# Linear Algebra Review

CMSC 435/634

## **Abstract Vectors**

 $(\vec{u}, \vec{v}, \vec{w} \text{ vectors}; a, b, c \text{ scalars})$ 

Abstract Vectors

- Addition:  $\vec{u} + \vec{v}$  is a vector
- Scalar Multiplication: au is a vector
- Commutitive:  $\vec{u} + \vec{v} = \vec{v} + \vec{u}$
- Distributive:  $(a + b)\vec{u} = a\vec{u} + b\vec{u}$
- Associative:  $(\vec{u} + \vec{v}) + \vec{w} = \vec{u} + (\vec{v} + \vec{w})$

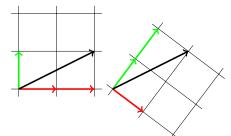
## Basis Vectors

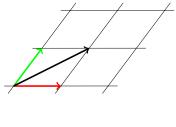
Vector as linear combination of basis vectors

• 
$$\vec{v} = 2\hat{i} + 1\hat{j} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

• 
$$\vec{v} = 1\hat{m} + 2\hat{n} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

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ho} + 1\hat{q} = \left[egin{array}{c} 1 \ 1 \end{array}
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## Notation

• Column: 
$$\vec{v} = \begin{bmatrix} v_0 \\ v_1 \end{bmatrix}$$

Some texts use columns for everything

• Row: 
$$\vec{v} = \begin{bmatrix} v_0 & v_1 \end{bmatrix}$$

- Some texts use rows for everything
- Results in transposes and swapped order from what we'll use
- I like columns for points/vectors, rows for normals

Abstract Vectors

## **Matrices**

• Matrix: 
$$A = \begin{bmatrix} a_{0,0} & a_{0,1} \\ a_{1,0} & a_{1,1} \end{bmatrix} = [a_{i,j}] = [a_{row,column}]$$

• Transpose: 
$$A^T = \begin{bmatrix} a_{0,0} & a_{1,0} \\ a_{0,1} & a_{1,1} \end{bmatrix} = [a_{j,i}]$$

• Multiply: 
$$AB = \begin{bmatrix} a_{0,0} & a_{0,1} \\ a_{1,0} & a_{1,1} \end{bmatrix} \begin{bmatrix} b_{0,0} & b_{0,1} \\ b_{1,0} & b_{1,1} \end{bmatrix} = \begin{bmatrix} a_{0,0}b_{0,0} + a_{0,1}b_{1,0} & a_{0,0}b_{0,1} + a_{0,1}b_{1,1} \\ a_{1,0}b_{0,0} + a_{1,1}b_{1,0} & a_{1,0}b_{0,1} + a_{1,1}b_{1,1} \end{bmatrix}$$

#### Matrix Code

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• Math: C = A B
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• Components:  $c_{i,j} = \sum_{\alpha} a_{i,\alpha} \ b_{\alpha,j}$ 

• Code:

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 \begin{array}{lll} & \textbf{for(int} & i = 0; & i < N; & + + i ) & \{ & \\ & \textbf{for(int} & j = 0; & j < M; & + + j ) & \{ & \\ & & c[i][j] & = 0; & \\ & & \textbf{for(int} & \alpha = 0; & \alpha < K; & \alpha + + ) & \{ & \\ & & c[i][j] & = c[i][j] & + a[i][\alpha] & * b[\alpha][j]; \\ & & \} & \\ & \} & \\ & \} & \\ & \} & \\ \end{array}
```

## Adjugate and Inverse

• Inverse:  $A^{-1}A = AA^{-1} = I$ 

• Determinant: |A|

$$\begin{vmatrix} a & b \\ c & d \end{vmatrix} = a |d| - b|c|$$

$$\begin{vmatrix} a & b & c \\ c & d & e \end{vmatrix} = a \begin{vmatrix} e & f \\ b & c \end{vmatrix} - b \begin{vmatrix} a & b & c \\ c & d & e \end{vmatrix}$$

$$\bullet \begin{vmatrix} a & b & c \\ d & e & f \\ g & h & i \end{vmatrix} = a \begin{vmatrix} e & f \\ h & i \end{vmatrix} - b \begin{vmatrix} d & f \\ g & i \end{vmatrix} + c \begin{vmatrix} d & e \\ g & h \end{vmatrix}$$

- Adjugate:  $A^* = cof(A)^T$  (matrix of cofactors cof(A))
  - Sometimes called Adjoint or Adjunct

$$\bullet \ A^{-1} = \frac{A^*}{|A|}$$

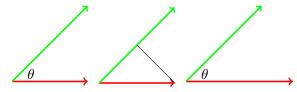
### Dot Product

Dot Product

- Also called inner product
  - $\vec{u} \bullet \vec{v}$  is a scalar
  - Commutitive:  $\vec{u} \bullet \vec{v} = \vec{v} \bullet \vec{u}$
  - Distributive:  $(a\vec{u}) \bullet \vec{v} = \vec{u} \bullet (a\vec{v}) = a(\vec{u} \bullet \vec{v})$
  - Associative:  $(\vec{u} + \vec{v}) \bullet \vec{w} = \vec{u} \bullet \vec{w} + \vec{v} \bullet \vec{w}$
  - $\vec{v} \bullet \vec{v} \geq 0$
  - $\vec{v} \bullet \vec{v} = 0 \leftrightarrow \vec{v} = \vec{0}$
- Equivalent notations
  - Vector:  $\vec{u} \bullet \vec{v}$
  - Matrix: U<sup>T</sup>V
  - $\sum_{\alpha} u_{\alpha} v_{\alpha}$

# Dot Defines Length and Angle

- $\vec{v} \bullet \vec{v} = |\vec{v}|^2$
- $\vec{u} \cdot \vec{v} = |\vec{u}||\vec{v}|\cos\theta$ 
  - **Defines** angle  $\theta$ !
  - If  $|\vec{v}| = 1$ , gives projection of  $\vec{u}$  onto  $\vec{v}$
  - If  $|\vec{u}| = |\vec{v}| = 1$ , gives just  $\cos \theta$



# Orthogonal & Normal

- Orthogonal = perpendicular:  $\vec{u} \bullet \vec{v} = 0$
- *Normal* (this usage) = unit-length:  $\vec{u} \cdot \vec{u} = 1$
- Orthonormal: set of vectors both orthogonal and normal
- Orthogonal matrix: rows (& columns) orthonormal
  - For orthogonal matrices,  $A^{-1} = A^T$

## 3D Cross Product

$$\vec{n} \times \vec{v}$$

Abstract Vectors

- length = area of parallelogram = twice area of triangle •  $|\vec{u} \times \vec{v}| = |\vec{u}||\vec{v}|\sin(\theta)$
- direction = perpendicular to  $\vec{u}$  and  $\vec{v}$  (right hand rule)

• 
$$\vec{w} = \vec{u} \times \vec{v} = \begin{vmatrix} \hat{i} \\ \hat{j} & U \end{vmatrix} V$$

$$\bullet \begin{bmatrix} w_0 \\ w_1 \\ w_2 \end{bmatrix} = \begin{bmatrix} u_1 v_2 - u_2 v_1 \\ u_2 v_0 - u_0 v_2 \\ u_0 v_1 - u_1 v_0 \end{bmatrix}$$

- Positive terms follow 012012 order
- Netative terms follow 210210 order

# Building an Orthogonal Basis

Vectors  $\vec{u}$ ,  $\vec{v}$ ,  $\vec{w}$ 

Gram-Schmidt (any number of dimensions)

• 
$$\vec{u'} = \vec{u}$$

$$\vec{\mathbf{v}}' = \vec{\mathbf{v}} - \hat{\mathbf{u}}' \quad (\vec{\mathbf{v}} \bullet \hat{\mathbf{u}}')$$

• 
$$\vec{v'} = \vec{v} - \frac{\vec{u'}}{|\vec{u'}|} \left( \vec{v} \bullet \frac{\vec{u'}}{|\vec{u'}|} \right)$$

• 
$$\vec{v'} = \vec{v} - \vec{u'} \vec{v} \bullet \vec{u'} / |\vec{u'}|^2$$

• 
$$\vec{v'} = \vec{v} - \vec{u'} \ \vec{v} \bullet \vec{u'} / \vec{u'} \bullet \vec{u'}$$

• 
$$\vec{w'} = \vec{w} - \vec{u'} \ \vec{w} \bullet \vec{u'} / \vec{u'} \bullet \vec{u'} - \vec{v'} \ \vec{w} \bullet \vec{v'} / \vec{v'} \bullet \vec{v'}$$

Cross-product (3D only)

• 
$$\vec{u'} = \vec{u}$$

• 
$$\vec{w'} = \vec{u'} \times \vec{v}$$

• 
$$\vec{v'} = \vec{w'} \times \vec{u'}$$