CX1104: Linear Algebra for Computing

$$\underbrace{\begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & a_{m3} & \dots & a_{mn} \end{bmatrix}}_{A} \underbrace{\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_n \end{bmatrix}}_{n \times n} = \underbrace{\begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}}_{m \times 1}$$

Chap. No: **6.1.4**

Lecture: Orthogonality

Topic: Dot Product

Concept: Important Inequalities

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Rev: 25th June 2020

Cauchy-Schwarz Inequality

DEFINITION 3 If \mathbf{u} and \mathbf{v} are nonzero vectors in R^2 or R^3 , and if θ is the angle between \mathbf{u} and \mathbf{v} , then the *dot product* (also called the *Euclidean inner product*) of \mathbf{u} and \mathbf{v} is denoted by $\mathbf{u} \cdot \mathbf{v}$ and is defined as

$$\mathbf{u} \cdot \mathbf{v} = \|\mathbf{u}\| \|\mathbf{v}\| \cos \theta \tag{12}$$

If $\mathbf{u} = \mathbf{0}$ or $\mathbf{v} = \mathbf{0}$, then we define $\mathbf{u} \cdot \mathbf{v}$ to be 0.

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

Formula (20) is not defined unless its argument satisfies the inequalities

$$-1 \le \frac{\mathbf{u} \cdot \mathbf{v}}{\|\mathbf{u}\| \|\mathbf{v}\|} \le 1 \tag{21}$$

Fortunately, these inequalities do hold for all nonzero vectors in \mathbb{R}^n as a result of the following fundamental result known as the *Cauchy–Schwarz inequality*.

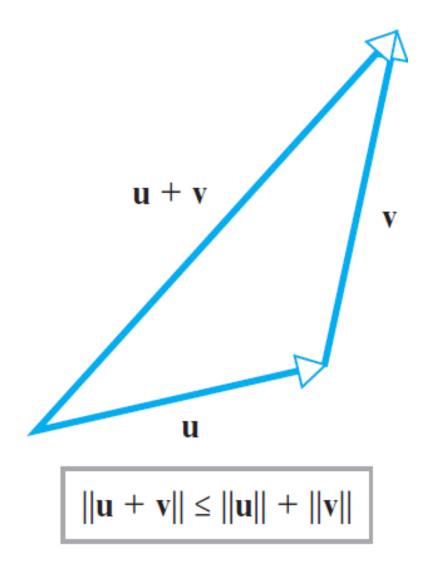
THEOREM 3.2.4 Cauchy–Schwarz Inequality

If
$$\mathbf{u} = (u_1, u_2, \dots, u_n)$$
 and $\mathbf{v} = (v_1, v_2, \dots, v_n)$ are vectors in \mathbb{R}^n , then
$$|\mathbf{u} \cdot \mathbf{v}| \le ||\mathbf{u}|| ||\mathbf{v}|| \tag{22}$$

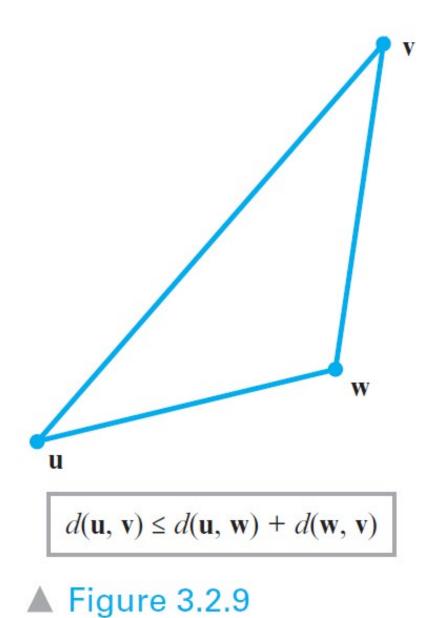
or in terms of components

$$|u_1v_1 + u_2v_2 + \dots + u_nv_n| \le (u_1^2 + u_2^2 + \dots + u_n^2)^{1/2} (v_1^2 + v_2^2 + \dots + v_n^2)^{1/2}$$
(23)

Triangle Inequality



▲ Figure 3.2.8



THEOREM 3.2.5 If \mathbf{u} , \mathbf{v} , and \mathbf{w} are vectors in \mathbb{R}^n , then:

(a)
$$\|\mathbf{u} + \mathbf{v}\| \le \|\mathbf{u}\| + \|\mathbf{v}\|$$

[Triangle inequality for vectors]

(b) $d(\mathbf{u}, \mathbf{v}) \le d(\mathbf{u}, \mathbf{w}) + d(\mathbf{w}, \mathbf{v})$ [Triangle inequality for distances]

Proof:

Proof (a)

$$\|\mathbf{u} + \mathbf{v}\|^{2} = (\mathbf{u} + \mathbf{v}) \cdot (\mathbf{u} + \mathbf{v}) = (\mathbf{u} \cdot \mathbf{u}) + 2(\mathbf{u} \cdot \mathbf{v}) + (\mathbf{v} \cdot \mathbf{v})$$

$$= \|\mathbf{u}\|^{2} + 2(\mathbf{u} \cdot \mathbf{v}) + \|\mathbf{v}\|^{2}$$

$$\leq \|\mathbf{u}\|^{2} + 2\|\mathbf{u}\|\|\mathbf{v}\| + \|\mathbf{v}\|^{2}$$

$$\leq \|\mathbf{u}\|^{2} + 2\|\mathbf{u}\|\|\mathbf{v}\| + \|\mathbf{v}\|^{2}$$

$$= (\|\mathbf{u}\| + \|\mathbf{v}\|)^{2}$$
Property of absolute value
$$\leq \|\mathbf{u}\|^{2} + 2\|\mathbf{u}\|\|\mathbf{v}\| + \|\mathbf{v}\|^{2}$$

$$= (\|\mathbf{u}\| + \|\mathbf{v}\|)^{2}$$

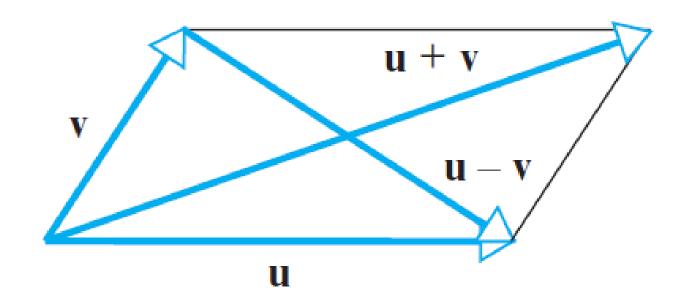
This completes the proof since both sides of the inequality in part (a) are nonnegative.

Proof (b) It follows from part (a) and Formula (11) that

$$d(\mathbf{u}, \mathbf{v}) = \|\mathbf{u} - \mathbf{v}\| = \|(\mathbf{u} - \mathbf{w}) + (\mathbf{w} - \mathbf{v})\|$$

$$\leq \|\mathbf{u} - \mathbf{w}\| + \|\mathbf{w} - \mathbf{v}\| = d(\mathbf{u}, \mathbf{w}) + d(\mathbf{w}, \mathbf{v})$$

Parallelogram Equation for Vectors



▲ Figure 3.2.10

THEOREM 3.2.6 Parallelogram Equation for Vectors

If **u** and **v** are vectors in \mathbb{R}^n , then

$$\|\mathbf{u} + \mathbf{v}\|^2 + \|\mathbf{u} - \mathbf{v}\|^2 = 2(\|\mathbf{u}\|^2 + \|\mathbf{v}\|^2)$$
 (24)

Proof:

THEOREM 3.2.6 Parallelogram Equation for Vectors

If **u** and **v** are vectors in \mathbb{R}^n , then

$$\|\mathbf{u} + \mathbf{v}\|^2 + \|\mathbf{u} - \mathbf{v}\|^2 = 2(\|\mathbf{u}\|^2 + \|\mathbf{v}\|^2)$$
 (24)

Proof

$$\|\mathbf{u} + \mathbf{v}\|^2 + \|\mathbf{u} - \mathbf{v}\|^2 = (\mathbf{u} + \mathbf{v}) \cdot (\mathbf{u} + \mathbf{v}) + (\mathbf{u} - \mathbf{v}) \cdot (\mathbf{u} - \mathbf{v})$$
$$= 2(\mathbf{u} \cdot \mathbf{u}) + 2(\mathbf{v} \cdot \mathbf{v})$$
$$= 2(\|\mathbf{u}\|^2 + \|\mathbf{v}\|^2)$$

THEOREM 3.2.7 If \mathbf{u} and \mathbf{v} are vectors in \mathbb{R}^n with the Euclidean inner product, then

$$\mathbf{u} \cdot \mathbf{v} = \frac{1}{4} \|\mathbf{u} + \mathbf{v}\|^2 - \frac{1}{4} \|\mathbf{u} - \mathbf{v}\|^2$$
 (25)

Proof

$$\|\mathbf{u} + \mathbf{v}\|^2 = (\mathbf{u} + \mathbf{v}) \cdot (\mathbf{u} + \mathbf{v}) = \|\mathbf{u}\|^2 + 2(\mathbf{u} \cdot \mathbf{v}) + \|\mathbf{v}\|^2$$
$$\|\mathbf{u} - \mathbf{v}\|^2 = (\mathbf{u} - \mathbf{v}) \cdot (\mathbf{u} - \mathbf{v}) = \|\mathbf{u}\|^2 - 2(\mathbf{u} \cdot \mathbf{v}) + \|\mathbf{v}\|^2$$

from which (25) follows by simple algebra.

Reference

Materials in these slides have been taken from:

Anton and Rorres, "Linear Algebra", 11th edition, Wiley.

Chapter: 3.1, 3.2

Euclidean Vector Spaces

- 3.1 Vectors in 2-Space, 3-Space, and *n*-Space 131
- 3.2 Norm, Dot Product, and Distance in \mathbb{R}^n 142

