

Data Structures & Algorithms for Geometry

⇒ Agenda:

- Introduce course
- Linear algebra primer
- Introduce bounding volumes
 - General BV characteristics
 - Axis-aligned bounding boxes

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- ⇒ Some knowledge of linear algebra / vector math.

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 - For most assignments you will need to implement classes that conform to a very specific interface.
- ⇒ Fundamental data structures
 - Most data structures for geometry are specialized versions of linked lists, binary trees, etc.
- ⇒ Some knowledge of linear algebra / vector math.
 - Can probably pick most of it up on the way, but be prepared to work a little harder.

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- ⇒ Creation and traversal of space partitions.
 - Doom (the *original*) made BSP trees popular...now you get to implement one.

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- ⇒ Representation and storage of polygons.
 - May seem trivial, but what if you want to operate on all the polygons that share a vertex?
- ⇒ Numerical stability and geometrical robustness.
 - Floating point math is inexact. This can cause problems. *Big* problems.

How will you be graded?

- ⇒ Bi-weekly quizzes worth 5 points each.
- ⇒ A final exam worth 50 points.
- ⇒ Bi-weekly programming assignments with 10 points each.
- ⇒ A term project worth 50 points.

How will programs be graded?

- ⇒ First and foremost, does the program produce the correct output?
- ⇒ Are appropriate algorithms and data-structures used?
- ⇒ Is the code readable and clear?

Linear algebra primer

⇒ Three important data types:

- Scalar values
- Row / column vectors
 - 1×4 and 4×1 are the sizes we'll most often encounter
- Square matrices
 - 4×4 is the size we'll most often encounter

Scalars

- ⇒ These are the numbers you know!
 - Example: 3.14, 5.0, 99.9, $\sqrt{2}$, etc.

Row vectors

- ⇒ These are special matrices that have multiple columns but only one row.
 - Example: $[5.0 \quad 3.14 \quad 37]$
- ⇒ Add and subtract the way you would expect.
 - Example: $[1 \quad 2 \quad 3] + [9 \quad 10 \quad 11] = [10 \quad 12 \quad 14]$
 - Both vectors *must* be the same size.
- ⇒ Operate with scalars the way you would expect.
 - Example: $3.2 \times [1 \quad 2 \quad 3] = [3.2 \quad 6.4 \quad 9.6]$
- ⇒ Notice that vector multiplication is missing...

Column vectors

⇒ These are special matrices that have multiple rows but only one column.

• Example: $\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$

⇒ Work just like row vectors.

Vector operations

- ⇒ There are only a few operations specific to vectors that are *really* important for us.

Dot Product

- ⇒ Noted as a “dot” between two vectors (e.g., $A \cdot B$)
- ⇒ Also known as “inner product.”
- ⇒ Multiply matching elements, sum all the results.
 - Example:

$$\begin{bmatrix} 2.3 & 1.2 \end{bmatrix} \cdot \begin{bmatrix} 1.7 & 6.5 \end{bmatrix} = (2.3 * 1.7) + (1.2 * 6.5) = 11.71$$

Why is the dot product so interesting?

- ⇒ In 3-space, the dot of two unit vectors is the cosine of the angle between the two vectors.
 - If the vectors are not already normalized (unit length), we can divide the dot product by the magnitudes.
 - Example:

$$\frac{a \cdot b}{|a||b|} = \cos \theta$$

Vector Magnitude

- ⇒ Noted by vertical bars, like absolute value.
- ⇒ Take the square root of the dot product of the vector with itself...like absolute value.
- ⇒ Result is the magnitude (a.k.a. length) of the vector.

- Example:
$$\left| \begin{bmatrix} \frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} \end{bmatrix} \right| = \sqrt{\begin{bmatrix} \frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} \end{bmatrix} \cdot \begin{bmatrix} \frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} \end{bmatrix}} =$$
$$\sqrt{\left(\frac{\sqrt{2}}{2}\right)^2 + \left(\frac{\sqrt{2}}{2}\right)^2} = \sqrt{\frac{2}{4} + \frac{2}{4}} = 1$$

Normalize

- ⇒ Noted by dividing a vector by its magnitude.
 - Example: $\frac{A}{|A|}$
- ⇒ Results in a vector with the same direction, but a magnitude of 1.0.
- ⇒ Works the same as with scalars.

Cross Product

- ⇒ Noted as an X between two vectors (e.g., $a \times b$)
- ⇒ Derivation of the cross product is not important.

The math is:

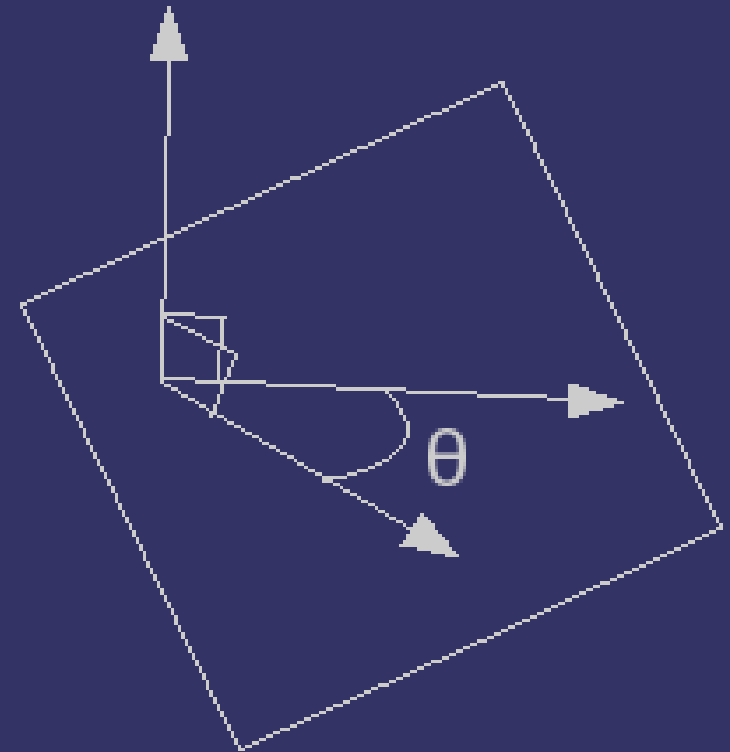
$$a \times b = \begin{bmatrix} a_y b_z - a_z b_y & a_z b_x - a_x b_z & a_x b_y - a_y b_x \end{bmatrix}$$

- ⇒ *Only* valid in 3-dimensions.

Why is the cross product so interesting?

- ⇒ Two really useful properties.
 - The result of the cross product between two vectors is a new vector that is perpendicular (also called normal) to both vectors.
 - If the source vectors are normalized:

$$|a \times b| = \sin \theta$$



Matrices

⇒ Like vectors, but have multiple rows and columns.

- Example:
$$\begin{bmatrix} 1.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 1.0 & 0.0 & 0.0 \\ 0.0 & 0.0 & 1.0 & 0.0 \\ 0.0 & 0.0 & 0.0 & 1.0 \end{bmatrix}$$

⇒ Add and subtract like you would expect.

- Like vectors, both matrices must be the same size...in both dimensions.

Matrix / vector multiplication

- ⇒ Special rules apply that make it different from scalar multiplication.
 - **Not** commutative! e.g., $M \times N \neq N \times M$
 - Is associative. e.g., $(NM)P = N(MP)$
 - Column count of first matrix must match row count of second matrix.
 - If M is a 4-by-3 matrix and N is a 3-by-1 matrix, we can do $M \times N$, but *not* $N \times M$.
 - If the source matrices are n-by-m and m-by-p, the resulting matrix will be n-by-p.

Matrix / vector multiplication (cont.)

- ⇒ To calculate an element of the matrix, C, resulting from AB:

$$C_{ij} = \sum_{r=1}^n a_{ir} b_{rj}$$

- ⇒ What does this look like?

Matrix / vector multiplication (cont.)

⇒ To calculate an element of the matrix, C, resulting from AB:

$$C_{ij} = \sum_{r=1}^n a_{ir} b_{rj}$$

⇒ What does this look like?

- The dot product of a row of A with a column of B.
- This is why the column count of A must match the row count of B...otherwise the dot product wouldn't work.

Transpose

- ⇒ Noted by a “T” in the exponent position (e.g., M^T).
- ⇒ The rows become the columns, and the columns become the rows.
 - Example:

$$\begin{bmatrix} 2 & 3 \\ 4 & 5 \\ 6 & 7 \end{bmatrix}^T = \begin{bmatrix} 2 & 4 & 6 \\ 3 & 5 & 7 \end{bmatrix}$$

References

http://en.wikipedia.org/wiki/Matrix_multiplication

http://en.wikipedia.org/wiki/Dot_product

http://en.wikipedia.org/wiki/Cross_product

Break

Bounding Volumes

⇒ From Wikipedia:

“...a bounding volume for a set of objects is a closed volume that completely contains the union of the objects in the set.”

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⇒ Why is this useful?

- We can represent complex geometry (a character model with 50,000 polygons) with a simplified approximation (a box with 6 polygons) that can be tested more quickly.
- Since the representation is inexact, so is the test result.

Desirable BV Characteristics

⇒ Inexpensive intersection tests

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 - The whole point of using a BV instead of the source geometry is to speed up rejection / acceptance tests between geometric objects.

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 - If the BV is a poor representation of the source geometry, tests between BVs will result in many false positives or false negatives.

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Desirable BV Characteristics

- ⇒ Inexpensive intersection tests
- ⇒ Tight fitting to source geometry
- ⇒ Inexpensive to compute
- ⇒ Easy to transform
 - Most interesting objects move. If the object moves, its BV needs to move with it. If it is too difficult / expensive to move, it may cancel out the speed-up.

Desirable BV Characteristics

- ⇒ Inexpensive intersection tests
- ⇒ Tight fitting to source geometry
- ⇒ Inexpensive to compute
- ⇒ Easy to transform
- ⇒ Inexpensive to store

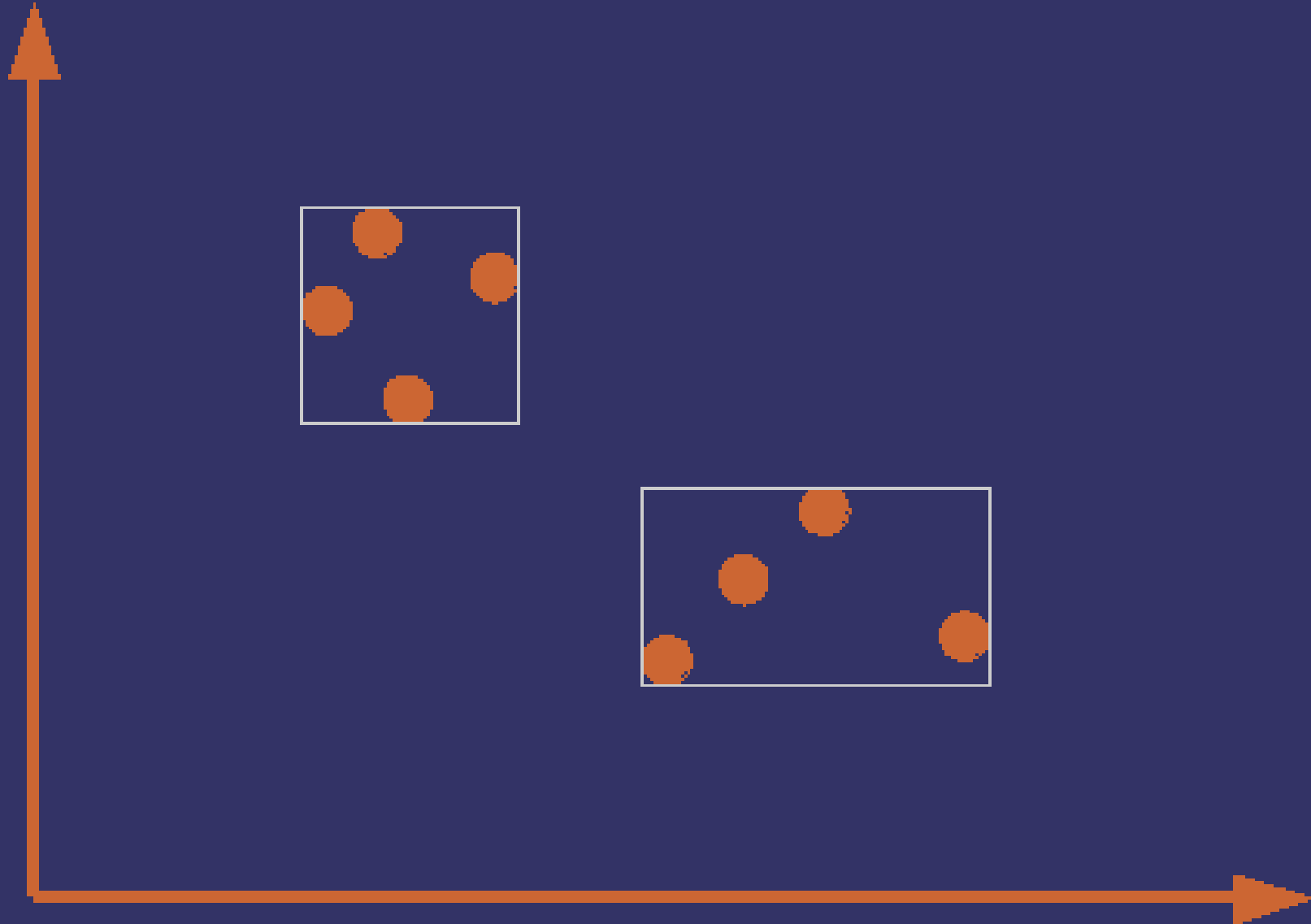
Desirable BV Characteristics

- ⇒ Inexpensive intersection tests
- ⇒ Tight fitting to source geometry
- ⇒ Inexpensive to compute
- ⇒ Easy to transform
- ⇒ Inexpensive to store
 - If the BV requires too much space to store or too much time to access, it can negatively impact performance.

Axis-aligned bounding box

- ⇒ One common BV is the *axis-aligned bounding box* (AABB).
- ⇒ Just an n-dimensional box whose sides are parallel to the axis and encloses all points.

AABB Example



AABB representation

- ⇒ Three common ways to represent an AABB
 - All are equivalent, and each can easily be converted to the other.
 - Depending on the data used to create the AABB, one representation may be easier to create than another.

Minimum / maximum point

```
class aabb_m_m {
    // Point such that for every
    // point P in the object:
    //     (min.x <= P.x <= max.x) &&
    //     (min.y <= P.y <= max.y) &&
    //     (min.z <= P.z <= max.z)
    point min;
    point max;
};
```

Min point / diameter

```
class aabb_m_d {  
    // Point such that for every  
    // point P in the object:  
    //     (min.x <= P.x) &&  
    //     (min.y <= P.y) &&  
    //     (min.z <= P.z)  
    point min;  
  
    // Size AABB in each dimension.  
    point diameter;  
};
```

Center / radius

```
class aabb_c_r {  
    // Center of the AABB.  
    point center;  
  
    // Distance from 'center' to each  
    // side of the AABB.  
    point radius;  
};
```

AABB-AABB Intersection

- ⇒ AABB-AABB intersection is just an interval overlap test extended to n -dimensions.
 - If there is *no* overlap in *any* dimension, the AABBs cannot intersect.
- ⇒ Examples:
 - Do $A = \{ c = \{ 1, 1 \}, r = \{ 1, 2 \} \}$ and $B = \{ c = \{ 3, 4 \}, r = \{ 1, 1 \} \}$ intersect?

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 - Do $A = \{ c = \{ 1, 1 \}, r = \{ 1, 2 \} \}$ and $B = \{ c = \{ 3, 4 \}, r = \{ 1, 1 \} \}$ intersect?
 - Do $A = \{ c = \{ 1, 1 \}, r = \{ 2, 2 \} \}$ and $B = \{ c = \{ 3, 4 \}, r = \{ 2, 2 \} \}$ intersect?

AABB-AABB Intersection (cont.)

```
bool aabb_c_r::intersect(aabb_c_r &box)
{
    const point dist = center - box.center;
    const point rad = radius + box.radius;

    if (abs(dist[0]) > (rad[0])) return false;
    if (abs(dist[1]) > (rad[1])) return false;
    if (abs(dist[2]) > (rad[2])) return false;

    return true;
}
```

AABB Creation

- ⇒ Creating an initial AABB from source data is a trivial $O(n)$ problem.

AABB Creation

- ⇒ Creating an initial AABB from source data is a trivial $O(n)$ problem.
 - Scan all of the points tracking the minimum and maximum value in each dimension. Convert the resulting values to the desired representation.
- ⇒ As the object moves, how can we update the AABB?
 - Rotations are the problem.

AABB from Bounding Sphere

⇒ For a given rotation:

1. Create a sphere centered at the point of rotation that encompasses the object.
2. Create an AABB from that sphere.

⇒ Merits and drawbacks of this technique?

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1. Create a sphere centered at the point of rotation that encompasses the object.
2. Create an AABB from that sphere.

⇒ Merits and drawbacks of this technique?

- Fast to compute.
- Only works well if the object has a single pivot point.
- Creates a *very* loose AABB.
- Why not just use the bounding sphere?!?

AABB from Source Data

- ⇒ Recalculate the AABB from the transformed source data.
 - Don't actually transform the points. Instead transform the axis use for the comparisons.
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AABB from Source Data

- ⇒ Recalculate the AABB from the transformed source data.
 - Don't actually transform the points. Instead transform the axis use for the comparisons.
- ⇒ Merits and drawbacks of this technique?
 - Creates a tight fitting AABB.
 - $O(n)$ per transformation can be too expensive.
 - Can be optimized using other search structures and / or a convex hull.
 - More about convex hulls next week.

AABB by Hill-climbing

- ⇒ Track the six points at the extrema of the AABB
- ⇒ To update, examine the neighboring points to search for the new extrema.
- ⇒ Merits and drawbacks of this technique?

AABB by Hill-climbing

- ⇒ Track the six points at the extrema of the AABB
- ⇒ To update, examine the neighboring points to search for the new extrema.
- ⇒ Merits and drawbacks of this technique?
 - Creates a tight fitting AABB.
 - Fast, but...
 - Requires precalculation of a convex hull
 - Requires a data structure that stores connectivity among points

AABB from Rotated AABB

- ⇒ Transform the original AABB and compute its AABB.
- ⇒ Merits and drawbacks of this technique?

AABB from Rotated AABB

- ⇒ Transform the original AABB and compute its AABB.
- ⇒ Merits and drawbacks of this technique?
 - Fast.
 - Not a very tight fitting AABB.
 - *Very* commonly used.

References

http://en.wikipedia.org/wiki/Bounding_volume

Next week...

- ⇒ More bounding volumes...
 - Bounding spheres
 - Oriented bounding boxes
 - k-DOPs
 - Convex hulls
- ⇒ Something we can do with our BVs!
- ⇒ First programming assignment will be assigned.

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