

CSC 433/533

Computer Graphics

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Credit: Joshua Levine

Lecture 18

Transformations and Science Graphs

Oct. 30, 2020

Geometric Transformations as Matrices

- Operations to arrange objects in a scene, view them with cameras, and get them on a screen all can be encoded with linear algebra
- Geometric operations: rotation, translation, scaling, projection, and more...
- These transforms operate differently on points, displacement vectors, and surface normals

2D Linear Transformations

- We can transform points in a 2D coordinate system by multiplying the point (a vector) by a matrix (the transformation):

- e.g. Multiply the matrix A by $\mathbf{x} = (x, y)$, or $A\mathbf{x}$:

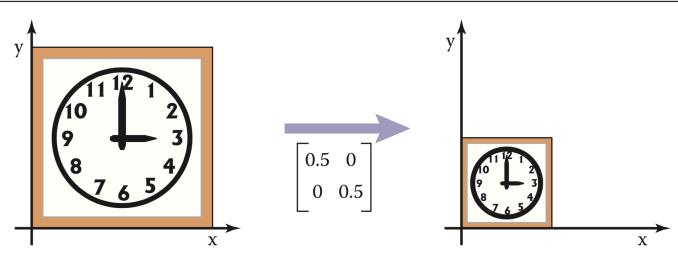
$$\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} a_{11}x + a_{12}y \\ a_{21}x + a_{22}y \end{bmatrix}$$

- Such transformations are called *linear* because they satisfy the relationship that $A(a\mathbf{x}_1 + \mathbf{x}_2) = aA\mathbf{x}_1 + A\mathbf{x}_2$

Scaling

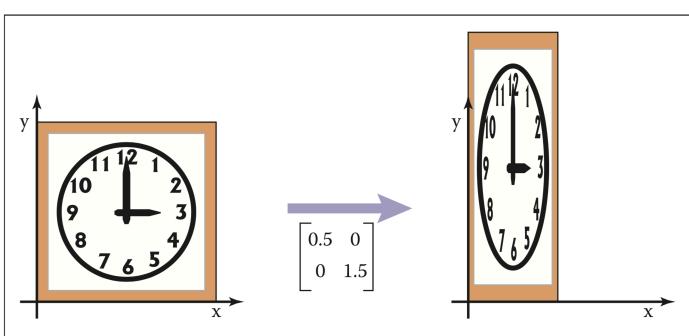
- Can be uniform, where $s_x = s_y$.

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



Scaling

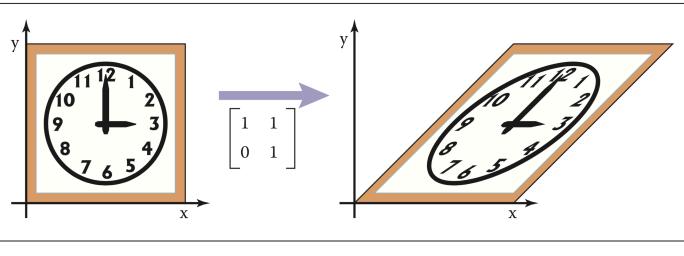
- Can also be nonuniform, where $s_x \neq s_y$.



Shearing

- Horizontal shearing shifts each row based on the y value.

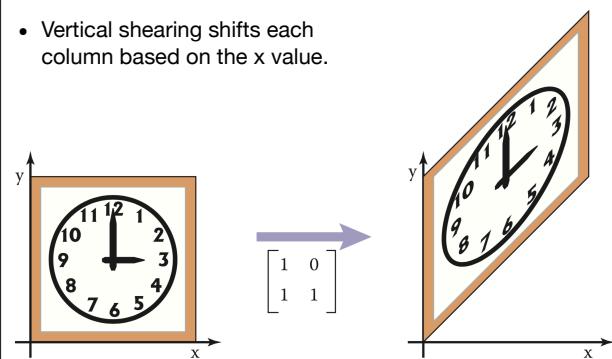
$$\text{shear-x}(s) = \begin{bmatrix} 1 & s \\ 0 & 1 \end{bmatrix}$$



Shearing

$$\text{shear-y}(s) = \begin{bmatrix} 1 & 0 \\ s & 1 \end{bmatrix}$$

- Vertical shearing shifts each column based on the x value.

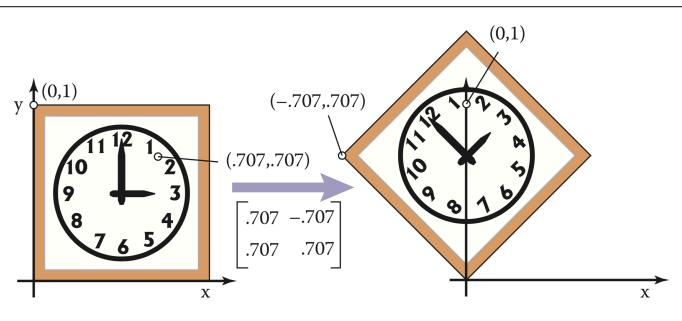


Rotation

- Rotate counterclockwise by an angle ϕ about the origin.

$$\text{rotate}(\phi) = \begin{bmatrix} \cos \phi & -\sin \phi \\ \sin \phi & \cos \phi \end{bmatrix}$$

$$\begin{bmatrix} \cos \frac{\pi}{4} & -\sin \frac{\pi}{4} \\ \sin \frac{\pi}{4} & \cos \frac{\pi}{4} \end{bmatrix} = \begin{bmatrix} 0.707 & -0.707 \\ 0.707 & 0.707 \end{bmatrix}$$

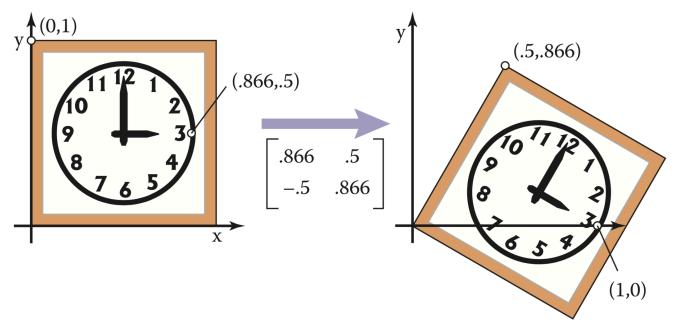


Rotation

- Clockwise rotations can also be represented by negative angles

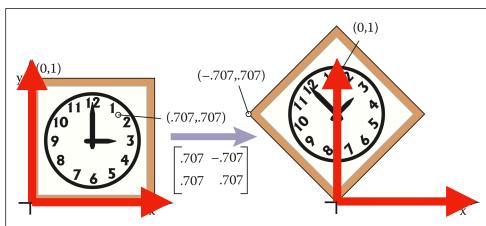
$$\text{rotate}(\phi) = \begin{bmatrix} \cos \phi & -\sin \phi \\ \sin \phi & \cos \phi \end{bmatrix}$$

$$\begin{bmatrix} \cos \frac{-\pi}{6} & -\sin \frac{-\pi}{6} \\ \sin \frac{-\pi}{6} & \cos \frac{-\pi}{6} \end{bmatrix} = \begin{bmatrix} 0.866 & 0.5 \\ -0.5 & 0.866 \end{bmatrix}$$



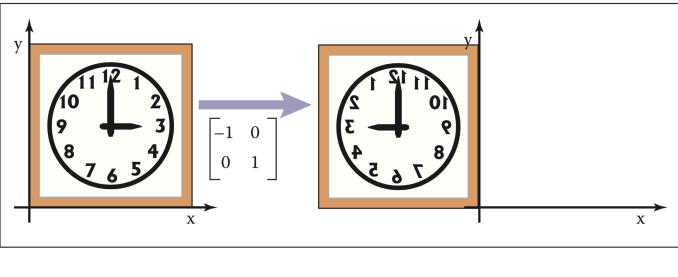
Rotations are Always Orthogonal Matrices

- Recall: An **orthogonal matrix** always has columns and rows that are orthogonal unit vectors
- Implication: Have the effect of rotating the coordinate axes

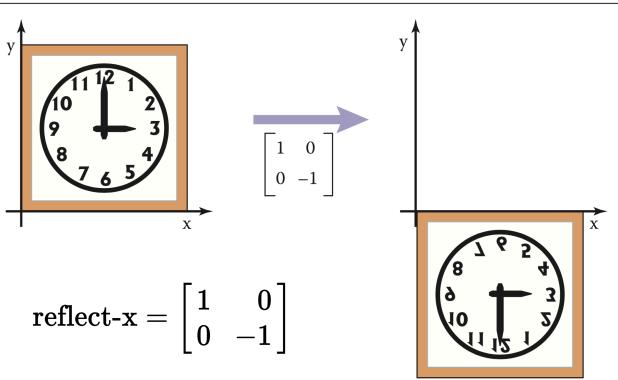


Reflection

$$\text{reflect-y} = \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix}$$



Reflection



Composition

- Transformations can be **composed** to perform combinations of transformations.
- For example, one could first rotate with matrix R and then scale with matrix S
- Applied to a point \mathbf{v}_1 , this would be
 - $\mathbf{v}_2 = \mathbf{R}\mathbf{v}_1$ to rotate and then
 - $\mathbf{v}_3 = \mathbf{S}\mathbf{v}_2 = \mathbf{S}\mathbf{R}\mathbf{v}_1$
 $= (\mathbf{SR})\mathbf{v}_1 = \mathbf{T}\mathbf{v}_1$ where $\mathbf{T} = \mathbf{SR}$

Order Matters for Composition

- First rotate, then non-uniform scale

- Rotate:

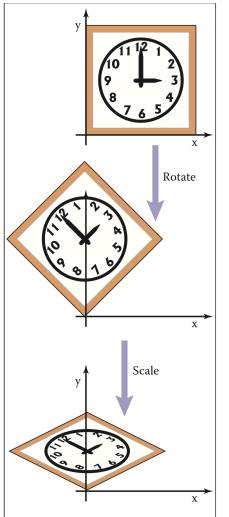
$$\begin{bmatrix} 0.707 & -0.707 \\ 0.707 & 0.707 \end{bmatrix}$$

- Scale:

$$\begin{bmatrix} 1 & 0 \\ 0 & 0.5 \end{bmatrix}$$

- Combined:

$$\begin{bmatrix} 1 & 0 \\ 0 & 0.5 \end{bmatrix} \begin{bmatrix} 0.707 & -0.707 \\ 0.707 & 0.707 \end{bmatrix} = \begin{bmatrix} 0.707 & -0.707 \\ 0.353 & 0.353 \end{bmatrix}$$



Order Matters for Composition

- First non-uniform scale, then rotate

- Rotate:

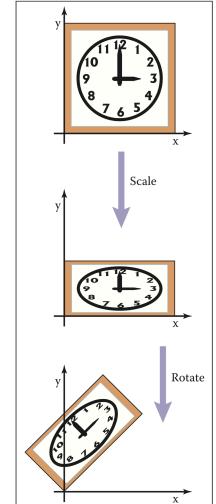
$$\begin{bmatrix} 0.707 & -0.707 \\ 0.707 & 0.707 \end{bmatrix}$$

- Scale:

$$\begin{bmatrix} 1 & 0 \\ 0 & 0.5 \end{bmatrix}$$

- Combined:

$$\begin{bmatrix} 0.707 & -0.707 \\ 0.707 & 0.707 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 0.5 \end{bmatrix} = \begin{bmatrix} 0.707 & -0.353 \\ 0.707 & 0.353 \end{bmatrix}$$



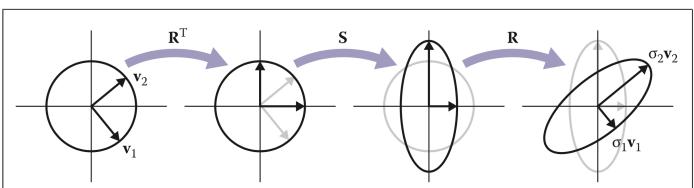
Decomposition

- Recall: a **symmetric matrix** is any matrix M such that $M = M^T$
- Let's consider a special type of composition, of the form \mathbf{RSR}^T
- Motivation: scaling on arbitrary axis, e.g. rotate, then scale, then rotate back
- Example: $\text{rotate}(-45^\circ)\text{scale}(1.5, 1)\text{rotate}(45^\circ)$

$$\begin{bmatrix} 1.25 & -0.25 \\ -0.25 & 1.25 \end{bmatrix}$$

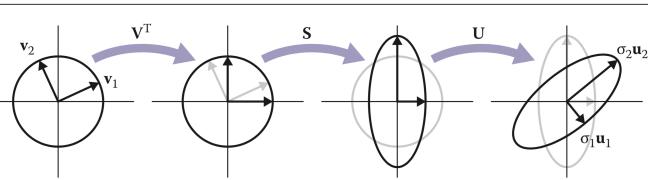
Decomposition

- \mathbf{RSR}^T is always a symmetric matrix
 - Why? $(\mathbf{RSR}^T)^T = \mathbf{R}^T \mathbf{S}^T \mathbf{R}^T = \mathbf{RSR}^T$
- Any symmetric metric A can be decomposed to $A = \mathbf{RSR}^T$
 - All rotations R, happen to be **orthogonal matrices**.
 - Any symmetric transformation matrix is a scale in some axis

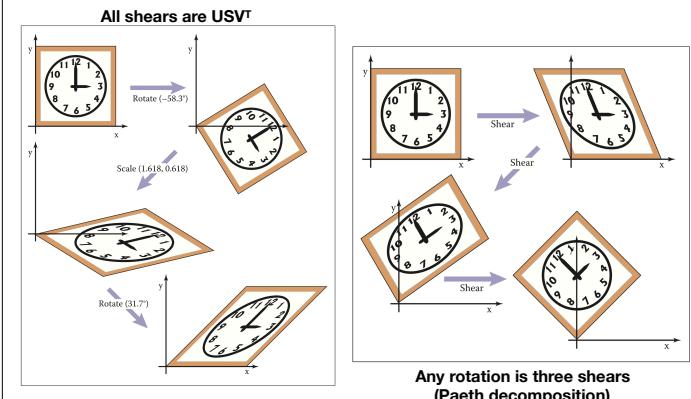


Decomposition

- Any arbitrary matrix can be decomposed to $A = USV^T$ where U and V are orthogonal matrices (but not necessarily rotations)
- This is called **singular value decomposition**
- This has a similar interpretation, but with different rotations before/after the scale



Some Consequences



Inversion

- Interpreting a matrix M as a geometric transformation, how might we undo the operation?
- We can apply the **inverse** transformation, it turns out by applying the inverse matrix, M^{-1}
 - Recall that $MM^{-1} = M^{-1}M = I$, the identity matrix
- This is true for all operations we've discussed, e.g. scale by s is undone by scale of 1/s, rotate by angle ϕ is undone by rotate of angle $-\phi$
- Key point:** the algebraic inverse is the geometric inverse

3D Linear Transformations

3D Linear Transformations

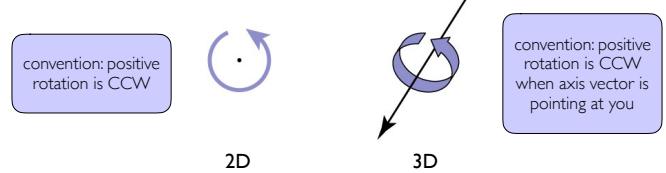
- We can transform points in a 3D coordinate system by multiplying the point (a vector) by a matrix (the transformation), just like in 2D!
- The only difference is we will use 3x3 matrices A by $\mathbf{x} = (x, y, z)$, or $A\mathbf{x}$, e.g. for scale and shear:

$$\text{scale}(s_x, s_y, s_z) = \begin{bmatrix} s_x & 0 & 0 \\ 0 & s_y & 0 \\ 0 & 0 & s_z \end{bmatrix}$$

$$\text{shear-x}(d_y, d_z) = \begin{bmatrix} 1 & d_y & d_z \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Rotations in 3D

- In 2D, a rotation is about a point
- In 3D, a rotation is about an axis



Rotations about 3D Axes

- In 3D, we need to pick an axis to rotate about

$$\text{rotate-z}(\phi) = \begin{bmatrix} \cos \phi & -\sin \phi & 0 \\ \sin \phi & \cos \phi & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

- And we can pick any of the three axes

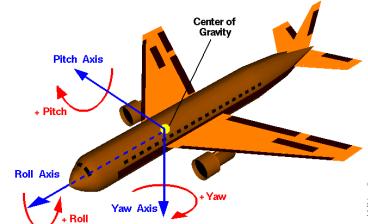
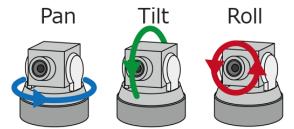
$$\text{rotate-x}(\phi) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \phi & -\sin \phi \\ 0 & \sin \phi & \cos \phi \end{bmatrix}$$

$$\text{rotate-y}(\phi) = \begin{bmatrix} \cos \phi & 0 & \sin \phi \\ 0 & 1 & 0 \\ -\sin \phi & 0 & \cos \phi \end{bmatrix}$$

Building Complex Rotations from Axis-Aligned Rotations

- Rotations about x , y , z are sometimes called **Euler angles**

- Build a combined rotation using matrix composition



Ishaya/Wikimedia Laboratory

Wikipedia

Arbitrary Rotations

- To rotate about any axis: we change the coordinate space we are working in, using orthogonal matrices.
- Consider orthogonal matrix R_{uvw} , form by taking three orthogonal vectors u , v , and w :

Property of orthogonal vectors:

$$\begin{aligned} \mathbf{u} \cdot \mathbf{u} &= \mathbf{v} \cdot \mathbf{v} = \mathbf{w} \cdot \mathbf{w} = 1 \\ \mathbf{u} \cdot \mathbf{v} &= \mathbf{v} \cdot \mathbf{w} = \mathbf{w} \cdot \mathbf{u} = 0 \end{aligned}$$

$$R_{uvw} = \begin{bmatrix} \mathbf{u} \\ \mathbf{v} \\ \mathbf{w} \end{bmatrix}$$

Arbitrary Rotations

- What happens when we apply R_{uvw} to any of the basis vectors, e.g.:

$$R_{uvw} \mathbf{u} = \begin{bmatrix} \mathbf{u} \cdot \mathbf{u} \\ \mathbf{v} \cdot \mathbf{u} \\ \mathbf{w} \cdot \mathbf{u} \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = \mathbf{x}$$

- But this means that if we apply R_{uvw}^T to the Cartesian coordinate vectors, e.g.:

$$R_{uvw}^T \mathbf{y} = \begin{bmatrix} x_u & x_v & x_w \\ y_u & y_v & y_w \\ z_u & z_v & z_w \end{bmatrix} \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} x_v \\ y_v \\ z_v \end{bmatrix} = \mathbf{v}$$

Arbitrary Rotations

- This means that if we want to rotate around an arbitrary axis, we need only to use a change of coordinates
- E.g. to rotate around a direction w , we
 - Compute orthogonal directions u , v , and w
 - Change the uvw axes to be xyz (R_{uvw})
 - Apply a $\text{rotate-z}()$
 - Finally, change the axes back to uvw (R_{uvw}^T)

$$\begin{bmatrix} x_u & x_v & x_w \\ y_u & y_v & y_w \\ z_u & z_v & z_w \end{bmatrix} \begin{bmatrix} \cos \phi & -\sin \phi & 0 \\ \sin \phi & \cos \phi & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_u & y_u & z_u \\ x_v & y_v & z_v \\ x_w & y_w & z_w \end{bmatrix}$$

R_{uvw}^T

$\text{rotate-z}()$

R_{uvw}

Transformations for Shapes in Computer Graphics

- These transformations also apply to transforming geometric objects
- Parametric forms:
 - We can transform the points directly, e.g. $M(\mathbf{p}(t))$ to transform the parametric positions $\mathbf{p}(t)$
- Implicit forms:
 - We invert the transform and test the predicate, e.g. if $f(M^{-1}(\mathbf{p})) = 0$ then \mathbf{p} is on the transformed implicit shape

Affine Transformations

Translation

- Using the mathematics we've described so far, what about moving objects in space.
- Problem (in 2D), we have:

$$\begin{aligned}x' &= m_{11}x + m_{12}y \\y' &= m_{21}x + m_{22}y\end{aligned}$$

- But we want:

$$\begin{aligned}x' &= m_{11}x + m_{12}y + x_t \\y' &= m_{21}x + m_{22}y + y_t\end{aligned}$$

Homogeneous Coordinates

- To put this into one system of linear equations, we **promote** (increase the dimensionality) by adding a component $w = 1$ for vectors
- $$\begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \begin{bmatrix} m_{11} & m_{12} & x_t \\ m_{21} & m_{22} & y_t \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} m_{11}x + m_{12}y + x_t \\ m_{21}x + m_{22}y + y_t \\ 1 \end{bmatrix}$$
- Implements a linear transformation followed by a translation (x_t, y_t)
 - These transformations are called **affine transformations**:
 - Like linear transformations, they keep straight lines straight and parallel lines parallel, but they do not preserve the origin

Homogeneous Coordinates

- We promote all points (x, y) to $(x, y, w=1)$, and similarly in 3D we promote (x, y, z) to $(x, y, z, w=1)$
- These new coordinates are called **homogeneous coordinates**
- Can be thought of as a clever bookkeeping scheme, but also have a geometric interpretation, compare the following matrix with a standard shear:

$$\begin{bmatrix} 1 & 0 & x_t \\ 0 & 1 & y_t \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} x + x_t z \\ y + y_t z \\ z \end{bmatrix}$$

Homogeneous Coordinates

- Composition works just as before, but using 3×3 multiplication instead of 2×2
- This approach is easier than keeping the linear transform and the translate separately stored

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

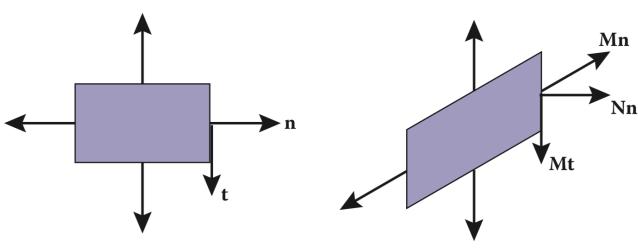
Transforming Points vs. Transforming Vectors

- Using homogeneous coordinates, we can differentiate between points and vectors
- Recall:
 - Vectors are just offsets (differences between points), and thus should be not affected by translation
 - Whereas points are represented by vectors offset from the origin
- Rule: vectors have $w=0$, whereas points have $w=1$:

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

Transforming Normals

- While differences between points transform OK, normals do not necessarily behave the same



Transforming Normals

- The problem is that the orthogonality constraint, that normals always point orthogonal to the surface, is not always preserved.
 - One can solve for the correct transformation by observing that tangent vectors, t , transform correctly and $t \cdot n = 0$.
- $$\mathbf{n}^T \mathbf{t} = \mathbf{n}^T \mathbf{I} \mathbf{t} = \mathbf{n}^T \mathbf{M}^{-1} \mathbf{M} \mathbf{t} = 0$$
- So, we can transform normals using the inverse matrix
- $$(\mathbf{M}^{-1})^T \mathbf{n}$$

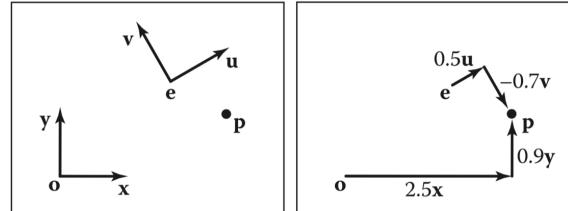
Coordinate Transformations

Coordinate Systems

- Points in space can be represented using an origin position and a set of orthogonal basis vectors:

$$\mathbf{p} = (x_p, y_p) \equiv \mathbf{0} + x_p \mathbf{x} + y_p \mathbf{y} \quad \mathbf{p} = (u_p, v_p) \equiv \mathbf{e} + u_p \mathbf{u} + v_p \mathbf{v}$$

- Any point can be described in either coordinate system



Matrices for Converting Coordinate Systems

- Using homogenous coordinates and affine transformations, we can convert between coordinate systems:

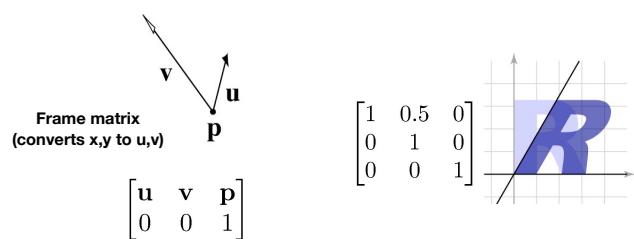
$$\begin{bmatrix} x_p \\ y_p \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & x_e \\ 0 & 1 & y_e \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_u & x_v & 0 \\ y_u & y_v & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} u_p \\ v_p \\ 1 \end{bmatrix} = \begin{bmatrix} x_u & x_v & x_e \\ y_u & y_v & y_e \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} u_p \\ v_p \\ 1 \end{bmatrix}$$

- More generally, any arbitrary coordinate system transform:

$$\mathbf{P}_{uv} = \begin{bmatrix} \mathbf{x}_{uv} & \mathbf{y}_{uv} & \mathbf{o}_{uv} \\ 0 & 0 & 1 \end{bmatrix} \mathbf{p}_{xy}$$

Affine Change of Coordinates

- It turns out this works even if u, v are not orthogonal.
- It also provides another way to interpret and construct transformation matrices



Lec19 Required Reading

- FOCG, Ch. 7

Reminder: Assignment 05

Assigned: Wednesday, Oct. 30
Written Due: Monday, Nov. 11, 4:59:59 pm
Program Due: Wednesday, Nov. 13, 4:59:59 pm