

**Reference Definitions**

**Inner products:** An inner product is a function that associates each pair of two vectors in a vector space  $V$  with a real number (called the inner product). For any  $\vec{x}, \vec{y}, \vec{z} \in V$  and  $c \in \mathbb{R}$ , the inner product satisfies the following three properties:

(a) **Symmetry:**  $\langle \vec{x}, \vec{y} \rangle = \langle \vec{y}, \vec{x} \rangle$

(b) **Linearity:**

i.  $\langle \vec{x}, \vec{y} + \vec{z} \rangle = \langle \vec{x}, \vec{y} \rangle + \langle \vec{x}, \vec{z} \rangle$

ii.  $\langle c\vec{x}, \vec{y} \rangle = c\langle \vec{x}, \vec{y} \rangle$

(c) **Positive-definiteness:**  $\langle \vec{x}, \vec{x} \rangle \geq 0$  with  $\langle \vec{x}, \vec{x} \rangle = 0$  if and only if  $\vec{x} = \vec{0}$

**Norm:** The norm of a vector  $\vec{x} \in V$  is defined to be:

$$\|\vec{x}\|^2 = \langle \vec{x}, \vec{x} \rangle \implies \|\vec{x}\| = \sqrt{\langle \vec{x}, \vec{x} \rangle}$$

**1. Packings**

(a) Can three vectors in the  $\mathbb{R}^2$  plane have only negative pairwise inner-products? That is, do there exist vectors  $\vec{u}, \vec{v}, \vec{w} \in \mathbb{R}^2$  such that  $\langle \vec{u}, \vec{v} \rangle < 0$ ,  $\langle \vec{v}, \vec{w} \rangle < 0$ , and  $\langle \vec{u}, \vec{w} \rangle < 0$ ?

*Hint:* Draw a picture!

(b) What about four vectors in  $\mathbb{R}^2$ ? That is, do there exist four vectors  $\vec{u}, \vec{v}, \vec{w}, \vec{x} \in \mathbb{R}^2$  such that for every pair of vectors  $\vec{a}, \vec{b}$ :  $\langle \vec{a}, \vec{b} \rangle < 0$ ?

*Bonus:* What about four vectors in  $\mathbb{R}^3$ ?

## 2. From Inner Products To Projections

Given that  $\langle \vec{x}, \vec{y} \rangle$  is a measure of similarity between two vectors, let's try to use this to find how much of one vector  $\vec{y}$  is in the direction of another vector  $\vec{x}$ .

- (a) Let's start with  $\langle \vec{x}, \vec{y} \rangle$ . We want a quantity that is independent of the norm of  $\vec{x}$ ,  $\|\vec{x}\|$ . Is  $\langle \vec{x}, \vec{y} \rangle$  independent of the norm? Consider  $\langle \vec{x}, \vec{y} \rangle$  for the examples below.

$$\vec{x} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \vec{y} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \vec{x} = \begin{bmatrix} 2 \\ 2 \end{bmatrix}, \vec{y} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

- (b) Suppose we divide  $\langle \vec{x}, \vec{y} \rangle$  by the norm of  $\vec{x}$ ,  $\|\vec{x}\|$ , to get  $\frac{\langle \vec{x}, \vec{y} \rangle}{\|\vec{x}\|}$ . Is this new quantity independent of the norm of  $\vec{x}$ ? Test it on the examples above.
- (c) We now have a scalar quantity that represents how much of  $\vec{y}$  is in the direction of  $\vec{x}$ . Let's try to find a vector that is how much of  $\vec{y}$  is in the  $\vec{x}$  direction. That is, we are looking for a vector  $\vec{z}$  that has a norm of  $\frac{\langle \vec{x}, \vec{y} \rangle}{\|\vec{x}\|}$  and points in the same direction as  $\vec{x}$ .
- (d) Given the projection between two vectors, defined as  $\text{proj}_{\vec{x}} \vec{y} = \frac{\langle \vec{x}, \vec{y} \rangle}{\|\vec{x}\|^2} \vec{x}$ , prove the Cauchy-Schwarz inequality,  $|\langle \vec{x}, \vec{y} \rangle| \leq \|\vec{x}\| \|\vec{y}\|$ .
- (e) Consider the quantity  $\frac{\langle \vec{x}, \vec{y} \rangle}{\|\vec{x}\| \|\vec{y}\|}$ . What is the maximum this quantity could be? When does this occur? What is the minimum this quantity could be? When does this occur?
- (f) We define the angle between two vectors as  $\cos(\theta) = \frac{\langle \vec{x}, \vec{y} \rangle}{\|\vec{x}\| \|\vec{y}\|}$ . When do two vectors have an angle of  $90^\circ$  between them? When do they have an angle of  $0^\circ$ ? When do they have an angle of  $180^\circ$ ?

## 3. Orthogonal Subspaces

Two vectors are  $\vec{x}$  and  $\vec{y}$  are said to be orthogonal if their inner product is zero. That is  $\langle \vec{x}, \vec{y} \rangle = 0$ .

Two subspaces  $\mathbb{S}_1$  and  $\mathbb{S}_2$  of  $\mathbb{R}^N$  are said to be orthogonal if all vectors in  $\mathbb{S}_1$  are orthogonal to all vectors in  $\mathbb{S}_2$ . That is,

$$\langle \vec{v}_1, \vec{v}_2 \rangle = 0 \quad \forall \vec{v}_1 \in \mathbb{S}_1, \vec{v}_2 \in \mathbb{S}_2.$$

- (a) Recall that the *row space* of an  $M \times N$  matrix  $\mathbf{A}$  is the subspace spanned by the rows of  $\mathbf{A}$  and that the *null space* of  $\mathbf{A}$  is the subspace of all vectors  $\vec{v}$  such that  $\mathbf{A}\vec{v} = \vec{0}$ .  
Prove that the row space and null space of any matrix are orthogonal subspaces. This can be denoted by  $\text{Col}(\mathbf{A}^T) \perp \text{Null}(\mathbf{A}) \quad \forall \mathbf{A} \in \mathbb{R}^{M \times N}$ .
- (b) Recall that the *column space* of an  $M \times N$  matrix  $\mathbf{A}$  is the subspace spanned by the columns of  $\mathbf{A}$  and that the *left null space* of  $\mathbf{A}$  is the subspace of all vectors  $\vec{v}$  such that  $\vec{v}^T \mathbf{A} = \vec{0}^T \iff \mathbf{A}^T \vec{v} = \vec{0}$ .  
Prove that the column space and left null space of any matrix are orthogonal subspaces. This can be denoted by  $\text{Col}(\mathbf{A}) \perp \text{Null}(\mathbf{A}^T) \quad \forall \mathbf{A} \in \mathbb{R}^{M \times N}$ .