MATH 304 Linear Algebra

Lecture 21: The Gram-Schmidt orthogonalization process. Eigenvalues and eigenvectors of a matrix.

### **Orthogonal sets**

Let V be a vector space with an inner product.

*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.

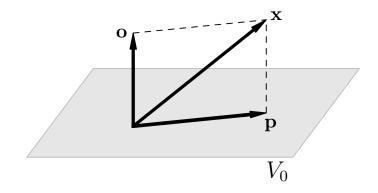
# **Orthogonal projection**

**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 the **orthogonal projection** of the vector **x** onto the subspace  $V_0$ . The distance from **x** to the subspace  $V_0$  is  $||\mathbf{o}||$ .

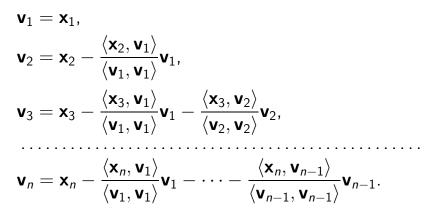
If 
$$\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$$
 is an orthogonal basis for  $V_0$  then  

$$\mathbf{p} = \frac{\langle \mathbf{x}, \mathbf{v}_1 \rangle}{\langle \mathbf{v}_1, \mathbf{v}_1 \rangle} \mathbf{v}_1 + \frac{\langle \mathbf{x}, \mathbf{v}_2 \rangle}{\langle \mathbf{v}_2, \mathbf{v}_2 \rangle} \mathbf{v}_2 + \dots + \frac{\langle \mathbf{x}, \mathbf{v}_n \rangle}{\langle \mathbf{v}_n, \mathbf{v}_n \rangle} \mathbf{v}_n.$$

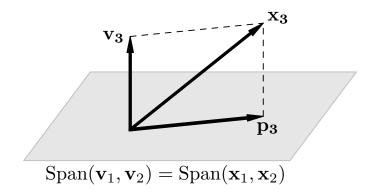


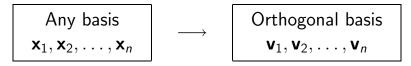
### The Gram-Schmidt orthogonalization process

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  $\mathbf{v}_1, \ldots, \mathbf{v}_k$  is the same as the span of  $\mathbf{x}_1, \ldots, \mathbf{x}_k$ ;

•  $\mathbf{v}_k$  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

$$\mathbf{v}_{1} = \mathbf{x}_{1}, \quad \mathbf{w}_{1} = \frac{\mathbf{v}_{1}}{\|\mathbf{v}_{1}\|},$$

$$\mathbf{v}_{2} = \mathbf{x}_{2} - \langle \mathbf{x}_{2}, \mathbf{w}_{1} \rangle \mathbf{w}_{1}, \quad \mathbf{w}_{2} = \frac{\mathbf{v}_{2}}{\|\mathbf{v}_{2}\|},$$

$$\mathbf{v}_{3} = \mathbf{x}_{3} - \langle \mathbf{x}_{3}, \mathbf{w}_{1} \rangle \mathbf{w}_{1} - \langle \mathbf{x}_{3}, \mathbf{w}_{2} \rangle \mathbf{w}_{2}, \quad \mathbf{w}_{3} = \frac{\mathbf{v}_{3}}{\|\mathbf{v}_{3}\|},$$

$$\dots$$

$$\mathbf{v}_{n} = \mathbf{x}_{n} - \langle \mathbf{x}_{n}, \mathbf{w}_{1} \rangle \mathbf{w}_{1} - \dots - \langle \mathbf{x}_{n}, \mathbf{w}_{n-1} \rangle \mathbf{w}_{n-1},$$

$$\mathbf{w}_{n} = \frac{\mathbf{v}_{n}}{\|\mathbf{v}_{n}\|}.$$
Then  $\mathbf{w}_{1}, \mathbf{w}_{2}, \dots, \mathbf{w}_{n}$  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). $\mathbf{v}_3 = \mathbf{x}_3 - \frac{\langle \mathbf{x}_3, \mathbf{v}_1 \rangle}{\langle \mathbf{v}_1, \mathbf{v}_1 \rangle} \mathbf{v}_1 - \frac{\langle \mathbf{x}_3, \mathbf{v}_2 \rangle}{\langle \mathbf{v}_2, \mathbf{v}_2 \rangle} \mathbf{v}_2$  $=(0,0,1)-\frac{2}{0}(1,2,2)-\frac{4/3}{4}(-4/3,-2/3,4/3)$ = (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  $\mathbf{x}_3 = (-3, 7, 1, 3)$ .

Let us apply the Gram-Schmidt process to vectors  $\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3, \mathbf{y}$ . We should obtain an orthogonal system  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3, \mathbf{v}_4$ . 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!

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. 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 z to the plane  $\Pi$  is the same as the distance from the point  $z - x_0$  to the plane  $\Pi_0$ .

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  $|\mathbf{w}_3|$ .

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

$$\mathbf{w}_1 = \mathbf{v}_1 = (1, -1, 1, -1),$$
  
 $\mathbf{w}_2 = \mathbf{v}_2 - \frac{\langle \mathbf{v}_2, \mathbf{w}_1 \rangle}{\langle \mathbf{w}_1, \mathbf{w}_1 \rangle} \mathbf{w}_1 = \mathbf{v}_2 = (0, 2, 2, 0) \text{ as } \mathbf{v}_2 \perp \mathbf{v}_1.$ 

$$\begin{split} \mathbf{w}_{3} &= (\mathbf{z} - \mathbf{x}_{0}) - \frac{\langle \mathbf{z} - \mathbf{x}_{0}, \mathbf{w}_{1} \rangle}{\langle \mathbf{w}_{1}, \mathbf{w}_{1} \rangle} \mathbf{w}_{1} - \frac{\langle \mathbf{z} - \mathbf{x}_{0}, \mathbf{w}_{2} \rangle}{\langle \mathbf{w}_{2}, \mathbf{w}_{2} \rangle} \mathbf{w}_{2} \\ &= (-1, 0, 1, 0) - \frac{0}{4} (1, -1, 1, -1) - \frac{2}{8} (0, 2, 2, 0) \\ &= (-1, -1/2, 1/2, 0). \\ |\mathbf{w}_{3}| &= \left| \left( -1, -\frac{1}{2}, \frac{1}{2}, 0 \right) \right| = \frac{1}{2} \left| (-2, -1, 1, 0) \right| = \frac{\sqrt{6}}{2} = \sqrt{\frac{3}{2}}. \end{split}$$

# Eigenvalues and eigenvectors of a matrix

Definition. Let A be an  $n \times n$  matrix. A number  $\lambda \in \mathbb{R}$  is called an **eigenvalue** of the matrix A if  $A\mathbf{v} = \lambda \mathbf{v}$  for a nonzero column vector  $\mathbf{v} \in \mathbb{R}^n$ . The vector  $\mathbf{v}$  is called an **eigenvector** of A belonging to (or associated with) the eigenvalue  $\lambda$ .

*Remarks.* • Alternative notation: eigenvalue = characteristic value, eigenvector = characteristic vector.

• The zero vector is never considered an eigenvector.

Example. 
$$A = \begin{pmatrix} 2 & 0 \\ 0 & 3 \end{pmatrix}.$$
$$\begin{pmatrix} 2 & 0 \\ 0 & 3 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 2 \\ 0 \end{pmatrix} = 2 \begin{pmatrix} 1 \\ 0 \end{pmatrix},$$
$$\begin{pmatrix} 2 & 0 \\ 0 & 3 \end{pmatrix} \begin{pmatrix} 0 \\ -2 \end{pmatrix} = \begin{pmatrix} 0 \\ -6 \end{pmatrix} = 3 \begin{pmatrix} 0 \\ -2 \end{pmatrix}.$$

Hence (1,0) is an eigenvector of A belonging to the eigenvalue 2, while (0,-2) is an eigenvector of A belonging to the eigenvalue 3.

Example. 
$$A = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}.$$
$$\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \quad \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ -1 \end{pmatrix} = \begin{pmatrix} -1 \\ 1 \end{pmatrix}.$$

Hence (1, 1) is an eigenvector of A belonging to the eigenvalue 1, while (1, -1) is an eigenvector of A belonging to the eigenvalue -1.

Vectors  $\mathbf{v}_1 = (1, 1)$  and  $\mathbf{v}_2 = (1, -1)$  form a basis for  $\mathbb{R}^2$ . Consider a linear operator  $L : \mathbb{R}^2 \to \mathbb{R}^2$ given by  $L(\mathbf{x}) = A\mathbf{x}$ . The matrix of L with respect to the basis  $\mathbf{v}_1, \mathbf{v}_2$  is  $B = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$ . Let A be an  $n \times n$  matrix. Consider a linear operator  $L : \mathbb{R}^n \to \mathbb{R}^n$  given by  $L(\mathbf{x}) = A\mathbf{x}$ . Let  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  be a nonstandard basis for  $\mathbb{R}^n$ and B be the matrix of the operator L with respect to this basis.

**Theorem** The matrix *B* is diagonal if and only if vectors  $\mathbf{v}_1, \mathbf{v}_2, \ldots, \mathbf{v}_n$  are eigenvectors of *A*. If this is the case, then the diagonal entries of the matrix *B* are the corresponding eigenvalues of *A*.

$$A\mathbf{v}_{i} = \lambda_{i}\mathbf{v}_{i} \iff B = \begin{pmatrix} \lambda_{1} & & O \\ & \lambda_{2} & \\ & & \ddots & \\ O & & & \lambda_{n} \end{pmatrix}$$