$$\lambda = (\frac{6}{a})^{\frac{1}{4}}$$
. So set $\chi^2 = \sqrt{\frac{2}{2}}$ in \otimes

$$2\langle u, y \rangle \leqslant \langle \langle u, u \rangle \langle y, y \rangle + \langle \langle u, u \rangle \langle y, y \rangle$$

$$=) \langle u, x \rangle \leqslant \sqrt{\langle u, u \rangle \langle v, x \rangle} = \|u\| \|v\|$$

$$\Rightarrow \sqrt{\langle x, x \rangle} \in 11011111 (**) \forall x, x \in V.$$

If $\langle u, y \rangle \gg 0$ we are done.

$$|\langle u, v \rangle| = -\langle u, v \rangle = \langle u, -v \rangle \leq ||u|| - ||u|| ||u||$$

So in either
$$\cos$$
 $|\langle u, x \rangle| \leq ||u|| ||x|||$

It follows immediately from the Cauchy-Schwarz inequality that

$$-1 \le \frac{\langle \mathbf{u}, \mathbf{v} \rangle}{\|\mathbf{u}\| \|\mathbf{v}\|} \le 1. \tag{7.4}$$

This then allows us to define the **angle** between pairs of vectors in real inner product spaces, by analogy with the usual definition for the Euclidean inner product.

Definition 7.2.2. Let V be a real inner product space & let $\mathbf{u}, \mathbf{v} \in V$. The **angle** between \mathbf{u} and \mathbf{v} is

$$\theta = \arccos\left(\frac{\langle \mathbf{u}, \mathbf{v} \rangle}{\|\mathbf{u}\| \|\mathbf{v}\|}\right).$$

Note that (7.4) guarantees that \arccos is only invoked on its domain, so θ is always well-defined.

E.g. 7.2.3. Consider $C[0, 2\pi]$ endowed with the inner product given in E.g. 7.1.7. Find the angle between $\sin(x)$ and $\cos(x)$.

$$\langle \sin, \cos \rangle = \int_{0}^{2\pi} \sin(x) \cos(x) dx = 0$$

 $\Rightarrow \theta = \arccos(0) = \frac{\pi}{2}$

7.2.2 Orthogonality

The angle between two elements of a real inner product space is $\pi/2$ iff their inner product is 0.

Definition 7.2.4. If V is a real inner product space and $\mathbf{u}, \mathbf{v} \in V$ then we say \mathbf{u} and \mathbf{v} are **orthogonal** if $\langle \mathbf{u}, \mathbf{v} \rangle = 0$.

E.g. 7.2.3 shows that \cos and \sin are orthogonal, on the space $C[0, 2\pi]$ with inner product (7.3).

E.g. 7.2.5. Consider $\mathbf{u} = (1, 1)$ and $\mathbf{v} = (1, -1)$.

- (a) Are **u** and **v** orthogonal with respect to the Euclidean inner product?
- (b) Are \mathbf{u} and \mathbf{v} orthogonal with respect to the weighted inner product $\langle \mathbf{u}, \mathbf{v} \rangle = 3u_1 v_1 + 2u_2 v_2$?

(a) u. v = 0 → u & x are orthogonal

(b) $\langle u, x \rangle = 3 - 2 = 1$ $\Rightarrow \quad \text{use } x \text{ are not orthogonal}$

Theorem 7.2.6 (Generalized Pythagorean theorem). If V is a real inner product space and $\mathbf{u}, \mathbf{v} \in V$ are orthogonal, then

$$\|\mathbf{u} + \mathbf{v}\|^2 = \|\mathbf{u}\|^2 + \|\mathbf{v}\|^2.$$

$$|u+y|^2 = \langle u+y, u+y \rangle$$

$$= \langle u+y, u \rangle + \langle u+y, y \rangle$$

$$= \langle u, u \rangle + \langle v, u \rangle + \langle u, y \rangle + \langle v, y \rangle$$

$$= |u|^2 + |v|^2$$



Definition 7.2.7. Let U and W be subspaces of an inner product space V. Then U and W are **orthogonal** if every vector of U is orthogonal to every vector of W. This is denoted $U \perp W$.

Lemma 7.2.8. Let U and W be subspaces of an inner product space V. If $U \perp W$, then $U \cap W = \{0\}$.

Identical proof to lemma 4.7.4 Proof.

Corollary 7.2.9. If U and W are orthogonal subspaces of an inner product space V, then their sum is direct, i.e. $U + W = U \oplus W$.

Definition 7.2.10. Let U be a subspace of an inner product space V. The orthogonal complement of U is the set U^{\perp} of all vectors in V that are orthogonal to every vector of U.

Lemma 7.2.11. Let U be a subspace of an inner product space V. Then U^{\perp} is a subspace of V.

Proof.

Identical proof to herma 4.7.6

Lemma 7.2.12. Let $U = \operatorname{span}\{\mathbf{u}_1, \dots, \mathbf{u}_r\} \subset V$, where V is an inner product space. Then

$$U^{\perp} = \{ \mathbf{v} \in V \text{ such that } \langle \mathbf{v}, \mathbf{u}_i \rangle = 0 \text{ for all } i = 1, \dots, r \}.$$

Proof.

Prob. Set

7.3 Gram-Schmidt Algorithm (A&R §6.3)

In an inner product space, certain bases are especially convenient to work with – orthonormal bases. The Gram-Schmidt algorithm shows us how to construct such bases.

Definition 7.3.1. A set of two or more vectors in a real inner product space are **orthogonal** if all pairs of vectors in the set are orthogonal. If, in addition, each vector has length 1, the set is **orthonormal**.

If v is a nonzero vector, then $\mathbf{v}/\|\mathbf{v}\|$ is a unit vector. The process of dividing a vector by its length is called **normalization**. Any orthogonal set of nonzero vectors can be normalized to obtain an orthonormal set.

E.g. 7.3.2. Verify that the set $S = \{(1,1,1), (0,1,-1), (2,-1,-1)\}$ is orthogonal in Euclidean 3-space. Is S orthonormal? If not, normalize each element of S to obtain an orthonormal set.

$$(1,1,1) \cdot (0,1,-1) = 0$$

$$(1,1,1) \cdot (2,-1,-1) = 0$$

$$(0,1,-1) \cdot (2,-1,-1) = 0$$

$$\|(1,1,1)\| = \sqrt{3}$$

$$\|(1,1,1)\| = \sqrt{3}$$

$$\|(1,1,1)\| = \sqrt{3}$$

$$\|(1,1,1)\| = \sqrt{2}$$

$$\|(2,-1,-1)\| = \sqrt{2}$$
But $\begin{cases} 1 \\ \sqrt{3} \end{cases} (1,1,1) , 1 (0,1,-1) , 1 (0,1,-1) \end{cases}$ is arthonormal

Theorem 7.3.3. If $S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ is an orthogonal set of non-zero vectors in an inner product space, then S is linearly independent.

Proof.

Then
$$\langle \underline{x}_i, \overline{y}_i = 0 \rangle \otimes \text{let } 1 \leq i \leq n$$
.

But $\langle \underline{x}_i, \underline{y}_i \rangle = 0$ (Prob Set), so linearity \Rightarrow

$$\widehat{\underline{y}}_i = \widehat{\underline{y}}_i \otimes \widehat{\underline{y}_i \otimes \widehat{\underline{y}}_i \otimes \widehat{\underline{y}_i \otimes \widehat{\underline{y}}_i \otimes \widehat{\underline{y}_i \otimes \underline{y}_i \otimes \widehat{\underline{y}}_i \otimes \widehat{\underline{y}_i \otimes \underline{y}_i \otimes \widehat{\underline{y}_i \otimes \underline{y}_i \otimes \widehat{\underline{y}_i \otimes \underline{y}_i \otimes \widehat{\underline{y}_i \otimes \underline{y}_i \otimes \widehat{\underline{$$

Corollary 7.3.4. If $S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ is an orthonormal subset of an inner product space, then S is linearly independent.

Proof. If S is orthonormal, then $\|Y_i\| = 1$ $\forall 1 \leq i \leq n \Rightarrow Y_i \neq Q \quad \forall 1 \leq i \leq n \quad So$ S is an orthogonal set of non zero vector. It is particularly easy to express a vector **u** in an inner product space as a linear combination of an orthonormal basis.

Theorem 7.3.6. If $B = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ is an orthonormal basis for an inner product space V, then for any $\mathbf{u} \in V$ we have

$$\mathbf{u} = \langle \mathbf{u}, \mathbf{v}_{1} \rangle \mathbf{v}_{1} + \langle \mathbf{u}, \mathbf{v}_{2} \rangle \mathbf{v}_{2} + \dots + \langle \mathbf{u}, \mathbf{v}_{n} \rangle \mathbf{v}_{n}.$$

Proof.

$$\langle \mathbf{u}, \mathbf{v}_{i} \rangle = \sum_{j=1}^{n} c_{j} \mathbf{v}_{j}, \mathbf{v}_{i} \rangle$$

$$= \sum_{j=1}^{n} c_{j} \langle \mathbf{v}_{j}, \mathbf{v}_{i} \rangle \qquad \text{by linearity}$$

$$= c_{i} \qquad \text{since} \qquad \langle \mathbf{v}_{j}, \mathbf{v}_{i} \rangle = \begin{cases} 0, & i \neq j \\ i, & i = j \end{cases}$$

$$\leq \mathbf{v}_{i} = \mathbf{v$$