MATH 323 Linear Algebra

Lecture 21: The Gram-Schmidt orthogonalization process.

# **Orthogonal sets**

Let V be an inner product space with an inner product  $\langle \cdot, \cdot \rangle$  and the induced norm  $\|\mathbf{v}\| = \sqrt{\langle \mathbf{v}, \mathbf{v} \rangle}$ .

*Definition.* Nonzero vectors  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k \in V$  form an **orthogonal set** if they are orthogonal to each other:  $\langle \mathbf{v}_i, \mathbf{v}_j \rangle = 0$  for  $i \neq j$ .

If, in addition, all vectors are of unit norm,  $\|\mathbf{v}_i\| = 1$ , then  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k$  is called an **orthonormal set**.

**Theorem** Any orthogonal set is linearly independent.

# **Orthogonal projection**

**Theorem** Let V be an inner product space and  $V_0$  be a finite-dimensional subspace of V. Then any vector  $\mathbf{x} \in V$  is uniquely represented as  $\mathbf{x} = \mathbf{p} + \mathbf{o}$ , where  $\mathbf{p} \in V_0$  and  $\mathbf{o} \perp V_0$ .

The component **p** is called the **orthogonal projection** of the vector **x** onto the subspace  $V_0$ .



The projection **p** is closer to **x** than any other vector in  $V_0$ . Hence the distance from **x** to  $V_0$  is  $||\mathbf{x} - \mathbf{p}|| = ||\mathbf{o}||$ . Let V be an inner product space. Let **p** be the orthogonal projection of a vector  $\mathbf{x} \in V$  onto a finite-dimensional subspace  $V_0$ .

If  $V_0$  is a one-dimensional subspace spanned by a vector  $\mathbf{v}$  then  $\mathbf{p} = \frac{\langle \mathbf{x}, \mathbf{v} \rangle}{\langle \mathbf{v}, \mathbf{v} \rangle} \mathbf{v}$ .

If  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  is an orthogonal basis for  $V_0$  then  $\mathbf{p} = \frac{\langle \mathbf{x}, \mathbf{v}_1 \rangle}{\langle \mathbf{v}_1, \mathbf{v}_1 \rangle} \mathbf{v}_1 + \frac{\langle \mathbf{x}, \mathbf{v}_2 \rangle}{\langle \mathbf{v}_2, \mathbf{v}_2 \rangle} \mathbf{v}_2 + \dots + \frac{\langle \mathbf{x}, \mathbf{v}_n \rangle}{\langle \mathbf{v}_n, \mathbf{v}_n \rangle} \mathbf{v}_n.$ Indeed,  $\langle \mathbf{p}, \mathbf{v}_i \rangle = \sum_{j=1}^n \frac{\langle \mathbf{x}, \mathbf{v}_j \rangle}{\langle \mathbf{v}_j, \mathbf{v}_j \rangle} \langle \mathbf{v}_j, \mathbf{v}_i \rangle = \frac{\langle \mathbf{x}, \mathbf{v}_i \rangle}{\langle \mathbf{v}_i, \mathbf{v}_i \rangle} \langle \mathbf{v}_i, \mathbf{v}_i \rangle = \langle \mathbf{x}, \mathbf{v}_i \rangle$   $\implies \langle \mathbf{x} - \mathbf{p}, \mathbf{v}_i \rangle = 0 \implies \mathbf{x} - \mathbf{p} \perp \mathbf{v}_i \implies \mathbf{x} - \mathbf{p} \perp V_0.$ 

### The Gram-Schmidt orthogonalization process

Let V be a vector space with an inner product. Suppose  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$  is a basis for V. Let



Then  $\mathbf{v}_1, \mathbf{v}_2, \ldots, \mathbf{v}_n$  is an orthogonal basis for V.





Properties of the Gram-Schmidt process:

• 
$$\mathbf{v}_k = \mathbf{x}_k - (\alpha_1 \mathbf{x}_1 + \dots + \alpha_{k-1} \mathbf{x}_{k-1}), \ 1 \le k \le n;$$

• the span of  $\mathbf{v}_1, \ldots, \mathbf{v}_{k-1}$  is the same as the span of  $\mathbf{x}_1, \ldots, \mathbf{x}_{k-1}$ ;

•  $\mathbf{v}_k$  is orthogonal to  $\mathbf{x}_1, \ldots, \mathbf{x}_{k-1}$ ;

•  $\mathbf{v}_k = \mathbf{x}_k - \mathbf{p}_k$ , where  $\mathbf{p}_k$  is the orthogonal projection of the vector  $\mathbf{x}_k$  on the subspace spanned by  $\mathbf{x}_1, \ldots, \mathbf{x}_{k-1}$ ;

•  $\|\mathbf{v}_k\|$  is the distance from  $\mathbf{x}_k$  to the subspace spanned by  $\mathbf{x}_1, \ldots, \mathbf{x}_{k-1}$ .

## Normalization

Let V be a vector space with an inner product. Suppose  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  is an orthogonal basis for V.

Let 
$$\mathbf{w}_1 = \frac{\mathbf{v}_1}{\|\mathbf{v}_1\|}$$
,  $\mathbf{w}_2 = \frac{\mathbf{v}_2}{\|\mathbf{v}_2\|}$ ,...,  $\mathbf{w}_n = \frac{\mathbf{v}_n}{\|\mathbf{v}_n\|}$ .

Then  $\mathbf{w}_1, \mathbf{w}_2, \ldots, \mathbf{w}_n$  is an orthonormal basis for V.

**Theorem** Any finite-dimensional vector space with an inner product has an orthonormal basis.

*Remark.* An infinite-dimensional vector space with an inner product may or may not have an orthonormal basis.

# **Orthogonalization / Normalization**

An alternative form of the Gram-Schmidt process combines orthogonalization with normalization.

Suppose  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$  is a basis for an inner product space V. Let

 $v_1 = x_1, \quad w_1 = \frac{v_1}{\|v_1\|},$  $\mathbf{v}_2 = \mathbf{x}_2 - \langle \mathbf{x}_2, \mathbf{w}_1 
angle \mathbf{w}_1$ ,  $\mathbf{w}_2 = rac{\mathbf{v}_2}{\|\mathbf{v}_2\|}$ ,  $\mathbf{v}_3 = \mathbf{x}_3 - \langle \mathbf{x}_3, \mathbf{w}_1 
angle \mathbf{w}_1 - \langle \mathbf{x}_3, \mathbf{w}_2 
angle \mathbf{w}_2$ ,  $\mathbf{w}_3 = rac{\mathbf{v}_3}{\|\mathbf{v}_3\|}$ ,  $\mathbf{v}_n = \mathbf{x}_n - \langle \mathbf{x}_n, \mathbf{w}_1 \rangle \mathbf{w}_1 - \cdots - \langle \mathbf{x}_n, \mathbf{w}_{n-1} \rangle \mathbf{w}_{n-1},$  $\mathbf{w}_n = \frac{\mathbf{v}_n}{\|\mathbf{v}_n\|}.$ Then  $\mathbf{w}_1, \mathbf{w}_2, \ldots, \mathbf{w}_n$  is an orthonormal basis for V.

# Problem. Let V<sub>0</sub> be a subspace of dimension k in R<sup>n</sup>. Let x<sub>1</sub>, x<sub>2</sub>, ..., x<sub>k</sub> be a basis for V<sub>0</sub>. (i) Find an orthogonal basis for V<sub>0</sub>. (ii) Extend it to an orthogonal basis for R<sup>n</sup>.

Approach 1. Extend  $\mathbf{x}_1, \ldots, \mathbf{x}_k$  to a basis  $\mathbf{x}_1, \mathbf{x}_2, \ldots, \mathbf{x}_n$  for  $\mathbb{R}^n$ . Then apply the Gram-Schmidt process to the extended basis. We shall obtain an orthogonal basis  $\mathbf{v}_1, \ldots, \mathbf{v}_n$  for  $\mathbb{R}^n$ . By construction,  $\operatorname{Span}(\mathbf{v}_1, \ldots, \mathbf{v}_k) = \operatorname{Span}(\mathbf{x}_1, \ldots, \mathbf{x}_k) = V_0$ . It follows that  $\mathbf{v}_1, \ldots, \mathbf{v}_k$  is a basis for  $V_0$ . Clearly, it is orthogonal.

Approach 2. First apply the Gram-Schmidt process to  $\mathbf{x}_1, \ldots, \mathbf{x}_k$  and obtain an orthogonal basis  $\mathbf{v}_1, \ldots, \mathbf{v}_k$  for  $V_0$ . Secondly, find a basis  $\mathbf{y}_1, \ldots, \mathbf{y}_m$  for the orthogonal complement  $V_0^{\perp}$  and apply the Gram-Schmidt process to it obtaining an orthogonal basis  $\mathbf{u}_1, \ldots, \mathbf{u}_m$  for  $V_0^{\perp}$ . Then  $\mathbf{v}_1, \ldots, \mathbf{v}_k, \mathbf{u}_1, \ldots, \mathbf{u}_m$  is an orthogonal basis for  $\mathbb{R}^n$ . Problem. Let Π be the plane in R<sup>3</sup> spanned by vectors x<sub>1</sub> = (1, 2, 2) and x<sub>2</sub> = (-1, 0, 2).
(i) Find an orthonormal basis for Π.
(ii) Extend it to an orthonormal basis for R<sup>3</sup>.

 $\mathbf{x}_1, \mathbf{x}_2$  is a basis for the plane  $\Pi$ . We can extend it to a basis for  $\mathbb{R}^3$  by adding one vector from the standard basis. For instance, vectors  $\mathbf{x}_1, \mathbf{x}_2$ , and  $\mathbf{x}_3 = (0, 0, 1)$  form a basis for  $\mathbb{R}^3$  because

$$\begin{vmatrix} 1 & 2 & 2 \\ -1 & 0 & 2 \\ 0 & 0 & 1 \end{vmatrix} = \begin{vmatrix} 1 & 2 \\ -1 & 0 \end{vmatrix} = 2 \neq 0$$

Using the Gram-Schmidt process, we orthogonalize the basis  $\mathbf{x}_1 = (1, 2, 2), \mathbf{x}_2 = (-1, 0, 2), \mathbf{x}_3 = (0, 0, 1)$ :  $\mathbf{v}_1 = \mathbf{x}_1 = (1, 2, 2).$  $\mathbf{v}_2 = \mathbf{x}_2 - \frac{\langle \mathbf{x}_2, \mathbf{v}_1 \rangle}{\langle \mathbf{v}_1, \mathbf{v}_1 \rangle} \mathbf{v}_1 = (-1, 0, 2) - \frac{3}{9} (1, 2, 2)$ = (-4/3, -2/3, 4/3). $\mathbf{v}_3 = \mathbf{x}_3 - \frac{\langle \mathbf{x}_3, \mathbf{v}_1 \rangle}{\langle \mathbf{v}_1, \mathbf{v}_1 \rangle} \mathbf{v}_1 - \frac{\langle \mathbf{x}_3, \mathbf{v}_2 \rangle}{\langle \mathbf{v}_2, \mathbf{v}_2 \rangle} \mathbf{v}_2$  $= (0,0,1) - \frac{2}{9}(1,2,2) - \frac{4/3}{4}(-4/3,-2/3,4/3)$ = (2/9, -2/9, 1/9).

Now  $\mathbf{v}_1 = (1, 2, 2)$ ,  $\mathbf{v}_2 = (-4/3, -2/3, 4/3)$ ,  $\mathbf{v}_3 = (2/9, -2/9, 1/9)$  is an orthogonal basis for  $\mathbb{R}^3$  while  $\mathbf{v}_1, \mathbf{v}_2$  is an orthogonal basis for  $\Pi$ . It remains to normalize these vectors.

 $\mathbf{w}_1, \mathbf{w}_2$  is an orthonormal basis for  $\Pi$ .  $\mathbf{w}_1, \mathbf{w}_2, \mathbf{w}_3$  is an orthonormal basis for  $\mathbb{R}^3$ . **Problem.** Find the distance from the point  $\mathbf{y} = (0, 0, 0, 1)$  to the subspace  $V \subset \mathbb{R}^4$  spanned by vectors  $\mathbf{x}_1 = (1, -1, 1, -1)$ ,  $\mathbf{x}_2 = (1, 1, 3, -1)$ , and  $\mathbf{x}_3 = (-3, 7, 1, 3)$ .

First we apply the Gram-Schmidt process to vectors  $\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3$  to obtain an orthogonal basis  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$  for the subspace V. Next we compute the orthogonal projection  $\mathbf{p}$  of the vector  $\mathbf{y}$  onto V:

$$\mathbf{p} = \frac{\langle \mathbf{y}, \mathbf{v}_1 \rangle}{\langle \mathbf{v}_1, \mathbf{v}_1 \rangle} \mathbf{v}_1 + \frac{\langle \mathbf{y}, \mathbf{v}_2 \rangle}{\langle \mathbf{v}_2, \mathbf{v}_2 \rangle} \mathbf{v}_2 + \frac{\langle \mathbf{y}, \mathbf{v}_3 \rangle}{\langle \mathbf{v}_3, \mathbf{v}_3 \rangle} \mathbf{v}_3.$$

Then the distance from **y** to V equals  $\|\mathbf{y} - \mathbf{p}\|$ .

Alternatively, we can apply the Gram-Schmidt process to vectors  $\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3, \mathbf{y}$ . We should obtain an orthogonal system  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3, \mathbf{v}_4$ . By construction,  $\mathbf{v}_4 = \mathbf{y} - \mathbf{p}$  so that the desired distance will be  $\|\mathbf{v}_4\|$ .

$$\begin{aligned} \mathbf{x}_{1} &= (1, -1, 1, -1), \ \mathbf{x}_{2} &= (1, 1, 3, -1), \\ \mathbf{x}_{3} &= (-3, 7, 1, 3), \ \mathbf{y} &= (0, 0, 0, 1). \end{aligned}$$
$$\mathbf{v}_{1} &= \mathbf{x}_{1} &= (1, -1, 1, -1), \\ \mathbf{v}_{2} &= \mathbf{x}_{2} - \frac{\langle \mathbf{x}_{2}, \mathbf{v}_{1} \rangle}{\langle \mathbf{v}_{1}, \mathbf{v}_{1} \rangle} \mathbf{v}_{1} &= (1, 1, 3, -1) - \frac{4}{4} (1, -1, 1, -1) \\ &= (0, 2, 2, 0), \\ \mathbf{v}_{3} &= \mathbf{x}_{3} - \frac{\langle \mathbf{x}_{3}, \mathbf{v}_{1} \rangle}{\langle \mathbf{v}_{1}, \mathbf{v}_{1} \rangle} \mathbf{v}_{1} - \frac{\langle \mathbf{x}_{3}, \mathbf{v}_{2} \rangle}{\langle \mathbf{v}_{2}, \mathbf{v}_{2} \rangle} \mathbf{v}_{2} \\ &= (-3, 7, 1, 3) - \frac{-12}{4} (1, -1, 1, -1) - \frac{16}{8} (0, 2, 2, 0) \\ &= (0, 0, 0, 0). \end{aligned}$$

The Gram-Schmidt process can be used to check linear independence of vectors! It failed because the vector  $\mathbf{x}_3$  is a linear combination of  $\mathbf{x}_1$  and  $\mathbf{x}_2$ . V is a plane, not a 3-dimensional subspace. To fix things, it is enough to drop  $\mathbf{x}_3$ , i.e., we should orthogonalize vectors  $\mathbf{x}_1, \mathbf{x}_2, \mathbf{y}$ .

$$\begin{split} \tilde{\mathbf{v}}_3 &= \mathbf{y} - \frac{\langle \mathbf{y}, \mathbf{v}_1 \rangle}{\langle \mathbf{v}_1, \mathbf{v}_1 \rangle} \mathbf{v}_1 - \frac{\langle \mathbf{y}, \mathbf{v}_2 \rangle}{\langle \mathbf{v}_2, \mathbf{v}_2 \rangle} \mathbf{v}_2 \\ &= (0, 0, 0, 1) - \frac{-1}{4} (1, -1, 1, -1) - \frac{0}{8} (0, 2, 2, 0) \\ &= (1/4, -1/4, 1/4, 3/4). \\ \tilde{\mathbf{v}}_3 &| = \left| \left( \frac{1}{4}, -\frac{1}{4}, \frac{1}{4}, \frac{3}{4} \right) \right| = \frac{1}{4} \left| (1, -1, 1, 3) \right| = \frac{\sqrt{12}}{4} = \frac{\sqrt{3}}{2}. \end{split}$$

**Problem.** Find the distance from the point  $\mathbf{z} = (0, 0, 1, 0)$  to the plane  $\Pi$  that passes through the point  $\mathbf{x}_0 = (1, 0, 0, 0)$  and is parallel to the vectors  $\mathbf{v}_1 = (1, -1, 1, -1)$  and  $\mathbf{v}_2 = (0, 2, 2, 0)$ .

The plane  $\Pi$  is not a subspace of  $\mathbb{R}^4$  as it does not pass through the origin. Let  $\Pi_0 = \text{Span}(\mathbf{v}_1, \mathbf{v}_2)$ . Then  $\Pi = \Pi_0 + \mathbf{x}_0$ .

Hence the distance from the point  $\mathbf{z}$  to the plane  $\Pi$  is the same as the distance from the point  $\mathbf{z} - \mathbf{x}_0$  to the plane  $\Pi - \mathbf{x}_0 = \Pi_0$ .

We shall apply the Gram-Schmidt process to vectors  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{z} - \mathbf{x}_0$ . This will yield an orthogonal system  $\mathbf{w}_1, \mathbf{w}_2, \mathbf{w}_3$ . The desired distance will be  $\|\mathbf{w}_3\|$ .

$${f v}_1=(1,-1,1,-1)$$
,  ${f v}_2=(0,2,2,0)$ ,  ${f z}-{f x}_0=(-1,0,1,0)$ .

$$\begin{split} \mathbf{w}_1 &= \mathbf{v}_1 = (1, -1, 1, -1), \\ \mathbf{w}_2 &= \mathbf{v}_2 - \frac{\langle \mathbf{v}_2, \mathbf{w}_1 \rangle}{\langle \mathbf{w}_1, \mathbf{w}_1 \rangle} \mathbf{w}_1 = \mathbf{v}_2 = (0, 2, 2, 0) \text{ as } \mathbf{v}_2 \perp \mathbf{v}_1. \\ \mathbf{w}_3 &= (\mathbf{z} - \mathbf{x}_0) - \frac{\langle \mathbf{z} - \mathbf{x}_0, \mathbf{w}_1 \rangle}{\langle \mathbf{w}_1, \mathbf{w}_1 \rangle} \mathbf{w}_1 - \frac{\langle \mathbf{z} - \mathbf{x}_0, \mathbf{w}_2 \rangle}{\langle \mathbf{w}_2, \mathbf{w}_2 \rangle} \mathbf{w}_2 \\ &= (-1, 0, 1, 0) - \frac{0}{4} (1, -1, 1, -1) - \frac{2}{8} (0, 2, 2, 0) \\ &= (-1, -1/2, 1/2, 0). \\ |\mathbf{w}_3| &= \left| \left( -1, -\frac{1}{2}, \frac{1}{2}, 0 \right) \right| = \frac{1}{2} \left| (-2, -1, 1, 0) \right| = \frac{\sqrt{6}}{2} = \sqrt{\frac{3}{2}}. \end{split}$$

**Problem.** Approximate the function  $f(x) = e^x$  on the interval [-1, 1] by a quadratic polynomial.

The best approximation would be a polynomial p(x) that minimizes the distance relative to the uniform norm:

$$\|f-p\|_\infty=\max_{|x|\leq 1}|f(x)-p(x)|.$$

However there is no analytic way to find such a polynomial. Instead, one can find a *"least squares"* approximation that minimizes the integral norm

$$||f - p||_2 = \left(\int_{-1}^1 |f(x) - p(x)|^2 dx\right)^{1/2}$$

The norm  $\|\cdot\|_2$  is induced by the inner product  $\langle g, h \rangle = \int_{-1}^1 g(x)h(x) \, dx.$ 

Therefore  $||f - p||_2$  is minimal if p is the orthogonal projection of the function f on the subspace  $\mathcal{P}_3$  of polynomials of degree at most 2.

We should apply the Gram-Schmidt process to the polynomials  $1, x, x^2$ , which form a basis for  $\mathcal{P}_3$ . This would yield an orthogonal basis  $p_0, p_1, p_2$ . Then

$$p(x) = rac{\langle f, p_0 
angle}{\langle p_0, p_0 
angle} p_0(x) + rac{\langle f, p_1 
angle}{\langle p_1, p_1 
angle} p_1(x) + rac{\langle f, p_2 
angle}{\langle p_2, p_2 
angle} p_2(x).$$