

## Section 6.7 Inner Product Spaces

**Definition:** An **inner product** on a vector space  $V$  is a function which assigns to each pair of vectors  $\mathbf{u}$  and  $\mathbf{v}$  in  $V$  a real number denoted by  $\langle \mathbf{u}, \mathbf{v} \rangle$  and that satisfies the following four axioms: For every  $\mathbf{u}, \mathbf{v}, \mathbf{w}$  in  $V$  and scalar  $c$

i  $\langle \mathbf{u}, \mathbf{v} \rangle = \langle \mathbf{v}, \mathbf{u} \rangle,$

ii  $\langle \mathbf{u} + \mathbf{v}, \mathbf{w} \rangle = \langle \mathbf{u}, \mathbf{w} \rangle + \langle \mathbf{v}, \mathbf{w} \rangle,$

iii  $\langle c\mathbf{u}, \mathbf{v} \rangle = c \langle \mathbf{u}, \mathbf{v} \rangle,$

iv  $\langle \mathbf{u}, \mathbf{u} \rangle \geq 0$  and  $\langle \mathbf{u}, \mathbf{u} \rangle = 0$  if and only if  $\mathbf{u} = \mathbf{0}$ .

A vector space with an inner product is called an **inner product space**.

# Norm, Distance, and Orthogonality

**Norm:** The norm of a vector  $\mathbf{v}$  is  $\|\mathbf{v}\| = \sqrt{\langle \mathbf{v}, \mathbf{v} \rangle}$ .

**A Unit Vector:** is a vector whose norm is 1.

**Distance:** The distance between two vectors  $\mathbf{u}$  and  $\mathbf{v}$  is  $\|\mathbf{u} - \mathbf{v}\|$ .

**Orthogonality:** Two vectors  $\mathbf{u}$  and  $\mathbf{v}$  are orthogonal if  $\langle \mathbf{u}, \mathbf{v} \rangle = 0$

**Orthogonal Projection:** The orthogonal projection of  $\mathbf{v}$  onto  $\mathbf{u}$  is the vector

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

**Pythagorean Theorem:** If  $\mathbf{u}$  and  $\mathbf{v}$  are orthogonal, then

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

## Example (using some calculus)

Consider the set  $C[a, b]$  of continuous functions defined on the interval  $[a, b]$ . For  $f$  and  $g$  in  $C[a, b]$

$$\langle f, g \rangle = \int_a^b f(t)g(t) dt$$

defines an inner product on  $C[a, b]$ .

(a) Show that  $f(x) = \sin x$  and  $g(x) = \cos x$  are orthogonal on  $[-\pi, \pi]$ .

$$\begin{aligned}\langle f, g \rangle &= \int_{-\pi}^{\pi} f(x) g(x) dx \\ &= \int_{-\pi}^{\pi} \sin x \cos x dx \\ &= \int_{-\pi}^{\pi} \frac{1}{2} \sin(2x) dx\end{aligned}$$

ID

$$\sin 2\theta = 2 \sin \theta \cos \theta$$

$$\begin{aligned}
 &= -\frac{1}{4} \cos(2x) \Big|_{-\pi}^{\pi} = -\frac{1}{4} (\cos(2\pi) - \cos(-2\pi)) \\
 &= -\frac{1}{4} (1 - 1) = 0
 \end{aligned}$$

So  $f(x)$  and  $g(x)$  are orthogonal.

## Example continued...

(b) Find the distance between  $f$  and  $g$  in  $C[-\pi, \pi]$ .

The square of the distance  $\|f - g\|^2$

$$\|f - g\|^2 = \langle f - g, f - g \rangle = \int_{-\pi}^{\pi} (f(x) - g(x))(f(x) - g(x)) dx$$

$$= \int_{-\pi}^{\pi} (\sin^2 x - 2 \sin x \cos x + \cos^2 x) dx$$

$$= \int_{-\pi}^{\pi} (\sin^2 x + \cos^2 x - 2 \sin x \cos x) dx$$

$$= \int_{-\pi}^{\pi} (1 - 2 \sin x \cos x) dx$$

$$= \int_{-\pi}^{\pi} dx - 2 \int_{-\pi}^{\pi} \sin x \cos x dx$$

0" by part (a)

$$= \int_{-\pi}^{\pi} dx = x \Big|_{-\pi}^{\pi} = \pi - (-\pi) = 2\pi$$

So the distance between  $\sin x$  and  $\cos x$   
is  $\sqrt{2\pi}$ .

# Some Key Inequalities

**Cauchy Schwarz:** In an inner product space  $V$ , for any  $\mathbf{u}, \mathbf{v}$  we have

$$| \langle \mathbf{u}, \mathbf{v} \rangle | \leq \|\mathbf{u}\| \|\mathbf{v}\|.$$

**Triangle:** For any  $\mathbf{u}, \mathbf{v}$

$$\|\mathbf{u} + \mathbf{v}\| \leq \|\mathbf{u}\| + \|\mathbf{v}\|.$$

## Gram Schmidt

Let  $\mathbb{P}_2[-1, 1]$  be the set of polynomials of degree at most 2 defined for  $-1 \leq t \leq 1$ . Using the following inner product

$$\langle \mathbf{p}, \mathbf{q} \rangle = \int_{-1}^1 \mathbf{p}(t) \mathbf{q}(t) dt$$

find an orthogonal basis for  $\mathbb{P}_2$  by starting with the elementary basis  $\{1, t, t^2\}$ .

We'll call the new basis  $\{\vec{v}_1, \vec{v}_2, \vec{v}_3\}$

$$\vec{v}_1 = 1$$

$$\vec{v}_2 = t - \frac{\langle t, 1 \rangle}{\langle 1, 1 \rangle} 1$$

$$\langle t, 1 \rangle = \int_{-1}^1 t \cdot 1 dt = \left. \frac{t^2}{2} \right|_{-1}^1 = \frac{1}{2} - \frac{(-1)^2}{2} = \frac{1}{2} - \frac{1}{2} = 0$$



So  $\vec{v}_1 = 1$  and  $\vec{v}_2 = t$

$$\vec{v}_3 = t^2 - \frac{\langle t^2, 1 \rangle}{\langle 1, 1 \rangle} 1 - \frac{\langle t^2, t \rangle}{\langle t, t \rangle} t$$

$$\langle t^2, 1 \rangle = \int_{-1}^1 t^2 \cdot 1 dt = \left. \frac{t^3}{3} \right|_{-1}^1 = \frac{1^3}{3} - \frac{(-1)^3}{3} = \frac{1}{3} + \frac{1}{3} = \frac{2}{3}$$

$$\langle 1, 1 \rangle = \int_{-1}^1 1 \cdot 1 dt = \int_{-1}^1 1 dt = \left. t \right|_{-1}^1 = 1 - (-1) = 2$$

$$\langle t^2, t \rangle = \int_{-1}^1 t^2 \cdot t dt = \int_{-1}^1 t^3 dt = \left. \frac{t^4}{4} \right|_{-1}^1 = \frac{1^4}{4} - \frac{(-1)^4}{4} = \frac{1}{4} - \frac{1}{4} = 0$$

$$\text{so } \vec{V}_3 = t^2 - \frac{2/3}{2} \cdot 1 = t^2 - \frac{1}{3}$$

An orthogonal basis is

$$\left\{ 1, t, t^2 - \frac{1}{3} \right\}$$

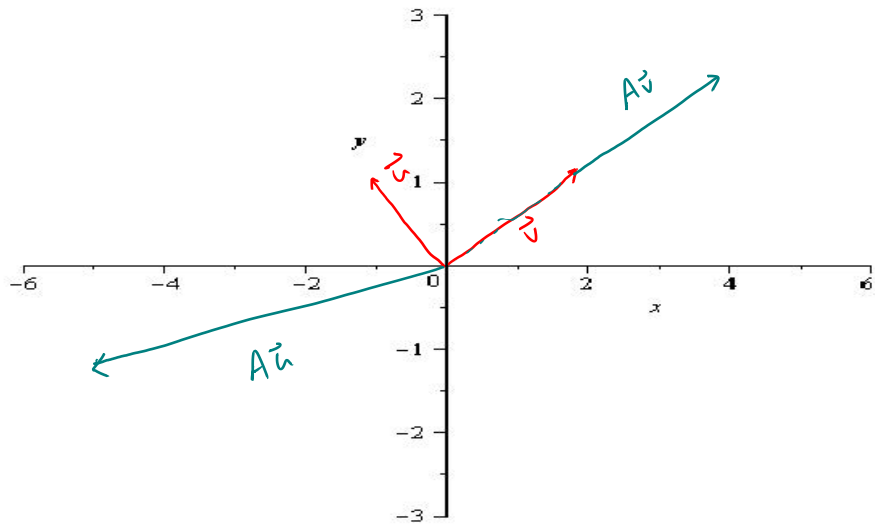
## Section 5.1: Eigenvectors and Eigenvalues

Consider the matrix  $A = \begin{bmatrix} 3 & -2 \\ 1 & 0 \end{bmatrix}$  and the vectors  $\mathbf{u} = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$  and  $\mathbf{v} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$ . Plot  $\mathbf{u}$ ,  $A\mathbf{u}$ ,  $\mathbf{v}$ , and  $A\mathbf{v}$  on the axis on the next slide.

$$A\vec{u} = \begin{bmatrix} 3 & -2 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} -1 \\ 1 \end{bmatrix} = \begin{bmatrix} -3-2 \\ -1+0 \end{bmatrix} = \begin{bmatrix} -5 \\ -1 \end{bmatrix}$$

$$A\vec{v} = \begin{bmatrix} 3 & -2 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 6-2 \\ 2+0 \end{bmatrix} = \begin{bmatrix} 4 \\ 2 \end{bmatrix}$$

# Example Plot



# Eigenvalues and Eigenvectors

Note that in this example, the matrix  $A$  seems to both stretch and rotate the vector  $\mathbf{u}$ . But the *action of  $A$*  on the vector  $\mathbf{v}$  is just a stretch/compress.

We wish to consider matrices with vectors that satisfy relationships such as

$$A\mathbf{x} = 2\mathbf{x}, \quad \text{or} \quad A\mathbf{x} = -4\mathbf{x}, \quad \text{or more generally} \quad A\mathbf{x} = \lambda\mathbf{x}$$

for constant  $\lambda$ —and for nonzero vector  $\mathbf{x}$ .

# Definition of Eigenvector and Eigenvalue

**Definition:** Let  $A$  be an  $n \times n$  matrix. A nonzero vector  $\mathbf{x}$  such that

$$A\mathbf{x} = \lambda\mathbf{x}$$

for some scalar  $\lambda$  is called an **eigenvector** of the matrix  $A$ .

A scalar  $\lambda$  such that there exists a nonzero vector  $\mathbf{x}$  satisfying  $A\mathbf{x} = \lambda\mathbf{x}$  is called an **eigenvalue** of the matrix  $A$ . Such a nonzero vector  $\mathbf{x}$  is an *eigenvector corresponding to  $\lambda$* .

Note that built right into this definition is that the eigenvector  $\mathbf{x}$  **must be nonzero!**

## Example

The number  $\lambda = -4$  is an eigenvalue of the matrix matrix

$A = \begin{bmatrix} 1 & 6 \\ 5 & 2 \end{bmatrix}$ . Find the corresponding eigenvectors.

We want  $\vec{x} \neq \vec{0}$  such that  $A\vec{x} = -4\vec{x}$

For  $\vec{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$  our equation is  $\begin{bmatrix} 1 & 6 \\ 5 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = -4 \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$

as a system

$$x_1 + 6x_2 = -4x_1$$

$$5x_1 + 2x_2 = -4x_2$$

we can subtract the stuff on the right

$$\left. \begin{aligned} (1 - (-4))x_1 + 6x_2 &= 0 \\ 5x_1 + (2 - (-4))x_2 &= 0 \end{aligned} \right\} \Rightarrow \begin{bmatrix} 5 & 6 \\ 5 & 6 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} 5 & 6 & 0 \\ 5 & 6 & 0 \end{bmatrix} \xrightarrow{\text{rref}} \begin{bmatrix} 1 & 6/5 & 0 \\ 0 & 0 & 0 \end{bmatrix} \quad \begin{array}{l} x_1 = -\frac{6}{5}x_2 \\ x_2 \text{ - free} \end{array}$$

so all eigen vectors look like

$$\vec{x} = x_2 \begin{bmatrix} -\frac{6}{5} \\ 1 \end{bmatrix} \quad \text{for } x_2 \neq 0.$$



# Eigenspace

**Definition:** Let  $A$  be an  $n \times n$  matrix and  $\lambda$  an eigenvalue of  $A$ . The set of all eigenvectors corresponding to  $\lambda$ —i.e. the set

$$\{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{x} \neq \mathbf{0} \text{ and } A\mathbf{x} = \lambda\mathbf{x}\},$$

is called the **eigenspace of  $A$  corresponding to  $\lambda$** .

**Remark:** The eigenspace is the same as the null space of the matrix  $A - \lambda I$ . It follows that the eigenspace is a subspace of  $\mathbb{R}^n$ .

In the  $2 \times 2$  case

$$A - \lambda I = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} - \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix}$$

## Example

The matrix  $A = \begin{bmatrix} 4 & -1 & 6 \\ 2 & 1 & 6 \\ 2 & -1 & 8 \end{bmatrix}$  has eigenvalue  $\lambda = 2$ . Find a basis for the eigenspace of  $A$  corresponding to  $\lambda$ .

The homogeneous system has coef. matrix  $A - 2I$

$$A - 2I = \begin{bmatrix} 4 & -1 & 6 \\ 2 & 1 & 6 \\ 2 & -1 & 8 \end{bmatrix} - \begin{bmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix} = \begin{bmatrix} 2 & -1 & 6 \\ 2 & -1 & 6 \\ 2 & -1 & 6 \end{bmatrix}$$

$$\begin{bmatrix} 2 & -1 & 6 \\ 2 & -1 & 6 \\ 2 & -1 & 6 \end{bmatrix} \xrightarrow{\text{ref}} \begin{bmatrix} 1 & -\frac{1}{2} & 3 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \quad \begin{array}{l} x_1 = \frac{1}{2}x_2 - 3x_3 \\ x_2, x_3 \text{ - free} \end{array}$$

Solution 5

$$\vec{x} = x_2 \begin{bmatrix} 1/2 \\ 1 \\ 0 \end{bmatrix} + x_3 \begin{bmatrix} -3 \\ 0 \\ 1 \end{bmatrix}$$

A basis for the eigen space is

$$\left\{ \begin{bmatrix} 1/2 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -3 \\ 0 \\ 1 \end{bmatrix} \right\}$$

# Matrices with Nice Structure

**Theorem:** If  $A$  is an  $n \times n$  triangular matrix, then the eigenvalues of  $A$  are its diagonal elements.

Find the eigenvalues of the matrix  $A = \begin{bmatrix} 3 & 0 & 0 \\ -2 & \pi & 0 \\ -1 & 0 & 1 \end{bmatrix}$

*A is lower triangular. The eigen values are*

$$\lambda_1 = 3, \lambda_2 = \pi, \lambda_3 = 1$$

## Example

Suppose  $\lambda = 0$  is an eigenvalue<sup>2</sup> of a matrix  $A$ . Argue that  $A$  is not invertible.

There must be an eigenvector  $\vec{x} \neq \vec{0}$  such that

$$A\vec{x} = 0\vec{x} = \vec{0}. \text{ Since } \vec{x} \neq \vec{0} \text{ we have a nontrivial}$$

solution to the homogeneous equation  $A\vec{x} = \vec{0}$ .

By the "invertible matrix theorem"  $A$  is  
not invertible.

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<sup>2</sup>Eigenvectors must be nonzero vectors, but it is perfectly legitimate to have a zero eigenvalue!

# Theorems

**Theorem:** A square matrix  $A$  is invertible if and only if zero is **not** an eigenvalue.

**Theorem:** If  $\mathbf{v}_1, \dots, \mathbf{v}_p$  are eigenvectors of a matrix  $A$  corresponding to distinct eigenvalues,  $\lambda_1, \dots, \lambda_p$ , then the set  $\{\mathbf{v}_1, \dots, \mathbf{v}_p\}$  is linearly independent.