

MAT 307 LECTURE 9

MIDTERM 1 IS MONDAY

COVERS THROUGH THIS

LECTURE

IN A GENERAL INNER PRODUCT SPACE, IF  $f, g \neq 0$ , DEFINE THE ANGLE BETWEEN  $f, g$

BY 
$$\cos \theta = \frac{\langle f, g \rangle}{\|f\| \cdot \|g\|} \quad \left( -1 \leq \frac{\langle f, g \rangle}{\|f\| \cdot \|g\|} \leq 1 \right)$$

BY  
(CAUCHY-SCHWARZ)

AN INNER PRODUCT SPACE IS

"GEOMETRIC" SINCE THERE IS A NOTION OF ANGLES BTWN VECTORS.

THEOREM: SUPPOSE  $V$  IS  
AN  $n$ -DIM'L VECTOR SPACE,  
 $B = \{\underline{u}_1, \dots, \underline{u}_n\}$  IS AN ORTHO-  
NORMAL SET. THEN  $B$  IS  
A BASIS FOR  $V$ , AND IF  
 $\underline{x} = u_1 \underline{u}_1 + \dots + u_n \underline{u}_n$  THEN  
 $u_j = \langle \underline{x}, \underline{u}_j \rangle$ .

THIS PROVES THAT THE  
SET IS A BASIS, AND  
VERIFIES

$$\underline{x} = c_1 \underline{u}_1 + \dots + c_n \underline{u}_n,$$

$$\forall i \in N \langle \underline{x}, \underline{u}_i \rangle = c_i.$$

□

THEOREM: LET  $S = \{u_1, \dots, u_n\}$

BE AN ORTHONORMAL SET  
IN A VECTOR SPACE  $V$ .

FOR ALL  $x \in V$ , THE

DISTANCE

$$\|x - (u_1 u_1 + \dots + u_n u_n)\|^2$$

IS MINIMIZED BY

CHOOSING  $u_i = \langle x, u_i \rangle$ .

$$\|x\|^2 \geq \sum_{i=1}^n \langle x, u_i \rangle^2.$$

"BESSEL'S INEQUALITY"

THE DISTANCE OF  $x$  FROM  
SPAN  $\{u_1, \dots, u_n\}$  IS

$$d^2 = \|x\|^2 - \sum_{i=1}^n \langle x, u_i \rangle^2.$$

PROOF:

$$\begin{aligned} & \| \underline{x} - (u_1 \underline{y}_1 + \dots + u_n \underline{y}_n) \|^2 \\ &= \langle \underline{x} - (u_1 \underline{y}_1 + \dots + u_n \underline{y}_n), \\ & \quad \underline{x} - (u_1 \underline{y}_1 + \dots + u_n \underline{y}_n) \rangle \\ &= \| \underline{x} \|^2 + \| u_1 \underline{y}_1 + \dots + u_n \underline{y}_n \|^2 \\ & \quad - 2 \langle \underline{x}, u_1 \underline{y}_1 + \dots + u_n \underline{y}_n \rangle \end{aligned}$$

$$\begin{aligned} \| u_1 \underline{y}_1 + \dots + u_n \underline{y}_n \|^2 &= \langle u_1 \underline{y}_1 + \dots + u_n \underline{y}_n, \\ & \quad u_1 \underline{y}_1 + \dots + u_n \underline{y}_n \rangle \\ &= \sum_{i=1}^n u_i^2. \end{aligned}$$

$$\begin{aligned} \text{THUS } \| \underline{x} - (u_1 \underline{y}_1 + \dots + u_n \underline{y}_n) \|^2 \\ &= \| \underline{x} \|^2 - 2 \sum_{k=1}^n u_k \langle \underline{x}, \underline{y}_k \rangle \\ & \quad + \sum_{k=1}^n u_k^2. \end{aligned}$$

$$\begin{aligned} \text{WRITE: } & -2 u_k \langle \underline{x}, \underline{y}_k \rangle + u_k^2 \\ &= (\langle \underline{x}, \underline{y}_k \rangle - u_k)^2 - \langle \underline{x}, \underline{y}_k \rangle^2 \end{aligned}$$

$$\begin{aligned} \text{THUS: } & \| \underline{x} - (u_1 \underline{y}_1 + \dots + u_n \underline{y}_n) \|^2 \\ &= \| \underline{x} \|^2 - \sum_{k=1}^n \langle \underline{x}, \underline{y}_k \rangle^2 + \sum_{k=1}^n (u_k - \langle \underline{x}, \underline{y}_k \rangle)^2 \end{aligned}$$

THE MINIMIZING CHOICE SETS  
THE COEFFICIENT  $u_k = \langle \underline{x}, \underline{y}_k \rangle$   
SO  $\textcircled{*} = 0$ .

THUS THE MINIMIZING  
DISTANCE IS OBTAINED BY

$$\sum_{k=1}^n \langle \underline{x}, \underline{y}_k \rangle \underline{y}_k \quad \text{WITH}$$

DISTANCE SQUARED

$$\| \underline{x} \|^2 - \sum_{k=1}^n \langle \underline{x}, \underline{y}_k \rangle^2.$$

## GRAM-SCHMIDT PROCESS:

GIVEN LINEARLY INDEPENDENT  
VECTORS  $\underline{u}_1, \underline{u}_2, \underline{u}_3, \underline{u}_4, \dots,$

AN ORTHONORMAL SEQUENCE

$\underline{v}_1, \underline{v}_2, \underline{v}_3, \underline{v}_4, \dots$  WITH,

FOR EACH  $k,$

$$\text{SPAN}\{\underline{u}_1, \dots, \underline{u}_k\} = \text{SPAN}\{\underline{v}_1, \dots, \underline{v}_k\}$$

CAN BE OBTAINED AS FOLLOWS.

IT REMAINS TO CHECK  
THAT

$$\text{SPAN} \{ \underline{u}_1, \dots, \underline{u}_k \} = \text{SPAN} \{ \underline{v}_1, \dots, \underline{v}_k \}$$

ALL  $k$ .

BY INDUCTION:

$$\underline{v}_1 = \frac{\underline{u}_1}{\|\underline{u}_1\|} \quad \text{SO SPANS EQUAL.}$$

$$\text{ASSUMING } \text{SPAN} \{ \underline{v}_1, \dots, \underline{v}_{k-1} \} \\ = \text{SPAN} \{ \underline{u}_1, \dots, \underline{u}_{k-1} \},$$

$$\underline{v}_k = \frac{\underline{u}_k - \sum_{j=1}^{k-1} \langle \underline{u}_k, \underline{v}_{j-1} \rangle \underline{v}_{j-1}}{\| \dots \dots \dots \|}.$$

$$\Rightarrow \underline{v}_k \in \text{SPAN} \{ \underline{u}_k, \underline{v}_1, \dots, \underline{v}_{k-1} \}$$

$$\subset \text{SPAN} \{ \underline{u}_1, \dots, \underline{u}_k \}.$$

THUS  $\{ \underline{v}_1, \dots, \underline{v}_k \}$  ARE  $k$

LINEARLY INDEP. VECTORS  
IN  $\text{SPAN} \{ \underline{u}_1, \dots, \underline{u}_k \}$ , SO  
A BASIS FOR THIS SPACE.  $\square$



PROJECTION IN THE  
DIRECTION OF  $\underline{u}_1$  IS  
GIVEN BY

$$\underline{u}_1 \cdot \underline{u}_1^t \underline{x} = \langle \underline{u}_1, \underline{x} \rangle \cdot \underline{u}_1.$$

REFLECTION IN THE PLANE  
 $\perp$  to  $\underline{u}_1$  IS GIVEN BY

$$R_{\underline{u}_1}(\underline{x}) = \underline{x} - 2(\underline{u}_1 \cdot \underline{x})\underline{u}_1$$

THE MATRIX IS  $\boxed{\underline{I} - 2 \underline{u}_1 \underline{u}_1^t}$ .

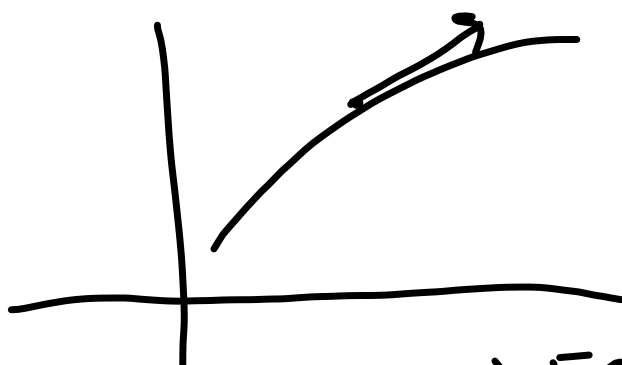
IF  $f$  IS A FUNCTION OF  
A SINGLE VARIABLE  $t$ ,

$$f(t) = \begin{pmatrix} f_1(t) \\ \vdots \\ f_n(t) \end{pmatrix}$$

"VECTOR  
VALUED  
FUNCTION!"

THE FUNCTION  $f$  IS CTS  
IF EACH COMPONENT FUNCTION  
IS CONTINUOUS.

THE IMAGE OF A VECTOR  
VALUED FUNCTION  $f(t)$   
IS CALLED A CURVE



THE TANGENT VECTOR TO THE  
CURVE AT  $f(t)$  IS  $f'(t)$ .

DEFINITION: A CURVE WITH  
PARAMETRIC REPRESENTATION

$$\underline{x}(t) = g(t)$$

IS SMOOTH IF  $g'(t)$

EXISTS FOR ALL  $t$  AND

IS NON-ZERO.

EXAMPLE: THE IMAGE OF  
THE CURVE

$$\underline{x}(t) = \begin{pmatrix} \cos t \\ \sin t \\ t \end{pmatrix} \text{ IS A}$$

HELIX, CONTAINED IN  $x^2 + y^2 = 1$

THE PRODUCT RULES OF  
 1-VARIABLE CALCULUS AND  
 MULTI-DIMENSIONAL  
 GENERALIZATIONS:

- IF  $\phi(t)$  IS A SCALAR FN,

$$\frac{d}{dt} (\phi(t) \underline{x}(t)) = \phi'(t) \underline{x}(t) + \phi(t) \underline{x}'(t)$$

- IF  $\underline{x}(t), \underline{y}(t)$  VECTOR VALUED FNS,

$$\begin{aligned} \frac{d}{dt} (\underline{x}(t) \cdot \underline{y}(t)) \\ = \underline{x}'(t) \cdot \underline{y}(t) + \underline{x}(t) \cdot \underline{y}'(t) \end{aligned}$$

$$\begin{aligned} \frac{d}{dt} (\underline{x}(t) \times \underline{y}(t)) \\ = \underline{x}'(t) \times \underline{y}(t) + \underline{x}(t) \times \underline{y}'(t) \end{aligned}$$

CHAIN RULE:  $u: \mathbb{R} \rightarrow \mathbb{R}$

$$\frac{d}{dt} \underline{x}(u(t)) = \underline{x}'(u(t)) \cdot u'(t).$$