



Now Playing:




My Mathematical Mind
Spoon
From *Gimme Fiction*
Released May 10, 2005

PetShop (BYU Students, SIGGRAPH 2006)



Geometric Objects in Computer Graphics



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COMP 575
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Last Time

- Overview of the second half of the semester
- Talked about real cameras and light transport
- Talked about how to turn those ideas into a ray-tracer
 - Generate rays
 - Intersect rays with objects
 - Determine pixel color

Time for some math

- Today we're going to review some of the basic mathematical constructs used in computer graphics
 - Scalars
 - Points
 - Vectors
 - Matrices
 - Other stuff (rays, planes, etc.)

Scalars

- A scalar is a quantity that does not depend on direction
- In other words, it's just a regular number
 - *i.e.* 7 is a scalar
 - so is 13.5
 - or -4

Points

- A point is a list of n numbers referring to a location in n-D
- The individual components of a point are often referred to as coordinates
- *i.e.* (2, 3, 4) is a point in 3-D space
 - This point's x-coordinate is 2, it's y-coordinate is 3, and it's z-coordinate is 4

Vectors

- A vector is a list of n numbers referring to a direction (and magnitude) in n-D
 - *i.e.*
- N.B. - From a data structures perspective, a vector looks exactly the same as a point
 - This will be important later

Rays

- A ray is just a vector with a starting point
- Ray = (Point, Vector)

Rays

- Let a ray be defined by point **p** and vector **d**
- The parametric form of a ray expresses it as a function as some scalar t, giving the set of all points the ray passes through:
 - $r(t) = \mathbf{p} + t\mathbf{d}, 0 \leq t \leq \infty$

Vectors

- We said that a vector encodes a direction and a magnitude in n-D
- How does it do this?
- Here are two ways to denote a vector in 2-D:

$$\mathbf{v} = \langle v_x, v_y \rangle$$
$$\mathbf{v} = \begin{bmatrix} v_x \\ v_y \end{bmatrix}$$

Vector Magnitude

- Geometrically, the magnitude of a vector is the Euclidean distance between its start and end points, or more simply, it's length

- Vector magnitude in n-D: $\|\mathbf{v}\| = \sqrt{\sum_{i=1}^n v_i^2}$
- Vector magnitude in 2-D: $\|\mathbf{v}\| = \sqrt{v_x^2 + v_y^2}$

Normalized Vectors

- Most of the time, we want to deal with normalized, or unit, vectors
- This means that the magnitude of the vector is 1:

- We can normalize a vector by dividing the vector by its magnitude:

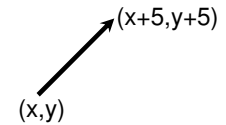
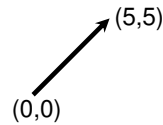
- N.B. The '^' denotes a normalized vector

$$\hat{V} = \frac{V}{\|V\|}$$

Question

- Are these two vectors the same?

- $(x,y) \neq (0,0)$



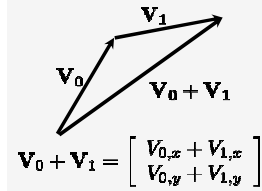
- A: Yes and no

- They are the same displacement vectors, which is what we will normally care about

Vector Addition

- Vectors are closed under addition
- Vector + Vector = Vector

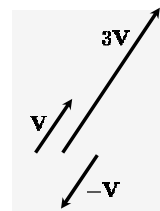
Vector Addition



$$V_0 + V_1 = \begin{bmatrix} V_{0,x} + V_{1,x} \\ V_{0,y} + V_{1,y} \end{bmatrix}$$

Vector Scaling

- Vectors are closed under multiplication with a scalar
- Scalar * Vector = Vector



Vector Scaling

$$aV = \begin{bmatrix} aV_x \\ aV_y \end{bmatrix}$$

Properties of Vector Addition & Scaling

Addition is Commutative

$$P + Q = Q + P$$

Addition is Associative

$$(P + Q) + R = P + (Q + R)$$

Scaling is Commutative and Associative

$$(ab)P = a(bP)$$

Scaling and Addition are Distributive

$$a(P + Q) = aP + aQ$$

$$(a + b)P = aP + bP$$

Points and Vectors

- Can define a vector by 2 points
 - Point - Point = Vector
- Can define a new point by a point and a vector
 - Point + Vector = Point

Linear Interpolation

- Can define a point in terms of 2 other points and a scalar
- Given points **P**, **R**, **Q** and a scalar **a**
 - $\mathbf{P} = a\mathbf{R} + (1 - a)\mathbf{Q}$
- How does this work?
 - It's really $\mathbf{P} = \mathbf{Q} + a\mathbf{V}$
 - $\mathbf{V} = \mathbf{R} - \mathbf{Q}$
 - Point + Vector = Point

Vector Multiplication?

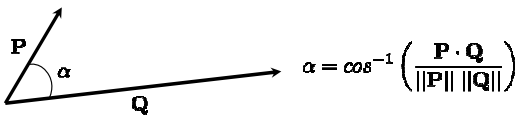
What does it mean to multiply two vectors?

- Not uniquely defined
- Two product operations are commonly used:
 - Dot (scalar, inner) product
 - Result is a scalar
 - Cross (vector, outer) product
 - Result is a new vector

Dot Product

$$\mathbf{P} \cdot \mathbf{Q} = \sum_{i=1}^n P_i Q_i = [P_1 \ P_2 \ \dots \ P_n] \begin{bmatrix} Q_1 \\ Q_2 \\ \dots \\ Q_n \end{bmatrix}$$

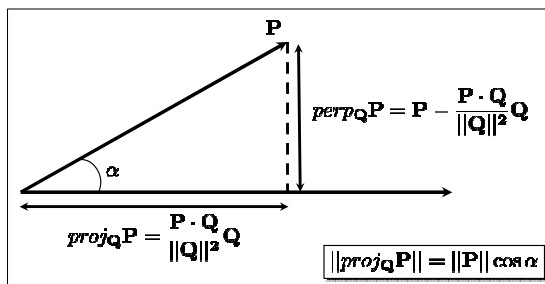
$$\mathbf{P} \cdot \mathbf{Q} = \|\mathbf{P}\| \|\mathbf{Q}\| \cos \alpha$$



Properties of Vector Dot Products

- | | |
|----------------------------|--|
| Commutative | $\mathbf{P} \cdot \mathbf{Q} = \mathbf{Q} \cdot \mathbf{P}$ |
| Associative with Scaling | $(a\mathbf{P}) \cdot \mathbf{Q} = a(\mathbf{P} \cdot \mathbf{Q})$ |
| Distributive with Addition | $\mathbf{P} \cdot (\mathbf{Q} + \mathbf{R}) = \mathbf{P} \cdot \mathbf{Q} + \mathbf{P} \cdot \mathbf{R}$ |
| | $\mathbf{P} \cdot \mathbf{P} = \ \mathbf{P}\ ^2$ |
| | $ \mathbf{P} \cdot \mathbf{Q} \leq \ \mathbf{P}\ \ \mathbf{Q}\ $ |

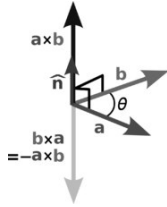
Perpendiculars and Projections



Dot Product Application: Lighting

- $\mathbf{P} \cdot \mathbf{Q} = \|\mathbf{P}\| \|\mathbf{Q}\| \cos a$
- So what does this mean if **P** and **Q** are normalized?
 - Can get $\cos a$ for just 3 multiplies and 2 adds
 - Very useful in lighting and shading calculations
 - Example: Lambert's cosine law

Cross Product



$$\mathbf{a} \times \mathbf{b} = \det \begin{bmatrix} \mathbf{i} & \mathbf{j} & \mathbf{k} \\ a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \end{bmatrix}$$

$$\mathbf{a} \times \mathbf{b} = \mathbf{i}(a_2b_3 - a_3b_2) - \mathbf{j}(a_1b_3 - a_3b_1) + \mathbf{k}(a_1b_2 - a_2b_1)$$

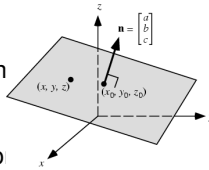
Cross Product

Application: Normals

- A normal (or surface normal) is a vector that is perpendicular to a surface at a given point
 - This is often used in lighting calculations
 - The cross product of 2 orthogonal vectors on the surface is a vector perpendicular to the surface
 - Can use the cross product to compute the normal

Planes

- How can we define a plane
 - 3 non-linear points
 - Use linear interpolation
 - A perpendicular vector and an incident point
 - $\mathbf{n} \cdot (\mathbf{x} - \mathbf{x}_0) = 0$
 - $ax + by + cz + d = 0$
 - Hessian normal form: Normalize \mathbf{n} first
 - $\mathbf{n} \cdot \mathbf{x} = -d$



Columns and Rows

- In this class, we will generally assume that a list forms a column vector:

$$(a, b, c, d) \Rightarrow \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix}$$

- The reason for this will become clear when we talk about matrices

Matrices

- Reminder: A matrix is a rectangular array of numbers
 - An $m \times n$ matrix has m rows and n columns
 - M_{ij} denotes the entry in the i -th row and j -th column of matrix M
 - These are generally thought of as 1-indexed (instead of 0-indexed)

Matrices

- Here, M is a 2×5 matrix:

$$\mathbf{M} = \begin{bmatrix} M_{11} & M_{12} & M_{13} & M_{14} & M_{15} \\ M_{21} & M_{22} & M_{23} & M_{24} & M_{25} \end{bmatrix}$$

Matrix Transposes

- The transpose of an $m \times n$ matrix is an $n \times m$ matrix
- Denoted M^T
- $M^T_{ij} = M_{ji}$

$$M^T = \begin{bmatrix} M_{11} & M_{12} & M_{13} & M_{14} & M_{15} \\ M_{21} & M_{22} & M_{23} & M_{24} & M_{25} \end{bmatrix}^T = \begin{bmatrix} M_{11} & M_{21} \\ M_{12} & M_{22} \\ M_{13} & M_{23} \\ M_{14} & M_{24} \\ M_{15} & M_{25} \end{bmatrix}$$

Matrix Addition

- Only well defined if the dimensions of the 2 matrices are the same
- That is, $m_1 = m_2$ and $n_1 = n_2$
- Here, M and G are both 2×5

$$(M + G)_{ij} = M_{ij} + G_{ij}$$

$$M + G = \begin{bmatrix} M_{11} + G_{11} & M_{12} + G_{12} & M_{13} + G_{13} & M_{14} + G_{14} & M_{15} + G_{15} \\ M_{21} + G_{21} & M_{22} + G_{22} & M_{23} + G_{23} & M_{24} + G_{24} & M_{25} + G_{25} \end{bmatrix}$$

Matrix Scaling

- Just like vector scaling
- Matrix * Scalar = Matrix

$$(aM)_{ij} = aM_{ij}$$

$$aM = \begin{bmatrix} aM_{11} & aM_{12} & aM_{13} & aM_{14} & aM_{15} \\ aM_{21} & aM_{22} & aM_{23} & aM_{24} & aM_{25} \end{bmatrix}$$

Properties of Matrix Addition and Scaling

Addition is Commutative

$$\mathbf{F} + \mathbf{G} = \mathbf{G} + \mathbf{F}$$

Addition is Associative

$$(\mathbf{F} + \mathbf{G}) + \mathbf{H} = \mathbf{F} + (\mathbf{G} + \mathbf{H})$$

Scaling is Associative

$$a(b\mathbf{F}) = (ab)\mathbf{F}$$

Scaling and Addition are Distributive

$$a(\mathbf{F} + \mathbf{G}) = a\mathbf{F} + a\mathbf{G}$$

$$(a + b)\mathbf{F} = a\mathbf{F} + b\mathbf{F}$$

Matrix Multiplication

- Only well defined if the number of columns of the first matrix and the number of rows of the second matrix are the same
- Matrix * Matrix = Matrix
- i.e.* if F is $m \times n$, and G is $n \times p$, then FG if $m \times p$
- Let's do an example

$$(FG)_{ij} = \sum_{k=1}^n F_{ik}G_{kj}$$

The Identity Matrix

- Defined such that the product of any matrix M and the identity matrix I is M
 - $\mathbf{IM} = \mathbf{MI} = \mathbf{M}$
- Let's derive it
- The identity matrix is a square matrix with ones on the diagonal and zeros elsewhere

The Identity Matrix

- Defined such that the product of any matrix M and the identity matrix I is M
 - $IM = MI = M$
- Let's derive it
- The identity matrix is a square matrix with ones on the diagonal and zeros elsewhere

$$(I_n)_{ij} = \begin{cases} 0 & i \neq j \\ 1 & i = j \end{cases}$$

$$I_3 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

Linear Systems, Matrix Inverses, etc.

- I'm not planning to cover this material in this course
- If there is any interest in going over this, let me know and I'll cover it on Tuesday

Next Time

- Going to cover coordinate systems and transforms, focusing on 2D