# Practical Linear Algebra: A GEOMETRY TOOLBOX

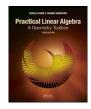
Third edition

Chapter 4: Changing Shapes: Linear Maps in 2D

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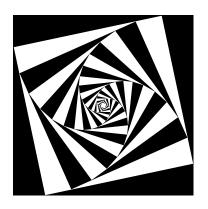
### Outline

- Introduction to Linear Maps in 2D
- Skew Target Boxes
- The Matrix Form
- 4 Linear Spaces
- Scalings
- 6 Reflections
- Rotations
- 8 Shears
- Projections
- 10 Areas and Linear Maps: Determinants
- Composing Linear Maps
- More on Matrix Multiplication
- Matrix Arithmetic Rules
- **14** WYSK



### Introduction to Linear Maps in 2D

2D linear maps (rotation and scaling) applied repeatedly to a square



Geometry has two parts

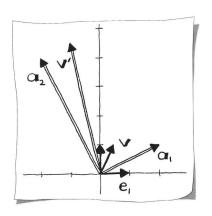
- description of the objects
- how these objects can be changed (transformed)

Transformations also called maps May be described using the tools of matrix operations: linear maps

Matrices first introduced by H. Grassmann in 1844
Became basis of linear algebra

### **Skew Target Boxes**

Revisit unit square to a rectangular target box mapping Examine part of mapping that is a linear map



Unit square defined by  $\mathbf{e}_1$  and  $\mathbf{e}_2$ Vector  $\mathbf{v}$  in  $[\mathbf{e}_1,\mathbf{e}_2]$ -system defined as

$$\mathbf{v}=v_1\mathbf{e}_1+v_2\mathbf{e}_2$$

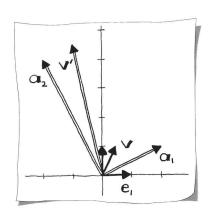
 $\boldsymbol{v}$  is now mapped to a vector  $\boldsymbol{v}'$  by

$$\mathbf{v}'=v_1\mathbf{a}_1+v_2\mathbf{a}_2$$

Duplicates the  $[\mathbf{e}_1, \mathbf{e}_2]$ -geometry in the  $[\mathbf{a}_1, \mathbf{a}_2]$ -system

### **Skew Target Boxes**

#### **Example:** linear combination



 $\left[a_{1},a_{2}\right]\text{-coordinate system: origin and}$ 

$$\mathbf{a}_1 = \begin{bmatrix} 2 \\ 1 \end{bmatrix} \qquad \mathbf{a}_2 = \begin{bmatrix} -2 \\ 4 \end{bmatrix}$$

Given 
$$\mathbf{v} = \begin{bmatrix} 1/2 \\ 1 \end{bmatrix}$$
 in  $[\mathbf{e}_1, \mathbf{e}_2]$ -system

$$\mathbf{v}' = \frac{1}{2} \times \begin{bmatrix} 2 \\ 1 \end{bmatrix} + 1 \times \begin{bmatrix} -2 \\ 4 \end{bmatrix} = \begin{bmatrix} -1 \\ 9/2 \end{bmatrix}$$

 $\mathbf{v}'$  has components  $\begin{bmatrix} 1/2\\1 \end{bmatrix}$  with respect to  $[\mathbf{a}_1, \mathbf{a}_2]$ -system

 $\mathbf{v}'$  has components  $\begin{bmatrix} -1\\ 9/2 \end{bmatrix}$  with respect to  $[\mathbf{e}_1, \mathbf{e}_2]$ -system

Components of a subscripted vector written with a double subscript

$$\mathbf{a}_1 = \begin{bmatrix} a_{1,1} \\ a_{2,1} \end{bmatrix}$$

The vector component index precedes the vector subscript

Components for  $\mathbf{v}'$  in  $[\mathbf{e}_1,\mathbf{e}_2]$ -system expressed as

$$\begin{bmatrix} -1\\ 9/2 \end{bmatrix} = \frac{1}{2} \times \begin{bmatrix} 2\\ 1 \end{bmatrix} + 1 \times \begin{bmatrix} -2\\ 4 \end{bmatrix}$$

Using matrix notation:

$$\begin{bmatrix} -1\\9/2 \end{bmatrix} = \begin{bmatrix} 2 & -2\\1 & 4 \end{bmatrix} \begin{bmatrix} 1/2\\1 \end{bmatrix}$$

 $2 \times 2$  matrix: 2 rows and 2 columns

— Columns are vectors  $\mathbf{a}_1$  and  $\mathbf{a}_2$ 



In general:

$$\mathbf{v}' = \begin{bmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = A\mathbf{v}$$

A is a  $2 \times 2$  matrix Elements  $a_{1,1}$  and  $a_{2,2}$  form the diagonal

- $\boldsymbol{v}'$  is the image of  $\boldsymbol{v}$
- ${f v}$  is the pre-image of  ${f v}'$
- $\mathbf{v}'$  is in the range of the map
- v is in the domain of the map

Product Av has two components:

$$A\mathbf{v} = \begin{bmatrix} v_1 \mathbf{a}_1 + v_2 \mathbf{a}_2 \end{bmatrix} = \begin{bmatrix} v_1 \mathbf{a}_{1,1} + v_2 \mathbf{a}_{1,2} \\ v_1 \mathbf{a}_{2,1} + v_2 \mathbf{a}_{2,2} \end{bmatrix}$$

Each component obtained as a dot product between the corresponding row of the matrix and  ${\bf v}$ 

#### Example:

$$\begin{bmatrix} 0 & -1 \\ 2 & 4 \end{bmatrix} \begin{bmatrix} -1 \\ 4 \end{bmatrix} = \begin{bmatrix} -4 \\ 14 \end{bmatrix}$$

Column space of A: all  $\mathbf{v}'$  formed as linear combination of the columns of A

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 $[\mathbf{e}_1,\mathbf{e}_2]$ -system can be interpreted as a matrix with columns  $\mathbf{e}_1$  and  $\mathbf{e}_2$ :

$$[\mathbf{e}_1,\mathbf{e}_2] \equiv egin{bmatrix} 1 & 0 \ 0 & 1 \end{bmatrix}$$

Called the  $2 \times 2$  identity matrix

Neat way to write matrix-times-vector:

$$\begin{array}{c|cccc} & 2 & \\ & 1/2 & \\ \hline 2 & -2 & 3 & \\ 1 & 4 & 4 & \end{array}$$

Interior dimensions (both 2) must be identical Outer dimensions (2 and 1) indicate the resulting vector or matrix size

9 / 51

Matrix addition:

$$\begin{bmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} \end{bmatrix} + \begin{bmatrix} b_{1,1} & b_{1,2} \\ b_{2,1} & b_{2,2} \end{bmatrix} = \begin{bmatrix} a_{1,1} + b_{1,1} & a_{1,2} + b_{1,2} \\ a_{2,1} + b_{2,1} & a_{2,2} + b_{2,2} \end{bmatrix}$$

Matrices must be of the same dimensions

Distributive law

$$A\mathbf{v} + B\mathbf{v} = (A+B)\mathbf{v}$$

Transpose matrix denoted by  $A^{\mathrm{T}}$ 

Formed by interchanging the rows and columns of A:

$$A = \begin{bmatrix} 1 & -2 \\ 3 & 5 \end{bmatrix} \quad \text{then} \quad A^{\mathrm{T}} = \begin{bmatrix} 1 & 3 \\ -2 & 5 \end{bmatrix}$$

May think of a vector  $\mathbf{v}$  as a matrix:

$$\mathbf{v} = \begin{bmatrix} -1 \\ 4 \end{bmatrix}$$
 then  $\mathbf{v}^{\mathrm{T}} = \begin{bmatrix} -1 & 4 \end{bmatrix}$ 

Identities:

$$[A+B]^{\mathrm{T}}=A^{\mathrm{T}}+B^{\mathrm{T}}$$
  $A^{\mathrm{T}^{\mathrm{T}}}=A$  and  $[cA]^{\mathrm{T}}=cA^{\mathrm{T}}$ 

Symmetric matrix:  $A = A^{T}$ 

### Example:

$$\begin{bmatrix} 5 & 8 \\ 8 & 1 \end{bmatrix}$$

No restrictions on diagonal elements

All other elements equal to element about the diagonal with reversed indices

For a 2 × 2 matrix:  $a_{2,1} = a_{1,2}$ 

 $2 \times 2$  zero matrix:

Matrix rank: number of linearly independent column (row) vectors

For  $2 \times 2$  matrix columns define an  $[\mathbf{a}_1, \mathbf{a}_2]$ -system Full rank=2:  $\mathbf{a}_1$  and  $\mathbf{a}_2$  are linearly independent

Rank deficient: matrix that does not have full rank If  $\mathbf{a}_1$  and  $\mathbf{a}_2$  are linearly dependent then matrix has rank 1 Also called a *singular* matrix

Only matrix with rank zero is zero matrix Rank of A and  $A^{T}$  are equal.

2D linear maps act on vectors in 2D linear spaces Also known as 2D vector spaces

Standard operations in a linear space are addition and scalar multiplication of vectors

$$\mathbf{v}' = v_1 \mathbf{a}_1 + v_2 \mathbf{a}_2$$
 — linearity property

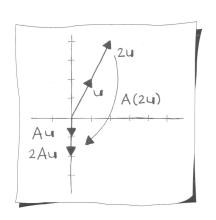
Linear maps – matrices – characterized by preservation of linear combinations:

$$A(a\mathbf{u} + b\mathbf{v}) = aA\mathbf{u} + bA\mathbf{v}.$$

Let's break this statement down into the two basic elements: scalar multiplication and addition

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#### Matrices preserve scalings



### Example:

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

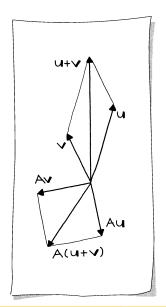
$$\mathbf{u} = \begin{bmatrix} 1 \\ 2 \end{bmatrix} \quad \mathbf{v} = \begin{bmatrix} -1 \\ 4 \end{bmatrix}$$

$$A(c\mathbf{u}) = cA\mathbf{u}$$

Let 
$$c=2$$

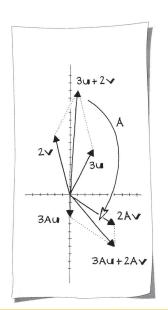
$$\begin{bmatrix} -1 & 1/2 \\ 0 & -1/2 \end{bmatrix} \left( 2 \times \begin{bmatrix} 1 \\ 2 \end{bmatrix} \right)$$

$$=2\times\begin{bmatrix}-1 & 1/2\\0 & -1/2\end{bmatrix}\begin{bmatrix}1\\2\end{bmatrix}=\begin{bmatrix}0\\-2\end{bmatrix}$$



Matrices preserve sums (distributive law):

$$A(\mathbf{u} + \mathbf{v}) = A\mathbf{u} + A\mathbf{v}$$



Matrices preserve linear combinations  $A(3\mathbf{u} + 2\mathbf{v})$ 

$$= \begin{bmatrix} -1 & 1/2 \\ 0 & -1/2 \end{bmatrix} \begin{pmatrix} 3 \begin{bmatrix} 1 \\ 2 \end{bmatrix} + 2 \begin{bmatrix} -1 \\ 4 \end{bmatrix} \end{pmatrix}$$

$$= \begin{bmatrix} 6 \\ -7 \end{bmatrix} 3A\mathbf{u} + 2A\mathbf{v}$$

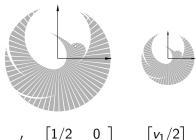
$$= 3 \begin{bmatrix} -1 & 1/2 \\ 0 & -1/2 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

$$+ 2 \begin{bmatrix} -1 & 1/2 \\ 0 & -1/2 \end{bmatrix} \begin{bmatrix} -1 \\ 4 \end{bmatrix}$$

$$= \begin{bmatrix} 6 \\ -7 \end{bmatrix}$$

# **Scalings**

### Uniform scaling:



$$\mathbf{v}' = \begin{bmatrix} 1/2 & 0 \\ 0 & 1/2 \end{bmatrix} \mathbf{v} = \begin{bmatrix} v_1/2 \\ v_2/2 \end{bmatrix}$$

# Scalings

General scaling: 
$$\mathbf{v}' = \begin{bmatrix} s_{1,1} & 0 \\ 0 & s_{2,2} \end{bmatrix} \mathbf{v}$$
 Example:  $\mathbf{v}' = \begin{bmatrix} 1/2 & 0 \\ 0 & 2 \end{bmatrix} \mathbf{v}$ 

Example: 
$$\mathbf{v}' = \begin{bmatrix} 1/2 & 0 \\ 0 & 2 \end{bmatrix} \mathbf{v}$$





Scaling affects the area of the object:

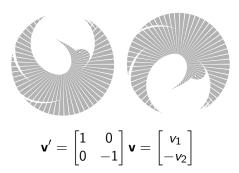
- Scale by  $s_{1,1}$  in  $e_1$ -direction, then area changes by a factor  $s_{1,1}$
- Similarly for  $s_{2,2}$  and  $\mathbf{e}_2$ -direction

Total effect: factor of  $s_{1,1}s_{2,2}$ 

Action of matrix: action ellipse

### Reflections

### Special scaling:



 ${f v}$  reflected about  ${f e}_1$ -axis or the line  $x_1=0$ 

Reflection maps each vector about a line through the origin

### Reflections

Reflection about line  $x_1 = x_2$ :





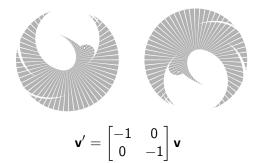
$$\mathbf{v}' = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \mathbf{v} = \begin{bmatrix} v_2 \\ v_1 \end{bmatrix}$$

Reflections change the sign of the area due to a change in orientation

- Rotate  $\mathbf{e}_1$  into  $\mathbf{e}_2$ : move in a counterclockwise
- Rotate  $a_1$  into  $a_2$ : move in a clockwise

### Reflections

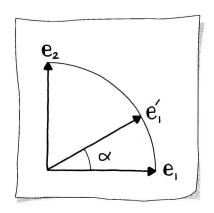
#### Reflection?



Check  ${\bf a}_1$  rotate to  ${\bf a}_2$  orientation: counterclockwise — same as  ${\bf e}_1, {\bf e}_2$  orientation This is a  $180^\circ$  rotation

Action ellipse: circle

### Rotations



Rotate  $\boldsymbol{e}_1$  and  $\boldsymbol{e}_2$  around the origin to

$$\mathbf{e}_1' = \begin{bmatrix} \cos \alpha \\ \sin \alpha \end{bmatrix} \quad \text{and} \quad \mathbf{e}_2' = \begin{bmatrix} -\sin \alpha \\ \cos \alpha \end{bmatrix}$$

These are the column vectors of the rotation matrix

$$R = \begin{bmatrix} \cos \alpha & -\sin \alpha \\ \sin \alpha & \cos \alpha \end{bmatrix}$$

### Rotations



Rotation matrix for  $\alpha = 45^{\circ}$ :

$$R = \begin{bmatrix} \sqrt{2}/2 & -\sqrt{2}/2 \\ \sqrt{2}/2 & \sqrt{2}/2 \end{bmatrix}$$

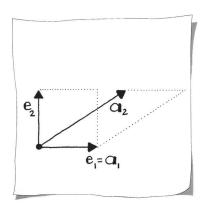
Rotations: special class of transformations called rigid body motions

Action ellipse: circle

Rotations do not change areas



#### Map a rectangle to a parallelogram



#### Example:

$$\mathbf{v} = egin{bmatrix} 0 \ 1 \end{bmatrix} \quad \longrightarrow \quad \mathbf{v}' = egin{bmatrix} d_1 \ 1 \end{bmatrix}$$

A shear in matrix form:

$$\begin{bmatrix} d_1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & d_1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

Application: generate italic fonts from standard ones.

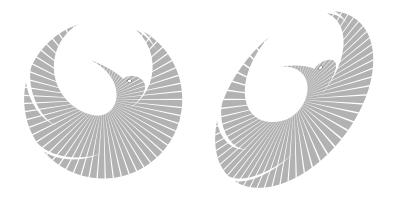
Shear along the  $e_1$ -axis applied to an arbitrary vector:





$$\mathbf{v}' = \begin{bmatrix} 1 & d_1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} v_1 + v_2 d_1 \\ v_2 \end{bmatrix}$$

#### Shear along the $e_2$ -axis:



$$\mathbf{v}' = \begin{bmatrix} 1 & 0 \\ d_2 & 1 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} v_1 \\ v_1 d_2 + v_2 \end{bmatrix}$$

What is the shear that achieves

$$\mathbf{v} = egin{bmatrix} v_1 \ v_2 \end{bmatrix} &\longrightarrow & \mathbf{v}' = egin{bmatrix} v_1 \ 0 \end{bmatrix}$$
?

A shear parallel to the  $e_2$ -axis:

$$\mathbf{v}' = egin{bmatrix} v_1 \ 0 \end{bmatrix} = egin{bmatrix} 1 & 0 \ -v_2/v_1 & 1 \end{bmatrix} egin{bmatrix} v_1 \ v_2 \end{bmatrix}$$

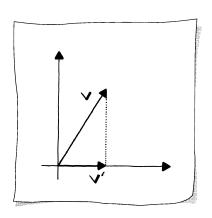
Shears do not change areas

(See rectangle to parallelogram sketch: both have the same base and the same height)

28 / 51

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Parallel projections: all vectors are projected in a parallel direction 2D: all vectors are projected onto a line



### Example:

$$\begin{bmatrix} 3 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 3 \\ 1 \end{bmatrix}$$

Orthogonal projection: angle of incidence with the line is 90° Otherwise: oblique projection

Perspective projection: projection direction is not constant — not a linear map

Orthogonal projections important for best approximation Oblique projections important to applications in fields such as computer graphics and architecture



Main property of a projection: *reduces dimensionality* Action ellipse: straight line segment which is covered twice

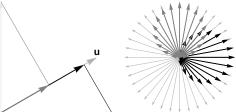
#### Construction:

Choose unit vector **u** to define a line onto which to project Projections of  $e_1$  and  $e_2$  are column vectors of matrix A

$$\mathbf{a}_1 = \frac{\mathbf{u} \cdot \mathbf{e}_1}{\|\mathbf{u}\|^2} \mathbf{u} = u_1 \mathbf{u}$$

$$\mathbf{a}_2 = \frac{\mathbf{u} \cdot \mathbf{e}_2}{\|\mathbf{u}\|^2} \mathbf{u} = u_2 \mathbf{u}$$

$$A = \begin{bmatrix} u_1 \mathbf{u} & u_2 \mathbf{u} \end{bmatrix} = \mathbf{u} \mathbf{u}^T$$



**Example:**  $\mathbf{u} = [\cos 30^{\circ} \sin 30^{\circ}]^{\mathrm{T}}$ 

Projection matrix 
$$A = \begin{bmatrix} u_1 \mathbf{u} & u_2 \mathbf{u} \end{bmatrix} = \mathbf{u} \mathbf{u}^T$$

Columns of A linearly dependent  $\Rightarrow$  rank one Map reduces dimensionality  $\Rightarrow$  area after map is zero

Projection matrix is idempotent: A = AA

Geometrically: once a vector projected onto a line, application of same projection leaves result unchanged

**Example:** 
$$\mathbf{u} = \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}$$
 then  $A = \begin{bmatrix} 0.5 & 0.5 \\ 0.5 & 0.5 \end{bmatrix}$ 

Action of projection matrix on a vector **x**:

$$A\mathbf{x} = \mathbf{u}\mathbf{u}^{\mathrm{T}}\mathbf{x} = (\mathbf{u} \cdot \mathbf{x})\mathbf{u}$$

Same result as orthogonal projections in Chapter 2

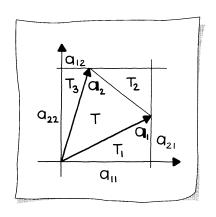
Let  $\mathbf{y}$  be projection of  $\mathbf{x}$  onto  $\mathbf{u}$  then  $\mathbf{x} = \mathbf{y} + \mathbf{y}^{\perp}$ 

$$A\mathbf{x} = \mathbf{u}\mathbf{u}^{\mathrm{T}}\mathbf{y} + \mathbf{u}\mathbf{u}^{\mathrm{T}}\mathbf{y}^{\perp}$$

Since  $\boldsymbol{u}^{\mathrm{T}}\boldsymbol{y}=\|\boldsymbol{y}\|$  and  $\boldsymbol{u}^{\mathrm{T}}\boldsymbol{y}^{\perp}=0$ 

$$A\mathbf{x} = \|\mathbf{y}\|\mathbf{u}$$

# Areas and Linear Maps: Determinants

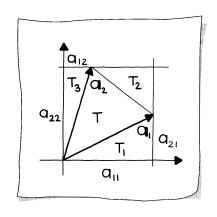


2D linear map takes  $[e_1, e_2]$  to  $[a_1, a_2]$  How does linear map affect area?

$$\begin{split} & \mathrm{area}(\boldsymbol{e}_1,\boldsymbol{e}_2) = 1 \\ & \text{(Square spanned by } [\boldsymbol{e}_1,\boldsymbol{e}_2]) \end{split}$$

P= area of parallelogram spanned by  $[\mathbf{a}_1,\mathbf{a}_2]$  P=2T

# Areas and Linear Maps: Determinants



Area T of triangle formed by  $\mathbf{a}_1$  and  $\mathbf{a}_2$ :

$$T = a_{1,1}a_{2,2} - T_1 - T_2 - T_3$$

Observe that

$$T_1 = \frac{1}{2} a_{1,1} a_{2,1}$$

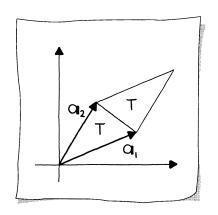
$$T_2 = \frac{1}{2} (a_{1,1} - a_{1,2}) (a_{2,2} - a_{2,1})$$

$$T_3 = \frac{1}{2} a_{1,2} a_{2,2}$$

$$T = \frac{1}{2} a_{1,1} a_{2,2} - \frac{1}{2} a_{1,2} a_{2,1}$$

$$P = a_{1,1} a_{2,2} - a_{1,2} a_{2,1}$$

# Areas and Linear Maps: Determinants



P: (signed) area of the parallelogram spanned by  $A = [\mathbf{a}_1, \mathbf{a}_2]$ This is the determinant of A— Notation: det A or |A|

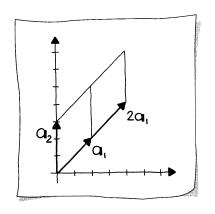
$$|A| = \begin{vmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} \end{vmatrix} = a_{1,1}a_{2,2} - a_{1,2}a_{2,1}$$

#### Determinant characterizes a linear map:

- ullet If |A|=1 then linear map does not change areas
- ullet If  $0 \le |A| < 1$  then linear map shrinks areas
- If |A| = 0 then matrix is rank deficient
- If |A| > 1 then linear map expands areas
- If |A| < 0 then linear map changes the orientation of objects

$$\begin{vmatrix} 1 & 5 \\ 0 & 1 \end{vmatrix} = (1)(1) - (5)(0) = 1$$
$$\begin{vmatrix} 1 & 0 \\ 0 & -1 \end{vmatrix} = (1)(-1) - (0)(0) = -1$$
$$\begin{vmatrix} 0.5 & 0.5 \\ 0.5 & 0.5 \end{vmatrix} = (0.5)(0.5) - (0.5)(0.5) = 0$$

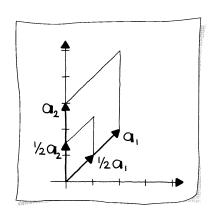
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$$|c\mathbf{a}_1,\mathbf{a}_2|=c|\mathbf{a}_1,\mathbf{a}_2|=c|A|$$

If one column of A scaled by c then A's determinant scaled by c

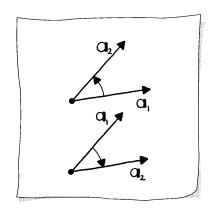
Sketch: 
$$c = 2$$



$$|c\mathbf{a}_1, c\mathbf{a}_2| = c^2 |\mathbf{a}_1, \mathbf{a}_2| = c^2 |A|$$

If both columns of A scaled by c then A's determinant scaled by  $c^2$ 

Sketch: 
$$c = 1/2$$



Two 2D vectors whose determinant is positive: right-handed Standard example: **e**<sub>1</sub> and **e**<sub>2</sub>

Two 2D vectors whose determinant is negative are called left-handed

Area sign change when columns interchanged:  $|\mathbf{a}_1, \mathbf{a}_2| = -|\mathbf{a}_2, \mathbf{a}_1|$  Verified using the definition of a determinant:

$$|\mathbf{a}_2, \mathbf{a}_1| = a_{1,2}a_{2,1} - a_{2,2}a_{1,1}$$

Matrix product used to compose linear maps:

$$\mathbf{v}'' = B\mathbf{v}' = B(A\mathbf{v}) = BA\mathbf{v} = C\mathbf{v}$$

$$C = \begin{bmatrix} b_{1,1} & b_{1,2} \\ b_{2,1} & b_{2,2} \end{bmatrix} \begin{bmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} \end{bmatrix} = \begin{bmatrix} b_{1,1}a_{1,1} + b_{1,2}a_{2,1} & b_{1,1}a_{1,2} + b_{1,2}a_{2,2} \\ b_{2,1}a_{1,1} + b_{2,2}a_{2,1} & b_{2,1}a_{1,2} + b_{2,2}a_{2,2} \end{bmatrix}$$

 $\mathbf{v}' = A\mathbf{v}$ 

Element  $c_{i,j}$  computed as dot product of B's  $i^{\mathrm{th}}$  row and A's  $j^{\mathrm{th}}$  column

41 / 51

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#### **Example:**

$$\mathbf{v} = \begin{bmatrix} 2 \\ -1 \end{bmatrix}, \quad A = \begin{bmatrix} -1 & 2 \\ 0 & 3 \end{bmatrix}, \quad B = \begin{bmatrix} 0 & -2 \\ -3 & 1 \end{bmatrix}$$
$$\mathbf{v}'A\mathbf{v} == \begin{bmatrix} -1 & 2 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 2 \\ -1 \end{bmatrix} = \begin{bmatrix} -4 \\ -3 \end{bmatrix}$$
$$\mathbf{v}'' = B\mathbf{v}' = \begin{bmatrix} 0 & -2 \\ -3 & 1 \end{bmatrix} \begin{bmatrix} -4 \\ -3 \end{bmatrix} = \begin{bmatrix} 6 \\ 9 \end{bmatrix}$$

Compute  $\mathbf{v}''$  using the matrix product BA:

$$C = BA = \begin{bmatrix} 0 & -2 \\ -3 & 1 \end{bmatrix} \begin{bmatrix} -1 & 2 \\ 0 & 3 \end{bmatrix} = \begin{bmatrix} 0 & -6 \\ 3 & -3 \end{bmatrix}$$

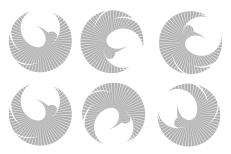
Verify that  $\mathbf{v}'' = C\mathbf{v}$ 

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Neat way to arrange two matrices when forming their product

$$\begin{array}{c|cccc}
 & -1 & 2 \\
\hline
 0 & 3 \\
\hline
 0 & -2 \\
 -3 & 1 & 3 \\
\hline
 & -1 & 2 \\
 0 & 3 \\
\hline
 0 & -2 & 0 & -6 \\
 -3 & 1 & 3 & -3 \\
\end{array}$$

Linear map composition is order dependent



Top: rotate by  $-120^{\circ}$ , then reflect about the (rotated)  ${\bf e}_1$ -axis Bottom: reflect, then rotate

Matrix products differs significantly from products of real numbers: Matrix products are not *commutative* 

$$AB \neq BA$$

Some maps to commute – example: 2D rotations

Farin & Hansford Practical Linear Algebra 44

Rank of a composite map:

$$rank(AB) \le min\{rank(A), rank(B)\}$$

Matrix multiplication does not increase rank

Special composition: idempotent matrix A = AA or  $A = A^2$ Thus  $A\mathbf{v} = AA\mathbf{v}$ 

### More on Matrix Multiplication

Vectors as matrices:  $\mathbf{u}^{\mathrm{T}}\mathbf{v} = \mathbf{u} \cdot \mathbf{v}$ 

**Example:** Let 
$$\mathbf{u} = \begin{bmatrix} 3 \\ 4 \end{bmatrix}$$
 and  $\mathbf{v} = \begin{bmatrix} -3 \\ 6 \end{bmatrix}$ 

$$\begin{aligned} (\mathbf{u}^{\mathrm{T}}\mathbf{v})^{\mathrm{T}} &= \mathbf{v}^{\mathrm{T}}\mathbf{u} \\ [\mathbf{u}^{\mathrm{T}}\mathbf{v}]^{\mathrm{T}} &= (\begin{bmatrix} 3 & 4 \end{bmatrix} \begin{bmatrix} -3 \\ 6 \end{bmatrix})^{\mathrm{T}} = \begin{bmatrix} 15 \end{bmatrix}^{\mathrm{T}} = 15 \\ \mathbf{v}^{\mathrm{T}}\mathbf{u} &= \begin{bmatrix} -3 & 6 \end{bmatrix} \begin{bmatrix} 3 \\ 4 \end{bmatrix} = \begin{bmatrix} 15 \end{bmatrix} = 15 \end{aligned}$$

## More on Matrix Multiplication

$$(AB)^{\mathrm{T}} = B^{\mathrm{T}}A^{\mathrm{T}}$$

$$(AB)^{\mathrm{T}} = \begin{pmatrix} \begin{bmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} \end{bmatrix} \begin{bmatrix} \mathbf{b}_{1,1} & b_{1,2} \\ \mathbf{b}_{2,1} & b_{2,2} \end{bmatrix} \end{pmatrix}^{\mathrm{T}} = \begin{bmatrix} c_{1,1} & \mathbf{c}_{1,2} \\ c_{2,1} & c_{2,2} \end{bmatrix}$$
$$B^{\mathrm{T}}A^{\mathrm{T}} = \begin{bmatrix} \mathbf{b}_{1,1}^{\mathrm{T}} & \mathbf{b}_{1,2}^{\mathrm{T}} \\ b_{2,1}^{\mathrm{T}} & b_{2,2}^{\mathrm{T}} \end{bmatrix}) \begin{bmatrix} a_{1,1}^{\mathrm{T}} & \mathbf{a}_{1,2}^{\mathrm{T}} \\ a_{2,1}^{\mathrm{T}} & \mathbf{a}_{2,2}^{\mathrm{T}} \end{bmatrix} = \begin{bmatrix} c_{1,1} & \mathbf{c}_{1,2} \\ c_{2,1} & c_{2,2} \end{bmatrix}$$

Since  $b_{i,j} = b_{j,i}^{\mathrm{T}}$  identical dot product calculated to form  $c_{1,2}$ 

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#### More on Matrix Multiplication

Determinant of a product matrix

$$|AB| = |A||B|$$

B scales objects by |B| and A scales objects by |A|Composition of the maps scales by the product of the individual scales **Example:** two scalings

$$A = \begin{bmatrix} 1/2 & 0 \\ 0 & 1/2 \end{bmatrix} \qquad B = \begin{bmatrix} 4 & 0 \\ 0 & 4 \end{bmatrix}$$

|A| = 1/4 and  $|B| = 16 \Rightarrow A$  scales down, and B scales up Effect of B's scaling greater than A's

$$AB = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$$
 scales up:  $|AB| = |A||B| = 4$ 

Exponents for matrices:
$$A^r = \underbrace{A \cdot \ldots \cdot A}$$

Some rules: 
$$A^{r+s} = A^r A^s$$
  $A^{rs} = (A^r)^s$   $A^0 = I$ 

$$A^{r \text{ times}} = (A^r)^s$$

#### Matrix Arithmetic Rules

Matrix sizes must be compatible for operations to be performed

- matrix addition: matrices to have the same dimensions
- matrix multiplication: "inside" dimensions to be equal Let A's dimensions be  $m \times r$  and B's are  $r \times n$  Product C = AB is permissible since inside dimension r is shared Resulting matrix C dimension  $m \times n$

Commutative Law for Addition: A+B=B+AAssociative Law for Addition: A+(B+C)=(A+B)+CNo Commutative Law for Multiplication:  $AB \neq BA$ Associative Law for Multiplication: A(BC)=(AB)CDistributive Law: A(B+C)=AB+ACDistributive Law: (B+C)A=BA+CA

#### Matrix Arithmetic Rules

Rules involving scalars:

$$a(B+C)=aB+aC$$

$$(a+b)C = aC + bC$$

$$(ab)C = a(bC)$$

$$a(BC) = (aB)C = B(aC)$$

Rules involving the transpose:

$$(A+B)^{\mathrm{T}} = A^{\mathrm{T}} + B^{\mathrm{T}}$$

$$(bA)^{\mathrm{T}} = bA^{\mathrm{T}}$$

$$(AB)^{\mathrm{T}} = B^{\mathrm{T}}A^{\mathrm{T}}$$

$$A^{\mathrm{T}^{\mathrm{T}}} = A$$

#### **WYSK**

- linear combination
- matrix form
- pre-image and image
- domain and range
- column space
- identity matrix
- matrix addition
- distributive law
- transpose matrix

- symmetric matrix
- rank of a matrix
- rank deficient
- singular matrix
- linear space or vector space
- subspace
- linearity property
- scalings

- action ellipse
- reflections
- rotations
- rigid body motions
- shears
- projections
- parallel projection
- oblique projection
- dyadic matrix
- idempotent ar map

- determinant
- signed area

matrix

- multiplication
- composite map
- noncommutative property of matrix multiplication
- transpose of a product or m of matrices
- rules of matrix arithmetic