

(Q1) Discuss dot product, norm, normalizing a vector, scaled version of a vector.

Dot Product

Ref 6.1.3, pg 2

// Ref 6.1.3, pg 5  
derived from Law of cosine

// Ref 6.1.3 pg 8  
row, col<sup>2</sup> vector form

Ref  
// 6.1.3  
pg 2

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

$\mathbf{u}, \mathbf{v} \in \mathbb{R}^n$

$$\mathbf{u} \cdot \mathbf{v} = \|\mathbf{u}\| \|\mathbf{v}\| \cos \theta. \quad (1)$$

$\downarrow$  norm of  $\mathbf{v}$        $\downarrow$   $0 \dots \pi$   
 $=$  length of  $\mathbf{u}$       between  $\mathbf{u}$  and  $\mathbf{v}$

$$= u_1 v_1 + u_2 v_2 + \dots + u_n v_n$$

↪ component form of dot product,

$$= \sum_{i=1}^n u_i v_i$$

$$= \mathbf{u}^T \mathbf{v} = \mathbf{v}^T \mathbf{u}$$

↪ matrix form of dot product

Norm of vector

Ref 6.1.2 pg 2

Length / magnitude of vector

$$\|\mathbf{v}\| = \sqrt{\sum_{i=1}^n (v_i)^2}; \mathbf{v} \in \mathbb{R}^n$$



Normalizing a vector

Ref 6.1.2 pg 4

Making a vector to be unit length.

$$\mathbf{u} = \frac{\mathbf{v}}{\|\mathbf{v}\|}$$

//  $\mathbf{u}$  is a vector in same direction with  $\mathbf{v}$  and has norm = 1.



Scaled Version of a vector

Ref 6.2.2 pg 3

$$\mathbf{w}_1 = \text{Proj}_{\mathbf{a}} \mathbf{u}$$

$$= \left( \frac{\mathbf{u} \cdot \mathbf{a}}{\|\mathbf{a}\|^2} \right) \mathbf{a}$$

$$= \left( \frac{\mathbf{u} \cdot \mathbf{a}}{\mathbf{a} \cdot \mathbf{a}} \right) \mathbf{a}$$

↑  
scalar      ↑  
along direction of  $\mathbf{a}$ .

$$1) \quad u = \begin{bmatrix} -1 \\ 2 \end{bmatrix}, \quad v = \begin{bmatrix} 4 \\ 6 \end{bmatrix}$$

Ref 6.1.3 pg 2

$$\text{Find } u \cdot u, v \cdot u, \frac{v \cdot u}{u \cdot u}$$

$$a) \quad u \cdot u = [-1 \ 2] \begin{bmatrix} -1 \\ 2 \end{bmatrix} = \sum_{i=1}^2 (u_i)^2$$

$$\therefore 1+4 = 5 = (-1)^2 + (2)^2$$

$$b) \quad v \cdot u = \|u\| \|v\| \cos \theta. \quad [\text{Definition of dot product}]$$

$$= \sum_{i=1}^2 v_i u_i$$

$$= [4 \ 6] \begin{bmatrix} -1 \\ 2 \end{bmatrix} = -4+12 = 8$$

$$c) \quad \frac{v \cdot u}{u \cdot u} = \frac{8}{5}$$

$$4) \quad \frac{u}{u \cdot u} = \frac{1}{u \cdot u} \begin{bmatrix} -1 \\ 2 \end{bmatrix} = \frac{1}{5} \begin{bmatrix} -1 \\ 2 \end{bmatrix} = \begin{bmatrix} -\frac{1}{5} \\ \frac{2}{5} \end{bmatrix}$$

To normalize, we should

$$\frac{u}{\|u\|} = \frac{1}{\sqrt{u \cdot u}} = \frac{1}{\sqrt{5}} \begin{bmatrix} -1 \\ 2 \end{bmatrix} = \begin{bmatrix} \frac{-1}{\sqrt{5}} \\ \frac{2}{\sqrt{5}} \end{bmatrix} \Rightarrow$$

$$\left\| \frac{u}{\|u\|} \right\| = \left( \frac{-1}{\sqrt{5}} \right)^2 + \left( \frac{2}{\sqrt{5}} \right)^2 = \frac{1}{5} + \frac{4}{5} = 1$$

$$5) \quad \left( \frac{u \cdot v}{v \cdot v} \right) v = \left( \frac{8}{16+36} \right) \begin{bmatrix} 4 \\ 6 \end{bmatrix}$$

$$= \left( \frac{2}{13} \right) \begin{bmatrix} 4 \\ 6 \end{bmatrix} = \begin{bmatrix} \frac{8}{13} \\ \frac{12}{13} \end{bmatrix}$$

This is the formula of  $\text{Proj}_v u$  [project  $u$  onto  $v$ ].

- The closest point along  $v$  to  $u$ .
- The orthogonal component  $e = u - \text{Proj}_v u$  is  $\perp$  to  $v$

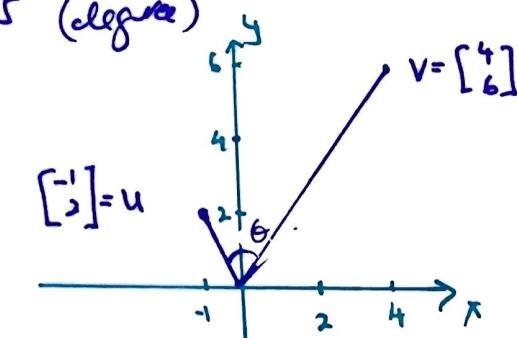
(b extra) What is  $\theta$ ?  $\angle$  between  $u$  and  $v$ .

$$\cos \theta = \frac{v \cdot u}{\|u\| \|v\|} = \frac{8}{\sqrt{5} \sqrt{52}} = 0.4961$$

(rad).  
0.11.

$$\theta = 1.0517 \text{ rad.}$$

$$\therefore \theta = 60.25^\circ \text{ (degree)}$$



2a)  $\underline{v} \cdot \underline{v} = \|\underline{v}\|^2 = \text{True}$

Definition of  $\|\cdot\|$  = length of vector  $\underline{v} \in \mathbb{R}^n$   
 (Sec 6.1.2 pg 2)

does the same thing

$$\begin{aligned} &= \sqrt{v_1^2 + v_2^2 + \dots + v_n^2} \\ &= \sqrt{\sum_{i=1}^n v_i^2} = \sqrt{\underline{v}^T \underline{v}} \\ &= \sqrt{\underline{v} \cdot \underline{v}} \end{aligned}$$

2b) for any scalar  $c$ ,  $\underline{u} \cdot (c \underline{v}) = c(\underline{u} \cdot \underline{v})$

// Note to differentiate  
 Scalar and vector, we put  $\underline{u}$  here.  
 If it is clear from context, then we do not.  
 In text book, vectors are in bold.

Theorem 1c (pg 331 4th edition)

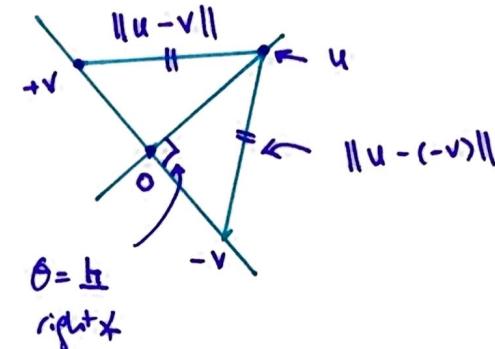
$$(c \underline{u}) \cdot \underline{v} = c(\underline{u} \cdot \underline{v}) = \underline{u} \cdot (c \underline{v})$$

$$\text{and } \underline{u} \cdot \underline{v} = \underline{v} \cdot \underline{u}$$

2c) if dist of  $\underline{u}$  to  $\underline{v} = \underline{u}$  to  $(-\underline{v})$   
 then  $\underline{u}$  and  $\underline{v}$  are orthogonal.

Ans: True

see pg 333, 4th edition, fig 5.



$$(\text{dist } (\underline{u}, \underline{-v}))^2$$

$$= \|\underline{u} - (\underline{-v})\|^2$$

$$= \|\underline{u} + \underline{v}\|^2$$

$$= (\underline{u} + \underline{v}) \cdot (\underline{u} + \underline{v})$$

$$= \underline{u} \cdot \underline{u} + \underline{v} \cdot \underline{u} + \underline{u} \cdot \underline{v} + \underline{v} \cdot \underline{v}$$

$$= \|\underline{u}\|^2 + \|\underline{v}\|^2 + 2 \underline{u} \cdot \underline{v}$$

$$(\text{dist } (\underline{u}, \underline{v}))^2$$

$$= \|\underline{u} - \underline{v}\|^2$$

$$= (\underline{u} - \underline{v}) \cdot (\underline{u} - \underline{v})$$

$$= \underline{u}^2 - \underline{v} \cdot \underline{u} - \underline{u} \cdot \underline{v} + \underline{v}^2$$

$$= \|\underline{u}\|^2 + \|\underline{v}\|^2 - 2 \underline{u} \cdot \underline{v}$$

$$= \|\underline{u}\|^2 + \|\underline{v}\|^2 - 2 \underline{u} \cdot \underline{v}$$

LHS == RHS

only when  $\underline{u} \cdot \underline{v} = 0$

which is when  $\theta = \frac{\pi}{2}$

2d) For a square matrix  $A$ , vectors in  $\text{Col}(A)$  are orthogonal to  $\text{Null}(A)$ .

Ans: [False].

(A) Counter Example:  $A = \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$ .

$$u = \text{Col}(A) = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$v = \text{Null}(A) = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$u \cdot v = \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} = 1 \Rightarrow u \cdot v \text{ is NOT } \perp.$$

(B)

More explanation:

If  $A$  is  $m \times n$  matrix,  
where  $(m \neq n)$ .

then  $\text{Col}(A) \in \mathbb{R}^m$

$\text{NULL}(A) \in \mathbb{R}^n$

$\therefore$  cannot even take dot product as vectors are not same size.

(C) More explanation:

Row Space  $\perp$  Null Space.  
dimension  $r$  dimension  $(n-r)$

Col Space  $\perp$  left Null Space  $N(A^T)$ .  
dimension  $r$  dimension  $(m-r)$ .

Ref: lecture 14 : Orthogonal Vectors and Subspace MIT  
18.06SC Linear Algebra.  
Fall 2011

2e) If vector  $\{v_1, \dots, v_p\}$  span a subspace  $W$ ,  
if  $x$  is  $\perp$  to each  $v_j, j=1 \dots p$ ,  
then  $x \in W^\perp$ ;  $v_j \in \mathbb{R}^n$

Ans: True

see proof in (Q3).

also  $W^\perp \oplus W = \mathbb{R}^n$

direct sum =  
 $\downarrow$  open ( $W^\perp \cup W$ )  
direct sum of  
sub-space.

// See Maths Stack Exchange

"Proof vector space is a direct sum of subspace and its orthogonal complement".

(Q3) Let  $\omega \subset \mathbb{R}^n$ ,

$\omega^\perp$  be set of vectors  $\perp$  to  $\omega$ .

Show that  $\omega^\perp$  is a subspace of  $\mathbb{R}^n$ .

Ans:

After we need to proof the 3 properties  
that a subspace must have.

a)  $\underline{0} \in \omega^\perp$

b) closed under addition

b) closed under multiplication  
scalar

a)  $\underline{0}$  is  $\perp$  to  $\omega$

Since  $u \in \omega$ ,  $u \cdot \underline{0} = \underline{0}$ .

$\therefore$  actually  $\underline{0} \in \omega$  for  $\omega$  to be

$$\Rightarrow \omega \cap \omega^\perp = \{\underline{0}\}$$

contains the zero vector.

Note: This question is related to  
row space  $C(A^T)$  vs  $N(A)$

that each is a subspace of  $\mathbb{R}^n$ .

b) if  $z \in \omega^\perp$ ,  $u \in \omega$

,  $c \in \mathbb{R}$  (scalar)

then  $(cz) \cdot u = c(z \cdot u) = c\underline{0} = \underline{0}$

since  $u$  is any element in  $\omega$ ,

$$\therefore cz \in \omega^\perp$$

c) let  $z_1, z_2 \in \omega^\perp$   
 $u \in \omega$ ,

$$(z_1 + z_2) \cdot u = z_1 \cdot u + z_2 \cdot u$$

$$= \underline{0} + \underline{0}$$

$$= \underline{0}$$

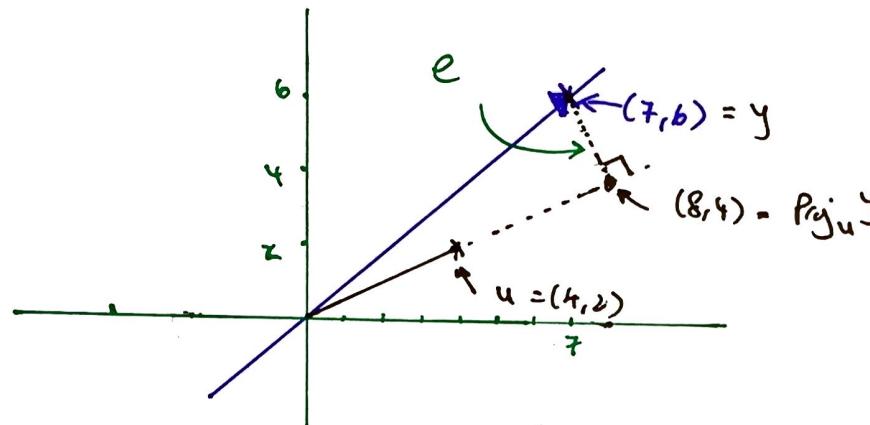
$$\therefore (z_1 + z_2) \in \omega^\perp$$

$\therefore$  closed under addition.

$\therefore \omega^\perp$  is a subspace.

Q Let  $y = \begin{bmatrix} 7 \\ 6 \end{bmatrix}$ ,  $u = \begin{bmatrix} 4 \\ 2 \end{bmatrix}$ , find orthogonal proj of  $y$  onto  $u$ . Write  $y$  as sum of 2 orthogonal vectors, one in  $\text{Span}\{u\}$  and one orthogonal to  $u$ .

Ans:



orthogonal projection of  $y$  onto  $u$  =  $\text{Proj}_u y$

$$\begin{aligned} \text{Proj}_u y &= \hat{y} = \frac{y \cdot u}{u \cdot u} u = \frac{\begin{bmatrix} 7 & 6 \end{bmatrix} \begin{bmatrix} 4 & 2 \end{bmatrix}}{16 + 4} \cdot \begin{bmatrix} 4 \\ 2 \end{bmatrix} \\ &= \left( \frac{40}{20} \right) \begin{bmatrix} 4 \\ 2 \end{bmatrix} \\ &= \begin{bmatrix} 8 \\ 4 \end{bmatrix} \quad \blacksquare \end{aligned}$$

$$e \perp u = y - \text{Proj}_u y$$

$$= \begin{bmatrix} 7 \\ 6 \end{bmatrix} - \begin{bmatrix} 8 \\ 4 \end{bmatrix} = \begin{bmatrix} -1 \\ 2 \end{bmatrix} \quad \blacksquare$$

Sanity check  $e \perp \text{Proj}_u y$

$$= [-1 \ 2] \begin{bmatrix} 8 \\ 4 \end{bmatrix} = -8 + 8 = 0.$$

$$\begin{aligned} \therefore y &= \begin{bmatrix} 7 \\ 6 \end{bmatrix} = \hat{y} + e \\ &= \begin{bmatrix} 8 \\ 4 \end{bmatrix} + \begin{bmatrix} -1 \\ 2 \end{bmatrix} \end{aligned}$$

Key idea:  $\hat{y} = \text{Proj}_u y$  is closest point and can be projected along  $u$  to  $y$ . by the error  $e = y - \hat{y}$  is orthogonal to  $\{u\}$ .

Ref:  
see 6.2.2 pg 4

5a) Determine if the follow set of vectors is orthonormal, if a set is only orthogonal, normalize the vectors to produce an orthonormal set.

Ans: orthogonal means  $u \cdot v = 0$ .

orthonormal means  $u \cdot v = 0$

and  $\|u\| = 1, \|v\| = 1$ .

$$u = \begin{bmatrix} y_3 \\ y_3 \\ y_3 \end{bmatrix}, v = \begin{bmatrix} -y_2 \\ 0 \\ +y_2 \end{bmatrix}$$

$$\therefore u \cdot v = [y_3 \ y_3 \ y_3] \begin{bmatrix} -y_2 \\ 0 \\ +y_2 \end{bmatrix} = -y_6 + 0 + y_6 = 0$$

$\Rightarrow$  orthogonal.

$$\text{But } \|u\| = \sqrt{u \cdot u} = \sqrt{y_3^2 + y_3^2 + y_3^2} = \sqrt{3y_3^2}$$

$$\|v\| = \sqrt{v \cdot v} = \sqrt{(-y_2)^2 + 0^2 + (y_2)^2} = \sqrt{2y_2^2}$$

To produce a vector with unit length.

$$\hat{u} = \frac{u}{\|u\|} = \frac{1}{\sqrt{3}} \begin{bmatrix} y_3 \\ y_3 \\ y_3 \end{bmatrix}; \quad \frac{1}{\sqrt{3}} = \sqrt{\frac{1}{3}}$$

$$= \begin{bmatrix} \sqrt{\frac{1}{3}} \\ \sqrt{\frac{1}{3}} \\ \sqrt{\frac{1}{3}} \end{bmatrix}.$$

$$\text{Length check } \|\hat{u}\|^2 = \left[ \sqrt{\frac{1}{3}} \ \sqrt{\frac{1}{3}} \ \sqrt{\frac{1}{3}} \right] \begin{bmatrix} \sqrt{\frac{1}{3}} \\ \sqrt{\frac{1}{3}} \\ \sqrt{\frac{1}{3}} \end{bmatrix}$$

$$= \frac{3}{9} + \frac{3}{9} + \frac{3}{9} = 1$$

To produce  $\hat{v}$ , normalize  $v$ ,

$$\hat{v} = \frac{v}{\|v\|} = \frac{1}{\sqrt{2}} \begin{bmatrix} -y_2 \\ 0 \\ +y_2 \end{bmatrix} = \begin{bmatrix} -\frac{y_2}{\sqrt{2}} \\ 0 \\ \frac{y_2}{\sqrt{2}} \end{bmatrix}$$

—————

b) Is  $\begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ -1 \\ 0 \end{bmatrix}$  orthogonal, orthonormal?

$u \quad \quad v$

$$\|u\| = 1, \|v\| = 1.$$

$$\text{But } u \cdot v = [0 \ 1 \ 0] \begin{bmatrix} 0 \\ -1 \\ 0 \end{bmatrix} = -1$$

$\neq 0 \therefore$  these two vectors are not orthogonal.

$\{u, v\}$  is not an orthogonal set.

Q6) a) Not every linearly independent set in  $\mathbb{R}^n$  is an orthogonal set.

Ans: TRUE  
 example:  $\left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$  is independent for  $\mathbb{R}^2$  BUT not orthogonal.

6b) If  $y$  is a linear combination of non-zero vectors from an orthogonal set, then the weights in the linear combination can be computed w/o row operations on a matrix.

Ans: TRUE.

let  $y = Ax$  where  $A$  has orthogonal col

$$= \begin{bmatrix} \uparrow & \uparrow & \uparrow \\ u_1 & u_2 & \dots & u_n \\ \downarrow & \downarrow & & \downarrow \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

:  $A \in \mathbb{R}^{M \times n}$

:  $x \in \mathbb{R}^n$

:  $y \in \mathbb{R}^M$

; col  $A$  forms an orthogonal set

Ans: TRUE

Ref: 6.2.5 pg 2 Theorem 8

Theorem 5 of Lay (pg 339)  
 $\frac{4}{11}$  edit

$\{u_1, u_2, \dots, u_n\}$  is an orthogonal set.

$y = u_1x_1 + u_2x_2 + \dots + u_nx_n$ , to find  $x_i$ ,  
 since  $u_i \cdot u_j = 0$  when  $i \neq j$ ; q by  $x_i$  on both sides.

$\therefore$  each weight can be found by:

$$x_i = \frac{u_i \cdot y}{u_i \cdot u_i} \quad (i = 1 \dots n)$$

6e) If the vectors in an orthogonal set is normalized, some of the new vectors may not be orthogonal.

Ans: False : if  $\{v_1, v_2, \dots, v_n\}$  are orthogonal, but each col vector is not unit length, then

$$\{\hat{v}_1, \hat{v}_2, \dots, \hat{v}_n\} = \left\{ \frac{v_1}{\|v_1\|}, \frac{v_2}{\|v_2\|}, \dots, \frac{v_n}{\|v_n\|} \right\}$$

normalized.

$$\text{and } \frac{v_i}{\|v_i\|} \cdot \frac{v_j}{\|v_j\|} = \left( \frac{1}{\|v_i\|} \frac{1}{\|v_j\|} \right) (v_i \cdot v_j)$$

$$= 0 \text{ if } i \neq j$$

Scaling each vector by its norm to form unit norm vector does not change its direction.

6d) A matrix with orthonormal col is an orthogonal matrix.

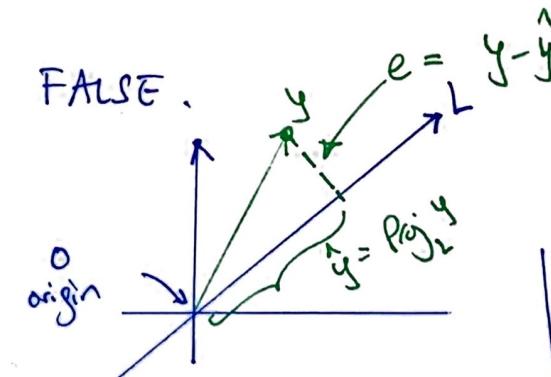
Ans: False.

Reason: orthogonal matrix MUST BE SQUARE.

A matrix with orthonormal col need NOT be square.

6e) If  $L$  is a line through  $O$  and if  $\hat{y}$  is the orthogonal projection of  $y$  onto  $L$ , then  $\|\hat{y}\|$  gives the distance from  $y$  to  $L$ .

Ans: FALSE.



see 6.2.5/pg 4

$$\begin{aligned} \text{distance } y \text{ to } L &= \|e\| \\ &= \|y - \hat{y}\| \end{aligned}$$

$e$  = vector perp to  $L$ .

$\hat{y}$  = Nearest pt on  $L$  to  $y$ .

$\|\hat{y}\|$  = length of approx.  $y$  using vector along  $L$ .

24a) Not every orthogonal set in  $\mathbb{R}^n$  is independent.

Ans: **TRUE**. = "bcz can contain  $\underline{0}$  vector".

But every orthogonal set of "Non-zero" vectors is linearly independent.

Note: Once a set contains the Null vector,  $\swarrow$  is orthogonal to every vector, the set of vectors is linearly dependent.

Why? If  $\{v_1, v_2, \dots, v_n\}$  are linearly independent,

$$c_1 v_1 + c_2 v_2 + \dots + c_n v_n = 0$$

$$\text{iff } c_1, c_2, \dots, c_n = 0.$$

$\therefore$  if one of the vector, e.g.  $v_i = \underline{0}$   
then  $c_i \neq 0, v_i = 0$

$$c_2 \dots c_n = 0 \therefore \text{rhs} = 0$$

24b) If a set  $S = \{u_1, u_2, \dots, u_p\}$  has the property  $u_i \cdot u_j = 0$  whenever  $i \neq j$  then  $S$  is an orthonormal set:

Ans: **False**. It is an orthogonal set.

To be orthonormal, each  $\|u_i\| = 1$

Note:

An **Orthogonal matrix** is a square matrix, and its col norm = 1 and is orthogonal to other col. | (orthonormal)

see 6.2.4 pg 3+4

24c) If the col of  $m \times n$  matrix are orthonormal, then the linear mapping  $x \rightarrow Ax$  preserve lengths.

Ans: **TRUE**

see Theorem 7 of Lay Pg 343.  
Theorem 7a)  $\|Ux\| = \|x\|$ .

Note  $(Ux) \cdot (Uy) = [Ux]^T [Uy]$

row vector                                  col vector

$$\begin{aligned} &= x^T U^T U y \\ &= x^T y. \\ &= x \cdot y. \end{aligned}$$

Hence if  $x = y \therefore (Ux) \cdot (Ux) = x \cdot x$

see 6.2.4  
pg 41  
Theorem 7

24d) The orthogonal projection of  $y$  onto  $v$  is the same as orthogonal projection of  $y$  onto  $c v$  whenever  $c \neq 0$ .

Ans: **TRUE**

Proof: suppose  $\text{Proj}_v y = \frac{y \cdot v}{v \cdot v} v$

replace  $v$  by  $c v \Rightarrow \frac{y \cdot cv}{cv \cdot cv} cv$

$$= \frac{c(y \cdot v)}{c^2(v \cdot v)} cv$$

$$= \frac{y \cdot v}{v \cdot v} v \quad \square$$

$\therefore \text{Proj}_v y$  does not depend on

$c$ , i.e. only need non-zero  $v$ .  
 $\xrightarrow{\text{see 6.2.4/pg 3}}$

24e) An orthogonal matrix is invertible.

Ans: **TRUE**, since  $Q^T Q = I = Q Q^T$   
 $Q = \text{orthogonal matrix } Q^{-1} = Q^T$