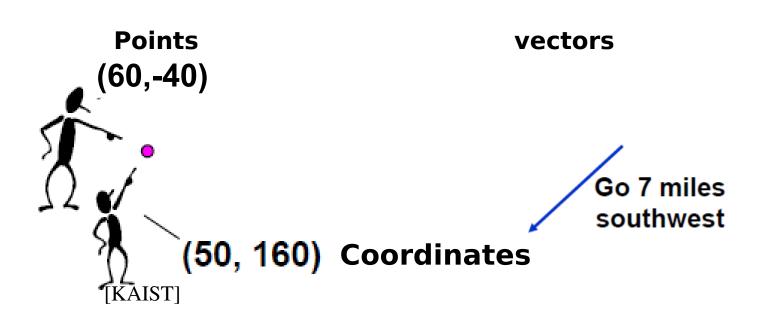
Vector Algebra Transformations

Lecture 4

Geometry

- A part of mathematics concerned with questions of size, shape, and relative positions of figures
- Coordinates are used to represent points and vectors
 - a naming scheme
 - The same point can be described by different coordinates



Points

- Conceptually, points and vectors are very different
 - A point p is a place in space
 - A vector V
 describes a direction independent
 of position (pay attentions notations)



Geometry

- Linear Algebra
 - Scalar
 - Vector
 - Linear independence
 - Linear transformations
- Frames (=Coordinates)
 - Points
 - Vectors
- Affine transformations
 - Translation
 - Rotation

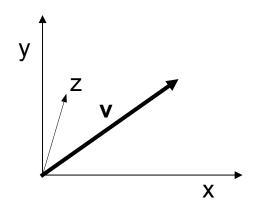
Vector Spaces

- A vector (or linear) space V over a scalar field S consists of a set on which the following two operators are defined and the following conditions hold:
- Two operators for vectors:
 - Vector-vector addition $\forall \vec{u}, \vec{v} \in V \quad \vec{u} + \vec{v} \in V$
 - Scalar-vector multiplication

$$\forall \vec{u} \in V, \forall a \in S \quad a\vec{u} \in V$$

Notation:

Notation:
• Vector
$$\vec{v} = \begin{bmatrix} a \\ b \end{bmatrix} = [a \ b \ c]^T$$





Vector Spaces

- Vector-vector addition
 - Commutes and associates

$$U + V = V + U \quad U + (V + W) = (U + V) + W$$

 An additive identity and an additive inverse for each vector

$$U + \overline{0} = U \quad U + (-U) = \overline{0}$$

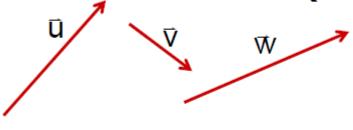
Scalar-vector multiplication distributes

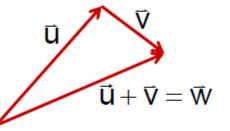
$$(a+b)\vec{u} = a\vec{u} + b\vec{u}$$
 $a(\vec{u} + \vec{v}) = a\vec{u} + a\vec{v}$

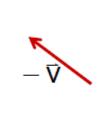


Example Vector Spaces

Geometric vectors (directed segments)







2ü

N-tuples of scalars

$$\bar{u} = (1,3,7)^t$$
 $\bar{u} + \bar{v} = (3,5,4)^t = \bar{w}$
 $\bar{v} = (2,2,-3)^t$ $2\bar{u} = (2,6,14)^t$
 $\bar{w} = (3,5,4)^t$ $-\bar{v} = (-2,-2,3)^t$

t denotes "transpose"

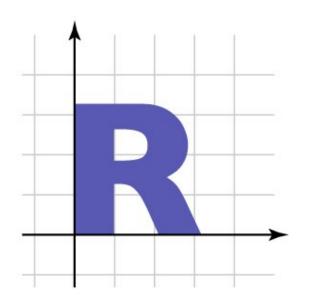
We can use N-tuples to represent vectors

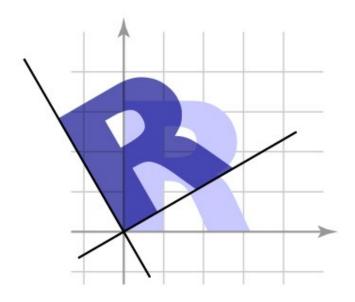


Transforming geometry

 Move a subset of the space (in 2D case, plane) using a mapping from the plane to itself

$$S \to \{T(\mathbf{v}) \mid \mathbf{v} \in S\}$$





Linear transformations

 One way to define a transformation is by matrix multiplication:

$$T(\mathbf{v}) = M\mathbf{v}$$

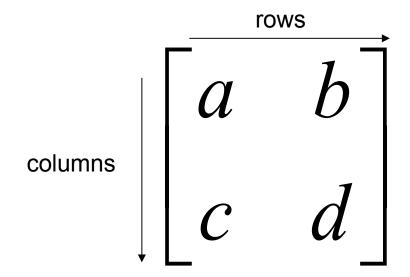
 Such transformations are *linear*, which is to say:

$$T(a\mathbf{u} + \mathbf{v}) = aT(\mathbf{u}) + T(\mathbf{v})$$

(and in fact all linear transformations can be written this way)

What is a Matrix?

A matrix is a set of elements, organized into rows and columns



Matrix operating on vectors

- Matrix is like a <u>function</u> that <u>transforms the vectors on a plane</u>
- Matrix operating on a general point => transforms x and y components
- System of linear equations: matrix is just the bunch of coefficients

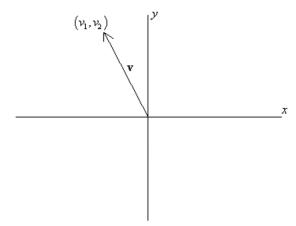
•
$$x' = ax + by$$

• $y' = cx + dy$

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} x' \\ y' \end{bmatrix}$$

Bases & Orthonormal Bases

• Basis vectors (or axes): frame of reference



Any point in a space is a *linear combination* of the basis vectors Usually, orthonormal matrices are used for defining coordinate frames *Ortho-Normal: orthogonal + normal*

[Orthogonal: dot product is zero

Normal: magnitude is one]

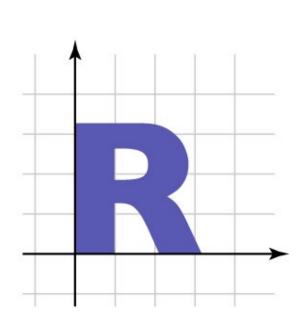
$$x = \begin{bmatrix} 1 & 0 & 0 \end{bmatrix}^T \qquad x \cdot y = 0$$

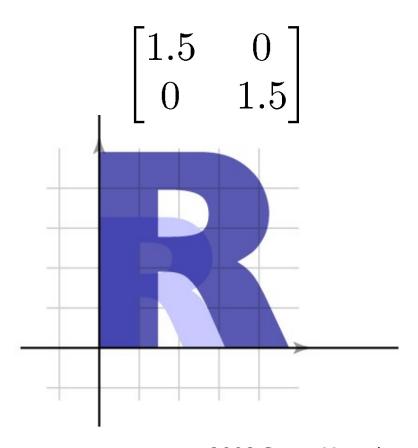
$$y = \begin{bmatrix} 0 & 1 & 0 \end{bmatrix}^T \qquad x \cdot z = 0$$
$$z = \begin{bmatrix} 0 & 0 & 1 \end{bmatrix}^T \qquad y \cdot z = 0$$

Geometry of 2D linear trans.

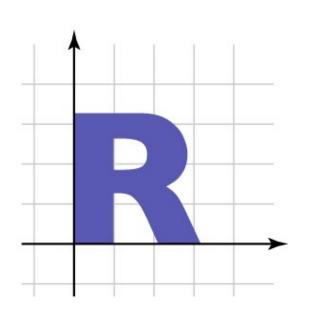
- 2x2 matrices have simple geometric interpretations
 - uniform scale
 - non-uniform scale
 - rotation
 - shear
 - reflection

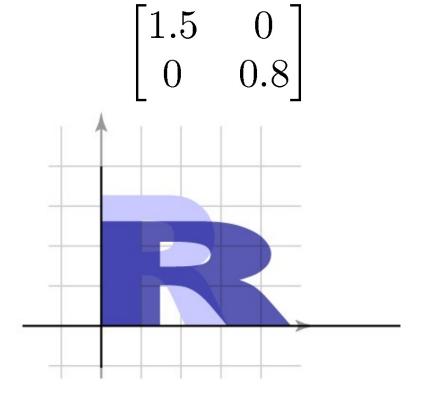
• Uniform scale
$$\begin{bmatrix} s & 0 \\ 0 & s \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} sx \\ sy \end{bmatrix}$$



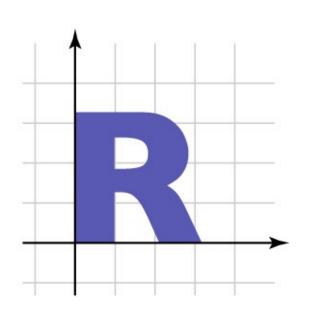


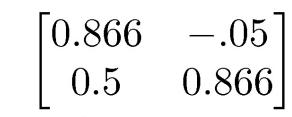
• Nonuniform scale
$$\begin{bmatrix} s_x & 0 \\ 0 & s_y \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} s_x x \\ s_y y \end{bmatrix}$$

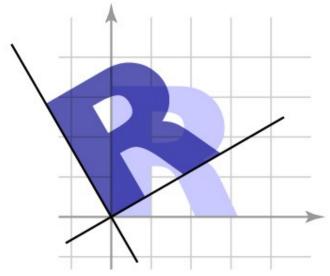




• Rotation $\begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} x \cos \theta - y \sin \theta \\ x \sin \theta + y \cos \theta \end{bmatrix}$

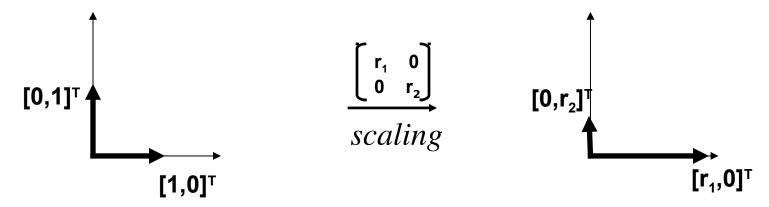


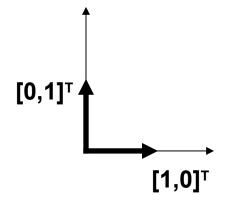


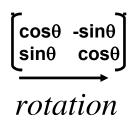


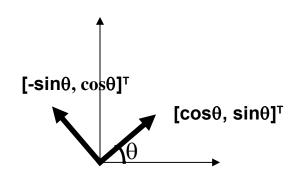
Matrices: Scaling, Rotation, Identity

- Scaling without rotation => "<u>diagonal</u> matrix"
- Rotation without stretching => "orthonormal matrix" O
- <u>Identity</u> ("do nothing") matrix = unit scaling, no rotation

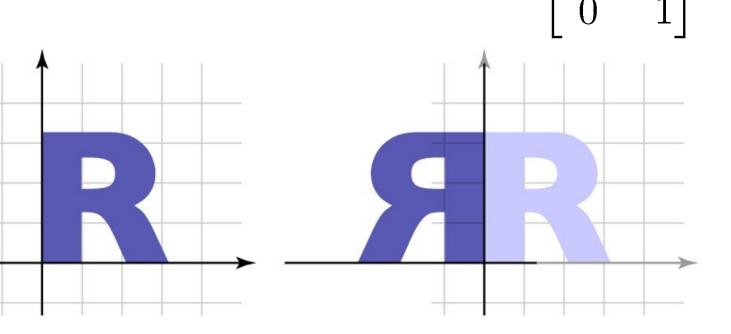




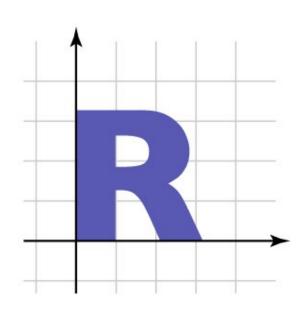


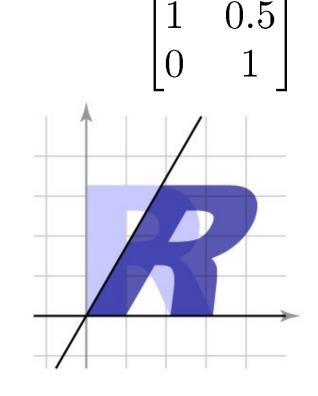


- Reflection
 - can consider it a special case of nonuniform scale



• Shear
$$\begin{bmatrix} 1 & a \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} x + ay \\ y \end{bmatrix}$$





Composing transformations

Want to move an object, then move it some more

-
$$\mathbf{p} \to T(\mathbf{p}) \to S(T(\mathbf{p})) = (S \circ T)(\mathbf{p})$$

- We need to represent S o T ("S compose T")
 - and would like to use the same representation as for S and T

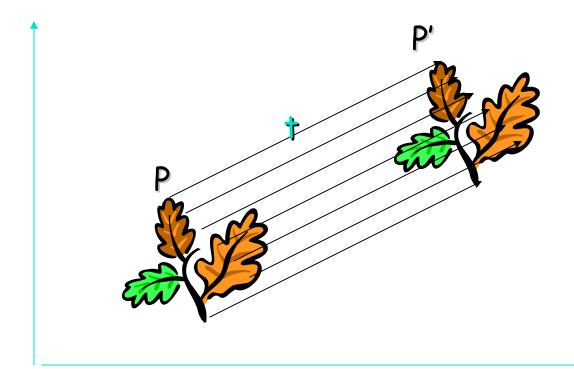
Composing transformations

 Composing linear transformations is straightforward

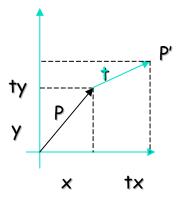
$$T(\mathbf{p}) = M_T \mathbf{p}; S(\mathbf{p}) = M_S \mathbf{p}$$
$$(S \circ T)(\mathbf{p}) = M_S M_T \mathbf{p}$$

- Transforming first by M_T then by M_S is the same as transforming by M_SM_T
 - only sometimes commutative
 - e.g. 2D rotations & uniform scales
 - e.g. non-uniform scales w/o rotation
- Note M_SM_T , or S o T, is T first, then S

Translation



$$\mathbf{P'} = (x + t_x, y + t_y) = \mathbf{P} + \mathbf{t}$$



Composing translations

Composing translations is easy

$$\mathbf{p} \to T(\mathbf{p}) \to S(T(\mathbf{p})) = (S \circ T)(\mathbf{p})$$

- Translation by \mathbf{u}_T then by \mathbf{u}_S is translation by $\mathbf{u}_T + \mathbf{u}_S$
- commutative!

$$T(\mathbf{p}) = \mathbf{p} + \mathbf{u}_T; S(\mathbf{p}) = \mathbf{p} + \mathbf{u}_S$$

 $(S \circ T)(\mathbf{p}) = \mathbf{p} + (\mathbf{u}_T + \mathbf{u}_S)$

Is translation linear transformation?

- Translation is the simplest transformation:
- Inverse:

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

$$T^{-1}(\mathbf{v}) = \mathbf{v} - \mathbf{u}$$

Is translation linear transformation?

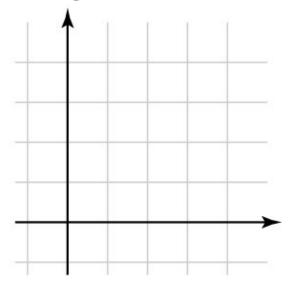
• No! because T(v) = Iv + u! = Iv

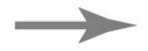
Affine transformation

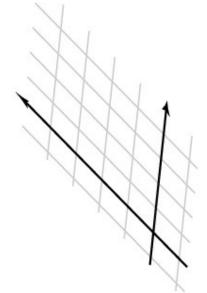
$$T(v)=Mv+u$$

Affine transformations

- straight lines preserved; parallel lines preserved
- ratios of lengths along lines preserved (midpoints preserved)
- Origin







Combining linear with translation

- Need to use both in single framework
- Can represent arbitrary seq. as

$$T(\mathbf{p}) = M_T \mathbf{p} + \mathbf{u}_T$$

$$S(\mathbf{p}) = M_S \mathbf{p} + \mathbf{u}_S$$

$$(S \circ T)(\mathbf{p}) = M_S (M_T \mathbf{p} + \mathbf{u}_T) + \mathbf{u}_S$$

$$= (M_S M_T) \mathbf{p} + (M_S \mathbf{u}_T + \mathbf{u}_S)$$

- Transforming by M_T and \mathbf{u}_T , then by M_S and \mathbf{u}_S , is the same as transforming by M_SM_T and $M_S\mathbf{u}_T+\mathbf{u}_S$
- This will work but is a little awkward

Homogeneous coordinates

- A trick for representing the foregoing more elegantly
- Extra component w for vectors, extra row/column for matrices
 - for affine, can always keep w = 1
- Represent linear transformations with dummy extra row and column

$$\begin{bmatrix} a & b & 0 \\ c & d & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} ax + by \\ cx + dy \\ 1 \end{bmatrix}$$

Homogeneous coordinates

 Represent translation using the extra column

$$egin{bmatrix} 1 & 0 & t \ 0 & 1 & s \ 0 & 0 & 1 \end{bmatrix} egin{bmatrix} x \ y \ 1 \end{bmatrix} = egin{bmatrix} x+t \ y+s \ 1 \end{bmatrix}$$

Homogeneous coordinates

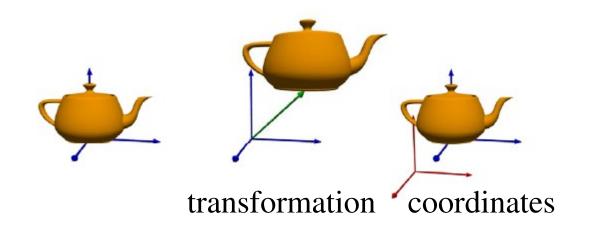
Composition just works, by 3x3 matrix multiplication

$$egin{bmatrix} M_S & \mathbf{u}_S \ 0 & 1 \end{bmatrix} egin{bmatrix} M_T & \mathbf{u}_T \ 0 & 1 \end{bmatrix} egin{bmatrix} \mathbf{p} \ 1 \end{bmatrix} \ = egin{bmatrix} (M_S M_T) \mathbf{p} + (M_S \mathbf{u}_T + \mathbf{u}_S) \ 1 \end{bmatrix}$$

- This is exactly the same result
 - but cleaner
 - and generalizes in useful ways as we'll see later

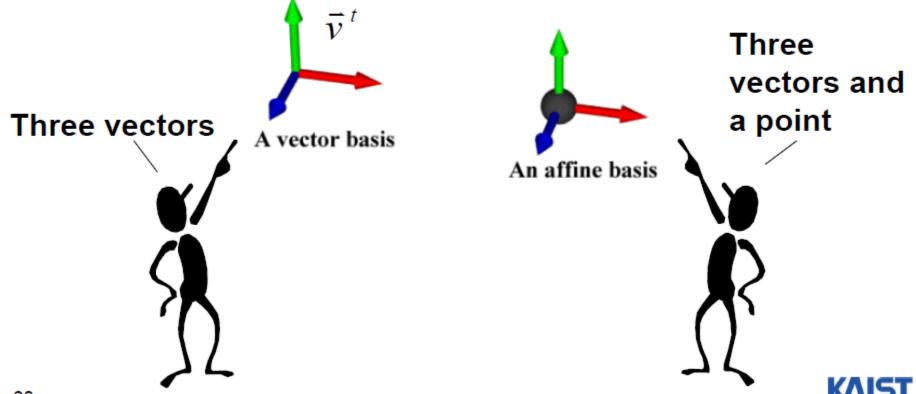
Homogeneous coordinates in 3D (Affine transformation)

$$\begin{vmatrix} v_{1x} & v_{2x} & v_{3x} & o_{x} \\ v_{1y} & v_{2y} & v_{3y} & o_{y} \\ v_{1z} & v_{2z} & v_{3z} & o_{z} \\ 0 & 0 & 1 \end{vmatrix} \begin{vmatrix} c_{1} \\ c_{2} \\ c_{3} \\ 1 \end{vmatrix} = \begin{vmatrix} v_{x} \cdot c + o_{x} \\ v_{y} \cdot c + o_{y} \\ v_{z} \cdot c + o_{z} \\ 1 \end{vmatrix} = \begin{pmatrix} V c + o \\ 1 \end{pmatrix}$$



Pictures of Frames

 Graphically, we will distinguish between vector bases and affine bases (frames) using the following convention



Coordinates:

A Basis for Points

- Key distinction between vectors and points: points are absolute, vectors are relative
- Vector space is completely defined by a set of basis vectors
- The space that points live in requires the specification of an absolute origin

$$p = \dot{o} + \sum_{j} \dot{V}_{j} c_{j} = \begin{bmatrix} \dot{V}_{1} & \dot{V}_{2} & \dot{V}_{3} & \dot{o} \end{bmatrix} \begin{bmatrix} c_{1} \\ c_{2} \\ c_{3} \\ 1 \end{bmatrix}$$

Notice how 4 scalars (one of which is 1) are required to identify a 3D point

Frames

- Points live in Affine spaces
- Affine-basis-sets are called frames

$$\dot{\mathbf{f}}^{\mathsf{t}} = \begin{bmatrix} \vec{\mathbf{v}}_1 & \vec{\mathbf{v}}_2 & \vec{\mathbf{v}}_3 & \dot{\mathbf{o}} \end{bmatrix}$$

Frames can describe vectors as well as points

$$\dot{p} = \begin{bmatrix} \vec{V}_1 & \vec{V}_2 & \vec{V}_3 & \dot{o} \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ c_3 \\ 1 \end{bmatrix} \qquad \dot{X} = \begin{bmatrix} \vec{V}_1 & \vec{V}_2 & \vec{V}_3 & \dot{o} \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ c_3 \\ 0 \end{bmatrix}$$



A Consistent Model

- Behavior of affine frame coordinates is completely consistent with our intuition
 - Subtracting two points yields a vector
 - Adding a vector to a point produces a point
 - If you multiply a vector by a scalar you still get a vector
 - Scaling points gives a nonsense 4th coordinate element in most cases

$$\begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ 1 \end{bmatrix} - \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ 1 \end{bmatrix} = \begin{bmatrix} a_1 - b_1 \\ a_2 - b_2 \\ a_3 - b_3 \\ 0 \end{bmatrix}$$

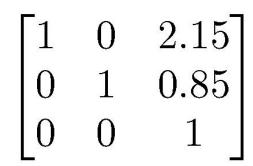
$$\begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ 1 \end{bmatrix} - \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ 1 \end{bmatrix} = \begin{bmatrix} a_1 - b_1 \\ a_2 - b_2 \\ a_3 - b_3 \\ 0 \end{bmatrix} \qquad \begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ 1 \end{bmatrix} + \begin{bmatrix} v_1 \\ v_2 \\ v_3 \\ 0 \end{bmatrix} = \begin{bmatrix} a_1 + v_1 \\ a_2 + v_2 \\ a_3 + v_3 \\ 1 \end{bmatrix}$$

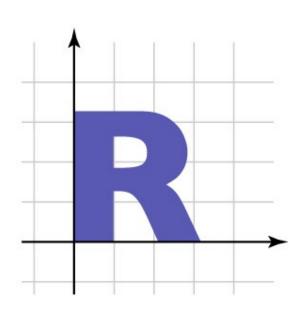


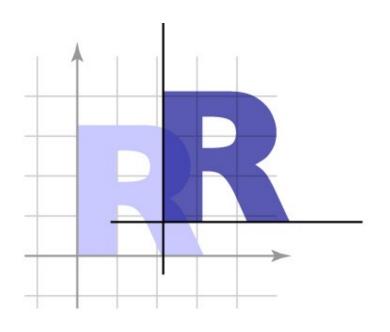
2D Affine transformation gallery

Translation

$\lceil 1 \rceil$	0	t_x
0	1	t_y
0	0	$1 \rfloor$

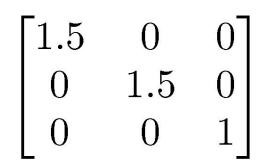


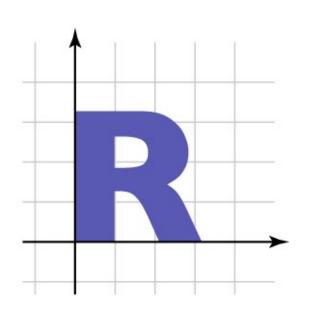


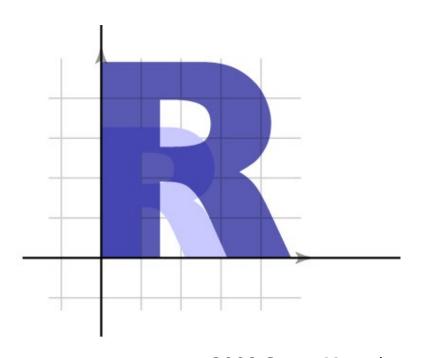


Uniform scale

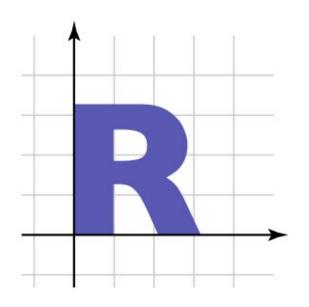
$\lceil s \rceil$	0	$\begin{bmatrix} 0 \end{bmatrix}$	
0	s	0	
0	0	1	

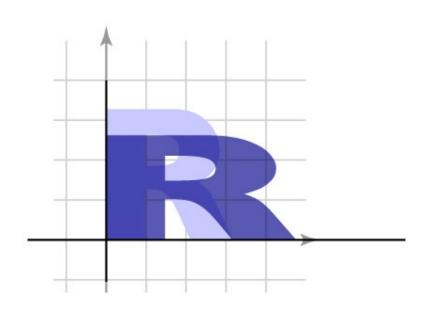


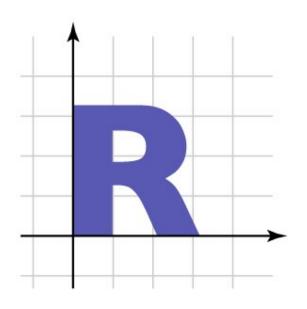


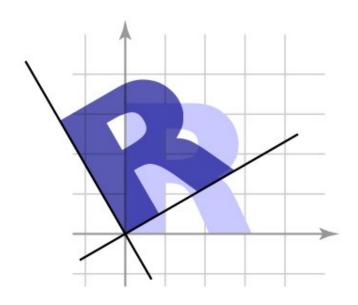


• Nonuniform $\mbox{sca} \begin{bmatrix} s_x & 0 & 0 \\ 0 & s_y & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad \begin{bmatrix} 1.5 & 0 & 0 \\ 0 & 0.8 & 0 \\ 0 & 0 & 1 \end{bmatrix}$



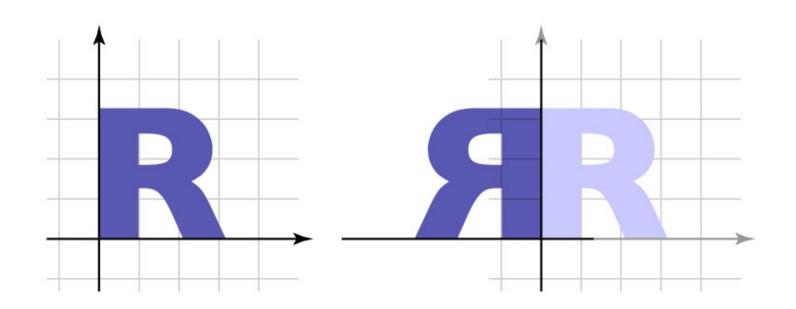






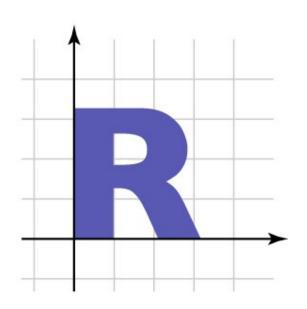
- Reflection
 - can consider it a special case of nonuniform scale

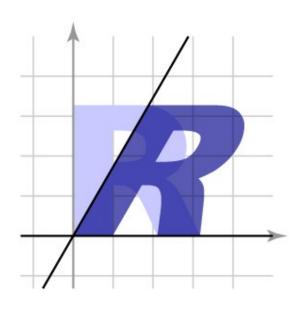
$\lceil -1 \rceil$	0	0
0	1	0
0	0	1_



• Shear

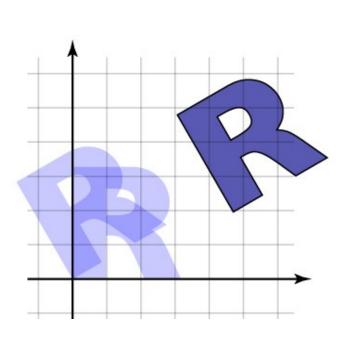
	a		Γ1	0.5	0
0	1	0	0	1	0
0	0	$\begin{bmatrix} 1 \end{bmatrix}$	0	0	$1 \rfloor$



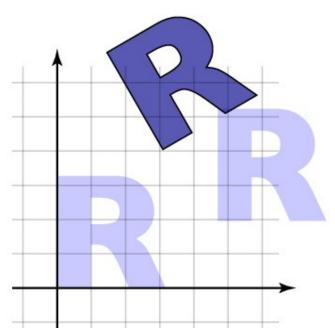


Composite affine transformations

In general not commutative: order matters!



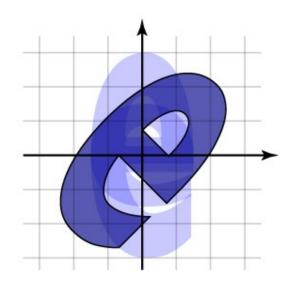
rotate, then translate



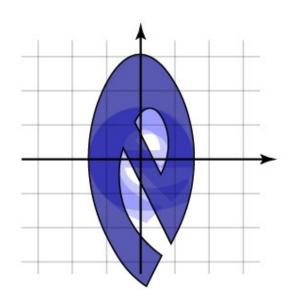
translate, then rotate

Composite affine transformations

Another example

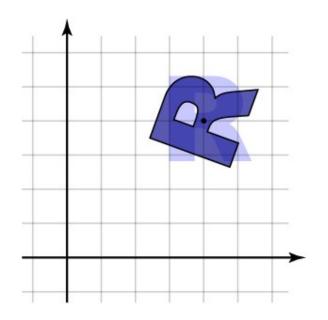


scale, then rotate



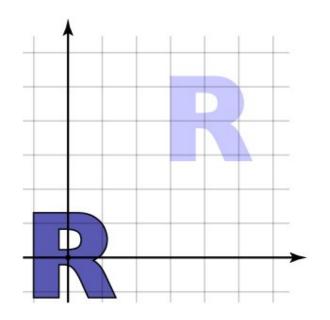
rotate, then scale

- Want to rotate about a particular point
 - could work out formulas directly...
- Know how to rotate about the origin
 - so translate that point to the origin



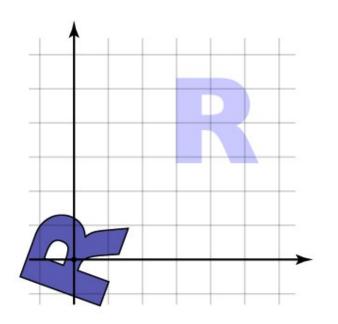
$$M = T^{-1}RT$$

- Want to rotate about a particular point
 - could work out formulas directly...
- Know how to rotate about the origin
 - so translate that point to the origin



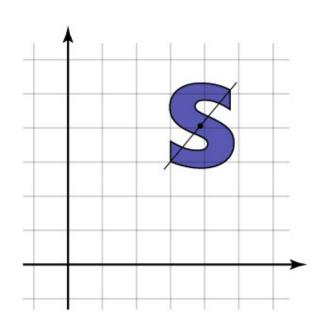
$$M = T^{-1}RT$$

- Want to rotate about a particular point
 - could work out formulas directly...
- Know how to rotate about the origin
 - so translate that point to the origin



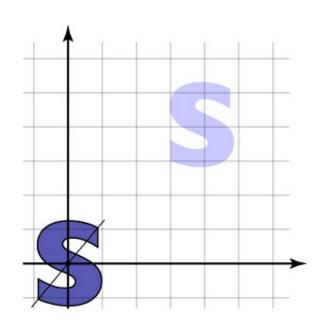
$$M = T^{-1}RT$$

- Want to scale along a particular axis and point
- Know how to scale along the y axis at the origin



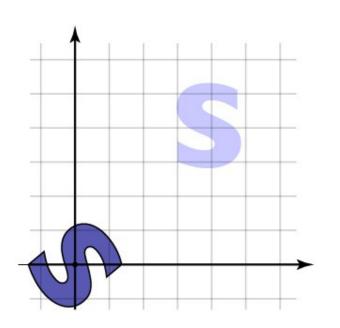
$$M = T^{-1}R^{-1}SRT$$

- Want to scale along a particular axis and point
- Know how to scale along the y axis at the origin



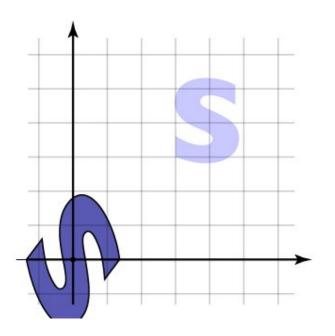
$$M = T^{-1}R^{-1}SRT$$

- Want to scale along a particular axis and point
- Know how to scale along the y axis at the origin



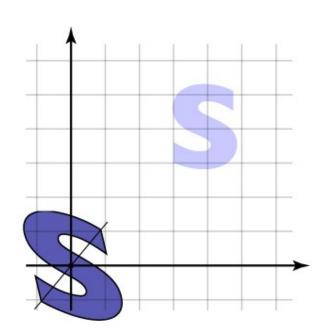
$$M = T^{-1}R^{-1}SRT$$

- Want to scale along a particular axis and point
- Know how to scale along the y axis at the origin



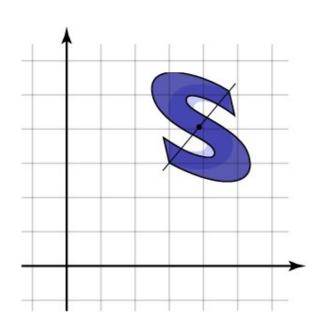
$$M = T^{-1}R^{-1}SRT$$

- Want to scale along a particular axis and point
- Know how to scale along the y axis at the origin



$$M = T^{-1}R^{-1}SRT$$

- Want to scale along a particular axis and point
- Know how to scale along the y axis at the origin



$$M = T^{-1}R^{-1}SRT$$

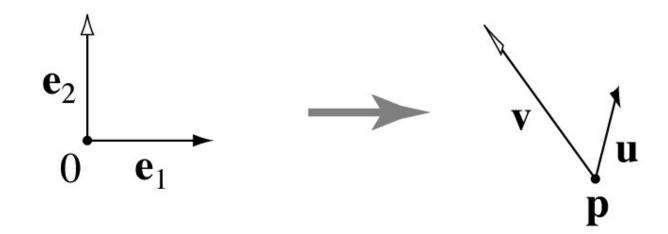
More math background

- Coordinate systems
 - Expressing vectors with respect to bases
 - Linear transformations as changes of basis

Affine change of coordinates

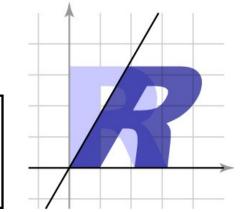
Six degrees of freedom

$$\begin{bmatrix} a_1 & a_2 & a_3 \\ a_4 & a_5 & a_6 \\ 0 & 0 & 1 \end{bmatrix} \quad \text{or} \quad \begin{bmatrix} \mathbf{u} & \mathbf{v} & \mathbf{p} \\ 0 & 0 & 1 \end{bmatrix}$$



Affine change of coordinates

- A new way to "read off" the matri
 - e.g. shear from earlier
 - can look at picture, see ef on basis vectors, write down matrix



- Also an easy way to construct transforms
 - e. g. scale by 2 across direction (1,2)

Rigid motions (Proper Euclidean space)

- A transform made up of only translation and rotation is a rigid motion or a rigid body transformation
- The linear part is an orthonormal matrix

$$R = \begin{bmatrix} Q & \mathbf{u} \\ 0 & 1 \end{bmatrix}$$

- Inverse of orthonormal matrix is transpose
 - so inverse of rigid motion is easy:

$$R^{-1}R = \begin{bmatrix} Q^T & -Q^T\mathbf{u} \\ 0 & 1 \end{bmatrix} \begin{bmatrix} Q & \mathbf{u} \\ 0 & 1 \end{bmatrix}$$

Transforming points and vectors

- Recall distinction points vs. vectors
 - vectors are just offsets (differences between points)
 - points have a location
 - represented by vector offset from a fixed origin
- Points and vectors transform differently
 - points respond to translation; vectors do not

$$\mathbf{v}=\mathbf{p}-\mathbf{q}$$
 $T(\mathbf{x})=M\mathbf{x}+\mathbf{t}$ $T(\mathbf{p}-\mathbf{q})=M\mathbf{p}+\mathbf{t}-(M\mathbf{q}+\mathbf{t})$ $=M(\mathbf{p}-\mathbf{q})+(\mathbf{t}-\mathbf{t})=M\mathbf{v}$ 8 Steve Marschner $oldsymbol{\cdot}_{57}$

Transforming points and vectors

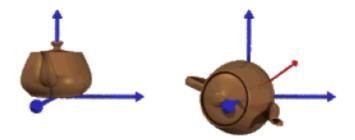
- Homogeneous coords. let us exclude translation
 - just put 0 rather than 1 in the last place

$$\begin{bmatrix} M & \mathbf{t} \\ \mathbf{0}^T & 1 \end{bmatrix} \begin{bmatrix} \mathbf{p} \\ 1 \end{bmatrix} = \begin{bmatrix} M\mathbf{p} + \mathbf{t} \\ 1 \end{bmatrix} \quad \begin{bmatrix} M & \mathbf{t} \\ \mathbf{0}^T & 1 \end{bmatrix} \begin{bmatrix} \mathbf{v} \\ 0 \end{bmatrix} = \begin{bmatrix} M\mathbf{v} \\ 0 \end{bmatrix}$$

 and note that subtracting two points cancels the extra coordinate, resulting in a vector!

3D Rotations

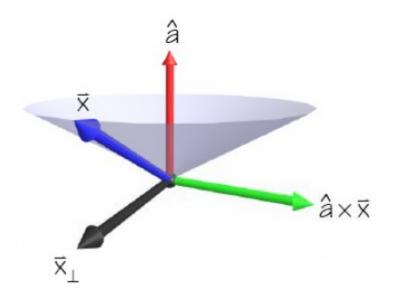
- More complicated than 2D rotations
 - Rotate objects along a rotation axis



- Several approaches
 - Compose three canonical rotations about the axes
 - Quaternions

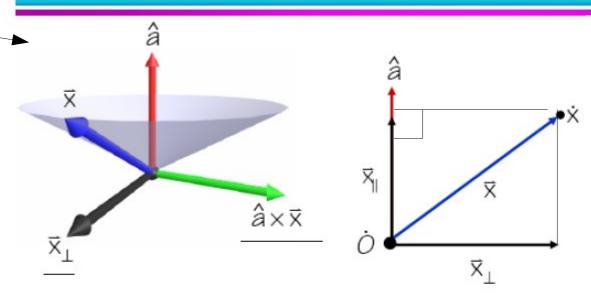


- Natural basis for rotation of a vector about a specified axis:
 - ° â rotation axis (normalized)
 - ° âxx vector perpendicular to
 - X
 _⊥ perpendicular component of X relative to â



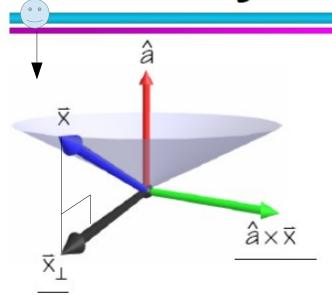


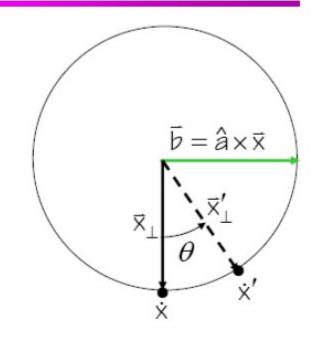




$$\vec{x}_{\parallel} = \hat{a}(\hat{a} \cdot \vec{x})$$

$$\vec{\mathbf{x}}_\perp = \vec{\mathbf{x}} - \vec{\mathbf{x}}_\parallel$$





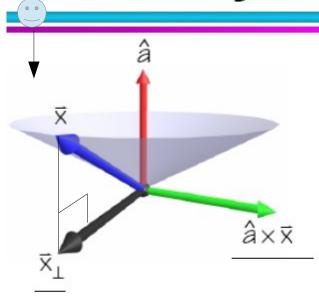
$$\dot{\mathbf{x}}' = \dot{O} + \mathbf{x}_{\parallel} + \mathbf{\overline{x}}_{\perp}'$$
$$\mathbf{\overline{x}}_{\perp}' = \cos\theta \,\mathbf{\overline{x}}_{\perp} + \sin\theta \,\mathbf{\overline{b}}$$

$$\vec{x}_{\parallel} = \hat{a}(\hat{a} \cdot \vec{x})$$

$$\vec{\mathbf{x}}_\perp = \vec{\mathbf{x}} - \vec{\mathbf{x}}_\parallel$$

$$\dot{\mathbf{x}}' = \dot{O} + \cos\theta\,\mathbf{\overline{x}} + (1 - \cos\theta)(\hat{\mathbf{a}}(\hat{\mathbf{a}}\cdot\mathbf{\overline{x}})) + \sin\theta(\hat{\mathbf{a}}\times\mathbf{\overline{x}})$$

Note that
$$\|\vec{x}_{\perp}\| = \|\hat{a} \times \vec{x}\|$$



$$\dot{\mathbf{x}}' = \dot{O} + \mathbf{x}_{\parallel} + \mathbf{\bar{x}}'_{\perp}$$

$$\mathbf{\bar{x}}'_{\perp} = \cos\theta \,\mathbf{\bar{x}}_{\perp} + \sin\theta \,\mathbf{\bar{b}}$$

$$\mathbf{\bar{x}}_{\parallel} = \hat{\mathbf{a}}(\hat{\mathbf{a}} \cdot \mathbf{\bar{x}})$$

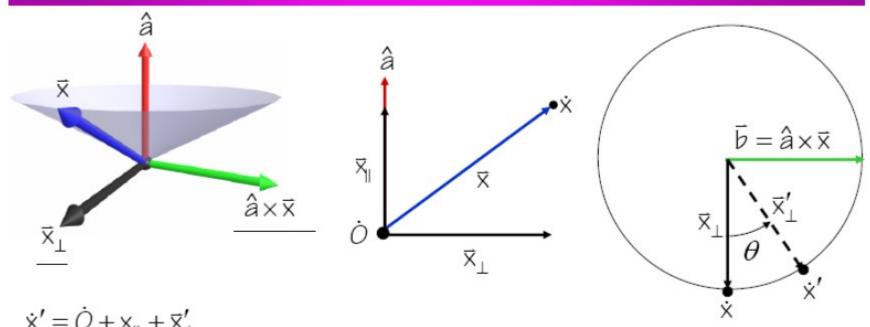
$$\vec{\mathbf{x}}^\top = \vec{\mathbf{x}} - \vec{\mathbf{x}}^\parallel$$

$$\dot{\mathbf{x}}' = \dot{O} + \cos\theta \, \bar{\mathbf{x}} + (1 - \cos\theta)(\hat{\mathbf{a}}(\hat{\mathbf{a}} \cdot \bar{\mathbf{x}})) + \sin\theta(\hat{\mathbf{a}} \times \bar{\mathbf{x}})$$

$$\mathbf{c}_{\dot{\mathbf{x}}'} = \mathbf{M}\mathbf{c}_{\dot{\mathbf{x}}}$$

$$\mathbf{M} = \operatorname{diag}(\dot{O}) + \cos\theta \operatorname{diag}([1 \ 1 \ 1 \ O]^{t})$$

$$+ (1 - \cos\theta)\mathbf{A}_{\otimes} + \sin\theta \mathbf{A}_{\times}$$



$$\dot{\mathbf{x}}' = \dot{O} + \mathbf{x}_{\parallel} + \mathbf{\bar{x}}'_{\perp}$$

$$\mathbf{\bar{x}}'_{\perp} = \cos\theta \,\mathbf{\bar{x}}_{\perp} + \sin\theta \,\mathbf{\bar{b}}$$

$$\mathbf{\bar{x}}_{\parallel} = \hat{\mathbf{a}}(\hat{\mathbf{a}} \cdot \mathbf{\bar{x}})$$

$$\vec{\mathbf{x}}_{\perp} = \vec{\mathbf{x}} - \vec{\mathbf{x}}_{\parallel}$$

$$\dot{\mathbf{x}}' = \dot{O} + \cos\theta \, \bar{\mathbf{x}} + (1 - \cos\theta)(\hat{\mathbf{a}}(\hat{\mathbf{a}} \cdot \bar{\mathbf{x}})) + \sin\theta(\hat{\mathbf{a}} \times \bar{\mathbf{x}})$$

$$\mathbf{c}_{\dot{\mathbf{x}}'} = \mathbf{M}\mathbf{c}_{\dot{\mathbf{x}}}$$

$$\mathbf{M} = \operatorname{diag}(\dot{O}) + \cos\theta \operatorname{diag}([1 \ 1 \ 1 \ O]^{t})$$

$$+ (1 - \cos\theta)\mathbf{A}_{\otimes} + \sin\theta \mathbf{A}_{\times}$$

Vector Algebra

- We already saw vector addition and multiplications by a scalar
- Will study three kinds of vector multiplications
 - Dot product (·)
 - Cross product (×)
 - Tensor product (⊗)

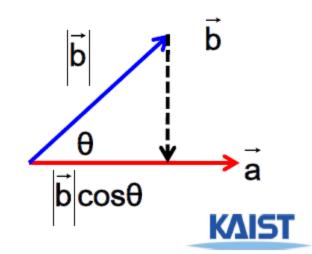
- returns a scalar
- returns a vector
- returns a matrix



Dot Product (·)

$$\vec{a} \cdot \vec{b} = \vec{a}^T \vec{b} = \begin{bmatrix} a_x & a_y & a_z & 0 \end{bmatrix} \begin{bmatrix} b_x \\ b_y \\ b_z \\ 0 \end{bmatrix} = s, \qquad \vec{a} \cdot \vec{b} = \vec{a}^T \vec{b} = \begin{bmatrix} a_x & a_y & a_z & 0 \end{bmatrix} \begin{bmatrix} b_x \\ b_y \\ b_z \\ 1 \end{bmatrix} = s$$

- Returns a scalar s
- Geometric interpretations s:
 - $\vec{a} \cdot \vec{b} = |\vec{a}| |\vec{b}| \cos \theta$
 - Length of b projected onto and a or vice versa
 - Distance of b from the origin in the direction of a

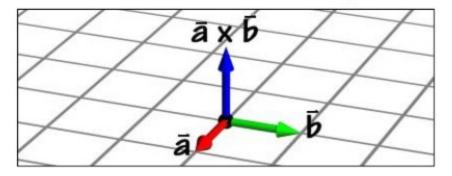


Cross Product (×)

$$\vec{a} \times \vec{b} \equiv \begin{bmatrix} 0 & -a_z & a_y & 0 \\ a_z & 0 & -a_x & 0 \\ -a_y & a_x & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} b_x \\ b_y \\ b_z \\ 0 \end{bmatrix} = \vec{c} \qquad \vec{b} \cdot \vec{c} = 0$$

$$\vec{c} = \begin{bmatrix} a_y b_z - a_z b_y & a_z b_x - a_x b_z & a_x b_y - a_y b_x \end{bmatrix}$$

- Return a vector \vec{c} that is perpendicular to both \vec{a} and \vec{b} , oriented according to the right-hand rule
- The matrix is called the skew-symmetric matrix of a





Cross Product (×)

 A mnemonic device for remembering the cross-product

$$\vec{a} \times \vec{b} \equiv \det \begin{bmatrix} \vec{i} & \vec{j} & \vec{k} \\ a_x & a_y & a_z \\ b_x & b_y & b_z \end{bmatrix}$$

$$= (a_y b_z - a_z b_y) \vec{i} + (a_z b_x - a_x b_z) \vec{j} + (a_x b_y - a_y b_x) \vec{k}$$

$$\vec{i} = \begin{bmatrix} 1 & O & O \end{bmatrix}$$

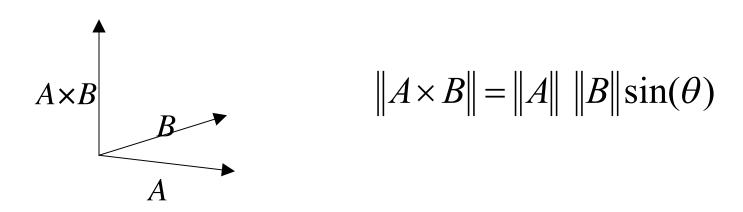
$$\vec{j} = \begin{bmatrix} O & 1 & O \end{bmatrix}$$

$$\vec{k} = \begin{bmatrix} O & O & 1 \end{bmatrix}$$

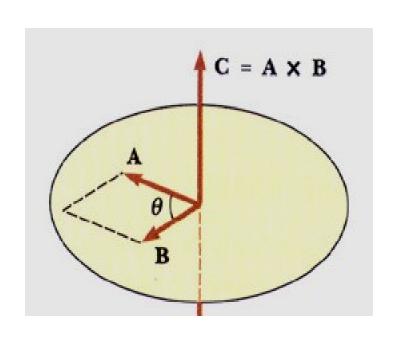


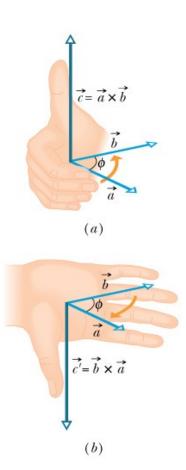
Vectors: Cross Product

- The cross product of vectors A and B is a vector C which is perpendicular to A and B
- The magnitude of \boldsymbol{C} is proportional to the sin of the angle between \boldsymbol{A} and \boldsymbol{B}
- The direction of *C* follows the **right hand rule** if we are working in a right-handed coordinate system



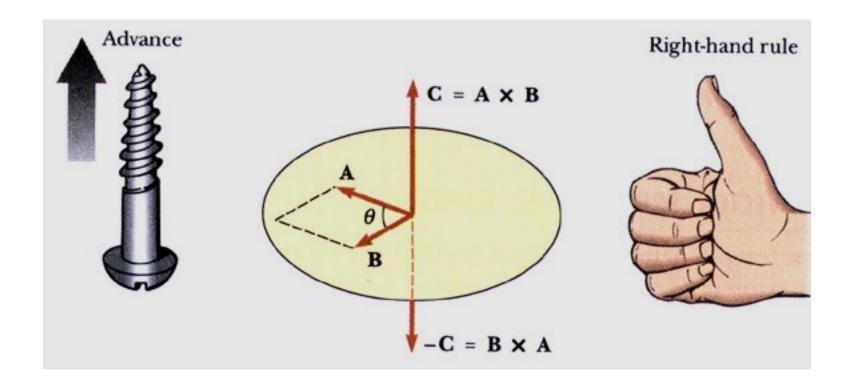
MAGNITUDE OF THE CROSS PRODUCT





DIRECTION OF THE CROSS PRODUCT

 The right hand rule determines the direction of the cross product



Tensor Product (⊗)

$$\vec{a} \otimes \vec{b} = \vec{a} \vec{b}^{t} = \begin{bmatrix} a_{x} \\ a_{y} \\ a_{z} \\ O \end{bmatrix} \begin{bmatrix} b_{x} & b_{y} & b_{z} & O \end{bmatrix} = \begin{bmatrix} a_{x}b_{x} & a_{x}b_{y} & a_{x}b_{z} & O \\ a_{y}b_{x} & a_{y}b_{y} & a_{y}b_{z} & O \\ a_{z}b_{x} & a_{z}b_{y} & a_{z}b_{z} & O \\ O & O & O & O \end{bmatrix}$$

$$(\vec{a} \otimes \vec{b}) \vec{c} = \begin{bmatrix} (b_x c_x + b_y c_y + b_z c_z) a_x \\ (b_x c_x + b_y c_y + b_z c_z) a_y \\ (b_x c_x + b_y c_y + b_z c_z) a_z \end{bmatrix} = \vec{a} (\vec{b} \cdot \vec{c})$$

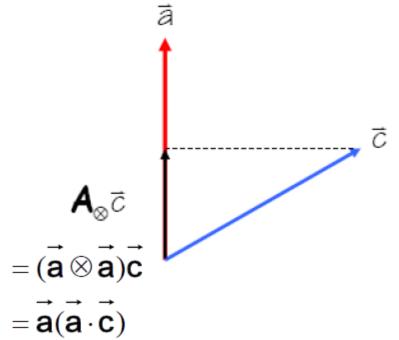
Creates a matrix that when applied to a vector c return a scaled by the project of c onto b



Tensor Product (⊗)

- Useful when $\vec{b} = \vec{a}$
- The matrix a⊗a is called







Sanity Check

Consider a rotation by about the x-axis

$$= \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & \cos \theta & -\sin \theta & 0 \\ 0 & \sin \theta & \cos \theta & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

 You can check it in any computer graphics book, but you don't need to memorize it

