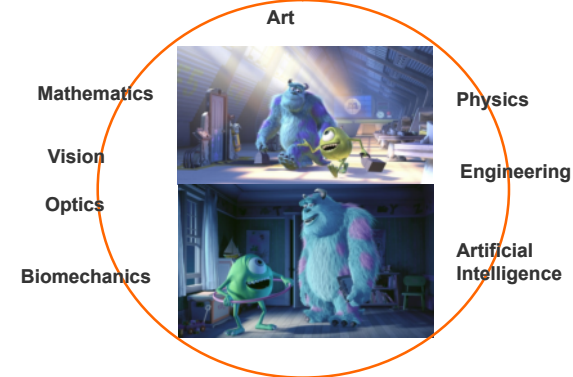


CS174A – Lecture 2

Linear Algebra:
The Algebra of Vectors and Matrices
(and Scalars)

Last Lecture: Computer Graphics

The Art and Science of creating imagery by computer



Why Study 3D Computer Graphics?

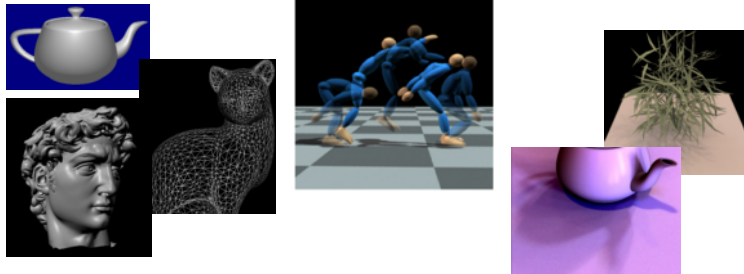
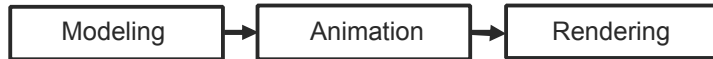
- CG has applications in many industries:
 - entertainment, scientific visualization, education
- Intellectually Challenging
 - Create and interact with realistic virtual world
 - Requires understanding of all aspects of physical world
 - New computing methods, displays, technologies
- Technically Challenging
 - Math of (perspective) projections, curves, surfaces
 - Physics of lighting and shading
 - 3D graphics software programming and hardware

Course Goals

- **Systems:** Write complex 3D graphics programs (real-time scene in OpenGL, offline raytracer)
- **Theory:** Mathematical aspects and algorithms underlying modern 3D graphics systems
- This course is **not** about the specifics of 3D graphics programs and APIs like Maya, Alias, DirectX but about the concepts underlying them.

Course Outline

3D Graphics Pipeline



Sub-areas of CG

Modeling

- How do we model (mathematically represent) objects?
- How do we construct models of specific objects?

Animation

- How do we represent the motions of objects?
- How do we give animators control of this motion?

Rendering

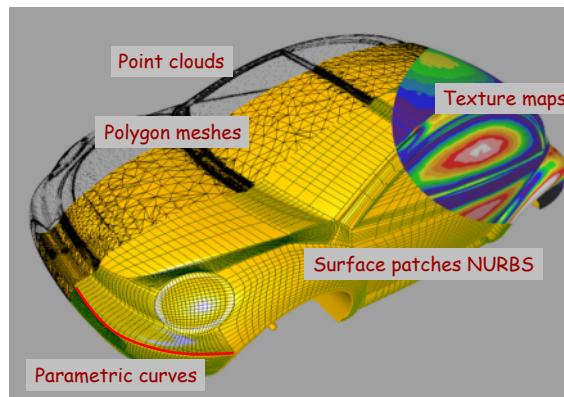
- How do we simulate the realworld behavior of light?
- How do we simulate the formation of images?

Interaction

- How do we enable humans and computers to interact?
- How do we design human+computer interfaces?

Modeling

Representing objects geometrically on a computer



Modeling

Primitives

- 3D points
- 3D lines and curves
- surfaces (BREPs): polygons, patches
- volumetric representations
- image-based representations

Attributes

- Color, texture maps
- Lighting properties

Geometric transformations



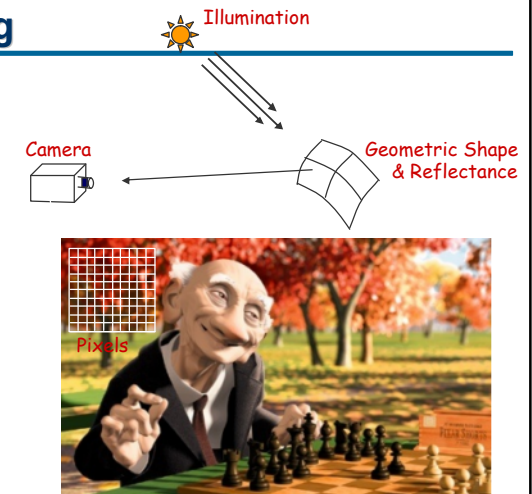
Animation

- Keyframe animation
- Motion capture
- Physics-based animation
- Autonomous motion planning



Rendering

Key elements



Linear Algebra Review: The Algebra of Vectors and Matrices (and Scalars)

Motivation

- Many graphics concepts need basic math, such as linear algebra
 - Vectors (dot products, cross products,...)
 - Matrices (matrix-matrix, matrix-vector mult., ...)
- Operations like translation, or rotation of the points that form an object are most efficiently performed as a matrix-vector multiply
- Thus, we will go out of our way to write everything as a matrix-vector multiply
- Chapters 2.4 (vectors) and 5.2 (matrices)
 - *Fundamentals of Computer Graphics* by Shirley and Marschner
 - Worthwhile to read all of chapters 2 and 5
- Should be refresher on very basic material for most of you

Lecture 2 - Outline

- Vectors,
 - Vector Basics
 - Vector Addition, Subtraction
 - Vector Multiplication
 - Dot product (2.4.3)
 - Cross product (2.4.4)
 - Constructing Coordinate Frames (2.4.5,6)
- Matrix (5.2)
 - Matrix Basics
 - Matrix Addition
 - Matrix Multiplication
 - Matrix Properties
 - Vector Multiplication in Matrix Form
- Points vs Vectors
- Lines and Planes

Notation

- Scalar
- Vector
- Matrix
- Tensor

$$a$$

$$\mathbf{a} = \begin{bmatrix} a_1 \\ \vdots \\ a_N \end{bmatrix}$$

$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1N} \\ a_{21} & & \ddots & \\ & & & a_{NN} \end{bmatrix}$$

$$\mathcal{A} = \begin{bmatrix} a_{111} & a_{112} & \cdots & a_{11N} \\ a_{211} & & \ddots & \\ & & & a_{NNN} \end{bmatrix}$$

Vectors

N-tuple of scalar elements

$$\mathbf{a} = \begin{bmatrix} a_1 \\ \vdots \\ a_n \\ \vdots \\ a_N \end{bmatrix}$$

Vector:
Bold lower-case

$a_n \in \mathbb{R}, \quad 1 \leq n \leq N$

Scalar:
Italic lower-case

- **Length and direction. Absolute position is not important**
- **Use to store offsets, displacements**
 - Note: Positions/locations are also represented by N-tuples, but strictly speaking, positions are not vectors and cannot be added. A location implicitly involves an origin, while an offset does not.

Column vs. Row Vectors

- vector: $\mathbf{a} = \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix}$ $\mathbf{a} = \begin{bmatrix} a_1 & a_2 & \cdots & a_n \end{bmatrix}^T$

Note: This is a column

- row vector: $\mathbf{a}^T = \begin{bmatrix} a_1 & a_2 & \cdots & a_n \end{bmatrix}$

(denoted by a vector symbol with a transpose)

Vectors

N-tuple: $\mathbf{x} = [x_1 \ \cdots \ x_n \ \cdots \ x_N]^T \quad x_n \in \mathfrak{R}$

Magnitude: $|\mathbf{x}| = \sqrt{x_1^2 + x_2^2 + \cdots + x_N^2}$

Unit vectors $\hat{\mathbf{x}} : |\hat{\mathbf{x}}| = 1$

Normalizing a vector $\hat{\mathbf{x}} = \frac{\mathbf{x}}{|\mathbf{x}|}$

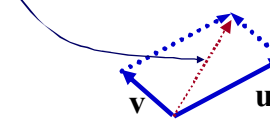
Vector-Vector Addition

- Parallelogram rule
 - tail to head, complete the triangle

geometric interpretation:
parallelogram rule

algebraic:
add coordinates

$$\mathbf{u} + \mathbf{v} = \mathbf{v} + \mathbf{u}$$



$$\mathbf{u} + \mathbf{v} = \begin{bmatrix} u_1 + v_1 \\ u_2 + v_2 \\ u_3 + v_3 \end{bmatrix}$$

examples:

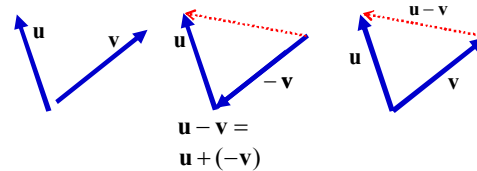
$$\begin{bmatrix} 3 \\ 6 \end{bmatrix} + \begin{bmatrix} 6 \\ 4 \end{bmatrix} =$$

$$\begin{bmatrix} 2 \\ 5 \\ 1 \end{bmatrix} + \begin{bmatrix} 3 \\ 1 \\ -1 \end{bmatrix} =$$

Vector-Vector Subtraction

- Adding a negatively scaled vector

$$\mathbf{u} - \mathbf{v} = \mathbf{u} + (-\mathbf{v}) = \begin{bmatrix} u_1 - v_1 \\ u_2 - v_2 \\ u_3 - v_3 \end{bmatrix}$$



- Example

$$\begin{bmatrix} 3 \\ 6 \end{bmatrix} - \begin{bmatrix} 6 \\ 4 \end{bmatrix} =$$

$$\begin{bmatrix} 2 \\ 5 \\ 1 \end{bmatrix} - \begin{bmatrix} 3 \\ 1 \\ -1 \end{bmatrix} =$$

Operations with Vectors

- Addition $\mathbf{x} + \mathbf{y} = [x_1 + y_1 \ \cdots \ x_n + y_n \ \cdots \ x_N + y_N]^T \quad 1 \leq n \leq N$

- Multiplication with scalar (scaling)

$$a\mathbf{x} = [ax_1 \ \cdots \ ax_n \ \cdots \ ax_N]^T \quad a, x_n \in \mathfrak{R}$$

- Properties

$$\mathbf{u} + \mathbf{v} = \mathbf{v} + \mathbf{u}$$

commutative

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

associative

$$a(\mathbf{u} + \mathbf{v}) = a\mathbf{u} + a\mathbf{v}, \quad a \in \mathfrak{R}$$

distributive

$$\mathbf{u} - \mathbf{u} = \mathbf{0}$$

inverse

Examples:

Linear Combination of Vectors

■ Definition

A linear combination of the m vectors $\mathbf{v}_1, \dots, \mathbf{v}_m$ is a vector of the form:

$$\mathbf{w} = a_1 \mathbf{v}_1 + \dots + a_m \mathbf{v}_m, \quad a_1, \dots, a_m \text{ in } \mathbb{R}$$

Special Cases

■ Linear combination

$$\mathbf{w} = a_1 \mathbf{v}_1 + \dots + a_m \mathbf{v}_m, \quad a_1, \dots, a_m \text{ in } \mathbb{R}$$

■ Affine combination:

■ A linear combination of vectors for which $a_1 + \dots + a_m = 1$

■ Convex combination

■ An affine combination for which $a_i \geq 0$ for $i = 1, \dots, m$

Linear Independence

■ Definition:

- Vectors $\mathbf{v}_1, \dots, \mathbf{v}_m$ are called *linearly independent*, if $a_1 \mathbf{v}_1 + \dots + a_m \mathbf{v}_m = \mathbf{0}$ if and only if $a_1 = a_2 = \dots = a_m = 0$

Vector Multiplication

■ Dot product

■ Cross product

■ Orthonormal bases and coordinate frames

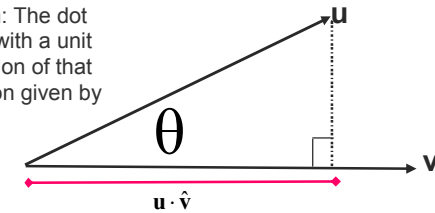
Dot Product

Motivation:

- Find angle between two vectors
 - e.g. cosine of angle between light source and surface for shading
- Finding projection of one vector on another
 - e.g. coordinates of a point in various coordinate frameworks -- object coordinate framework, world coordinate framework, camera coordinate framework, image coordinate framework
- Advantage: computed easily in cartesian components

Dot (Scalar) Product – Angle Computation

- Geometric definition: The dot product of a vector with a unit vector is the projection of that vector in the direction given by the unit vector.



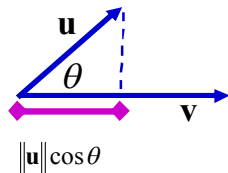
$$\mathbf{u} \cdot \frac{\mathbf{v}}{\|\mathbf{v}\|} = \|\mathbf{u}\| \cos(\theta)$$

$$\theta = \cos^{-1} \left(\frac{\mathbf{u} \cdot \mathbf{v}}{\|\mathbf{u}\| \|\mathbf{v}\|} \right)$$

$$\mathbf{u} \cdot \mathbf{v} = \|\mathbf{u}\| \|\mathbf{v}\| \cos(\theta)$$

Dot Product Geometry – Projection (of u on v)

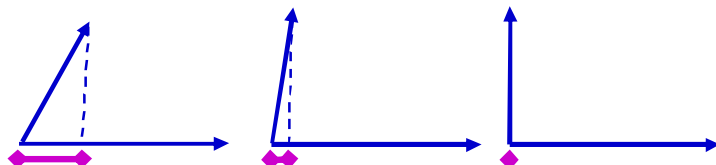
- The dot product of a vector with a unit vector is the projection of that vector in the direction given by the unit vector.



$$\mathbf{u} \cdot \frac{\mathbf{v}}{\|\mathbf{v}\|} = \|\mathbf{u}\| \cos \theta$$

$$\mathbf{u} \cdot \mathbf{v} = \|\mathbf{u}\| \|\mathbf{v}\| \cos \theta$$

- as vectors become perpendicular, $\mathbf{u} \cdot \mathbf{v} = \|\mathbf{u}\| \|\mathbf{v}\| \cos \theta \rightarrow 0$



Dot (Scalar) Product

- Linear algebra definition of the dot product, also known as the scalar product:

$$\begin{bmatrix} u_1 & u_2 & u_3 \end{bmatrix} \cdot \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix} =$$

$$\begin{bmatrix} 6 \\ 1 \\ 2 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 7 \\ 3 \end{bmatrix} =$$

Dot (Scalar) Product

- multiply: vector * vector = scalar
- dot product, aka inner product

$$\mathbf{u} \bullet \mathbf{v}$$

$$\begin{bmatrix} u_1 & & \\ & v_1 & \\ & & \end{bmatrix}$$

$$\begin{bmatrix} u_2 & & \\ & v_2 & \\ & & \end{bmatrix} = (u_1 * v_1) + (u_2 * v_2) + (u_3 * v_3)$$

$$\begin{bmatrix} u_3 & & \\ & v_3 & \\ & & \end{bmatrix}$$

$$\begin{bmatrix} 6 \\ 1 \\ 2 \end{bmatrix} \bullet \begin{bmatrix} 1 \\ 7 \\ 3 \end{bmatrix} = (6*1) + (1*7) + (2*3) = 6 + 7 + 6 = 19$$

Dot Product

Definition: $\mathbf{a}, \mathbf{b} \in \mathbb{R}^N$

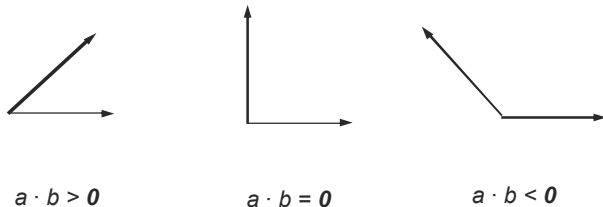
$$\mathbf{a} \cdot \mathbf{b} = \sum_{n=1}^N a_n b_n = \|\mathbf{a}\| \|\mathbf{b}\| \cos \theta$$

Properties

1. Symmetry: $\mathbf{a} \cdot \mathbf{b} = \mathbf{b} \cdot \mathbf{a}$
2. Linearity: $(\mathbf{a} + \mathbf{b}) \cdot \mathbf{c} = \mathbf{a} \cdot \mathbf{c} + \mathbf{b} \cdot \mathbf{c}$
3. Homogeneity: $(s\mathbf{a}) \cdot \mathbf{b} = s(\mathbf{a} \cdot \mathbf{b})$
4. $\|\mathbf{b}\|^2 = \mathbf{b} \cdot \mathbf{b}$
5. $\mathbf{a} \cdot \mathbf{b} = \|\mathbf{a}\| \|\mathbf{b}\| \cos(\theta)$

Dot Product and Perpendicularity

From Property 5. $\mathbf{a} \cdot \mathbf{b} = \|\mathbf{a}\| \|\mathbf{b}\| \cos(\theta)$



Perpendicular Vectors

The dot product can be employed to determine if two vectors are perpendicular.

- Vectors \mathbf{a} and \mathbf{b} are perpendicular iff $\mathbf{a} \cdot \mathbf{b} = 0$
- Also called orthogonal vectors

It is easy to see that the standard unit vectors form an orthogonal basis:

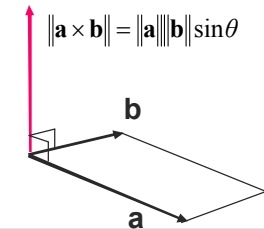
$$\mathbf{i} \cdot \mathbf{j} = 0, \quad \mathbf{j} \cdot \mathbf{k} = 0, \quad \mathbf{i} \cdot \mathbf{k} = 0$$

Vector Multiplication

- Dot product
- **Cross product**
- Orthonormal bases and coordinate frames

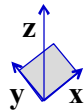
Cross (vector) product

- Motivation: Useful in constructing coordinate systems (later)
 - Problem: Construct a orthonormal camera coordinate frame into which to transform world objects, given \mathbf{a} (viewing direction), and a second vector \mathbf{b} (up direction of camera)
- Cross product is a vector orthogonal to the two initial vectors
- Direction determined by *right-hand rule*



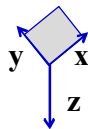
RHS vs. LHS Coordinate Systems

- right-handed coordinate system **convention**



right hand rule:
index finger x, second finger y;
right thumb points up
 $\mathbf{z} = \mathbf{x} \times \mathbf{y}$

- left-handed coordinate system



left hand rule:
index finger x, second finger y;
left thumb points down
 $\mathbf{z} = \mathbf{x} \times \mathbf{y}$

Cross Product

Defined only for 3D vectors and with respect to the standard unit vectors

Algebraic Definition:

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

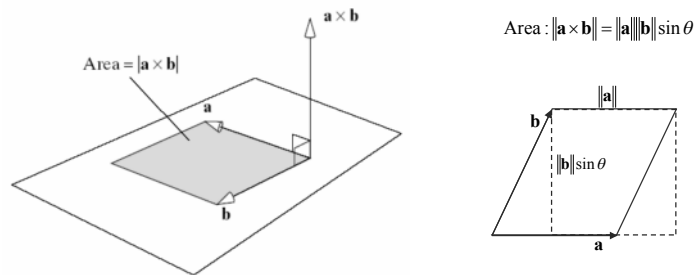
$$= (a_2 b_3 - a_3 b_2) \mathbf{i} + (a_3 b_1 - a_1 b_3) \mathbf{j} + (a_1 b_2 - a_2 b_1) \mathbf{k}$$

Cross Product

Geometric Definition: $\|\mathbf{a} \times \mathbf{b}\| = \|\mathbf{a}\| \|\mathbf{b}\| \sin \theta$

$\|\mathbf{a} \times \mathbf{b}\|$ parallelogram area

$\mathbf{a} \times \mathbf{b}$ perpendicular to parallelogram



Properties of the Cross Product

1. $\mathbf{i} \times \mathbf{j} = \mathbf{k}$, $\mathbf{i} \times \mathbf{k} = -\mathbf{j}$, $\mathbf{j} \times \mathbf{k} = \mathbf{i}$
2. Antisymmetry: $\mathbf{a} \times \mathbf{b} = -\mathbf{b} \times \mathbf{a}$
3. Linearity: $\mathbf{a} \times (\mathbf{b} + \mathbf{c}) = \mathbf{a} \times \mathbf{b} + \mathbf{a} \times \mathbf{c}$
4. Homogeneity: $(s\mathbf{a}) \times \mathbf{b} = s(\mathbf{a} \times \mathbf{b})$
5. The cross product is normal to both vectors: $(\mathbf{a} \times \mathbf{b}) \cdot \mathbf{a} = (\mathbf{a} \times \mathbf{b}) \cdot \mathbf{b} = 0$
6. $|\mathbf{a} \times \mathbf{b}| = |\mathbf{a}| |\mathbf{b}| \sin(\theta)$

Vector Multiplication

- Dot product
- Cross product
- **Orthonormal bases and coordinate frames**

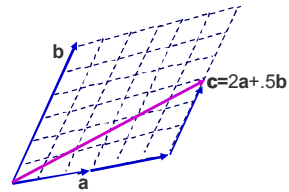
Orthonormal bases/coordinate frames

- Important for representing points, locations
- Often, many sets of coordinate systems
 - Global, local, world, model, parts of model (head, hands, ...)
- Critical issue is transforming between these systems/bases
 - Topic of next 3 lectures

Basis Vectors

- Given any two vectors that are **linearly independent** (nonzero and nonparallel)
 - Can define any vector in the plane as a linear combination of these two vectors

$$\mathbf{c} = w_1 \mathbf{a} + w_2 \mathbf{b}$$



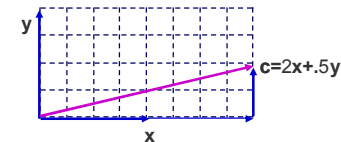
Orthonormal Basis Vectors

- If the basis vectors are **orthonormal** (perpendicular/orthogonal and unit length)

$$\mathbf{x} \bullet \mathbf{y} = 0$$

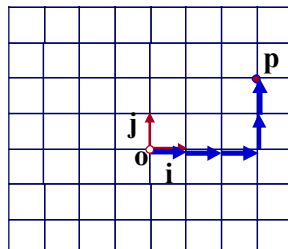
$$\|\mathbf{x}\| = \|\mathbf{y}\| = 1$$

- we have Cartesian coordinate system
- familiar Pythagorean definition of distance



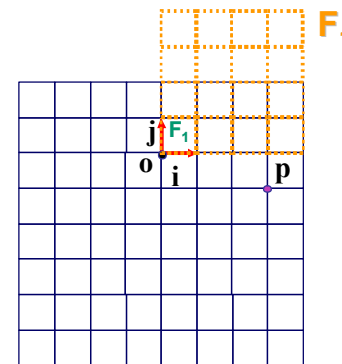
Basis Vectors and Origins

- Cordinate system**: just basis vectors
 - can only specify offset: vectors
- Coordinate frame**: basis vectors and origin
 - can specify location as well as offset: points & vectors



$$\mathbf{p} = \mathbf{o} + x\mathbf{i} + y\mathbf{j}$$

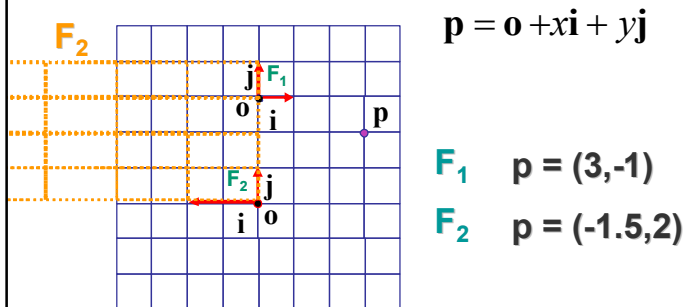
Working with Frames



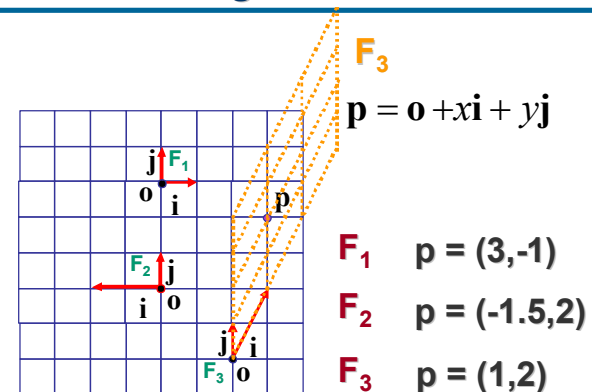
$$\mathbf{p} = \mathbf{o} + x\mathbf{i} + y\mathbf{j}$$

$$\mathbf{F}_1 \quad \mathbf{p} = (3, -1)$$

Working with Frames



Working with Frames



Generators and Base Vectors

How many vectors are needed to generate a vector space?

- Given a vector space \mathbf{R}^n we can prove that we need minimum n vectors to generate all vectors \mathbf{v} in \mathbf{R}^n
- Any set of vectors that generate a vector space is called a **generator set**
- A generator set with minimum size is called a **basis** for the given vector space

Standard Unit Vectors

For any vector space \mathbf{R}^n :

$$\begin{aligned} \mathbf{i}_1 &= (1, 0, 0, \dots, 0, 0)^T \\ \mathbf{i}_2 &= (0, 1, 0, \dots, 0, 0)^T \\ &\dots \\ \mathbf{i}_n &= (0, 0, 0, \dots, 0, 1)^T \end{aligned}$$

The elements of a vector \mathbf{v} in \mathbf{R}^n are the scalar coefficients of the linear combination of the basis vectors

Standard Unit Vectors

$$\mathbf{v} = (x_1, \dots, x_n)^T, \quad x_i \in \mathbb{R}$$

$$\begin{aligned}(x_1, x_2, \dots, x_n)^T &= x_1(1, 0, 0, \dots, 0, 0)^T \\ &\quad + x_2(0, 1, 0, \dots, 0, 0)^T \\ &\quad \dots \\ &\quad + x_n(0, 0, 0, \dots, 0, 1)^T\end{aligned}$$

Representation of Vectors Through Basis Vectors

Given a vector space \mathbb{R}^n , a set of basis vectors \mathbf{B} $\{\mathbf{b}_i \text{ in } \mathbb{R}^n, i=1, \dots, n\}$ and a vector \mathbf{v} in \mathbb{R}^n we can always find scalar coefficients such that:

$$\mathbf{v} = a_1\mathbf{b}_1 + \dots + a_n\mathbf{b}_n$$

So, vector \mathbf{v} expressed with respect to B is:

$$\mathbf{v}_B = (a_1, \dots, a_n)$$

Outline

- Vectors,
 - Vector Basics
 - Vector Addition, Subtraction
 - Vector Multiplication
 - Dot product
 - Cross product
 - Coordinate Frames
- Matrix
 - Matrix Basics
 - Matrix Addition
 - Matrix Multiplication
 - Matrix Properties
 - Vector Multiplication in Matrix Form
- Points vs Vectors
- Lines and Planes

Matrices

- Can be used to transform points (vectors)
 - Translation, rotation, shear, scale (more detail next lecture)
- Section 5.2.1 and 5.2.2 of text
 - Instructive to read all of 5 but not that relevant to course

What is a Matrix

- It is a two-way array ($m \times n$ = m rows, n columns) that maps one vector space into another

$$\begin{pmatrix} 13 \\ 52 \\ 04 \end{pmatrix}$$

- Can be used to transform points (vectors)
 - Translation, rotation, shear, scale (more detail next lecture)

Matrix-Matrix Addition

- add: matrix + matrix = matrix

$$\begin{bmatrix} m_{11} & m_{12} \\ m_{21} & m_{22} \end{bmatrix} + \begin{bmatrix} n_{11} & n_{12} \\ n_{21} & n_{22} \end{bmatrix} = \begin{bmatrix} n_{11} + m_{11} & n_{12} + m_{12} \\ n_{21} + m_{21} & n_{22} + m_{22} \end{bmatrix}$$

- example

$$\begin{bmatrix} 1 & 3 \\ 2 & 4 \end{bmatrix} + \begin{bmatrix} -2 & 5 \\ 7 & 1 \end{bmatrix} =$$

Matrix Addition

Addition:

$$\mathbf{A}_{m \times n} + \mathbf{B}_{m \times n} = (a_{ij} + b_{ij})$$

Properties:

- $\mathbf{A} + \mathbf{B} = \mathbf{B} + \mathbf{A}$
- $\mathbf{A} + (\mathbf{B} + \mathbf{C}) = (\mathbf{A} + \mathbf{B}) + \mathbf{C}$
- $f(\mathbf{A} + \mathbf{B}) = f\mathbf{A} + f\mathbf{B}$
- Transpose: $\mathbf{A}^T = (a_{ij})^T = (a_{ji})$

Scalar-Matrix Multiplication

- multiply: scalar * matrix = matrix

$$a \begin{bmatrix} m_{11} & m_{12} \\ m_{21} & m_{22} \end{bmatrix} = \begin{bmatrix} a * m_{11} & a * m_{12} \\ a * m_{21} & a * m_{22} \end{bmatrix}$$

- example

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

Matrix-Matrix Multiplication

- row by column

$$\begin{bmatrix} m_{11} & m_{12} \\ m_{21} & m_{22} \end{bmatrix} \begin{bmatrix} n_{11} & n_{12} \\ n_{21} & n_{22} \end{bmatrix} = \begin{bmatrix} p_{11} & p_{12} \\ p_{21} & p_{22} \end{bmatrix}$$

$$p_{11} = m_{11}n_{11} + m_{12}n_{21}$$

Matrix-Matrix Multiplication

- row by column

$$\begin{bmatrix} m_{11} & m_{12} \\ m_{21} & m_{22} \end{bmatrix} \begin{bmatrix} n_{11} & n_{12} \\ n_{21} & n_{22} \end{bmatrix} = \begin{bmatrix} p_{11} & p_{12} \\ p_{21} & p_{22} \end{bmatrix}$$

$$p_{11} = m_{11}n_{11} + m_{12}n_{21}$$

$$p_{21} = m_{21}n_{11} + m_{22}n_{21}$$

$$p_{12} = m_{11}n_{12} + m_{12}n_{22}$$

$$p_{22} = m_{21}n_{12} + m_{22}n_{22}$$

Matrix-Matrix Multiplication

- row by column

$$\begin{bmatrix} m_{11} & m_{12} \\ m_{21} & m_{22} \end{bmatrix} \begin{bmatrix} n_{11} & n_{12} \\ n_{21} & n_{22} \end{bmatrix} = \begin{bmatrix} p_{11} & p_{12} \\ p_{21} & p_{22} \end{bmatrix}$$

$$p_{11} = m_{11}n_{11} + m_{12}n_{21}$$

$$p_{21} = m_{21}n_{11} + m_{22}n_{21}$$

$$p_{12} = m_{11}n_{12} + m_{12}n_{22}$$

Matrix-Matrix Multiplication

- row by column

$$\begin{bmatrix} m_{11} & m_{12} \\ m_{21} & m_{22} \end{bmatrix} \begin{bmatrix} n_{11} & n_{12} \\ n_{21} & n_{22} \end{bmatrix} = \begin{bmatrix} p_{11} & p_{12} \\ p_{21} & p_{22} \end{bmatrix}$$

$$p_{11} = m_{11}n_{11} + m_{12}n_{21}$$

$$p_{21} = m_{21}n_{11} + m_{22}n_{21}$$

$$p_{12} = m_{11}n_{12} + m_{12}n_{22}$$

$$p_{22} = m_{21}n_{12} + m_{22}n_{22}$$

Matrix-Matrix Multiplication

- Number of columns in first must = rows in second

$$\begin{pmatrix} 1 & 3 \\ 2 & 2 \\ 0 & 4 \end{pmatrix} \begin{pmatrix} 3 & 6 & 9 & 4 \\ 2 & 7 & 8 & 3 \end{pmatrix} = \begin{pmatrix} 9 & 27 & 33 & 13 \\ 19 & 44 & 61 & 26 \\ 8 & 28 & 32 & 12 \end{pmatrix}$$

- Element (i,j) in product is the matrix product of row i of first matrix and column j of second matrix

$$\begin{pmatrix} \mathbf{a}_1^T \\ \mathbf{a}_2^T \\ \mathbf{a}_3^T \end{pmatrix} \begin{pmatrix} \mathbf{b}_1 & \mathbf{b}_2 & \mathbf{b}_3 & \mathbf{b}_4 \end{pmatrix} = \begin{pmatrix} \mathbf{a}_1^T \mathbf{b}_1 & \mathbf{a}_1^T \mathbf{b}_2 & \mathbf{a}_1^T \mathbf{b}_3 & \mathbf{a}_1^T \mathbf{b}_4 \\ \mathbf{a}_2^T \mathbf{b}_1 & \mathbf{a}_2^T \mathbf{b}_2 & \mathbf{a}_2^T \mathbf{b}_3 & \mathbf{a}_2^T \mathbf{b}_4 \\ \mathbf{a}_3^T \mathbf{b}_1 & \mathbf{a}_3^T \mathbf{b}_2 & \mathbf{a}_3^T \mathbf{b}_3 & \mathbf{a}_3^T \mathbf{b}_4 \end{pmatrix}$$

Dot Product in Matrix Form

*A vector is a column matrix

$$\begin{aligned} \mathbf{a} \cdot \mathbf{b} &= \sum_{n=1}^N a_n b_n \\ &= a_1 b_1 + a_2 b_2 + a_3 b_3 \\ &= (a_1, a_2, a_3) \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix} \\ &= \mathbf{a}^T \mathbf{b} \end{aligned}$$

Cross Product in Matrix Form

$$\begin{aligned} \mathbf{a} \times \mathbf{b} &= \begin{vmatrix} \mathbf{i} & \mathbf{j} & \mathbf{k} \\ a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \end{vmatrix} = \begin{bmatrix} a_2 b_3 - a_3 b_2 \\ a_3 b_1 - a_1 b_3 \\ a_1 b_2 - a_2 b_1 \end{bmatrix} \\ &= \begin{bmatrix} 0 & -a_3 & a_2 \\ a_3 & 0 & -a_1 \\ -a_2 & a_1 & 0 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} \\ &= \mathbf{A} \mathbf{b} \end{aligned}$$

Rewrite

\mathbf{A} is the dual matrix of vector \mathbf{a}

Summary: Multiplication with Matrices

Definition:

$$\begin{aligned} \mathbf{C}_{m \times r} &= \mathbf{A}_{m \times n} \mathbf{B}_{n \times r} \\ (\mathbf{C}_{ij}) &= \left(\sum_{k=1}^n a_{ik} b_{kj} \right) \end{aligned}$$

Properties:

- $\mathbf{AB} \neq \mathbf{BA}$
- $\mathbf{A}(\mathbf{BC}) = (\mathbf{AB})\mathbf{C}$
- $f(\mathbf{AB}) = (f\mathbf{A})\mathbf{B}$
- $\mathbf{A}(\mathbf{B} + \mathbf{C}) = \mathbf{AB} + \mathbf{AC}$,
 $(\mathbf{B} + \mathbf{C})\mathbf{A} = \mathbf{BA} + \mathbf{CA}$
- $(\mathbf{AB})^T = \mathbf{B}^T \mathbf{A}^T$
- $(\mathbf{AB})^{-1} = \mathbf{B}^{-1} \mathbf{A}^{-1}$

Outline

- Vectors,
 - Vector Basics
 - Vector Addition, Subtraction
 - Vector Multiplication
 - Dot product
 - Cross product
 - Constructing Coordinate Frames
- Matrix
 - Matrix Basics
 - Matrix Addition
 - Matrix Multiplication
 - Matrix Properties
 - Vector Multiplication in Matrix Form
- Points vs Vectors
- Lines and Planes

Homogeneous Coordinates

Vectors and Points are represented as column matrices

$$\mathbf{v} = \begin{pmatrix} v_1 \\ v_2 \\ v_3 \\ 0 \end{pmatrix} \quad P = \begin{pmatrix} p_1 \\ p_2 \\ p_3 \\ 1 \end{pmatrix}$$

Lines and Planes

In addition to vectors and points, lines and planes are fundamental geometric entities in computer graphics

- Recall how we represent them mathematically...

Lines

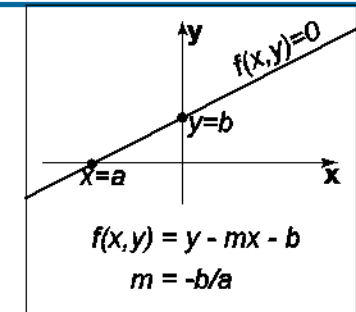
- **Explicit**
 - slope-intercept form

$$y = mx + b$$

- **Implicit**

$$\begin{aligned} y - mx - b &= 0 \\ Ax + By + C &= 0 \\ f(x, y) &= 0 \end{aligned}$$

- find where function is 0
- plug in (x,y), check if
 - 0: on line
 - < 0: inside
 - > 0: outside



2D Parametric Lines

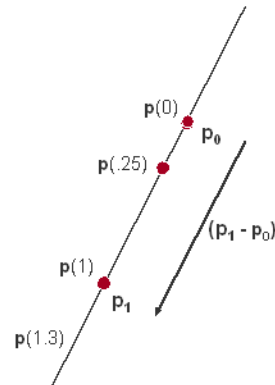
- Start at point p_0 , go towards p_1 , according to parameter t

- $p(0) = p_0, p(1) = p_1$

- $p(t) = p_0 + t(p_1 - p_0)$

- $p(t) = o + t(d)$

- $$\begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} x_0 + t(x_1 - x_0) \\ y_0 + t(y_1 - y_0) \end{bmatrix}$$



Summary: Lines

Representations of a line (in 2D)

- Explicit - we have one dependent variable on the left-hand side of an equation, and all the independent variables and constants on the right-hand side of the equation

$$y = \frac{dy}{dx}(x - x_0) + y_0$$

- Implicit - An implicit equation is an algebraic line/curve formed by points that satisfy an equation. Implicits TEST their (x, y) input values and return true if the implicit relationship is satisfied. The general solution of implicit equations requires a search!

$$F(x, y) = (x - x_0)dy - (y - y_0)dx$$

if $F(x, y) = 0$ then (x, y) is on line
 $F(x, y) > 0$ (x, y) is below line
 $F(x, y) < 0$ (x, y) is above line

- Parametric - also called generating functions, they generate coordinate pairs as output, given parameter values as input.

$$x(t) = x_0 + t(x_1 - x_0)$$

$$y(t) = y_0 + t(y_1 - y_0)$$

$$t \in [0, 1]$$

$$P(t) = P_0 + t(P_1 - P_0), \text{ or}$$

$$P(t) = (1 - t)P_0 + tP_1$$

Planes

Plane equations

Implicit $F(x, y, z) = Ax + By + Cz + D = 0 = N \cdot P + D$
 Points on Plane $F(x, y, z) = 0$

Explicit $z = -(A/C)x - (B/C)y - D/C, C \neq 0$

Parametric

$$Plane(s, t) = P_0 + s(P_1 - P_0) + t(P_2 - P_0)$$

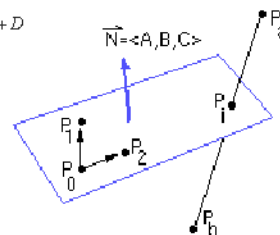
P_0, P_1, P_2 not collinear

or

$Plane(s, t) = P_0 + sV_1 + tV_2$ where V_1, V_2 are basis vectors

convex combination defines a triangle:

$$Plane(s, t) = (1 - s - t)P_0 + sP_1 + tP_2$$



Implicit Functions

- find where function is 0

- plug in (x, y) , check if

- 0: on line
 - < 0: inside
 - > 0: outside

- analogy: terrain

- sea level: $f=0$
 - altitude: function value
 - topo map: equal-value contours (level sets)

