangle = \langle u, w \rangle + \langle v, w \rangle$ for all $u, v, w \in V$

Homogeneity in First Slot:

- $\langle au, v \rangle = a\langle u, v \rangle$ for all $a \in \mathbb{R}$ and all $u, v \in V$

In another definition of the Inner Product, the concepts of "additivity" and "homogeneity" are combined into a concept called "linearity". **Bilinearity** is when there is linearity in both the First and Second slots. Additionally, there is a concept called **Symmetry** for all real numbers.

For $x, y, z \in V$ and $a, b \in \mathbb{R}$:

Bilinearity:

- Additivity and Homogeneity in First and Second Slot:
- $\langle ax + by, z\rangle = a\langle x, z\rangle + b\langle y, z\rangle$
- $\langle x, ay + bz\rangle = a\langle x, y\rangle + b\langle x, z\rangle$

Symmetry:

- $\langle x, y\rangle = \langle y, x\rangle$

Norm

The Norm of a vector x is defined as the square root inner product of x with itself:

$$\|x\| = \sqrt{\langle x, x\rangle}$$

The Euclidean Norm, also called 2-norm, is defined:

$$\|x\|_2 = \sqrt{{x_1}^2 + \ldots + {x_n}^2}$$

which has the following properties:

Positivity:

- $\|x\| \geq 0$
- $\|x\| = 0$ if and only if $x = 0$

Homogeneity:

- $\|ax\| = |a|\|x\|$ for all $a \in \mathbb{R}$

Triangle Inequality:

- $\|x + y\| \leq \|x\| + \|y\|$

Orthogonal

Two vectors $u, v \in V$ are called **orthogonal** if the inner product between them is 0,

$$\langle u, v\rangle = 0$$

you could also say "u is orthogonal to v". Orthogonal is another way of saying "at right angles to each other", or "perpendicular".

Linear Map

A linear map from vector space V to vector space W is a function $T : V \to W$ with the following properties:

Additivity

- $T(u + v) = Tu + Tv$ for all vectors $u, w \in V$

Homogeneity

- $T(av) = a(Tv)$ for all $a \in \mathbb{R}$ and all $v \in V$

Linear Maps and Matrices

Suppose M is a linear map $f : \mathbb{R}^a \to \mathbb{R}^b$, then M can be written as b-by-a matrix:

$$\begin{bmatrix} x_{1,1} & \cdots & x_{1,a} \\ \vdots & \vdots & \vdots \\ x_{b,1} & \cdots & x_{b,a} \end{bmatrix}$$

Proofs

Law of Cosines

TODO

Triangle Inequality

TODO

Cauchy-Schwartz Inequality

TODO

Other

For vectors $u, v \in V$ such that:

$$\langle u, v\rangle = \|u\|\|v\| \cos\theta$$

Show that u, v are orthogonal when $\theta = 0$.