

# Lecture Four

## Section 4.1 – Matrix Transformations from $R^n$ to $R^m$

### Definition

If  $V$  and  $W$  are vector spaces, and if  $f$  is a function with domain  $V$  and codomain  $W$ , then we say that  $f$  is a transformation from  $V$  to  $W$  or that  $f$  maps  $V$  to  $W$ , which we denote by writing

$$f : V \rightarrow W$$

In the special case where  $V = W$ , the transformation is also called an operator on  $V$ .

### Matrix Transformation

$$\begin{aligned} w_1 &= a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n \\ w_2 &= a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n \\ &\vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \\ w_m &= a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n \end{aligned}$$

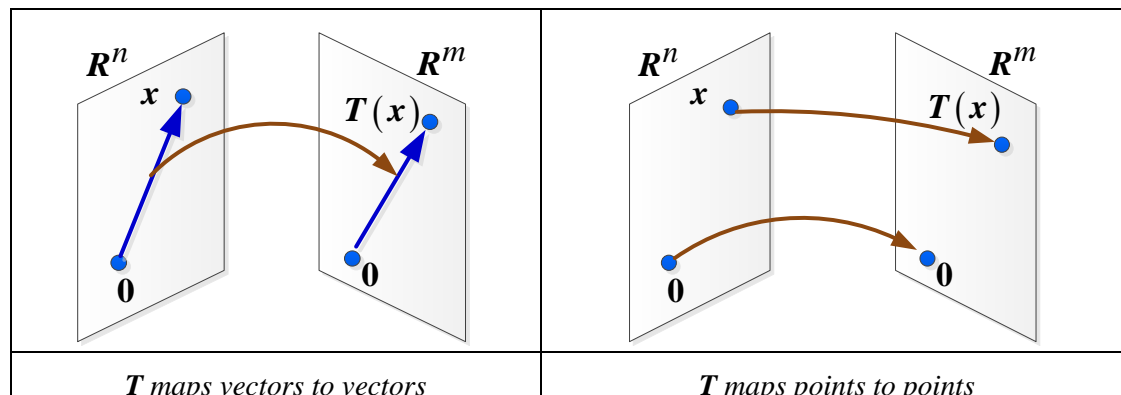
Which we can write in matrix formation

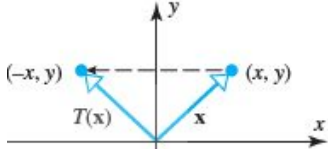
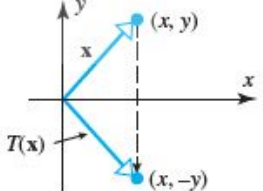
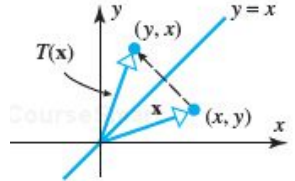
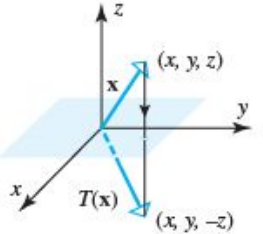
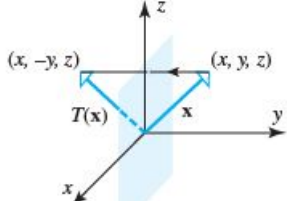
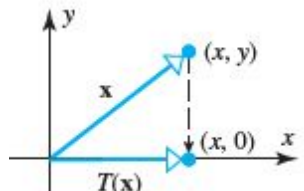
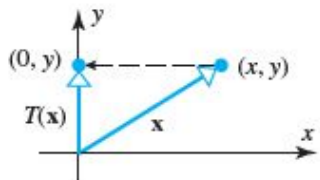
$$\begin{bmatrix} w_1 \\ w_2 \\ w_3 \\ w_4 \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & a_{1n} \\ a_{21} & a_{22} & a_{2n} \\ & & \\ a_{m1} & a_{m2} & a_{mn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}$$

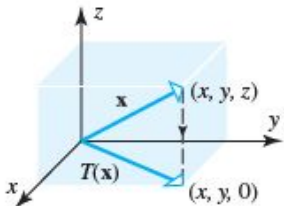
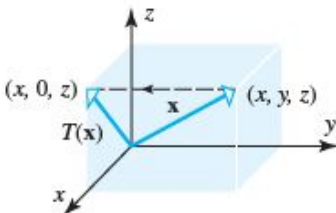
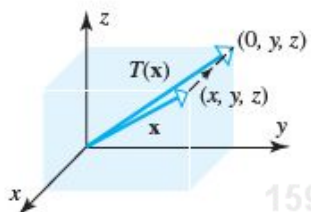
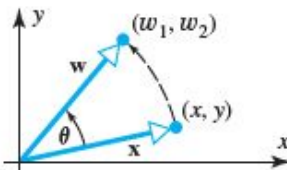
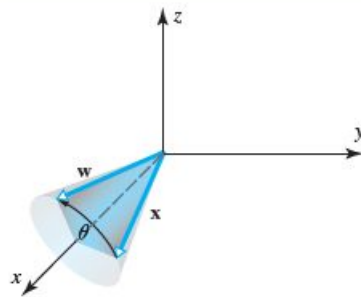
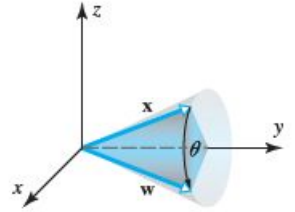
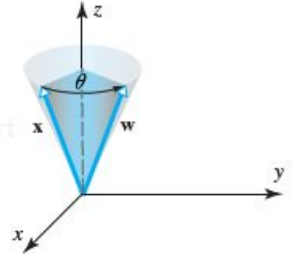
$$w = Ax$$

Although we could view this as a linear system, we will view it instead as a transformation that maps the column vector  $x$  in  $R^n$  into the column vector  $w$  in  $R^m$  by multiplying  $x$  on the left by  $A$ . We call this a **matrix transformation** (or **matrix operator** if  $m = n$ ) and we denote it by

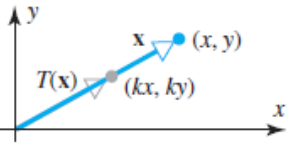
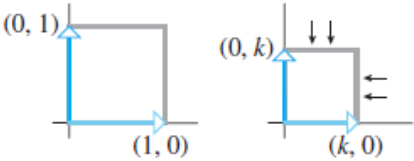
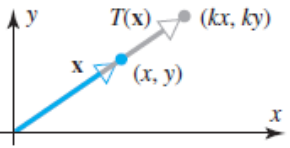
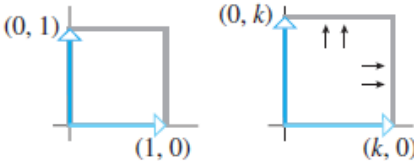
$$T_A : R^n \rightarrow R^m$$

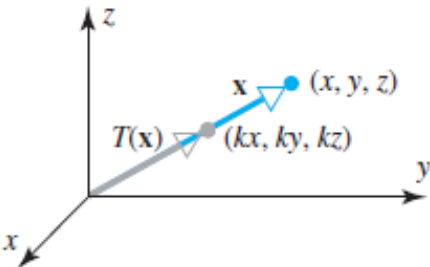
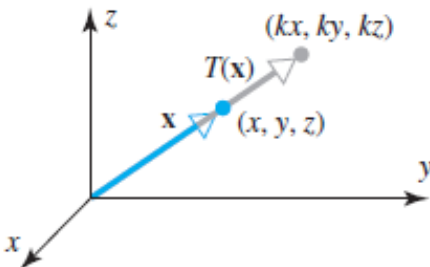


<p>Reflection about the <math>y</math>-axis</p> $T(x, y) = (-x, y)$		$T(e_1) = T(1, 0) = (-1, 0)$ $T(e_2) = T(0, 1) = (0, 1)$	$\begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix}$
<p>Reflection about the <math>x</math>-axis</p> $T(x, y) = (x, -y)$		$T(e_1) = T(1, 0) = (1, 0)$ $T(e_2) = T(0, 1) = (0, -1)$	$\begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$
<p>Reflection about the line <math>y = x</math></p> $T(x, y) = (y, x)$		$T(e_1) = T(1, 0) = (0, 1)$ $T(e_2) = T(0, 1) = (1, 0)$	$\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$
<p>Reflection about the <math>xy</math>-plane</p> $T(x, y, z) = (x, y, -z)$		$T(e_1) = T(1, 0, 0) = (1, 0, 0)$ $T(e_2) = T(0, 1, 0) = (0, 1, 0)$ $T(e_3) = T(0, 0, 1) = (0, 0, -1)$	$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{bmatrix}$
<p>Reflection about the <math>yz</math>-plane</p> $T(x, y, z) = (-x, y, z)$		$T(e_1) = T(1, 0, 0) = (-1, 0, 0)$ $T(e_2) = T(0, 1, 0) = (0, 1, 0)$ $T(e_3) = T(0, 0, 1) = (0, 0, 1)$	$\begin{bmatrix} -1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$
<p>Orthogonal projection on the <math>x</math>-axis</p> $T(x, y) = (x, 0)$		$T(e_1) = T(1, 0) = (1, 0)$ $T(e_2) = T(0, 1) = (0, 0)$	$\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$
<p>Orthogonal projection on the <math>y</math>-axis</p> $T(x, y) = (0, y)$		$T(e_1) = T(1, 0) = (0, 0)$ $T(e_2) = T(0, 1) = (0, 1)$	$\begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$

<p>Orthogonal projection on the <math>xy</math>-Plane <math>T(x, y, z) = (x, y, 0)</math></p>		$T(1,0,0) = (1,0,0)$ $T(0,1,0) = (0,1,0)$ $T(0,0,1) = (0,0,0)$	$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}$
<p>Orthogonal projection on the <math>xz</math>-Plane <math>T(x, y, z) = (x, 0, z)</math></p>		$T(1,0,0) = (1,0,0)$ $T(0,1,0) = (0,0,0)$ $T(0,0,1) = (0,0,1)$	$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}$
<p>Orthogonal projection on the <math>yz</math>-Plane <math>T(x, y, z) = (0, y, z)</math></p>		$T(1,0,0) = (0,0,0)$ $T(0,1,0) = (0,1,0)$ $T(0,0,1) = (0,0,1)$	$\begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$
<b>Rotation Operators</b>			
<p>Rotation through an angle <math>\theta</math></p>		$w_1 = x \cos \theta - y \sin \theta$ $w_2 = x \sin \theta + y \cos \theta$	$\begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}$
<p>Counterclockwise rotation about the positive <math>x</math>-axis through an angle <math>\theta</math></p>		$w_1 = x$ $w_2 = y \cos \theta - z \sin \theta$ $w_3 = x \sin \theta + z \cos \theta$	$\begin{pmatrix} 1 & 0 & 0 \\ 0 & \cos \theta & -\sin \theta \\ 0 & \sin \theta & \cos \theta \end{pmatrix}$
<p>Counterclockwise rotation about the positive <math>y</math>-axis through an angle <math>\theta</math></p>		$w_1 = x \cos \theta + z \sin \theta$ $w_2 = y$ $w_3 = -x \sin \theta + z \cos \theta$	$\begin{pmatrix} \cos \theta & 0 & \sin \theta \\ 0 & 1 & 0 \\ -\sin \theta & 0 & \cos \theta \end{pmatrix}$
<p>Counterclockwise rotation about the positive <math>z</math>-axis through an angle <math>\theta</math></p>		$w_1 = x \cos \theta - y \sin \theta$ $w_2 = x \sin \theta + y \cos \theta$ $w_3 = z$	$\begin{pmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{pmatrix}$

# Dilations and Contractions

<p>Contraction with factor <math>k</math> on <math>R^2</math></p> <p><math>(0 \leq k &lt; 1)</math></p>			$\begin{bmatrix} k & 0 \\ 0 & k \end{bmatrix}$
<p>Dilation with factor <math>k</math> on <math>R^2</math></p> <p><math>(k &gt; 1)</math></p>			

<p>Contraction with factor <math>k</math> on <math>R^3</math></p> <p><math>(0 \leq k \leq 1)</math></p>		$\begin{bmatrix} k & 0 & 0 \\ 0 & k & 0 \\ 0 & 0 & k \end{bmatrix}$
<p>Dilation with factor <math>k</math> on <math>R^3</math></p> <p><math>(k \geq 1)</math></p>		

## Expansion or Compression

<p>Compression of <math>R^2</math> in the <math>x</math>-direction with factor <math>k</math></p> <p><math>(0 \leq k &lt; 1)</math></p>			$\begin{bmatrix} k & 0 \\ 0 & 1 \end{bmatrix}$
<p>Expansion of <math>R^2</math> in the <math>x</math>-direction with factor <math>k</math></p> <p><math>(k &gt; 1)</math></p>			
<p>Compression of <math>R^2</math> in the <math>y</math>-direction with factor <math>k</math></p> <p><math>(0 \leq k &lt; 1)</math></p>			$\begin{bmatrix} 1 & 0 \\ 0 & k \end{bmatrix}$
<p>Expansion of <math>R^2</math> in the <math>y</math>-direction with factor <math>k</math></p> <p><math>(k &gt; 1)</math></p>			

## Shear

<p>Shear of <math>R^2</math> in the <math>x</math>-direction with factor <math>k</math></p> <p><math>T(x, y) = (x + ky, y)</math></p>		$\begin{bmatrix} 1 & k \\ 0 & 1 \end{bmatrix}$
<p>Shear of <math>R^2</math> in the <math>y</math>-direction with factor <math>k</math></p> <p><math>T(x, y) = (x, y + kx)</math></p>		$\begin{bmatrix} 1 & 0 \\ k & 1 \end{bmatrix}$

## Orthogonal Projections on Lines through the Origin

$$T(e_1) = \begin{bmatrix} \cos^2 \theta \\ \sin \theta \cos \theta \end{bmatrix} \quad T(e_2) = \begin{bmatrix} \sin \theta \cos \theta \\ \sin^2 \theta \end{bmatrix}$$

$$P_0 = \begin{pmatrix} \cos^2 \theta & \sin \theta \cos \theta \\ \sin \theta \cos \theta & \sin^2 \theta \end{pmatrix}$$

### ***Example***

Find the orthogonal projection of the vector  $\mathbf{x} = (1, 5)$  on the line through the origin that makes an angle of  $\frac{\pi}{6}$  ( $= 30^\circ$ ) with the  $x$ -axis

### **Solution**

$$P_0 = \begin{pmatrix} \cos^2 \theta & \sin \theta \cos \theta \\ \sin \theta \cos \theta & \sin^2 \theta \end{pmatrix}$$

$$= \begin{pmatrix} \cos^2\left(\frac{\pi}{6}\right) & \sin\left(\frac{\pi}{6}\right)\cos\left(\frac{\pi}{6}\right) \\ \sin\left(\frac{\pi}{6}\right)\cos\left(\frac{\pi}{6}\right) & \sin^2\left(\frac{\pi}{6}\right) \end{pmatrix}$$

$$= \begin{pmatrix} \left(\frac{\sqrt{3}}{2}\right)^2 & \left(\frac{1}{2}\right)\left(\frac{\sqrt{3}}{2}\right) \\ \left(\frac{1}{2}\right)\left(\frac{\sqrt{3}}{2}\right) & \left(\frac{1}{2}\right)^2 \end{pmatrix}$$

$$= \begin{pmatrix} \frac{3}{4} & \frac{\sqrt{3}}{4} \\ \frac{\sqrt{3}}{4} & \frac{1}{4} \end{pmatrix}$$

$$P_0 \mathbf{x} = \begin{pmatrix} \frac{3}{4} & \frac{\sqrt{3}}{4} \\ \frac{\sqrt{3}}{4} & \frac{1}{4} \end{pmatrix} \begin{pmatrix} 1 \\ 5 \end{pmatrix} = \begin{pmatrix} \frac{3+5\sqrt{3}}{4} \\ \frac{\sqrt{3}+5}{4} \end{pmatrix} \approx \begin{pmatrix} 2.91 \\ 1.68 \end{pmatrix}$$

## Four Fundamental Subspaces

1. The **row space** is  $C(A^T)$ , a subspace of  $\mathbf{R}^n$ .
2. The **column space** is  $C(A)$ , a subspace of  $\mathbf{R}^m$ .
3. The **nullspace** is  $N(A)$ , a subspace of  $\mathbf{R}^n$ .
4. The **left nullspace** is  $N(A^T)$ , a subspace of  $\mathbf{R}^m$ .

## The Four Subspaces for $\mathbf{R}$

Consider the matrix 3 by 5:

$$\begin{bmatrix} 1 & 3 & 5 & 0 & 9 \\ 0 & 0 & 0 & 1 & 8 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \quad \begin{array}{ll} m=3 & \text{pivot rows 1 and 2} \\ n=5 & \\ r=2 & \text{pivot columns 1 and 4} \end{array}$$

1. Rows 1 and 2 are a basis. The row space contains combination of all 3 rows.

The **row space** of  $\mathbf{R}$  has dimension 2 (= **rank**).

**The dimension of the row space is  $r$ .** The nonzero rows of  $\mathbf{R}$  form a basis.

2. The **column space** of  $\mathbf{R}$  has dimension  $r = 2$ .

The pivot columns 1 and 4 form a basis. They are independent because they start with the  $r$  by  $r$  identity matrix.

There are 3 special solutions:

$$C_2 = 3C_1 \quad \text{The special solution is } (-3, 1, 0, 0, 0)$$

$$C_3 = 5C_1 \quad \text{The special solution is } (-5, 0, 1, 0, 0)$$

$$C_5 = 9C_1 + 8C_2 \quad \text{The special solution is } (-9, 0, 0, -8, 1)$$

**The dimension of the column space is  $r$ .** The pivot columns form a basis.

3. The **nullspace** has dimension  $n - r = 5 - 2 = 3$  (free variables). Here  $x_2, x_3, x_5$  are free (no pivots in those columns). They yield the three special solutions to  $R\mathbf{x} = 0$ . Set a free variable to 1, and solve for  $x_1$  and  $x_4$ .

$$s_2 = \begin{bmatrix} -3 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} \quad s_3 = \begin{bmatrix} -5 \\ 0 \\ 1 \\ 0 \\ 0 \end{bmatrix} \quad s_5 = \begin{bmatrix} -9 \\ 0 \\ 0 \\ -8 \\ 1 \end{bmatrix}$$

$R\mathbf{x} = 0$  has the complete solution:  $\mathbf{x} = x_2 s_2 + x_3 s_3 + x_5 s_5$

**The nullspace has dimension  $n - r$ .** The special solutions form a basis.

4. The **nullspace** of  $R^T$  has dimension  $m - r = 3 - 2 = 1$

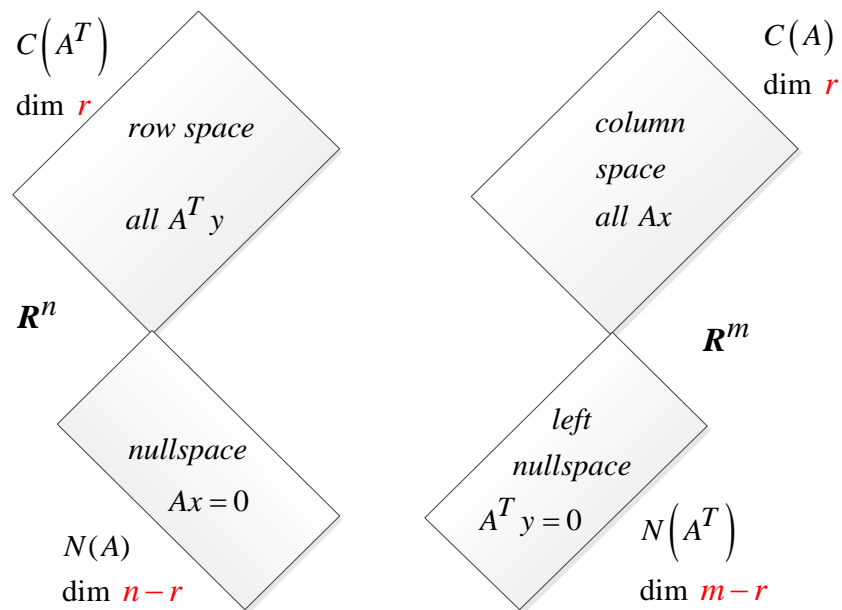
$$\text{The equation } R^T y = 0: \begin{pmatrix} 1 & 0 & 0 \\ 3 & 0 & 0 \\ 5 & 0 & 0 \\ 0 & 1 & 0 \\ 9 & 8 & 0 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \Rightarrow \begin{cases} y_1 = 0 \\ y_2 = 0 \\ y_3 \text{ anything} \end{cases}$$

The nullspace of  $R^T$  contains all vectors  $y = (0, 0, y_3)$  and it is the line of the basis vector  $(0, 0, 1)$ .

**The left nullspace has dimension  $m - r$ .** The solutions are  $y = (0, \dots, y_{r+1}, \dots, y_m)$

✚ In  $\mathbf{R}^n$  the row space and nullspace have dimensions  $r$  and  $n - r$  (adding to  $n$ )

✚ In  $\mathbf{R}^m$  the column space and left nullspace have dimensions  $r$  and  $m - r$  (total  $m$ )





## ***The Four Subspaces for $A$***

***The subspace dimensions for  $A$  are the same as for  $R$ .***

These matrices are connected by an invertible matrix  $E$ .  $EA = R$  and  $A = E^{-1}R$

**1.**  $A$  has the same row space as  $R$ . Same dimension  $r$  and same basis

Every row of  $A$  is a combination of the rows of  $R$ . Also every row of  $R$  is a combination of the rows of  $A$ .

**2.** The column space of  $A$  has dimension  $r$ . The number of independent columns equals the number of independent rows.

**3.**  $A$  has the same nullspace as  $R$ . Dimension  $n - r$  and same basis.

$$(\text{dimension of column space}) + (\text{dimension of nullspace}) = \text{dimension of } R^n$$

**4.** The left nullspace  $A$  (the nullspace of  $A^T$ ) has dimension  $m - r$ .

## **Fundamental Theorem of Linear Algebra, (Part 1)**

The column space and row space both have dimension  $r$ .

The nullspaces have dimensions  $n - r$  and  $m - r$ .

## ***Example***

Consider  $A = \begin{bmatrix} 1 & 2 & 3 \end{bmatrix}$

$A$  has  $m = 1$ ,  $n = 3$ , and rank:  $r = 1$ .

The row space is a line in  $R^3$ .

The nullspace is the plane  $Ax = x_1 + 2x_2 + 3x_3 = 0$ . This plane has dimension 2 (which is  $3 - 1$ ).

The columns of this 1 by 3 matrix are in  $R^1$ . The column space is all of  $R^1$ .

The left nullspace contains only the zero vector.

The only solution to  $A^T y = 0$  is  $y = 0$ , the only combination of the row that gives the zero row. Thus

$N(A^T)$  is  $Z$ , the zero space with dimension 0 ( $m - r$ ). In  $R^m$  the dimensions  $(1 + 0) = 1$ .

### ***Example***

Consider  $A = \begin{pmatrix} 1 & 2 & 3 \\ 1 & 2 & 3 \end{pmatrix}$

$A$  has  $m = 2$ ,  $n = 3$ , and rank:  $r = 1$ .

The row space is a line in  $\mathbf{R}^3$ .

The nullspace is the plane  $x_1 + 2x_2 + 3x_3 = 0$ . This plane has dimension 3 (1 + 2).

The columns are multiples of the first column (1, 1).

The left nullspace contains more than one zero vector. The solution to  $A^T y = 0$  has the solution  $y = (1, -1)$ .

The column space and nullspace are perpendicular lines in  $\mathbf{R}^2$ . Their dimensions are 1 and 1 = 2.

Column space = line through  $\begin{pmatrix} 1 \\ 1 \end{pmatrix}$

Left nullspace=line through  $\begin{pmatrix} 1 \\ -1 \end{pmatrix}$

## Exercises

## Section 4.1 – Matrix Transformations from $R^n$ to $R^m$

1. Find the standard matrix for the transformation defined by the equations

$$a) \begin{cases} w_1 = 2x_1 - 3x_2 + 4x_4 \\ w_2 = 3x_1 + 5x_2 - x_4 \end{cases}$$

$$b) \begin{cases} w_1 = 7x_1 + 2x_2 - 8x_3 \\ w_2 = -x_2 + 5x_3 \\ w_3 = 4x_1 + 7x_2 - x_3 \end{cases}$$

$$c) \begin{cases} w_1 = x_1 \\ w_2 = x_1 + x_2 \\ w_3 = x_1 + x_2 + x_3 \\ w_4 = x_1 + x_2 + x_3 + x_4 \end{cases}$$

2. Find the standard matrix for the operator  $T$  defined by the formula

$$a) T(x_1, x_2) = (2x_1 - x_2, x_1 + x_2)$$

$$b) T(x_1, x_2, x_3) = (x_1 + 2x_2 + x_3, x_1 + 5x_2, x_3)$$

$$c) T(x_1, x_2, x_3) = (4x_1, 7x_2, -8x_3)$$

3. Find the standard matrix for the transformation  $T$  defined by the formula

$$a) T(x_1, x_2) = (x_2, -x_1, x_1 + 3x_2, x_1 - x_2)$$

$$b) T(x_1, x_2, x_3) = (0, 0, 0, 0, 0)$$

$$c) T(x_1, x_2, x_3, x_4) = (x_4, x_1, x_3, x_2, x_1 - x_3)$$

$$d) T(x_1, x_2, x_3, x_4) = (7x_1 + 2x_2 - x_3 + x_4, x_2 + x_3, -x_1)$$

## Section 4.2 – General Linear Transformations

### Definition

A transformation  $T$  assigns an output  $T(v)$  to each input vector  $v$ . The transformation is **linear** if it meets these requirements for all  $v$  and  $w$ :

$$\begin{cases} T(cv) = cT(v) \\ T(v + w) = T(v) + T(w) \end{cases}$$

We can combine both into one:  $T(cv + dw) = cT(v) + dT(w)$

### Theorem

If  $T : V \rightarrow W$  is a linear transformation, then:

1.  $T(0) = 0$
2.  $T(u - v) = T(u) - T(v)$  for all  $u$  and  $v$  in  $V$ .

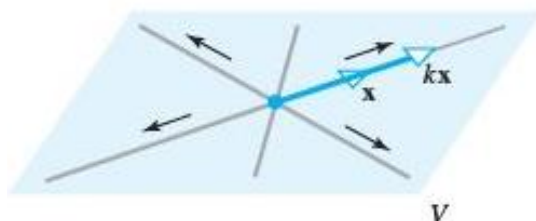
### Example

If  $V$  is a vector space and  $k$  is any scalar, then the mapping  $T : V \rightarrow W$  given by  $T(x) = kx$  is a linear operator on  $V$ , for if  $c$  is any scalar and if  $u$  and  $v$  are any vectors in  $V$ , then

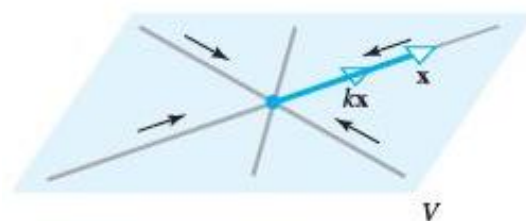
$$T(cu) = k(cu) = c(ku) = cT(u)$$

$$T(u + v) = k(u + v) = ku + kv = T(u) + T(v)$$

If  $0 < k < 1$ , then  $T$  is called **contraction** of  $V$  with factor  $k$ , and if  $k > 1$ , then  $T$  is called **dilation** of  $V$  with factor  $k$



*Dilation of  $V$*



*Contraction of  $V$*

### Example

Determine if the given function  $T$  is a linear transformation. Also give the domain and range of  $T$ ; if  $T$  is linear, find the  $A$  such  $T = f_A$ .  $T(x, y, z) = (z - x, z - y)$

### Solution

Let  $\mathbf{u} = (x_1, y_1, z_1)$  and  $\mathbf{v} = (x_2, y_2, z_2)$

$$\begin{aligned} T(\mathbf{u} + \mathbf{v}) &= T(x_1 + x_2, y_1 + y_2, z_1 + z_2) \\ &= (z_1 + z_2 - (x_1 + x_2), z_1 + z_2 - (y_1 + y_2)) \\ &= (z_1 + z_2 - x_1 - x_2, z_1 + z_2 - y_1 - y_2) \\ &= (z_1 - x_1, z_1 - y_1) + (z_2 - x_2, z_2 - y_2) \\ &= T(x_1, y_1, z_1) + T(x_2, y_2, z_2) \\ &= T(\mathbf{u}) + T(\mathbf{v}) \end{aligned}$$

$$\begin{aligned} T(r\mathbf{u}) &= T(rx_1, ry_1, rz_1) \\ &= (rz_1 - rx_1, rz_1 - ry_1) \\ &= r(z_1 - x_1, z_1 - y_1) \\ &= rT(\mathbf{u}) \end{aligned}$$

Since  $T(\mathbf{u} + \mathbf{v}) = T(\mathbf{u}) + T(\mathbf{v})$  and  $T(r\mathbf{u}) = rT(\mathbf{u})$ , then function  $T$  is a linear transformation.

**Domain:**  $T: \mathbb{R}^3 \rightarrow \mathbb{R}^2$

$$T(x, y, z) = (z - x, z - y) \Rightarrow \begin{pmatrix} -x + z \\ -y + z \end{pmatrix}$$

$$A = \begin{matrix} \begin{matrix} x & y & z \end{matrix} \\ \begin{bmatrix} -1 & 0 & 1 \\ 0 & -1 & 1 \end{bmatrix} \end{matrix}$$

### Example – the Zero Transformations

Let  $V$  and  $W$  be any vector spaces. The mapping  $T: V \rightarrow W$  such that  $T(\mathbf{v}) = 0$  for every  $\mathbf{v}$  in  $V$  is a linear transformation called the zero transformation. To see that  $T$  is linear, observe that:

$$T(\mathbf{u} + \mathbf{v}) = 0, \quad T(\mathbf{u}) = 0, \quad T(\mathbf{v}) = 0, \quad \text{and} \quad T(k\mathbf{u}) = 0$$

Therefore;  $T(\mathbf{u} + \mathbf{v}) = T(\mathbf{u}) + T(\mathbf{v})$  and  $T(k\mathbf{u}) = kT(\mathbf{u})$

### Example

Choose a fixed vector  $\mathbf{a} = (1, 3, 4)$ , and let  $T(v)$  be the dot product  $\mathbf{a} \cdot \mathbf{v}$ :

### Solution

$$\text{Let } v = (v_1, v_2, v_3)$$

$$T(v) = \mathbf{a} \cdot \mathbf{v} = (1, 3, 4) \cdot (v_1, v_2, v_3) = v_1 + 3v_2 + 4v_3$$

This is linear. The inputs  $v$  come from three dimensional space, so  $V = \mathbb{R}^3$ . The output just numbers, so the output space is  $W = \mathbb{R}^1$ . We are multiplying by the row matrix  $A = [1, 3, 4]$ .

$$\text{Then } T(v) = Av$$

### Example

Show that the length  $T(v) = \|v\|$  is not linear.

### Solution

$$\|v + w\| \stackrel{?}{=} \|v\| + \|w\| \quad \text{There are not equal because the sides of a triangle satisfy an inequality}$$

$$\|v + w\| \leq \|v\| + \|w\|$$

$$\|cw\| \stackrel{?}{=} c\|v\| \quad \text{Not - because the length } \|-v\| \neq -\|v\|$$

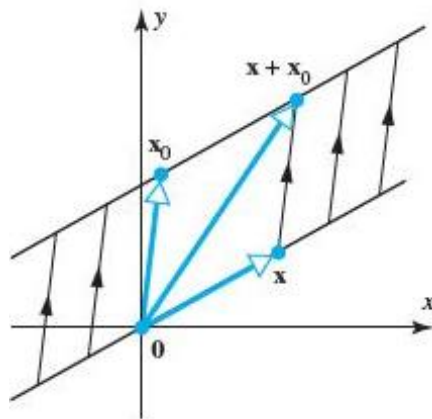
### Example

If  $x_0$  is a fixed nonzero vector in  $\mathbb{R}^2$ , then the transformation

$$T(x) = x + x_0$$

It has a geometric effect of translating each point  $x$  in a direction parallel to  $x_0$  through a distance of

$\|x_0\|$ . This cannot be a linear transformation since  $T(0) = x_0$



### ***Theorem***

Let  $T : V \rightarrow W$  be the linear transformation, where  $V$  is finite dimensional. If  $S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  is a basis for  $V$ , then the image of any vector  $\mathbf{v}$  in  $V$  can be expressed as

$$T(\mathbf{v}) = c_1 T(\mathbf{v}_1) + c_2 T(\mathbf{v}_2) + \dots + c_n T(\mathbf{v}_n)$$

Where  $c_1, c_2, \dots, c_n$  are the coefficients required to express  $\mathbf{v}$  as a linear combination of the vectors in  $S$ .

### ***Example***

Consider the basis  $S = \{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  for  $R^3$ , where  $\mathbf{v}_1 = (1, 1, 1)$   $\mathbf{v}_2 = (1, 1, 0)$   $\mathbf{v}_3 = (1, 0, 0)$

Let  $T : R^3 \rightarrow R^2$  be the linear transformation for which

$$T(\mathbf{v}_1) = (1, 0), \quad T(\mathbf{v}_2) = (2, -1), \quad T(\mathbf{v}_3) = (4, 3)$$

Find a formula for  $T(\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3)$ , and then use that formula to compute  $T(2, -3, 5)$

### **Solution**

$$(\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3) = c_1 (1, 1, 1) + c_2 (1, 1, 0) + c_3 (1, 0, 0)$$

$$\begin{aligned} c_1 + c_2 + c_3 &= x_1 \\ c_1 + c_2 &= x_2 \\ c_1 &= x_3 \end{aligned} \Rightarrow \begin{cases} c_3 = x_1 - x_2 \\ c_2 = x_2 - x_3 \\ c_1 = x_3 \end{cases}$$

$$\begin{aligned} (\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3) &= x_3 (1, 1, 1) + (x_2 - x_3)(1, 1, 0) + (x_1 - x_2)(1, 0, 0) \\ &= x_3 \mathbf{v}_1 + (x_2 - x_3) \mathbf{v}_2 + (x_1 - x_2) \mathbf{v}_3 \end{aligned}$$

$$\begin{aligned} T(\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3) &= x_3 T(\mathbf{v}_1) + (x_2 - x_3) T(\mathbf{v}_2) + (x_1 - x_2) T(\mathbf{v}_3) \\ &= x_3 (1, 0) + (x_2 - x_3) (2, -1) + (x_1 - x_2) (4, 3) \\ &= (4x_1 - 2x_2 + x_3, 3x_1 - 4x_2 + x_3) \end{aligned}$$

$$T(2, -3, 5) = (4(2) - 2(-3) + (5), 3(2) - 4(-3) + (5)) = \underline{(9, 23)}$$

### Example

$T$  is the transformation that rotates every vector by  $30^\circ$ , the domain is the  $xy$ -plane (where the input vector  $\mathbf{v}$  is). The range is also the  $xy$ -plane (where the rotated  $T(\mathbf{v})$  is). Is the rotation linear?

### Solution

Yes it is. We can rotate two vectors and add the results. The sum of rotation  $T(\mathbf{v}) + T(\mathbf{w})$  is the same as the rotation  $T(\mathbf{v} + \mathbf{w})$  of the sum.

The whole plane is turning together, in this linear transformation.

### Definition

If  $T : V \rightarrow W$  is a linear transformation, then the set of vectors in  $V$  that  $T$  maps into  $\mathbf{0}$  is called **kernel** of  $T$  and is denoted by  $\ker(T)$ . The set of all vectors in  $W$  that are images under  $T$  of at least one vector in  $V$  is called the **range** of  $T$  and is denoted by  $R(T)$ .

### Note:

Transformations have a language of their own. Where there is no matrix, we can't talk about a column space. But the idea can be rescued and used. The column space consisted of all outputs  $A\mathbf{v}$ . The nullspace consisted of all inputs for which  $A\mathbf{v} = \mathbf{0}$ . Translate those into "range" and "kernel"

**Range** of  $T$  = set of all outputs  $T(\mathbf{v})$ : corresponds to column space

**Kernel** of  $T$  = set of all inputs for which  $T(\mathbf{v}) = \mathbf{0}$ : corresponds to nullspace

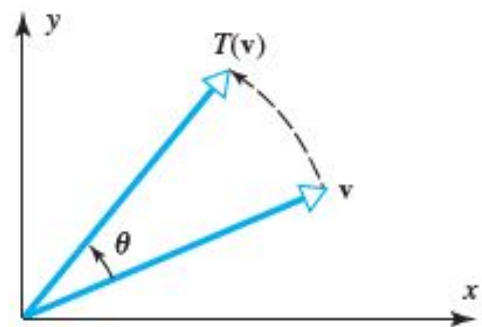
### Example

Project every 3-dimensional vector down onto the  $xy$  plane.

The range is that plane, which contains every  $T(\mathbf{v})$ . The kernel is the  $z$  axis (which projects down to zero). This projection is linear.

### Example – Kernel and Range of a Rotation

Let  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  be the linear operator that rotates each vector in the  $xy$ -plane through the angle  $\theta$ . Since every vector in the  $xy$ -plane can be obtained by rotating some vector through the angle  $\theta$ , it follows that  $R(T) = \mathbb{R}^2$ . Moreover, the only vector that rotates into  $\mathbf{0}$  is  $\mathbf{0}$ , so  $\ker(T) = \{\mathbf{0}\}$





### ***Theorem***

If  $T : V \rightarrow W$  is a linear transformation, then:

1. The *kernel* of  $T$  is a subspace of  $V$
2. The *range* of  $T$  is a subspace of  $W$

### ***Theorem***

If  $T : V \rightarrow W$  is a linear transformation from an  $n$ -dimensional vector space  $V$  to a vector space  $W$ , then

$$\text{rank}(T) + \text{nullity}(T) = n$$

### ***Example***

Project every 3-dimensional vector down onto horizontal plane  $z = 1$ .

The vector  $\mathbf{v} = (x, y, z)$  is transformed to  $T(\mathbf{v}) = (x, y, 1)$ . This transformation is not linear, it doesn't even transform  $\mathbf{v} = 0$  into  $T(\mathbf{v}) = 0$ .

Multiply every 3-dimensional vector by a 3 by 3 matrix  $A$ . This is definitely a linear transformation

$$T(\mathbf{v} + \mathbf{w}) = A(\mathbf{v} + \mathbf{w}) \quad \text{which does equal} \quad A\mathbf{v} + A\mathbf{w} = T(\mathbf{v}) + T(\mathbf{w})$$

### ***Example***

Suppose  $A$  is an invertible matrix. The kernel of  $T$  is the zero vector; the range  $W$  equals the domain  $V$ .

Another linear transformation is multiplication by  $A^{-1}$ . This is the inverse transformation  $T^{-1}$ , which brings every vector  $T(\mathbf{v})$  back to  $\mathbf{v}$ :

$$T^{-1}(T(\mathbf{v})) = \mathbf{v} \quad \text{matches the matrix multiplication} \quad A^{-1}(A\mathbf{v}) = \mathbf{v}$$

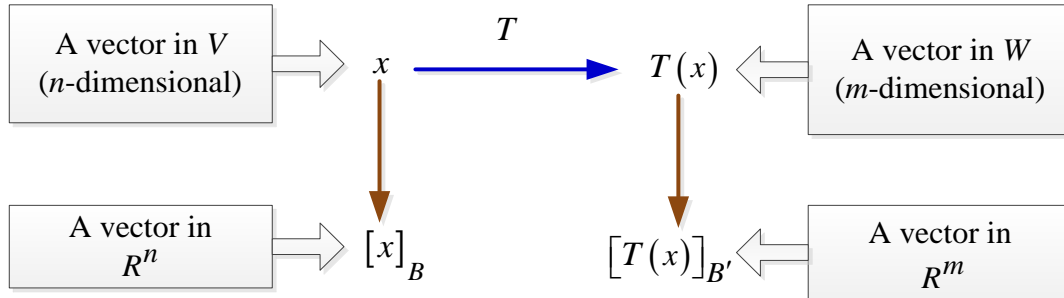
### ***Are all linear transformation produced by matrices?***

Each  $m$  by  $n$  matrix does produce a linear transformation from  $V = R^n$  to  $W = R^m$ . When a linear  $T$  is described as a “rotation” or “projection” or “...” is there always a matrix hiding behind  $T$ ?

The answer is yes. This is an approach to linear algebra that doesn't start with matrices. The next section shows that we still end up with matrices.

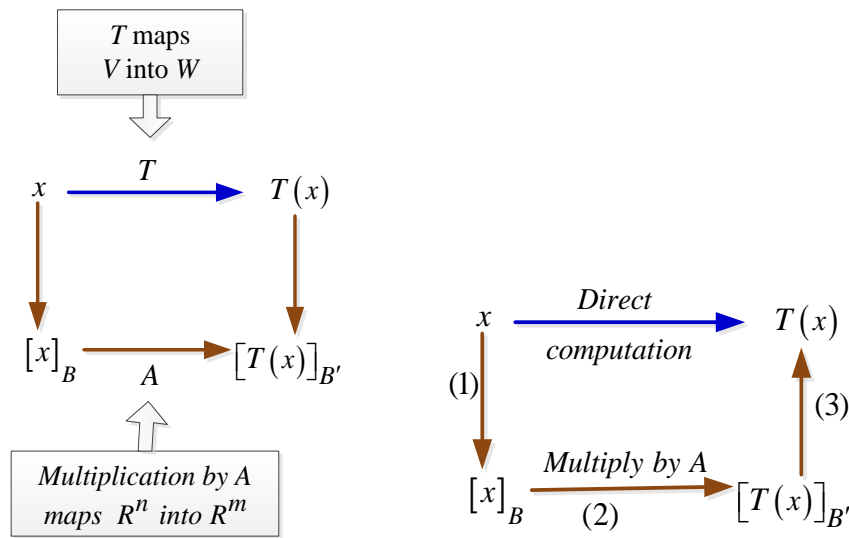
## Matrices for General Linear Transformations

Suppose that  $V$  is an  $n$ -dimensional vector space,  $W$  is an  $m$ -dimensional vector space, and that  $T: V \rightarrow W$  is a linear transformation. Suppose further that  $B$  is a basis for  $V$ , that  $B'$  is a basis for  $W$ , and that for each  $\mathbf{x}$  in  $V$ , the coordinate matrices for  $\mathbf{x}$  and  $T(\mathbf{x})$  are  $[\mathbf{x}]_B$  and  $[T(\mathbf{x})]_{B'}$ , respectively



By using matrix multiplication, we can execute the linear transformation and the following indirect procedure:

1. Compute the coordinate vector  $[\mathbf{x}]_B$
2. Multiply  $[\mathbf{x}]_B$  on the left by  $A$  to produce  $[T(\mathbf{x})]_{B'}$
3. Reconstruct  $T(\mathbf{x})$  from its coordinate vector  $[T(\mathbf{x})]_{B'}$



$$A[\mathbf{x}]_B = [T(\mathbf{x})]_{B'}$$

### Example

Let  $T : P_1 \rightarrow P_2$  be the linear transformation defined by  $T(p(x)) = xp(x)$

Find the matrix for  $T$  with respect to the standard bases

$$B = \{u_1, u_2\} \quad \text{and} \quad B' = \{v_1, v_2, v_3\}$$

Where  $u_1 = 1, u_2 = x; v_1 = 1, v_2 = x, v_3 = x^2$

### Solution

$$T(u_1) = T(1) = x$$

$$T(u_2) = T(x) = x(x) = x^2$$

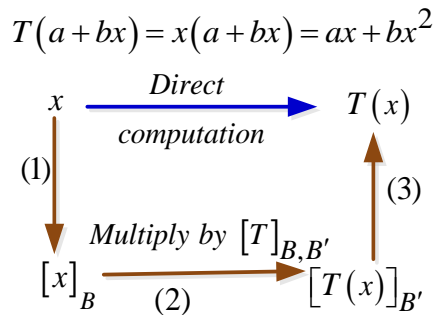
$$\left[ T(u_1) \right]_{B'} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \quad \left[ T(u_2) \right]_{B'} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

The matrix for  $T$  with respect to  $B$  and  $B'$  is

$$[T]_{B, B'} = \left[ \left[ T(u_1) \right]_{B'} \mid \left[ T(u_2) \right]_{B'} \right] = \begin{pmatrix} 0 & 0 \\ 1 & 0 \\ 0 & 1 \end{pmatrix}$$

### Example

Let  $T : P_1 \rightarrow P_2$  be the linear transformation defined by  $T(p(x)) = xp(x)$  describe in the following figure to perform the computation



$$B = \{1, x\} \quad \text{and} \quad B' = \{1, x, x^2\}$$

### Solution

**Step 1:** The coordinates matrix for  $x = ax + b$  relative to the basis  $B = \{1, x\}$  is

$$[x]_B = \begin{bmatrix} a \\ b \end{bmatrix}$$

**Step 2:** Multiply  $[x]_B$  by the matrix  $[T]_{B, B'}$  found in previous example, we obtain

$$[T]_{B,B'} [\mathbf{x}]_B = \begin{bmatrix} 0 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} 0 \\ a \\ b \end{bmatrix} = [T(\mathbf{x})]_{B'}$$

**Step 3:** Reconstructing  $T(\mathbf{x}) = T(ax + b)$  from  $[T(\mathbf{x})]_{B'}$  we obtain

$$T(ax + b) = 0 + ax + bx^2 = ax + bx^2$$

## Exercises Section 4.2 – General Linear Transformations

1. The matrix  $\begin{pmatrix} 1 & 0 \\ 3 & 1 \end{pmatrix}$  gives a shearing transformation  $T(x, y) = (x, 3x + y)$ .

What happens to  $(1, 0)$  and  $(2, 0)$  on the  $x$ -axis.

What happens to the points on the vertical lines  $x = 0$  and  $x = a$ ?

2. A nonlinear transformation  $T$  is invertible if every  $\mathbf{b}$  in the output space comes from exactly one  $\mathbf{x}$  in the input space.  $T(\mathbf{x}) = \mathbf{b}$  always has exactly one solution. Which of these transformation (on real numbers  $x$ ) is invertible and what is  $T^{-1}$ ? None are linear, not even  $T_3$ . When you solve  $T(\mathbf{x}) = \mathbf{b}$ , you are inverting  $T$ :

$$T_1(x) = x^2 \quad T_2(x) = x^3 \quad T_3(x) = x + 9 \quad T_4(x) = e^x \quad T_5(x) = \frac{1}{x} \quad \text{for nonzero } x's$$

3. If  $S$  and  $T$  are linear transformations, is  $S(T(\mathbf{v}))$  linear or quadratic?

a) If  $S(\mathbf{v}) = \mathbf{v}$  and  $T(\mathbf{v}) = \mathbf{v}$ , then  $S(T(\mathbf{v})) = \mathbf{v}$  or  $\mathbf{v}^2$ ?

b)  $S(\mathbf{w}_1 + \mathbf{w}_2) = S(\mathbf{w}_1) + S(\mathbf{w}_2)$  and  $T(\mathbf{v}_1 + \mathbf{v}_2) = T(\mathbf{v}_1) + T(\mathbf{v}_2)$  combine into

$$S(T(\mathbf{v}_1 + \mathbf{v}_2)) = S(\text{_____}) = \text{_____} + \text{_____}$$

4. Find the range and kernel (like the column space and nullspace) of  $T$ :

a)  $T(v_1, v_2) = (v_2, v_1)$

c)  $T(v_1, v_2) = (0, 0)$

b)  $T(v_1, v_2, v_3) = (v_1, v_2)$

d)  $T(v_1, v_2) = (v_1, v_1)$

5.  $M$  is any 2 by 2 matrix and  $A = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix}$ . The transformation  $T$  is defined by  $T(M) = AM$ . What rules of matrix multiplication show that  $T$  is linear?

6. Which of these transformations satisfy  $T(v + w) = T(v) + T(w)$  and which satisfy  $T(cv) = cT(v)$ ?

a)  $T(v) = \frac{v}{\|v\|}$

c)  $T(v) = (v_1, 2v_2, 3v_3)$

b)  $T(v) = v_1 + v_2 + v_3$

d)  $T(v) = \text{largest component of } v$ .

7. Consider the basis  $S = \{v_1, v_2, v_3\}$  for  $R^3$ , where  $v_1 = (1, 1, 1)$   $v_2 = (1, 1, 0)$   $v_3 = (1, 0, 0)$  and let  $T: R^3 \rightarrow R^3$  be the linear transformation for which

$$T(v_1) = (2, -1, 4), \quad T(v_2) = (3, 0, 1), \quad T(v_3) = (-1, 5, 1)$$

Find a formula for  $T(\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3)$ , and then use that formula to compute  $T(2, 4, -1)$

8. Consider the basis  $S = \{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  for  $R^3$ , where

$\mathbf{v}_1 = (1, 2, 1)$   $\mathbf{v}_2 = (2, 9, 0)$   $\mathbf{v}_3 = (3, 3, 4)$  and let  $T: R^3 \rightarrow R^2$  be the linear transformation for which

$$T(\mathbf{v}_1) = (1, 0), \quad T(\mathbf{v}_2) = (-1, 1), \quad T(\mathbf{v}_3) = (0, 1)$$

Find a formula for  $T(\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3)$ , and then use that formula to compute  $T(7, 13, 7)$

9. let  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$  be vectors in a vector space  $V$ , and let  $T: V \rightarrow R^3$  be the linear transformation for which  $T(\mathbf{v}_1) = (1, -1, 2)$ ,  $T(\mathbf{v}_2) = (0, 3, 2)$ ,  $T(\mathbf{v}_3) = (-3, 1, 2)$ . Find  $T(2\mathbf{v}_1 - 3\mathbf{v}_2 + 4\mathbf{v}_3)$

10. Let  $T: R^2 \rightarrow R^2$  be the linear operation given by the formula  $T(x, y) = (2x - y, -8x + 4y)$

Which of the following vectors are in  $R(T)$

a)  $(1, -4)$  b)  $(5, 0)$  c)  $(-3, 12)$

11. Let  $T: R^2 \rightarrow R^2$  be the linear operation given by the formula  $T(x, y) = (2x - y, -8x + 4y)$

Which of the following vectors are in  $\ker(T)$

a)  $(5, 10)$  b)  $(3, 2)$  c)  $(1, 1)$

12. Let  $T: R^4 \rightarrow R^3$  be the linear operation given by the formula

$$T(\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3, \mathbf{x}_4) = (4x_1 + x_2 - 2x_3 - 3x_4, 2x_1 + x_2 + x_3 - 4x_4, 6x_1 - 9x_3 + 9x_4)$$

Which of the following vectors are in  $R(T)$

a)  $(0, 0, 6)$  b)  $(1, 3, 0)$  c)  $(2, 4, 1)$

13. Let  $T: R^4 \rightarrow R^3$  be the linear operation given by the formula

$$T(\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3, \mathbf{x}_4) = (4x_1 + x_2 - 2x_3 - 3x_4, 2x_1 + x_2 + x_3 - 4x_4, 6x_1 - 9x_3 + 9x_4)$$

Which of the following vectors are in  $\ker(T)$

a)  $(3, -8, 2, 0)$  b)  $(0, 0, 0, 1)$  c)  $(0, -4, 1, 0)$

14. Determine if the given function  $T$  is a linear transformation

$$T: M_{22} \rightarrow M_{22} \text{ by } T \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} 2ab & 3cd \\ 0 & 0 \end{bmatrix}$$

15. Determine if the given function  $T$  is a linear transformation

$$T : M_{22} \rightarrow M_{22} \text{ by } T \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} a+d & 0 \\ 0 & b+c \end{bmatrix}$$

16. Determine if the given function  $T$  is a linear transformation where  $A$  is fixed  $2 \times 3$  matrix

$$T : M_{22} \rightarrow M_{23} \text{ by } T(B) = BA$$

17. Determine if the given function  $T$  is a linear transformation. Also give the domain and range of  $T$ ; if  $T$  is linear, find the  $A$  such  $T = f_A$ .  $T(x, y, z) = (2x + y, x - y + z)$

18. Determine if the given function  $T$  is a linear transformation. Also give the domain and range of  $T$ .

$$T(x, y) = (x^2, y)$$

19. Determine if the given function  $T$  is a linear transformation. Also give the domain and range of  $T$ ; if  $T$  is linear, find the  $A$  such  $T = f_A$ .

$$T(x, y, z) = (z - x, z - y)$$

20. Show that the function  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  given the formula

$$T(x_1, x_2, x_3) = (2x_1 - x_2 + x_3, x_2 - 4x_3) \text{ is linear transformation}$$

## Section 4.3 – LU-Decompositions

The goal is to describe Gaussian elimination in the most useful way by looking at them closely, which are factorizations of a matrix.

*The factors are triangular matrices.*

*The factorization that comes from elimination is  $A = LU$ .*

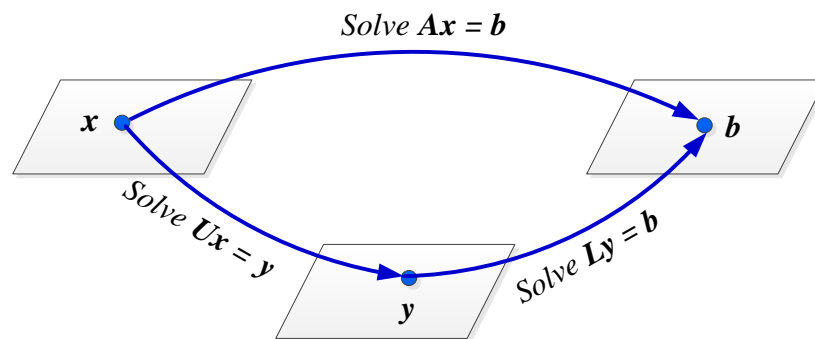
### The Method of LU-Decomposition

**Step 1:** Rewrite the system  $Ax = b$  as  $LUx = b$

**Step 2:** Define a new  $n \times 1$  matrix  $y$  by  $Ux = y$

**Step 3:** Use  $Ux = y$  to rewrite  $LUx = b$  as  $Ly = b$  and solve this system for  $y$ .

**Step 4:** Substitute  $y$  in  $Ux = y$  and solve for  $x$ .



### Example

Given 2 by 2 matrix  $A = \begin{pmatrix} 2 & 1 \\ 6 & 8 \end{pmatrix}$

To make **row 2 column 1** is **zero** then we need to subtract 3 times **row 1** from **row 2**

$$\begin{pmatrix} 2 & 1 \\ 6 & 8 \end{pmatrix} R_2 - 3R_1$$

That step is  $E_{21} = \begin{pmatrix} 1 & 0 \\ -3 & 1 \end{pmatrix}$  in the forward direction such that:

$$E_{21}A = \begin{pmatrix} 1 & 0 \\ -3 & 1 \end{pmatrix} \begin{pmatrix} 2 & 1 \\ 6 & 8 \end{pmatrix} = \begin{pmatrix} 2 & 1 \\ 0 & 5 \end{pmatrix} = U$$



The return step from  $U$  to  $A$  is  $L = E_{21}^{-1}$

Back from  $U$  to  $A$ :  $\underline{E_{21}^{-1}U} = \begin{pmatrix} 1 & 0 \\ 3 & 1 \end{pmatrix} \begin{pmatrix} 2 & 1 \\ 0 & 5 \end{pmatrix} = \begin{pmatrix} 2 & 1 \\ 6 & 8 \end{pmatrix} \underline{= A}$

Therefore;  $A = LU$

### Example

What matrix  $L$  and  $U$  puts  $A$  into triangular form  $A = LU$  where

$$A = \begin{pmatrix} 2 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 2 \end{pmatrix}$$

### Solution

$$\begin{pmatrix} 2 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 2 \end{pmatrix} R_2 - \frac{1}{2}R_1 : \ell_{21} \rightarrow \begin{pmatrix} 2 & 1 & 0 \\ 0 & \frac{3}{2} & 1 \\ 0 & 1 & 2 \end{pmatrix}$$

$$\begin{pmatrix} 2 & 1 & 0 \\ 0 & \frac{3}{2} & 1 \\ 0 & 1 & 2 \end{pmatrix} R_3 - \frac{2}{3}R_2 : \ell_{32} \rightarrow \begin{pmatrix} 2 & 1 & 0 \\ 0 & \frac{3}{2} & 1 \\ 0 & 0 & \frac{4}{3} \end{pmatrix} = U$$

The lower triangular  $L$  has all **1's** on its diagonal. The multipliers  $\ell_{ij}$  are **below** the diagonal of  $L$

$$L = \begin{pmatrix} 1 & 0 & 0 \\ \frac{1}{2} & 1 & 0 \\ 0 & \frac{2}{3} & 1 \end{pmatrix}$$

$$\begin{pmatrix} 2 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 2 \end{pmatrix} \overset{A}{=} \begin{pmatrix} 1 & 0 & 0 \\ \frac{1}{2} & 1 & 0 \\ 0 & \frac{2}{3} & 1 \end{pmatrix} \overset{L}{\left( \right)} \begin{pmatrix} 2 & 1 & 0 \\ 0 & \frac{3}{2} & 1 \\ 0 & 0 & \frac{4}{3} \end{pmatrix} \overset{U}{\left( \right)}$$

$$\diamond \left( E_{32} E_{31} E_{21} \right) A = U \quad \text{becomes} \quad A = \begin{pmatrix} E^{-1} & E^{-1} & E^{-1} \\ 21 & 31 & 32 \end{pmatrix} U \quad \text{which is} \quad A = LU$$

The inverses go in opposite order.

❖  $(A = LU)$  This is **elimination without row exchanges**. The **upper triangular  $U$**  has the pivots on its diagonal. The **lower triangular  $L$**  has all 1's on its diagonal. **The multipliers  $\ell_{ij}$  are below the diagonal of  $L$ .**

## One Square System = Two Triangular Systems

**Factor:** into  $L$  and  $U$ , by forward elimination on  $A$ .

**Solve:** forward on  $b$  using  $L$ , then back substitution using  $U$ .

Solve  $Lc = b$  and then solve  $Ux = c$

### Example

Forward elimination on  $Ax = b$  ends at  $Ux = c$

$$\begin{array}{rcl} x + 2y = 5 & & x + 2y = 5 \\ 4x + 9y = 21 & \text{becomes} & y = 1 \end{array}$$

### Solution

The multiplier was 4.  $(R_2 - 4R_1)$

The lower triangular system:  $Lc = b$

$$\begin{bmatrix} 1 & 0 \\ 4 & 1 \end{bmatrix} [c] = \begin{bmatrix} 5 \\ 21 \end{bmatrix} \Rightarrow c = \begin{bmatrix} 5 \\ 1 \end{bmatrix}$$

The upper triangular system:  $Ux = c$

$$\begin{bmatrix} 1 & 2 \\ 0 & 1 \end{bmatrix} [x] = \begin{bmatrix} 5 \\ 1 \end{bmatrix} \Rightarrow x = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$$

## To solve 1000 equations on a PC

- ❖ Elimination on  $A$  requires about  $\frac{1}{3}n^3$  multiplications and  $\frac{1}{3}n^3$  subtractions.
- ❖ Each right side needs  $n^2$  multiplications and  $n^2$  subtractions.

## Exercises      Section 4.3 – LU-Decompositions

1. What matrix  $E$  puts  $A$  into triangular form  $EA = U$ ? Multiply by  $E^{-1} = L$  to factor  $A$  into  $LU$ :

$$A = \begin{bmatrix} 2 & 1 & 0 \\ 0 & 4 & 2 \\ 6 & 3 & 5 \end{bmatrix}$$

2. Solve  $Lc = b$  to find  $c$ . Then solve  $Ux = c$  to find  $x$ . What was  $A$ ?

$$L = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 1 \end{bmatrix} \quad U = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix} \quad b = \begin{bmatrix} 4 \\ 5 \\ 6 \end{bmatrix}$$

3. Find  $L$  and  $U$  for the symmetric matrix

$$A = \begin{bmatrix} a & a & a & a \\ a & b & b & b \\ a & b & c & c \\ a & b & c & d \end{bmatrix}$$

Find four conditions on  $a, b, c, d$  to get  $A = LU$  with four pivots

4. For which  $c$  is  $A = LU$  impossible – with three pivots?

$$A = \begin{pmatrix} 1 & 2 & 0 \\ 3 & c & 1 \\ 0 & 1 & 1 \end{pmatrix}$$

5. Find an  $LU$ -decomposition of the coefficient matrix, and then use to solve the system

$$a) \quad \begin{bmatrix} 2 & 8 \\ -1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -2 \\ -2 \end{bmatrix}$$

$$b) \quad \begin{bmatrix} -5 & -10 \\ 6 & 5 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -10 \\ 19 \end{bmatrix}$$

$$c) \quad \begin{bmatrix} 2 & -2 & -2 \\ 0 & -2 & 2 \\ -1 & 5 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -4 \\ -2 \\ 6 \end{bmatrix}$$

$$d) \quad \begin{bmatrix} -3 & 12 & -6 \\ 1 & -2 & 2 \\ 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -33 \\ 7 \\ -1 \end{bmatrix}$$

$$e) \quad \begin{bmatrix} -1 & 0 & 1 & 0 \\ 2 & 3 & -2 & 6 \\ 0 & -1 & 2 & 0 \\ 0 & 0 & 1 & 5 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 5 \\ -1 \\ 3 \\ 7 \end{bmatrix}$$

## Section 4.4 – Eigenvalues and Eigenvectors

In many problems in science and mathematics, linear equations  $A\mathbf{x} = \mathbf{b}$  come from steady state problems. Eigenvalues have their greatest importance in dynamic problems. The solution of

$Ax = \lambda x$  or  $\frac{dx}{dt} = Ax$  (is changing with time) has nonzero solutions. (*All matrices are square*)

### Definition

Suppose  $A$  is an  $n \times n$  matrix and

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

The values of  $\lambda$  are called eigenvalues of the matrix  $A$  and the nonzero vectors  $\mathbf{x}$  in  $\mathbf{R}^n$  are called the eigenvectors corresponding to that eigenvalue ( $\lambda$ ).

✚ One of the meanings of the word “*eigen*” in German is “*proper*”; eigenvalues are also called *proper values*, *characteristic values*, or *latent roots*.

### Example

The vector  $x = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$  is an eigenvector of  $A = \begin{bmatrix} 3 & 0 \\ 8 & -1 \end{bmatrix}$  corresponding to the eigenvalue  $\lambda = 3$  since

$$Ax = \begin{bmatrix} 3 & 0 \\ 8 & -1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 3 \\ 6 \end{bmatrix} = 3 \begin{bmatrix} 1 \\ 2 \end{bmatrix} = 3x$$

Eigenvalues and eigenvectors have a useful geometric interpretation in  $\mathbf{R}^2$  and  $\mathbf{R}^3$ .

## The equation for the *eigenvalues*

Let's rewrite the equation  $A\mathbf{x} = \lambda\mathbf{x}$ .

$$A\mathbf{x} - \lambda\mathbf{x} = 0$$

$\lambda$  : are the eigenvalues and not a vector

$$A\mathbf{x} - \lambda I\mathbf{x} = 0$$

$$(A - \lambda I)\mathbf{x} = 0$$

The matrix  $A - \lambda I$  times the eigenvectors  $\mathbf{x}$  is the zero vector. The eigenvectors makes up the nullspace of  $A - \lambda I$ .

### ***Definition***

The number  $\lambda$  is an eigenvalue of  $A$  if and only if  $A - \lambda I$  is singular:

$$\det(A - \lambda I) = 0$$

This is called ***characteristic equation*** of  $A$ ; the scalars satisfying this equation are the eigenvalues of  $A$ . when expanding the determinant  $\det(A - \lambda I)$  is a polynomial in  $\lambda$  called the ***characteristic polynomial*** of  $A$ .

### ***Example***

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

### **Solution**

$$\begin{aligned} \det(A - \lambda I) &= \det\left(\begin{bmatrix} 3 & 2 \\ -1 & 0 \end{bmatrix} - \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}\right) \\ &= \det\left(\begin{bmatrix} 3-\lambda & 2 \\ -1 & -\lambda \end{bmatrix}\right) \\ &= (3-\lambda)(-\lambda) + 2 \\ &= \lambda^2 - 3\lambda + 2 \end{aligned}$$

The characteristic equation of  $A$  is:

$$\lambda^2 - 3\lambda + 2 = 0 \Rightarrow \boxed{\lambda_1 = 1} \quad \boxed{\lambda_2 = 2}; \text{ these are the eigenvalues of } A.$$

### ***Theorem***

If  $\mathbf{A}$  is an  $n \times n$  triangular matrix (upper triangular, lower triangular, or diagonal), then the eigenvalues of  $\mathbf{A}$  are the entries on the main diagonal of  $\mathbf{A}$ .

### ***Example***

Find the eigenvalues of the lower triangular matrix

$$A = \begin{pmatrix} \frac{1}{2} & 0 & 0 \\ -1 & \frac{3}{2} & 0 \\ 5 & -8 & -\frac{1}{4} \end{pmatrix}$$

### **Solution**

The eigenvalues are:  $\lambda = \frac{1}{2}$ ,  $\lambda = \frac{3}{2}$ , and  $\lambda = -\frac{1}{4}$

### ***Theorem***

If  $\mathbf{A}$  is an  $n \times n$  matrix, the following are equivalent.

- a)*  $\lambda$  is an eigenvalue of  $\mathbf{A}$ .
- b)* The system of equations  $(\mathbf{A} - \lambda \mathbf{I})\mathbf{x} = \mathbf{0}$  has nontrivial solutions.
- c)* There is a nonzero vector  $\mathbf{x}$  in  $\mathbf{R}^n$  such that  $\mathbf{Ax} = \lambda\mathbf{x}$ .
- d)*  $\lambda$  is a real solution of the characteristic equation  $\det(\mathbf{A} - \lambda \mathbf{I}) = 0$

## Eigenvectors

To find the eigenvector  $\mathbf{x}$ , for each eigenvalue  $\lambda$  solve  $(A - \lambda I)x = 0$  *or*  $Ax = \lambda x$

From the eigenvalues, the eigenvectors, in the form  $\mathbf{V}_i = \begin{pmatrix} x_i \\ y_i \end{pmatrix}$ , of the system can be determined

by letting:

$$(A - \lambda_1 I)V_1 = 0 \quad \text{and} \quad (A - \lambda_2 I)V_2 = 0$$

### Example

Find the eigenvalues and the eigenvectors of the matrix  $A = \begin{pmatrix} 1 & 2 \\ 2 & 4 \end{pmatrix}$

### Solution

$$\begin{aligned} \det(A - \lambda I) &= \det \begin{pmatrix} 1-\lambda & 2 \\ 2 & 4-\lambda \end{pmatrix} \\ &= (1-\lambda)(4-\lambda) - 4 \\ &= \lambda^2 - 5\lambda + 4 - 4 \\ &= \lambda^2 - 5\lambda \\ &= \lambda(\lambda - 5) = \mathbf{0} \end{aligned}$$

The eigenvalues of  $A$  are:  $\lambda_1 = 0$   $\lambda_2 = 5$

For  $\lambda_1 = 0$ , we have:

$$(A - \lambda_1 I)V_1 = 0$$

$$\begin{pmatrix} 1-0 & 2 \\ 2 & 4-0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\begin{pmatrix} 1 & 2 \\ 2 & 4 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \rightarrow \begin{cases} x + 2y = 0 \\ 2x + 4y = 0 \end{cases} \Rightarrow x = -2y$$

If  $y = -1 \Rightarrow x = 2$ , therefore the eigenvector  $V_1 = \begin{pmatrix} 2 \\ -1 \end{pmatrix}$

$$\text{Or } \begin{pmatrix} -2y \\ y \end{pmatrix} = y \begin{pmatrix} -2 \\ 1 \end{pmatrix} \Rightarrow \mathbf{V_1 = \begin{pmatrix} -2 \\ 1 \end{pmatrix} \text{ or } \begin{pmatrix} 2 \\ -1 \end{pmatrix}}$$

For  $\lambda_2 = 5$ , we have:

$$(A - \lambda_2 I)V_2 = 0$$

$$\begin{pmatrix} 1-5 & 2 \\ 2 & 4-5 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\begin{pmatrix} -4 & 2 \\ 2 & -1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \rightarrow \begin{cases} -4x + 2y = 0 \\ 2x - y = 0 \end{cases} \Rightarrow 2x = y$$

$$\text{If } x = 1 \Rightarrow y = 2, \text{ therefore the eigenvector } V_2 = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$

## Power of a Matrix

### *Theorem*

If  $k$  is a positive integer,  $\lambda$  is an eigenvalue of a matrix  $A$ , and  $\mathbf{x}$  is a corresponding eigenvector, then  $\lambda^k$  is an eigenvalue of  $A^k$  and  $\mathbf{x}$  is a corresponding eigenvector.

### *Example*

Find the eigenvalues of  $A^7$  for  $A = \begin{pmatrix} 0 & 0 & -2 \\ 1 & 2 & 1 \\ 1 & 0 & 3 \end{pmatrix}$

### *Solution*

$$\det(A - \lambda I) = \begin{vmatrix} -\lambda & 0 & -2 \\ 1 & 2-\lambda & 1 \\ 1 & 0 & 3-\lambda \end{vmatrix} = \lambda^3 - 5\lambda^2 + 8\lambda - 4 = 0$$

The eigenvalues of  $A$ :  $\lambda = 1$  and  $\lambda = 2$

The eigenvalues of  $A^7$  are:  $\boxed{\lambda = 1^7 = 1}$  and  $\boxed{\lambda = 2^7 = 128}$

### *Theorem*

A square matrix  $A$  is invertible iff  $\lambda = 0$  is not an eigenvalue of  $A$ .



## Summary

To solve the eigenvalue problem for an  $n$  by  $n$  matrix:

1. Compute the determinant of  $A - \lambda I$ . With  $\lambda$  subtracted along the diagonal, this determinant starts with  $\lambda^n$  or  $-\lambda^n$ . It is a polynomial in  $\lambda$  of degree  $n$ .
2. Find the roots of this polynomial, by solving  $\det(A - \lambda I) = 0$ . The  $n$  roots are the  $n$  eigenvalues of  $A$ . They make  $A - \lambda I$  singular.
3. For each eigenvalue  $\lambda$ , solve  $(A - \lambda I)\mathbf{x} = \mathbf{0}$  to find an eigenvector  $\mathbf{x}$ .

## Imaginary Eigenvalues

### Example

Find the eigenvalues and the eigenvectors of the matrix  $A = \begin{pmatrix} -2 & -1 \\ 5 & 2 \end{pmatrix}$

### Solution

$$\begin{aligned}\det(A - \lambda I) &= \begin{vmatrix} -2 - \lambda & -1 \\ 5 & 2 - \lambda \end{vmatrix} \\ &= (-2 - \lambda)(2 - \lambda) + 5 \\ &= \lambda^2 - 4 + 5 \\ &= \lambda^2 + 1 = 0 \\ \Rightarrow \lambda^2 &= -1\end{aligned}$$

The solutions are:  $\lambda = \pm i$ .

$$\lambda_1 = i : (A - \lambda_1 I)V_1 = 0$$

$$\begin{pmatrix} -2 - i & -1 \\ 5 & 2 - i \end{pmatrix} \begin{pmatrix} x_1 \\ y_1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \rightarrow \begin{cases} (-2 - i)x_1 - y_1 = 0 \\ 5x_1 + (2 - i)y_1 = 0 \end{cases} \Rightarrow \begin{cases} (-2 - i)x_1 = y_1 \\ 5x_1 = -(2 - i)y_1 \end{cases}$$

Therefore the eigenvector  $V_1 = \begin{pmatrix} -1 \\ 2 + i \end{pmatrix}$

$$\lambda_1 = -i : (A - \lambda_2 I)V_2 = 0$$

$$\begin{pmatrix} -2 + i & -1 \\ 5 & 2 + i \end{pmatrix} \begin{pmatrix} x_2 \\ y_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \rightarrow \begin{cases} (-2 + i)x_2 - y_2 = 0 \\ 5x_2 + (2 + i)y_2 = 0 \end{cases} \Rightarrow \begin{cases} (-2 + i)x_2 = y_2 \\ 5x_2 = -(2 + i)y_2 \end{cases}$$

Therefore the eigenvector  $V_2 = \begin{pmatrix} 1 \\ -2 + i \end{pmatrix}$

### Example

Find the eigenvalues and the eigenvectors of the matrix  $A = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}$

### Solution

$$\det(A - \lambda I) = \begin{vmatrix} -\lambda & 1 \\ -1 & -\lambda \end{vmatrix} = \lambda^2 + 1 = 0$$
$$\Rightarrow \lambda^2 = -1 \rightarrow \lambda_1 = i \quad \lambda_2 = -i$$

The matrix  $A$  is a  $90^\circ$  rotation which has no real eigenvalues or eigenvectors.

No vector  $A\mathbf{x}$  stays in the same direction as  $\mathbf{x}$  (except the zero vector which is useless).

If we add the eigenvalues together the result is zero which is the trace of  $A$ .

$$\lambda_1 = i : (A - \lambda_1 I)V_1 = 0$$

$$\begin{pmatrix} -i & 1 \\ -1 & -i \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \rightarrow \begin{cases} -ix + y = 0 \\ -x - iy = 0 \end{cases} \Rightarrow x = -iy$$

$$\text{Therefore the eigenvector } V_1 = \begin{pmatrix} 1 \\ i \end{pmatrix}$$

$$\lambda_2 = -i : (A - \lambda_2 I)V_2 = 0$$

$$\begin{pmatrix} i & 1 \\ -1 & i \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \rightarrow \begin{cases} ix + y = 0 \\ -x + iy = 0 \end{cases} \Rightarrow x = iy$$

$$\text{Therefore the eigenvector } V_2 = \begin{pmatrix} i \\ 1 \end{pmatrix}$$

## Exercises      Section 4.4 – Eigenvalues and Eigenvectors

1. Find the eigenvalues and eigenvectors of  $A$ ,  $A^2$ ,  $A^{-1}$ , and  $A + 4I$ :

$$A = \begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix} \quad \text{and} \quad A^2 = \begin{bmatrix} 5 & -4 \\ -4 & 5 \end{bmatrix}$$

Check the trace  $\lambda_1 + \lambda_2$  and the determinant  $\lambda_1 \lambda_2$  for  $A$  and also  $A^2$ .

2. Show directly that the given vectors are eigenvectors of the given matrix. What are the corresponding eigenvalues

$$\mathbf{v}_1 = \begin{bmatrix} 1 \\ -3 \end{bmatrix}, \quad \mathbf{v}_2 = \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \quad A = \begin{bmatrix} 1 & -2 \\ -3 & 6 \end{bmatrix}$$

3. For which real numbers  $c$  does this matrix  $A$  have

$$A = \begin{pmatrix} 2 & -c \\ -1 & 2 \end{pmatrix}$$

- a) Two real eigenvalues and eigenvectors.
- b) A repeated eigenvalue with only one eigenvector
- c) Two complex eigenvalues and eigenvectors.

4. Find the eigenvalues of  $A$ ,  $B$ ,  $AB$ , and  $BA$ :

$$A = \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix} \quad B = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \quad AB = \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix} \quad BA = \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix}$$

- a) The eigenvalues of  $AB$  (are equal to) (are not equal to) eigenvalues of  $A$  times eigenvalues of  $B$ .
- b) The eigenvalues of  $AB$  (are equal to) (are not equal to) eigenvalues of  $BA$ .

5. When  $a + b = c + d$  show that  $(1, 1)$  is an eigenvector and find both eigenvalues of

$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

6. The eigenvalues of  $A$  equal to the eigenvalues of  $A^T$ . This is because  $\det(A - \lambda I)$  equals  $\det(A^T - \lambda I)$ . That is true because \_\_\_\_\_. Show by an example that the eigenvectors of  $A$  and  $A^T$  are not the same.

7. Let  $A = \begin{bmatrix} 2 & -1 \\ 1 & 2 \end{bmatrix}$ . Compute the eigenvalues and eigenvectors of  $A$ .

8. Let  $A = \begin{bmatrix} 2 & -2 & 3 \\ 1 & 1 & 1 \\ 1 & 3 & -1 \end{bmatrix}$

- What is the characteristic polynomial for  $A$  (i.e. compute  $\det(A - \lambda I)$ )?
- Verify that 1 is an eigenvalue of  $A$ . What is a corresponding eigenvector?
- What are the other eigenvalues of  $A$ ?

9. For the following matrices:

a)  $\begin{pmatrix} 3 & 0 \\ 8 & -1 \end{pmatrix}$

b)  $\begin{pmatrix} 10 & -9 \\ 4 & -2 \end{pmatrix}$

c)  $\begin{pmatrix} 0 & 3 \\ 4 & 0 \end{pmatrix}$

d)  $\begin{bmatrix} -2 & -7 \\ 1 & 2 \end{bmatrix}$

e)  $\begin{pmatrix} 4 & 0 & 1 \\ -2 & 1 & 0 \\ -2 & 0 & 1 \end{pmatrix}$

f)  $\begin{pmatrix} 3 & 0 & -5 \\ \frac{1}{5} & -1 & 0 \\ 1 & 1 & -2 \end{pmatrix}$

g)  $\begin{pmatrix} -2 & 0 & 1 \\ -6 & -2 & 0 \\ 19 & 5 & -4 \end{pmatrix}$

h)  $\begin{pmatrix} 0 & 0 & 2 & 0 \\ 1 & 0 & 1 & 0 \\ 0 & 1 & -2 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$

i)  $\begin{pmatrix} 10 & -9 & 0 & 0 \\ 4 & -2 & 0 & 0 \\ 0 & 0 & -2 & -7 \\ 0 & 0 & 1 & 2 \end{pmatrix}$

j)  $\begin{bmatrix} -1 & 0 & 1 \\ -1 & 3 & 0 \\ -4 & 13 & -1 \end{bmatrix}$

- Find the characteristic equation
- Find the eigenvalues
- Find the eigenvectors

10. Find the eigenvalues of  $A^9$  for  $A = \begin{pmatrix} 1 & 3 & 7 & 11 \\ 0 & \frac{1}{2} & 3 & 8 \\ 0 & 0 & 0 & 4 \\ 0 & 0 & 0 & 2 \end{pmatrix}$

11. Find the eigenvalues of the matrices

$$A = \begin{pmatrix} 0.7 & 0.4 \\ 0.3 & 0.6 \end{pmatrix}, \quad A^2 = \begin{pmatrix} 0.61 & 0.52 \\ 0.39 & 0.48 \end{pmatrix}, \quad A^\infty = \begin{pmatrix} 0.57143 & 0.57143 \\ 0.42857 & 0.42857 \end{pmatrix}, \quad \text{and} \quad B = \begin{pmatrix} 0.3 & 0.6 \\ 0.7 & 0.4 \end{pmatrix}$$

12. Given the matrix  $\begin{bmatrix} -1 & -3 \\ -3 & 7 \end{bmatrix}$
- a) Find the characteristic polynomial.
  - b) Find the eigenvalues
  - c) Find the bases for its eigenspaces
  - d) Graph the eigenspaces
  - e) Verify directly that  $A\mathbf{v} = \lambda\mathbf{v}$ , for all associated eigenvectors and eigenvalues.

13. Given the matrix  $\begin{bmatrix} 5 & 0 & -4 \\ 0 & -3 & 0 \\ -4 & 0 & -1 \end{bmatrix}$
- a) Find the characteristic polynomial.
  - b) Find the eigenvalues
  - c) Find the bases for its eigenspaces
  - d) Graph the eigenspaces
  - e) Verify directly that  $A\mathbf{v} = \lambda\mathbf{v}$ , for all associated eigenvectors and eigenvalues.

14. Given:  $A = \begin{bmatrix} -1 & 7 & -1 \\ 0 & 1 & 0 \\ 0 & 15 & -2 \end{bmatrix}$ . Compute  $A^{11}$

## Section 4.5 – Diagonalization

When  $\mathbf{x}$  is an eigenvector, multiplication by  $\mathbf{A}$  is just multiplication by a single number:  $\mathbf{Ax} = \lambda\mathbf{x}$ . The matrix  $\mathbf{A}$  turns into a diagonal matrix  $\mathbf{\Lambda}$  when we use the eigenvectors property.

### Diagonalization

Suppose the  $n$  by  $n$  matrix  $\mathbf{A}$  has  $n$  linearly independent eigenvectors  $x_1, \dots, x_n$ . Put them into the column of an **eigenvector matrix**  $\mathbf{P}$ . Then  $\mathbf{P}^{-1}\mathbf{AP}$  is the eigenvalue matrix  $\mathbf{\Lambda}$ :

$$\mathbf{P}^{-1}\mathbf{AP} = \mathbf{\Lambda} = \mathbf{D} = \begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix}$$

### Example

The projection matrix  $\mathbf{A} = \begin{pmatrix} .5 & .5 \\ .5 & .5 \end{pmatrix}$  has  $\lambda = 1$  and  $0$ .

The eigenvectors are:  $(1, 1)$  &  $(-1, 1)$  that are the value of  $\mathbf{P}$ .  $\mathbf{P} = \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix}$

$$\mathbf{P}^{-1} = \frac{1}{2} \begin{pmatrix} 1 & 1 \\ -1 & 1 \end{pmatrix} = \begin{pmatrix} .5 & .5 \\ -.5 & .5 \end{pmatrix}$$

$$\begin{matrix} \begin{pmatrix} .5 & .5 \\ -.5 & .5 \end{pmatrix} & \begin{pmatrix} .5 & .5 \\ .5 & .5 \end{pmatrix} & \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} & = & \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \\ \mathbf{P}^{-1} & \mathbf{A} & \mathbf{P} & = & \mathbf{D} \end{matrix}$$

### Definition

A square matrix  $\mathbf{A}$  is called **diagonalizable** if there is an invertible matrix  $\mathbf{P}$  such that  $\mathbf{P}^{-1}\mathbf{AP}$  is diagonal; the matrix  $\mathbf{P}$  is said to **diagonalize**  $\mathbf{A}$ .

### **Theorem**

**Independent  $x$  from different  $\lambda$**  - Eigenvectors  $x_1, \dots, x_n$  that correspond to distinct (all different) eigenvalues are linearly independent. An  $n$  by  $n$  matrix that has  $n$  different eigenvalues (no repeated  $\lambda$ 's) must be diagonalizable.

### **Proof**

$$\text{Suppose } c_1 x_1 + c_2 x_2 = 0 \quad (1)$$

$$\begin{pmatrix} c_1 x_1 & c_2 x_2 \end{pmatrix} \begin{pmatrix} \lambda_1 \\ \lambda_2 \end{pmatrix} = 0$$

$$c_1 \lambda_1 x_1 + c_2 \lambda_2 x_2 = 0 \quad (2)$$

Multiply (1) by  $\lambda_2$ , that implies to

$$c_1 \lambda_2 x_1 + c_2 \lambda_2 x_2 = 0 \quad (3)$$

$$(2) - (3)$$

$$c_1 \lambda_1 x_1 + c_2 \lambda_2 x_2 - (c_1 \lambda_2 x_1 + c_2 \lambda_2 x_2) = 0$$

$$c_1 \lambda_1 x_1 + c_2 \lambda_2 x_2 - c_1 \lambda_2 x_1 - c_2 \lambda_2 x_2 = 0$$

$$c_1 \lambda_1 x_1 - c_1 \lambda_2 x_1 = 0$$

$$c_1 (\lambda_1 - \lambda_2) x_1 = 0$$

Since  $x_i \neq 0$  and  $\lambda$ 's are different  $\lambda_1 - \lambda_2 \neq 0$ , we forced  $c_1 = 0$

$$\text{Similarly; Multiply (1) by } \lambda_1, \text{ that implies to } c_1 \lambda_1 x_1 + c_2 \lambda_1 x_2 = 0 \quad (4)$$

$$(2) - (4)$$

$$c_1 \lambda_1 x_1 + c_2 \lambda_2 x_2 - c_1 \lambda_1 x_1 - c_2 \lambda_1 x_2 = 0$$

$$c_2 (\lambda_2 - \lambda_1) x_2 = 0 \Rightarrow c_2 = 0$$

Therefore,  $x_1$  and  $x_2$  must be independent.

### **Theorem**

If  $v_1, \dots, v_n$  are eigenvectors of  $A$  corresponding to distinct eigenvalues  $\lambda_1, \dots, \lambda_n$ , then

$\{v_1, v_2, \dots, v_k\}$  is linearly independent set.

### ***Theorem***

If an  $n \times n$  matrix  $A$  has  $n$  distinct eigenvalues, then the following are equivalent:

- a)  $A$  is diagonalizable
- b)  $A$  has  $n$  linearly independent eigenvectors.

### ***Example***

Given the Markov matrix  $A = \begin{bmatrix} .8 & .3 \\ .2 & .7 \end{bmatrix}$

### **Solution**

$$\begin{aligned} |A - \lambda I| &= \begin{vmatrix} .8 - \lambda & .3 \\ .2 & .7 - \lambda \end{vmatrix} = (.8 - \lambda)(.7 - \lambda) - .06 \\ &= \lambda^2 - 1.5\lambda + .56 - .06 \\ &= \lambda^2 - 1.5\lambda + .5 = 0 \end{aligned}$$

The eigenvalues are:  $\lambda_1 = 1, \lambda_2 = .5$

For  $\lambda_1 = 1$ , we have:  $(A - \lambda_1 I)V_1 = 0$

$$\begin{pmatrix} -.2 & .3 \\ .2 & -.3 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \rightarrow \begin{cases} -.2x + .3y = 0 \\ .2x - .3y = 0 \end{cases} \Rightarrow .2x = .3y$$

If  $y = .2 \Rightarrow x = .3$ , therefore the eigenvector  $V_1 = \begin{pmatrix} .3 \\ .2 \end{pmatrix}$

For  $\lambda_2 = .5$ , we have:  $(A - \lambda_2 I)V_2 = 0$

$$\begin{pmatrix} .3 & .3 \\ .2 & .2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \rightarrow \begin{cases} .3x + .3y = 0 \\ .2x + .2y = 0 \end{cases} \Rightarrow x = -y$$

If  $y = -1 \Rightarrow x = 1$ , therefore the eigenvector  $V_2 = \begin{pmatrix} 1 \\ -1 \end{pmatrix}$

$$P = \begin{pmatrix} .3 & 1 \\ .2 & -1 \end{pmatrix} \Rightarrow P^{-1} = \frac{1}{-.5} \begin{pmatrix} -1 & -1 \\ -.2 & .3 \end{pmatrix} = \begin{pmatrix} 2 & 2 \\ .4 & -.6 \end{pmatrix}$$

$$\begin{matrix} \begin{pmatrix} .3 & 1 \\ .2 & -1 \end{pmatrix} & \begin{pmatrix} 1 & 0 \\ 0 & .5 \end{pmatrix} & \begin{pmatrix} 2 & 2 \\ .4 & -.6 \end{pmatrix} & = & \begin{pmatrix} .8 & .3 \\ .2 & .7 \end{pmatrix} \\ P & D & P^{-1} & & A \end{matrix}$$



## ***Eigenvalues of $AB$ and $A + B$***

An eigenvalue of  $A$  times an eigenvalue of  $B$  usually does not give an eigenvalue of  $AB$ .

$$\lambda_A \lambda_B \neq \lambda_{AB}$$

***Commuting matrices share eigenvectors:*** Suppose  $A$  and  $B$  can be diagonalized. They share the eigenvector matrix  $P$  if and only if  $AB = BA$ .

## **Matrix Powers $A^k$**

$$A^2 = PDP^{-1}PDP^{-1} = PD^2P^{-1}$$

$$A^k = (PDP^{-1}) \cdots (PDP^{-1}) = PD^kP^{-1}$$

The eigenvector matrix for  $A^k$  is still  $S$ , and the eigenvalue matrix is  $A^k$ . The eigenvectors don't change, and the eigenvalues are taken to the  $k^{th}$  power. When  $A$  is diagonalized,  $A^k u_0$  is easy.

Here are steps (taken from Fibonacci):

1. Find the eigenvalues of  $A$  and look for  $n$  independent eigenvectors.
2. Write  $u_0$  as a combination  $c_1 x_1 + \cdots + c_n x_n$  of the eigenvectors.
3. Multiply each eigenvector  $x_i$  by  $(\lambda_i)^k$ . Then

$$u_k = A^k u_0 = c_1 (\lambda_1)^k x_1 + \cdots + c_n (\lambda_n)^k x_n$$

## ***Example***

Compute  $A^k$  where  $A = \begin{pmatrix} 1 & 1 \\ 0 & 2 \end{pmatrix}$

### **Solution**

The matrix  $A$  has  $\lambda_1 = 1 \rightarrow x_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$   $\lambda_2 = 2 \rightarrow x_2 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$

$$A^k = PD^kP^{-1} = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1^k & \\ & 2^k \end{pmatrix} \begin{pmatrix} 1 & -1 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 2^k - 1 \\ 0 & 2^k \end{pmatrix}$$

## Similar Matrices

### Definition

If  $A$  and  $B$  are square matrices, then we say that  **$B$  is similar to  $A$**  if there exists an invertible matrix  $P$  such that  $B = P^{-1}AP$  or  $A = PBP^{-1}$

✚ Similar matrices  $B$  and  $M^{-1}AM$  have the same eigenvalues. If  $x$  is an eigenvector of  $A$  then  $M^{-1}x$  is an eigenvector of  $B = M^{-1}AM$ .

### Proof

Since  $B = M^{-1}AM \Rightarrow A = MBM^{-1}$

Suppose  $Ax = \lambda x$ :

$$MBM^{-1}x = \lambda x$$

$$BM^{-1}x = \lambda M^{-1}x$$

The eigenvalue of  $B$  is the same  $\lambda$ . The eigenvector is now  $M^{-1}x$

### Example

The projection  $A = \begin{bmatrix} .5 & .5 \\ .5 & .5 \end{bmatrix}$  is similar to  $D = P^{-1}AP = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$

Choose  $M = \begin{bmatrix} 1 & 0 \\ 1 & 2 \end{bmatrix}$ ; the similar matrix  $M^{-1}AM$  is  $\begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$

Also choose  $M = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$ ; the similar matrix  $M^{-1}AM$  is  $\begin{bmatrix} .5 & -.5 \\ -.5 & .5 \end{bmatrix}$

These matrices  $M^{-1}AM$  all have the same eigenvalues 1 and 0. **Every 2 by 2 matrix with those eigenvalues is similar to  $A$ .** The eigenvectors change with  $M$ .

### Example

$A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$  is similar to every matrix  $B = \begin{bmatrix} cd & d^2 \\ -c^2 & -cd \end{bmatrix}$  except  $B = 0$ .

These matrices  $B$  all have zero determinant (like  $A$ ). They all have rank one (like  $A$ ). Their trace is  $cd - cd = 0$ . Their eigenvalues are 0 and 0 (like  $A$ ).

Choose  $M = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$  with  $ad - bc = 1$  and  $B = M^{-1}AM$

Connections between similar matrices  $A$  and  $B$ :

<i>Not Changed</i>	<i>Changed</i>
Eigenvalues	Eigenvectors
Trace and determinant	Nullspace
Rank	Column space
Number of independent eigenvectors	Row space
Jordan form	Left nullspace
	Singular values

### **Example**

Jordan matrix  $J$  has triple eigenvalues 5, 5, 5. Its only eigenvectors are multiples of  $(1, 0, 0)$ . Algebraic multiplicity 3, geometric multiplicity 1:

$$\text{If } J = \begin{bmatrix} 5 & 1 & 0 \\ 0 & 5 & 1 \\ 0 & 0 & 5 \end{bmatrix} \text{ then } J - 5I = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix} \text{ has rank 2.}$$

Every similar matrix  $B = M^{-1}JM$  has the same triple eigenvalues 5, 5, 5. Also  $B - 5I$  must have the same rank 2. Its nullspace has dimension  $3 - 2 = 1$ . So each similar matrix  $B$  also has only one independent eigenvector.

The transpose matrix  $J^T$  has the same eigenvalues 5, 5, 5, and  $J^T - 5I$  has the same rank 2.

**Jordan's theory says that  $J^T$  is similar to  $J$ .** The matrix that produces the similarity happens to be the reverse identity  $M$ :

$$J^T = M^{-1}JM \quad \text{is} \quad \begin{bmatrix} 5 & 0 & 0 \\ 1 & 5 & 0 \\ 0 & 1 & 5 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} 5 & 1 & 0 \\ 0 & 5 & 1 \\ 0 & 0 & 5 \end{bmatrix} \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}$$

There is one line of eigenvectors  $(x_1, 0, 0)$  for  $J$  and another line  $(0, 0, x_3)$  for  $J^T$ .

## Fibonacci Numbers

Every new Fibonacci number is the sum of the two previous  $F$ 's.

The *sequence* 0, 1, 1, 2, 3, 5, 8, 13, ... comes from  $F_{k+2} = F_{k+1} + F_k$

### Problem

Find the Fibonacci number  $F_{100}$

We can apply the rule one step at a time, or just use Linear algebra.

Let consider the matrix equation:  $u_{k+1} = Au_k$ . Fibonacci rule gave us a two-step rule for scalars.

Let  $u_k = \begin{pmatrix} F_{k+1} \\ F_k \end{pmatrix}$ , the rule  $\begin{matrix} F_{k+2} = F_{k+1} + F_k \\ F_{k+1} = F_{k+1} \end{matrix}$  becomes  $u_{k+1} = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} u_k$ .

Every step multiplies by  $A = \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}$ , after 100 steps we reach  $u_{100} = \begin{pmatrix} F_{101} \\ F_{100} \end{pmatrix}$

$$u_0 = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad u_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \quad u_2 = \begin{pmatrix} 2 \\ 1 \end{pmatrix} \quad u_3 = \begin{pmatrix} 3 \\ 2 \end{pmatrix} \quad \dots \quad u_{100} = \begin{pmatrix} F_{101} \\ F_{100} \end{pmatrix}$$

$$|A - \lambda I| = \begin{vmatrix} 1-\lambda & 1 \\ 1 & -\lambda \end{vmatrix} = -\lambda(1-\lambda) - 1 = \lambda^2 - \lambda - 1$$

The characteristic equation is  $\lambda^2 - \lambda - 1 = 0$  and the solutions are

$$\lambda_1 = \frac{1+\sqrt{5}}{2} \approx 1.618 \quad \text{and} \quad \lambda_2 = \frac{1-\sqrt{5}}{2} \approx -.618$$

The eigenvectors are:

$$(A - \lambda_1 I)v_1 = \begin{bmatrix} 1-\lambda_1 & 1 \\ 1 & -\lambda_1 \end{bmatrix} \begin{bmatrix} x_1 \\ y_1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \rightarrow x_1 - \lambda_1 y_1 = 0 \Rightarrow v_1 = \begin{pmatrix} \lambda_1 \\ 1 \end{pmatrix}$$

$$(A - \lambda_2 I)v_2 = \begin{bmatrix} 1-\lambda_2 & 1 \\ 1 & -\lambda_2 \end{bmatrix} \begin{bmatrix} x_2 \\ y_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \rightarrow x_2 - \lambda_2 y_2 = 0 \Rightarrow v_2 = \begin{pmatrix} \lambda_2 \\ 1 \end{pmatrix}$$

$$S = \begin{pmatrix} \lambda_1 & \lambda_2 \\ 1 & 1 \end{pmatrix}$$

The combination of these eigenvectors that give  $u_0 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$ :

$$\begin{bmatrix} 1 \\ 0 \end{bmatrix} = \frac{1}{\lambda_1 - \lambda_2} \left( \begin{bmatrix} \lambda_1 \\ 1 \end{bmatrix} - \begin{bmatrix} \lambda_2 \\ 1 \end{bmatrix} \right) = \frac{v_1 - v_2}{\lambda_1 - \lambda_2}$$

$$\begin{aligned}
u_{100} &= \frac{(\lambda_1)^{100} v_1 - (\lambda_2)^{100} v_2}{\lambda_1 - \lambda_2} \\
F_{100} &= \frac{1}{\lambda_1 - \lambda_2} \left[ (\lambda_1)^{100} - (\lambda_2)^{100} \right] \\
&= \frac{1}{\frac{1+\sqrt{5}}{2} - \frac{1-\sqrt{5}}{2}} \left[ \left( \frac{1+\sqrt{5}}{2} \right)^{100} - \left( \frac{1-\sqrt{5}}{2} \right)^{100} \right] \\
&\approx 2.54 \times 10^{20}
\end{aligned}$$

## The Jordan Form

For every  $A$ , we want to choose  $M$  so that  $M^{-1}AM$  is as nearly diagonal as possible. When  $A$  has a full set of  $n$  eigenvectors, they go into the columns of  $M$ . Then  $M = P$ . The matrix  $P^{-1}AP$  is diagonal.

If  $A$  has  $s$  independent eigenvectors, it is similar to a matrix  $J$  that has  $s$  Jordan blocks on its diagonal. There is a matrix  $M$  such that

$$M^{-1}AM = \begin{bmatrix} J_1 & & \\ & \ddots & \\ & & J_s \end{bmatrix} = J$$

Each block in  $J$  has one eigenvalue  $\lambda_i$ , one eigenvector, and 1's above the diagonal:

$$J_i = \begin{bmatrix} \lambda_i & 1 & & \\ & \ddots & \ddots & \\ & & \ddots & 1 \\ & & & \lambda_i \end{bmatrix}$$

*$A$  is similar to  $B$  if they share the same Jordan form  $J$  – not otherwise.*

## Exercises Section 4.5 – Diagonalization

1. The Lucas numbers are like Fibonacci numbers except they start with  $L_1 = 1$  and  $L_2 = 3$ . Following the rule  $L_{k+2} = L_{k+1} + L_k$ . The next Lucas numbers are 4, 7, 11, 18. Show that the Lucas number  $L_{100} = \lambda_1^{100} + \lambda_2^{100}$ .

2. Find all eigenvector matrices  $S$  that diagonalize  $A$  (rank 1) to give  $S^{-1}AS = \Lambda$ :

$$A = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 \end{bmatrix}$$

What is  $A^n$ ? Which matrices  $B$  commute with  $A$  (so that  $AB = BA$ )

3. Determine whether the matrix is diagonalizable

a)  $\begin{bmatrix} 2 & 0 \\ 1 & 2 \end{bmatrix}$

b)  $\begin{bmatrix} -3 & 2 \\ -2 & 1 \end{bmatrix}$

c)  $\begin{bmatrix} -1 & 0 & 1 \\ -1 & 3 & 0 \\ -4 & 13 & -1 \end{bmatrix}$

d)  $\begin{bmatrix} 2 & -1 & 0 & 1 \\ 0 & 2 & 1 & -1 \\ 0 & 0 & 3 & 2 \\ 0 & 0 & 0 & 3 \end{bmatrix}$

4. Find a matrix  $P$  that diagonalizes  $A$ , and compute  $P^{-1}AP$

a)  $A = \begin{bmatrix} -14 & 12 \\ -20 & 17 \end{bmatrix}$

b)  $A = \begin{bmatrix} 1 & 0 \\ 6 & -1 \end{bmatrix}$

c)  $A = \begin{bmatrix} 3 & -2 & 0 \\ -2 & 3 & 0 \\ 0 & 0 & 5 \end{bmatrix}$

5. Determine if the matrices are diagonalizable. If so, find a matrix  $P$  that diagonalizes  $A$  and determine  $P^{-1}AP$ .

a)  $A = \begin{bmatrix} 3 & 0 & -2 \\ 0 & 3 & 0 \\ 0 & 0 & 3 \end{bmatrix}$

c)  $A = \begin{pmatrix} -2 & 0 & 0 & 0 \\ 0 & -2 & 0 & 0 \\ 0 & 0 & 3 & 0 \\ 0 & 0 & 1 & 3 \end{pmatrix}$

d)  $A = \begin{bmatrix} -2 & 0 & 0 & 0 \\ 0 & -2 & 5 & -5 \\ 0 & 0 & 3 & 0 \\ 0 & 0 & 0 & 3 \end{bmatrix}$

b)  $A = \begin{bmatrix} 19 & -9 & -6 \\ 25 & -11 & -9 \\ 17 & -9 & -4 \end{bmatrix}$

6. The 4 by 4 triangular Pascal matrix  $P_L$  and its inverse (alternating diagonals) are

$$P_L = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 2 & 1 & 0 \\ 1 & 3 & 3 & 1 \end{bmatrix} \quad \text{and} \quad P_L^{-1} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ -1 & 1 & 0 & 0 \\ 1 & -2 & 1 & 0 \\ -1 & 3 & -3 & 1 \end{bmatrix}$$

Check that  $P_L$  and  $P_L^{-1}$  have the same eigenvalues. Find a diagonal matrix  $D$  with alternating signs that gives  $P_L^{-1} = D^{-1}P_L D$ , so  $P_L$  is similar to  $P_L^{-1}$ . Show that  $P_L D$  with columns of alternating signs is its own inverse.

Since  $P_L$  and  $P_L^{-1}$  are similar they have the same Jordan form  $J$ . Find  $J$  by checking the number of independent eigenvectors of  $P_L$  with  $\lambda = 1$ .

7. If  $\mathbf{x}$  is in the nullspace of  $A$  show that  $M^{-1}\mathbf{x}$  is in the nullspace of  $M^{-1}AM$ .

The nullspaces of  $A$  and  $M^{-1}AM$  have the same (vectors) (basis) (dimension)

8. These Jordan matrices have eigenvalues 0, 0, 0, 0. They have two eigenvectors (one from each block). But the block sizes don't match and they are not similar:

$$J = \left( \begin{array}{cc|cc} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ \hline 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{array} \right) \quad \text{and} \quad K = \left( \begin{array}{ccc|c} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \\ \hline 0 & 0 & 0 & 0 \end{array} \right)$$

For any matrix  $M$  compare  $JM$  with  $MK$ . If they are equal show that  $M$  is not invertible. Then  $M^{-1}JM = K$  is Impossible;  $J$  is not similar to  $K$ .

9. Prove that  $A^T$  is always similar to  $A$  ( $\lambda$ 's are the same):

a) For one Jordan block  $J_i$ , find  $M_i$  so that  $M_i^{-1}J_i M_i = J_i^T$ .

b) For any  $J$  with blocks  $J_i$ , build  $M_0$  from blocks so that  $M_0^{-1}J M_0 = J^T$ .

c) For any  $A = M J M^{-1}$ : Show that  $A^T$  is similar to  $J^T$  and so to  $J$  and so to  $A$ .

10. Why are these statements all true?

a) If  $A$  is similar to  $B$  then  $A^2$  is similar to  $B^2$ .

b)  $A^2$  and  $B^2$  can be similar when  $A$  and  $B$  are not similar.

c)  $\begin{pmatrix} 3 & 0 \\ 0 & 4 \end{pmatrix}$  is similar to  $\begin{pmatrix} 3 & 1 \\ 0 & 4 \end{pmatrix}$

d)  $\begin{pmatrix} 3 & 0 \\ 0 & 3 \end{pmatrix}$  is not similar to  $\begin{pmatrix} 3 & 1 \\ 0 & 3 \end{pmatrix}$

e) If we exchange rows 1 and 2 of  $A$ , and then exchange columns 1 and 2 the eigenvalues stay the same. In this case  $M = ?$

11. If an  $n \times n$  matrix  $A$  has all eigenvalues  $\lambda = 0$  prove that  $A^n$  is the zero matrix.

12. If  $A$  is similar to  $A^{-1}$ , must all the eigenvalues equal to 1 or  $-1$ ?

13. Show that  $A$  and  $B$  are not similar matrices

a)  $A = \begin{pmatrix} 1 & 1 \\ 1 & 2 \end{pmatrix} \quad B = \begin{pmatrix} 1 & 0 \\ 3 & -2 \end{pmatrix}$

b)  $A = \begin{pmatrix} 4 & -1 \\ 2 & 4 \end{pmatrix} \quad B = \begin{pmatrix} 4 & 1 \\ 2 & 4 \end{pmatrix}$

c)  $A = \begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & 2 \\ 0 & 0 & 1 \end{pmatrix} \quad B = \begin{pmatrix} 1 & 2 & 0 \\ \frac{1}{2} & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$

14. Prove that two similar matrices have the same determinant. Explain geometrically why this is reasonable.

15. Prove that two similar matrices have the same characteristic polynomial and thus the same eigenvalues. Explain geometrically why this is reasonable.

16. Suppose that  $A$  is a matrix. Suppose that the linear transformation associated to  $A$  has two linearly independent eigenvectors. Prove that  $A$  is similar to a diagonal matrix.

17. Prove that if  $A$  is a  $2 \times 2$  matrix that has two distinct eigenvalues, then  $A$  is similar to a diagonal matrix.

18. Suppose that the characteristic polynomial of a matrix has a double root. Is it true that the matrix has two linearly independent eigenvectors? Consider the example  $\begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}$ . Is it true that matrices with equal characteristic polynomial are necessarily similar?

19. Show that the given matrix is not diagonalizable.  $\begin{bmatrix} 1 & 1 \\ -1 & -1 \end{bmatrix}$



- 20.** Determine if the given matrix is diagonalizable. If, so, find matrices  $S$  and  $\Lambda(D)$  such that the given matrix equals  $S\Lambda S^{-1}$

a)  $\begin{bmatrix} 3 & 3 \\ 4 & 2 \end{bmatrix}$

b)  $\begin{bmatrix} 1 & 0 & 1 \\ 0 & 2 & 0 \\ -1 & 0 & -1 \end{bmatrix}$

## Section 4.6 – Orthogonal Diagonalization

### Definition

A square matrix  $A$  is called orthogonally diagonalizable if there is an orthogonal matrix  $P$  such that  $P^{-1}AP (= P^T AP)$  is diagonal; the matrix  $P$  is said to orthogonally diagonalize  $A$ .

$$P^T AP = D$$

We say that  $A$  is orthogonally diagonalizable and that  $P$  orthogonally diagonalizes  $A$ .

### Theorem

If  $A$  is an  $n \times n$  matrix, then the following are equivalent.

- a)  $A$  is orthogonally diagonalizable
- b)  $A$  has an orthonormal set of  $n$  eigenvectors.
- c)  $A$  is symmetric.

### Theorem

If  $A$  is symmetric matrix, then:

- a) The eigenvalues of  $A$  are all real numbers.
- b) Eigenvectors from different eigenspaces are orthogonal.

### Example

Find an orthogonal matrix  $P$  that diagonalizes

$$A = \begin{pmatrix} 4 & 2 & 2 \\ 2 & 4 & 2 \\ 2 & 2 & 4 \end{pmatrix}$$

### Solution

$$\begin{aligned} \det(A - \lambda I) &= \begin{vmatrix} 4-\lambda & 2 & 2 \\ 2 & 4-\lambda & 2 \\ 2 & 2 & 4-\lambda \end{vmatrix} \\ &= (4-\lambda)^3 + 8 + 8 - 4(4-\lambda) - 4(4-\lambda) - 4(4-\lambda) \\ &= 64 - 48\lambda + 12\lambda^2 - \lambda^3 + 16 - 12(4-\lambda) \\ &= 64 - 48\lambda + 12\lambda^2 - \lambda^3 + 16 - 48 + 12\lambda \\ &= -\lambda^3 + 12\lambda^2 - 36\lambda + 32 = 0 \end{aligned}$$

The eigenvalues are:  $\boxed{\lambda = 2 \quad \text{and} \quad \lambda = 8}$

For  $\lambda_1 = 2$ , we have:  $(A - \lambda_1 I)V_1 = 0$

$$\begin{pmatrix} 2 & 2 & 2 \\ 2 & 2 & 2 \\ 2 & 2 & 2 \end{pmatrix} \begin{pmatrix} x_1 \\ y_1 \\ z_1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \rightarrow \begin{cases} 2x_1 + 2y_1 + 2z_1 = 0 \\ 2x_1 + 2y_1 + 2z_1 = 0 \\ 2x_1 + 2y_1 + 2z_1 = 0 \end{cases}$$

If  $z_1 = 0$  and  $y_1 = 1 \Rightarrow x_1 = -1$ , therefore the eigenvector  $V_1 = \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix}$

For  $\lambda_2 = 2$ , we have:  $(A - \lambda_2 I)V_2 = 0$

$$\begin{pmatrix} 2 & 2 & 2 \\ 2 & 2 & 2 \\ 2 & 2 & 2 \end{pmatrix} \begin{pmatrix} x_2 \\ y_2 \\ z_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \rightarrow \begin{cases} 2x_2 + 2y_2 + 2z_2 = 0 \\ 2x_2 + 2y_2 + 2z_2 = 0 \\ 2x_2 + 2y_2 + 2z_2 = 0 \end{cases}$$

If  $y_2 = 0$  and  $z_2 = 1 \Rightarrow x_2 = -1$ , therefore the eigenvector  $V_2 = \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix}$

For  $\lambda_3 = 8$ , we have:  $(A - \lambda_3 I)V_3 = 0$

$$\begin{pmatrix} -4 & 2 & 2 \\ 2 & -4 & 2 \\ 2 & 2 & -4 \end{pmatrix} \begin{pmatrix} x_3 \\ y_3 \\ z_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \rightarrow \begin{cases} -4x_3 + 2y_3 + 2z_3 = 0 \\ 2x_3 - 4y_3 + 2z_3 = 0 \\ 2x_3 + 2y_3 - 4z_3 = 0 \end{cases}$$

$$\begin{pmatrix} -4 & 2 & 2 \\ 2 & -4 & 2 \\ 2 & 2 & -4 \end{pmatrix} \xrightarrow{rref} \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & -1 \\ 0 & 0 & 0 \end{pmatrix} \begin{cases} x_3 = z_3 \\ y_3 = z_3 \end{cases}$$

If  $z_3 = 1 \Rightarrow y_3 = 1, x_3 = 1$ , therefore the eigenvector  $V_3 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$

$$u_1 = \frac{v_1}{\|v_1\|} = \frac{(-1, 1, 0)}{\sqrt{(-1)^2 + 1^2 + 0}} = \frac{(-1, 1, 0)}{\sqrt{2}} = \left( -\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}, 0 \right)$$

$$\begin{aligned} w_2 &= v_2 - (v_2 \cdot u_1)u_1 = (-1, 0, 1) - \left[ (-1, 0, 1) \cdot \left( -\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}, 0 \right) \right] \left( -\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}, 0 \right) \\ &= (-1, 0, 1) - \frac{1}{\sqrt{2}} \left( -\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}, 0 \right) \\ &= \left( -\frac{1}{2}, -\frac{1}{2}, 1 \right) \end{aligned}$$

$$\begin{aligned}
u_2 &= \frac{w_2}{\|w_2\|} = \frac{\left(-\frac{1}{2}, -\frac{1}{2}, 1\right)}{\sqrt{\frac{1}{4} + \frac{1}{4} + 1}} = \frac{\left(-\frac{1}{2}, -\frac{1}{2}, 1\right)}{\sqrt{\frac{6}{4}}} \\
&= \frac{2}{\sqrt{6}} \left(-\frac{1}{2}, -\frac{1}{2}, 1\right) \\
&= \left(-\frac{1}{\sqrt{6}}, -\frac{1}{\sqrt{6}}, \frac{2}{\sqrt{6}}\right)
\end{aligned}$$

$$\begin{aligned}
w_3 &= v_3 - (v_3 \cdot u_1)u_1 - (v_3 \cdot u_2)u_2 \\
&= (1, 1, 1) - (0) \left(-\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}, 0\right) - (0) \left(-\frac{1}{\sqrt{6}}, -\frac{1}{\sqrt{6}}, \frac{2}{\sqrt{6}}\right) \\
&= (1, 1, 1)
\end{aligned}$$

$$\begin{aligned}
u_3 &= \frac{w_3}{\|w_3\|} = \frac{1}{\sqrt{1^2 + 1^2 + 1^2}} (1, 1, 1) \\
&= \left(\frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}\right)
\end{aligned}$$

$$P = \begin{bmatrix} -\frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{6}} & \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{6}} & \frac{1}{\sqrt{3}} \\ 0 & \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{3}} \end{bmatrix}$$

$$P^T A P = \begin{pmatrix} -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \\ -\frac{1}{\sqrt{6}} & -\frac{1}{\sqrt{6}} & \frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} \end{pmatrix} \begin{pmatrix} 4 & 2 & 2 \\ 2 & 4 & 2 \\ 2 & 2 & 4 \end{pmatrix} \begin{pmatrix} -\frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{6}} & \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{6}} & \frac{1}{\sqrt{3}} \\ 0 & \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{3}} \end{pmatrix} = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 8 \end{bmatrix}$$

## Spectral Decomposition

The spectral decomposition of  $A$  is:

$$A = \lambda_1 \mathbf{u}_1 \mathbf{u}_1^T + \lambda_2 \mathbf{u}_2 \mathbf{u}_2^T + \cdots + \lambda_n \mathbf{u}_n \mathbf{u}_n^T$$

### Example

The matrix  $A = \begin{pmatrix} 1 & 2 \\ 2 & -2 \end{pmatrix}$

### Solution

$$\begin{aligned} \det(A - \lambda I) &= \begin{vmatrix} 1-\lambda & 2 \\ 2 & -2-\lambda \end{vmatrix} = (1-\lambda)(-2-\lambda) - 4 \\ &= \lambda^2 + \lambda - 6 = 0 \end{aligned}$$

The eigenvalues are:  $\boxed{\lambda = -3 \text{ and } \lambda = 2}$

The corresponding eigenvectors are:  $V_1 = \begin{pmatrix} 1 \\ -2 \end{pmatrix}$  and  $V_2 = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$

$$u_1 = \frac{v_1}{\|v_1\|} = \frac{(1, -2)}{\sqrt{1^2 + (-2)^2}} = \left( \frac{1}{\sqrt{5}}, -\frac{2}{\sqrt{5}} \right)$$

$$\begin{aligned} w_2 &= v_2 - (v_2 \cdot u_1) u_1 = (2, 1) - \left[ (2, 1) \cdot \left( \frac{1}{\sqrt{5}}, -\frac{2}{\sqrt{5}} \right) \right] \left( \frac{1}{\sqrt{5}}, -\frac{2}{\sqrt{5}} \right) \\ &= (2, 1) - (0) \left( \frac{1}{\sqrt{5}}, -\frac{2}{\sqrt{5}} \right) \\ &= (2, 1) \end{aligned}$$

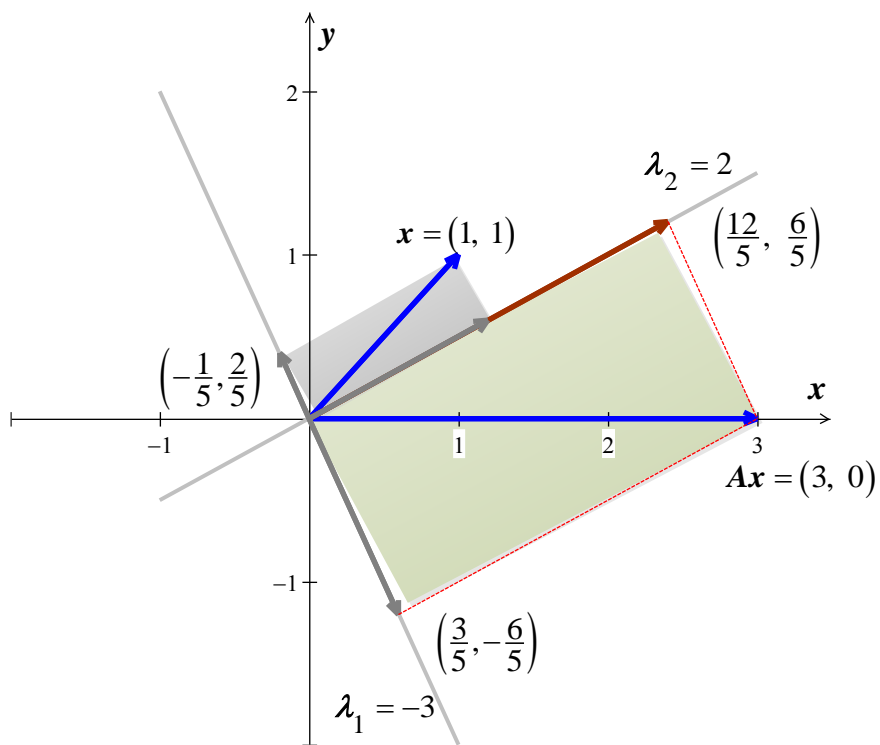
$$\begin{aligned} u_2 &= \frac{w_2}{\|w_2\|} = \frac{(2, 1)}{\sqrt{5}} \\ &= \left( \frac{2}{\sqrt{5}}, \frac{1}{\sqrt{5}} \right) \end{aligned}$$

$$\begin{aligned} \begin{pmatrix} 1 & 2 \\ 2 & -2 \end{pmatrix} &= \lambda_1 \mathbf{u}_1 \mathbf{u}_1^T + \lambda_2 \mathbf{u}_2 \mathbf{u}_2^T \\ &= -3 \begin{pmatrix} \frac{1}{\sqrt{5}} \\ -\frac{2}{\sqrt{5}} \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{5}} & -\frac{2}{\sqrt{5}} \end{pmatrix} + 2 \begin{pmatrix} \frac{2}{\sqrt{5}} \\ \frac{1}{\sqrt{5}} \end{pmatrix} \begin{pmatrix} \frac{2}{\sqrt{5}} & \frac{1}{\sqrt{5}} \end{pmatrix} \end{aligned}$$

$$= -3 \begin{pmatrix} \frac{1}{5} & -\frac{2}{5} \\ -\frac{2}{5} & \frac{4}{5} \end{pmatrix} + 2 \begin{pmatrix} \frac{4}{5} & \frac{2}{5} \\ \frac{2}{5} & \frac{1}{5} \end{pmatrix}$$

The spectral decomposition about the image of the vector  $\mathbf{x} = (1, 1)$

$$\begin{aligned} A\mathbf{x} &= \begin{pmatrix} 1 & 2 \\ 2 & -2 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = -3 \begin{pmatrix} \frac{1}{5} & -\frac{2}{5} \\ -\frac{2}{5} & \frac{4}{5} \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} + 2 \begin{pmatrix} \frac{4}{5} & \frac{2}{5} \\ \frac{2}{5} & \frac{1}{5} \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} \\ &= -3 \begin{pmatrix} -\frac{1}{5} \\ \frac{2}{5} \end{pmatrix} + 2 \begin{pmatrix} \frac{6}{5} \\ \frac{3}{5} \end{pmatrix} \\ &= \begin{pmatrix} \frac{3}{5} \\ -\frac{6}{5} \end{pmatrix} + \begin{pmatrix} \frac{12}{5} \\ \frac{6}{5} \end{pmatrix} \\ &= \begin{pmatrix} 3 \\ 0 \end{pmatrix} \end{aligned}$$



### Example

Consider a 2 by 2 symmetric matrix  $A = \begin{pmatrix} a & b \\ b & c \end{pmatrix}$

### Solution

The eigenvalues are:

$$\begin{aligned}\det(A - \lambda I) &= \begin{vmatrix} a - \lambda & b \\ b & c - \lambda \end{vmatrix} = (a - \lambda)(c - \lambda) - b^2 \\ &= \lambda^2 - (a + c)\lambda + ac - b^2 = 0\end{aligned}$$

$$\lambda = \frac{(a + c) \pm \sqrt{(a + c)^2 - 4(ac - b^2)}}{2} \quad \therefore (a + c)^2 - 4(ac - b^2) > 0$$

The eigenvectors are:

$$\begin{aligned}(A - \lambda_1 I)V_1 &= \begin{pmatrix} a - \lambda_1 & b \\ b & c - \lambda_1 \end{pmatrix} \begin{pmatrix} x_1 \\ y_1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \rightarrow \begin{cases} (a - \lambda_1)x_1 + by_1 = 0 \\ bx_1 + (c - \lambda_1)y_1 = 0 \end{cases} \Rightarrow \text{if } x_1 = b \quad y_1 = a - \lambda_1 \\ V_1 &= \begin{pmatrix} b \\ \lambda_1 - a \end{pmatrix}\end{aligned}$$

$$\begin{aligned}(A - \lambda_2 I)V_2 &= \begin{pmatrix} a - \lambda_2 & b \\ b & c - \lambda_2 \end{pmatrix} \begin{pmatrix} x_2 \\ y_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \rightarrow \begin{cases} (a - \lambda_2)x_2 + by_2 = 0 \\ bx_2 + (c - \lambda_2)y_2 = 0 \end{cases} \Rightarrow \text{if } y_2 = b \quad x_2 = c - \lambda_1 \\ V_2 &= \begin{pmatrix} \lambda_2 - a \\ c \end{pmatrix}\end{aligned}$$

$$\begin{aligned}\lambda_1 + \lambda_2 &= \frac{(a + c) - \sqrt{(a + c)^2 - 4(ac - b^2)} + (a + c) + \sqrt{(a + c)^2 - 4(ac - b^2)}}{2} \\ &= \frac{2(a + c)}{2} \\ &= a + c\end{aligned}$$

$$\begin{aligned}V_1 \cdot V_2 &= b(\lambda_1 - a) + b(\lambda_2 - c) \\ &= b(\lambda_1 + \lambda_2 - a - c) \\ &= b(a + c - a - c) \\ &= 0\end{aligned}$$

Therefore, these eigenvectors are perpendicular.

### ***Theorem***

**Orthogonal Eigenvectors:** Eigenvectors of a real symmetric matrix (when they correspond to different  $\lambda$ 's) are always perpendicular.

### ***Proof***

Suppose  $Ax = \lambda_1 x$ ,  $Ay = \lambda_2 y$  and  $A = A^T$ .

The dot products of the first equation with  $y$  and the second with  $x$ :

$$(\lambda_1 x)^T y = (Ax)^T y = x^T A^T y = x^T Ay = x^T \lambda_2 y$$

$$\Rightarrow x^T \lambda_1 y = x^T \lambda_2 y$$

Since  $\lambda_1 \neq \lambda_2$ , this proves that  $x^T y = 0$ .

The eigenvector  $x$  (for  $\lambda_1$ ) is perpendicular to the eigenvector  $y$  (for  $\lambda_2$ )

### ***Example***

Find the  $\lambda$ 's and  $v$ 's for this symmetric matrix with trace zero:  $A = \begin{bmatrix} -3 & 4 \\ 4 & 3 \end{bmatrix}$

### **Solution**

$$\begin{aligned} \det(A - \lambda I) &= \begin{vmatrix} -3 - \lambda & 4 \\ 4 & 3 - \lambda \end{vmatrix} \\ &= (-3 - \lambda)(3 - \lambda) - 16 \\ &= -9 + \lambda^2 - 16 \\ &= \lambda^2 - 25 \end{aligned}$$

The eigenvalues are:  $\lambda_1 = 5$   $\lambda_2 = -5$

The eigenvectors are:

$$(A - \lambda_1 I)v_1 = \begin{bmatrix} -8 & 4 \\ 4 & -2 \end{bmatrix} \begin{bmatrix} x_1 \\ y_1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \rightarrow \begin{cases} -8x_1 + 4y_1 = 0 \\ 4x_1 - 2y_1 = 0 \end{cases} \Rightarrow 2x_1 = y_1 \rightarrow v_1 = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$

$$(A - \lambda_2 I)v_2 = \begin{bmatrix} 2 & 4 \\ 4 & 8 \end{bmatrix} \begin{bmatrix} x_2 \\ y_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \rightarrow \begin{cases} 2x_2 + 4y_2 = 0 \\ 4x_2 + 8y_2 = 0 \end{cases} \Rightarrow x_2 = -2y_2 \rightarrow v_2 = \begin{pmatrix} -2 \\ 1 \end{pmatrix}$$

$$v_1 \cdot v_2 = (1)(-2) + (2)(1) = -2 + 2 = 0$$

Thus, the eigenvectors are perpendicular.

The unit vector of the eigenvectors by dividing by their length  $\sqrt{2^2 + 1^2} = \sqrt{5}$



The  $v_1$  and  $v_2$  are the columns of  $Q$ .

$$Q = \frac{\begin{bmatrix} 1 & -2 \\ 2 & 1 \end{bmatrix}}{\sqrt{5}} \rightarrow Q^{-1} = Q^T = \frac{\begin{bmatrix} 1 & 2 \\ -2 & 1 \end{bmatrix}}{\sqrt{5}}$$

$$A = QDQ^T = \frac{\begin{bmatrix} 1 & -2 \\ 2 & 1 \end{bmatrix}}{\sqrt{5}} \begin{bmatrix} 5 & 0 \\ 0 & -5 \end{bmatrix} \frac{\begin{bmatrix} 1 & 2 \\ -2 & 1 \end{bmatrix}}{\sqrt{5}}$$

➤ Every symmetric matrix  $A$  has a complete set of orthogonal eigenvectors:

$$A = PDP^{-1} \Rightarrow A = QDQ^T$$

## Complex Eigenvalues of Real Matrices

For real matrices, complex  $\lambda$ 's and  $x$ 's come in “conjugate pairs”

$$\text{if } Ax = \lambda x \text{ then } A\bar{x} = \bar{\lambda}\bar{x}$$

### Example

$$\text{Given } A = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

### Solution

The eigenvalues of  $A$ :

$$\begin{aligned} \det(A - \lambda I) &= \begin{vmatrix} \cos \theta - \lambda & -\sin \theta \\ \sin \theta & \cos \theta - \lambda \end{vmatrix} \\ &= (\cos \theta - \lambda)(\cos \theta - \lambda) + \sin^2 \theta \\ &= \cos^2 \theta - 2\lambda \cos \theta + \lambda^2 + \sin^2 \theta \\ &= \lambda^2 - 2\lambda \cos \theta + 1 \end{aligned}$$

$$\begin{aligned} \lambda &= \frac{2\cos \theta \pm \sqrt{4\cos^2 \theta - 4}}{2} \\ &= \frac{2\cos \theta \pm \sqrt{4(\cos^2 \theta - 1)}}{2} \\ &= \frac{2\cos \theta \pm 2\sqrt{-\sin^2 \theta}}{2} \\ &= \cos \theta \pm i \sin \theta \end{aligned}$$

$$\cos^2 \theta + \sin^2 \theta = 1 \rightarrow \cos^2 \theta - 1 = -\sin^2 \theta$$

The eigenvalues are conjugate to each other.

The eigenvectors:

$$(A - \lambda_1 I)v_1 = \begin{bmatrix} \cos \theta - (\cos \theta + i \sin \theta) & -\sin \theta \\ \sin \theta & \cos \theta - (\cos \theta + i \sin \theta) \end{bmatrix} \begin{bmatrix} x_1 \\ y_1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} -i \sin \theta & -\sin \theta \\ \sin \theta & -i \sin \theta \end{bmatrix} \begin{bmatrix} x_1 \\ y_1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \Rightarrow \begin{cases} -i \sin \theta x_1 - \sin \theta y_1 = 0 \\ \sin \theta x_1 - i \sin \theta y_1 = 0 \end{cases}$$

$$\Rightarrow x_1 - iy_1 = 0 \Rightarrow x_1 = iy_1 \rightarrow v_1 = \begin{pmatrix} 1 \\ -i \end{pmatrix}$$

$$\rightarrow v_2 = \begin{pmatrix} 1 \\ i \end{pmatrix}$$

The vector  $x = v_1$

$$Ax = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} 1 \\ -i \end{bmatrix} = (\cos \theta + i \sin \theta) \begin{bmatrix} 1 \\ -i \end{bmatrix}$$

$$A\bar{x} = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} 1 \\ i \end{bmatrix} = (\cos \theta - i \sin \theta) \begin{bmatrix} 1 \\ i \end{bmatrix}$$

$$|\lambda| = \sqrt{\cos^2 \theta + \sin^2 \theta} = 1$$

This fact holds for the eigenvalues of every orthogonal matrix.

### ***Theorem – Equivalent Statements***

If  $A$  is an  $n \times n$  matrix, then the following statements are equivalent.

- a)  $A$  is invertible
- b)  $A\mathbf{x} = \mathbf{0}$  has only the trivial solution
- c) The reduced row echelon form of  $A$  is  $I_n$
- d)  $A$  is expressible as a product of elementary matrices
- e)  $A\mathbf{x} = \mathbf{b}$  is consistent for every  $n \times 1$  matrix  $\mathbf{b}$
- f)  $A\mathbf{x} = \mathbf{b}$  has exactly one solution for every  $n \times 1$  matrix  $\mathbf{b}$
- g)  $\det(A) \neq 0$
- h) The column vectors of  $A$  are linearly independent
- i) The row vectors of  $A$  are linearly independent
- j) The column vectors of  $A$  span  $\mathbf{R}^n$
- k) The row vectors of  $A$  span  $\mathbf{R}^n$
- l) The column vectors of  $A$  form a basis for  $\mathbf{R}^n$
- m) The row vectors of  $A$  form a basis for  $\mathbf{R}^n$
- n)  $A$  has a rank  $n$ .
- o)  $A$  has nullity 0.
- p) The orthogonal complement of the null space of  $A$  is  $\mathbf{R}^n$
- q) The orthogonal complement of the row space of  $A$  is  $\{0\}$
- r) The range of  $T_A$  is  $\mathbf{R}^n$
- s)  $T_A$  is one-to-one.
- t)  $\lambda = 0$  is not an eigenvalue of  $A$ .
- u)  $A^T A$  is invertible,

## Exercises      Section 4.6 – Orthogonal Diagonalization

1. Find a matrix  $P$  that orthogonally diagonalizes  $A$ , and determine  $P^{-1}AP$

$$a) \quad A = \begin{pmatrix} 3 & 1 \\ 1 & 3 \end{pmatrix}$$

$$d) \quad A = \begin{pmatrix} 1 & 1 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

$$f) \quad A = \begin{pmatrix} -7 & 24 & 0 & 0 \\ 24 & 7 & 0 & 0 \\ 0 & 0 & -7 & 24 \\ 0 & 0 & 24 & 7 \end{pmatrix}$$

$$b) \quad A = \begin{pmatrix} 6 & 2\sqrt{3} \\ 2\sqrt{3} & 7 \end{pmatrix}$$

$$e) \quad A = \begin{pmatrix} 2 & -1 & -1 \\ -1 & 2 & -1 \\ -1 & -1 & 2 \end{pmatrix}$$

$$g) \quad A = \begin{bmatrix} 3 & 1 & 0 & 0 \\ 1 & 3 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$c) \quad A = \begin{pmatrix} -2 & 0 & -36 \\ 0 & -3 & 0 \\ -36 & 0 & -23 \end{pmatrix}$$

2. Find the eigenvalues of  $A$  and  $B$  and check the Orthogonality of their first two eigenvectors. Graph these eigenvectors to see the discrete sines and cosines:

$$A = \begin{bmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 2 \end{bmatrix} \quad B = \begin{bmatrix} 1 & -1 & 0 & 0 \\ -1 & 2 & -1 & 0 \\ 0 & -1 & 2 & -1 \\ 0 & 0 & -1 & 1 \end{bmatrix}$$

The  $-1, 2, -1$  pattern in both matrices is a “second derivative. Then  $Ax = \lambda x$  and  $Bx = \lambda x$  are like  $\frac{d^2x}{dt^2} = \lambda x$ . This has eigenvectors  $x = \sin kt$  and  $x = \cos kt$  that are the bases for Fourier series. The matrices lead to “discrete sines” and “discrete cosines” that are the bases for the discrete Fourier Transform. This DFT is absolutely central to all areas of digital signal processing.

3. Suppose  $Ax = \lambda x$  and  $Ay = 0y$  and  $\lambda \neq 0$ . Then  $y$  is in the nullspace and  $x$  is in the column space. They are perpendicular because \_\_\_\_\_. (why are these subspaces orthogonal?) If the second eigenvalue is a nonzero number  $\beta$ , apply this argument to  $A - \beta I$ . The eigenvalue moves to zero and the eigenvectors stay the same – so they are perpendicular.
4. True or false. Give a reason or a counterexample.
- A matrix with real eigenvalues and eigenvectors is symmetric.
  - A matrix with real eigenvalues and orthogonal eigenvectors is symmetric.
  - The inverse of a symmetric matrix is symmetric
  - The eigenvector matrix  $S$  of a symmetric matrix is symmetric.
  - A complex symmetric matrix has real eigenvalues.
  - If  $A$  is symmetric, then  $e^{iA}$  is symmetric.
  - If  $A$  is Hermitian, then  $e^{iA}$  is Hermitian.

5. Find a symmetric matrix  $\begin{pmatrix} 1 & b \\ b & 1 \end{pmatrix}$  that has a negative eigenvalue.
- How do you know it must have a negative pivot?
  - How do you know it can't have two negative eigenvalues?

6. Which of these classes of matrices do  $A$  and  $B$  belong to: Invertible, orthogonal, projection, permutation, diagonalizable, Markov?

$$A = \begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix} \quad B = \frac{1}{3} \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix}$$

Which of these factorizations are possible for  $A$  and  $B$ : LU, QR,  $ADP^{-1}$ ,  $QDQ^T$ ?

7. Construct a 3 by 3 matrix  $A$  with no zero entries whose columns are mutually perpendicular. Compute  $A^T A$ . Why is it a diagonal matrix?

8. Assuming that  $b \neq 0$ , find a matrix that orthogonally diagonalizes  $\begin{bmatrix} a & b \\ b & a \end{bmatrix}$