

## School of Statistics, University of the Philippines (Diliman) Linangan ng Estadistika, Unibersidad ng Pilipinas (Diliman)

26 of September 2015

Al-Ahmadgaid B. Asaad PS 1 | Stat 233

1.  $V^{\perp}$  and V|W are subspace.

Proof.

(a) Let  $\mathbf{w} \in V$  where  $V \subset \mathbb{R}^n$  is a subspace. And consider  $\mathbf{u}, \mathbf{v} \in V^{\perp}$  the orthogonal complement of V. Then  $\langle \mathbf{u}, \mathbf{w} \rangle = 0$  and  $\langle \mathbf{v}, \mathbf{w} \rangle = 0$ ,  $\forall \mathbf{u}, \mathbf{v} \in V^{\perp}$ . Now it follows that:

$$\langle \mathbf{u}, \mathbf{w} \rangle + \langle \mathbf{v}, \mathbf{w} \rangle = \langle \mathbf{u} + \mathbf{v}, \mathbf{w} \rangle = 0$$
, i.e.  $\mathbf{u} + \mathbf{v} \in V^{\perp}$ .

Also,

$$\langle k\mathbf{u}, \mathbf{w} \rangle = k \langle \mathbf{u}, \mathbf{w} \rangle = k \cdot 0 = 0$$
  
 $\langle k\mathbf{v}, \mathbf{w} \rangle = k \langle \mathbf{v}, \mathbf{w} \rangle = k \cdot 0 = 0.$ 

Implying,  $V^{\perp}$  is closed under addition and scalar multiplication. Therefore,  $V^{\perp}$  is a subspace.

- (b) By definition of quotient space,  $V|W=V\cap W^{\perp}$  and that  $W\subset V$ . Consider the following cases:
  - Case 1. Let  $\mathbf{w} \in W$  and  $\mathbf{v}_1, \mathbf{v}_2 \in V|W$ , then  $\langle \mathbf{w}, \mathbf{v}_1 \rangle = 0$  and  $\langle \mathbf{w}, \mathbf{v}_2 \rangle = 0$  since  $W \perp V|W$  by definition of quotient space  $(\cdot \mod \cdot)$ . So that,

$$\langle \mathbf{w}, \mathbf{v}_1 \rangle + \langle \mathbf{w}, \mathbf{v}_2 \rangle = \langle \mathbf{w}, \mathbf{v}_1 + \mathbf{v}_2 \rangle = 0.$$

And

$$\langle k\mathbf{w}, \mathbf{v}_1 \rangle = k \langle \mathbf{w}, \mathbf{v}_1 \rangle = 0$$
  
 $\langle k\mathbf{w}, \mathbf{v}_2 \rangle = k \langle \mathbf{w}, \mathbf{v}_2 \rangle = 0.$ 

Hence V|W is closed under addition and scalar multiplication.

Case 2. Let  $\mathbf{u} \in V^{\perp}$  and  $\mathbf{v}_1, \mathbf{v}_2 \in V|W$ , then  $\langle \mathbf{u}, \mathbf{v}_1 \rangle = 0$  and  $\langle \mathbf{u}, \mathbf{v}_2 \rangle = 0$  since  $V^{\perp}$  is orthogonal to V|W. It follows that

$$\langle \mathbf{u}, \mathbf{v}_1 \rangle + \langle \mathbf{u}, \mathbf{v}_2 \rangle = \langle \mathbf{u}, \mathbf{v}_1 + \mathbf{v}_2 \rangle = 0.$$

Further,

$$\langle k\mathbf{u}, \mathbf{v}_1 \rangle = k \langle \mathbf{u}, \mathbf{v}_1 \rangle = 0$$
  
 $\langle k\mathbf{u}, \mathbf{v}_2 \rangle = k \langle \mathbf{u}, \mathbf{v}_2 \rangle = 0.$ 

Then again V|W is closed under addition and scalar multiplication.

From two cases above, one can conclude that V|W is a subspace.

2. Let  $W \subset V \subset \mathbb{R}^n$ . Then  $\dim(V|W) = \dim V - \dim W$ .

To prove this problem, let's formally define what the dimension (dim) of a subspace first.

**Definition 1.** If S is a subspace of  $\mathbb{R}^n$ , then the number of vectors in a basis for S is called the dimension of S, denoted by  $\dim(S)$ .

*Proof.* By definition of quotient space, if  $\mathbf{x} \in V|W$  and  $\mathbf{y} \in W$ , then  $\langle \mathbf{x}, \mathbf{y} \rangle = 0$  since  $V|W \perp W$ . Now consider  $\mathbf{u}_1, \dots, \mathbf{u}_p$  be mutually orthogonal nonzero vectors that span V|W. Also, let  $\mathbf{u}_{p+1}, \dots, \mathbf{u}_{p+k}$  be mutually orthogonal nonzero vectors that span W. If  $\mathbf{X} = [\mathbf{u}_1, \dots, \mathbf{u}_p]$  and  $\mathbf{Y} = [\mathbf{u}_{p+1}, \dots, \mathbf{u}_{p+k}]$  be matrices with column vectors  $\mathbf{u}_i$  and  $\mathbf{u}_j$ , respectively,  $i = 1, \dots, p$  and  $j = p+1, \dots, p+k$ . Then  $\mathbf{X}$  and  $\mathbf{Y}$  forms the basis of V|W and W since  $\mathbf{u}_i$ s and  $\mathbf{u}_j$ s are linearly independent vectors  $\forall i, j$  by Lemma 1.6. So that,

$$\dim(V|W) = \# \text{ of columns in } \mathbf{X} = p,$$

$$\dim(W) = \#$$
 of columns in  $\mathbf{Y} = k$ 

Given that  $W \subset V$ , then  $W = V \cap W$ . Thus,

$$\dim(V|W) = \dim(V \cap W^{\perp}) = \# \text{ of columns in } \mathbf{X} = p,$$

$$\dim(W) = \dim(V \cap W) = \# \text{ of columns in } \mathbf{Y} = k.$$

Stat 233 PS 1

Because subspace is a set of all vectors that span the said space, then from set theory, it follows that

$$V = \{V \cap W\} \cup \{V \cap W^{\perp}\}, \text{ for } W \subset V.$$

What is left to show now is that,

$$\mathcal{L}(\{V \cap W\} \cup \{V \cap W^{\perp}\}) = V,$$

that is, the span of the vectors in the set  $\{V \cap W\} \cup \{V \cap W^{\perp}\}$  is the subspace V. Let  $\mathbf{v} \in V$ , then by Theorem 1.4(f),

$$P_{V|W}\mathbf{v} = P_V\mathbf{v} - P_W\mathbf{v}$$
$$P_V\mathbf{v} = P_{V|W}\mathbf{v} + P_W\mathbf{v}$$

It follows that  $P_W \mathbf{v} \in W$ , and because  $W = \{\mathbf{w} : \mathbf{X}\mathbf{a} = \mathbf{w}\}$  for  $\mathbf{X} = [\mathbf{u}_{p+1}, \dots, \mathbf{u}_{p+k}]$  the basis matrix of W is a subset of V. We have,

$$P_W \mathbf{v} = \sum_{j=p+1}^{p+k} x_j a_j \text{ spans } W.$$

Also  $P_{V|W}\mathbf{v} \in V|W$  since  $V|W = {\mathbf{v} : \mathbf{Ya} = \mathbf{v}}$  for  $\mathbf{Y} = [\mathbf{u}_1, \dots, \mathbf{u}_p]$  the basis matrix of V|W is a subset of V. So

$$P_{V|W}\mathbf{v} = \sum_{i=1}^{p} x_i a_i \text{ spans } V|W.$$

Implying

$$P_V \mathbf{v} = P_W \mathbf{v} + P_{V|W} \mathbf{v} = \sum_{i=1}^p x_i a_i + \sum_{j=p+1}^{p+k} x_j a_j = \sum_{i=1}^{p+k} x_i a_i,$$

that is,  $P_V \mathbf{v} \in V$  and that  $V = \{\mathbf{v} : [\mathbf{X}, \mathbf{Y}] \mathbf{b} = \mathbf{v}\}$ . And because  $\mathbf{X}$  and  $\mathbf{Y}$  are bases of orthogonal subspaces  $(W \perp V|W)$  then the matrix  $\mathbf{Z} = [\mathbf{X}, \mathbf{Y}]$  say, forms a basis matrix for V since the columns of  $\mathbf{Z}$  are linearly independent. Therefore from the definition of dimension of a matrix,

$$\dim(V) = \dim(W) + \dim(V|W)$$
$$\dim(V|W) = \dim(V) - \dim(W)$$

3. Let  $W \subset V$ . Then V|(V|W) = W.

*Proof.* We need to show the following first:

- (a)  $W \subset V|(V|W)$ ; and,
- (b)  $\dim(W) = \dim(V|(V|W))$

So that by result from Linear Algebra, V|(V|W) = W.

(a) Let  $\mathbf{x} \in W$ , then  $\mathbf{x} \in V$ . If  $\mathbf{y} \in V|W$ , then  $\langle \mathbf{x}, \mathbf{y} \rangle = 0$ , since  $(V|W = V \cap W^{\perp}) \perp W$ . Implies,  $V|(V|W) = V \cap (V|W)^{\perp}$ . Now since a subspace is a set of all vectors that span the said space, then by set theory, it follows that

$$V \cap (V|W)^{\perp} = V \cap \{V \cap W^{\perp}\}^{\perp}$$
$$= V \cap \{V^{\perp} \cup W\}$$
$$= \{V \cap V^{\perp}\} \cup \{V \cap W\}$$
$$= V \cap W = W.$$

Hence, if  $\mathbf{x} \in W$ , then  $\mathbf{x} \in V|(V|W)$ . And therefore, we conclude that  $W \subset V|(V|W)$ .

(b) From problem 2, if  $W \subset V \subset \mathbb{R}^n$ , then  $\dim(V|W) = \dim V - \dim W$ . Thus,

$$\dim(V|(V|W)) = \dim V - \dim V|W$$

$$= \dim V - (\dim V - \dim W)$$

$$= \dim W.$$

With results from part (a)  $W \subset V|(V|W)$ ; and part (b)  $\dim(W) = \dim(V|(V|W))$ . V|(V|W) = W.

4. Show that  $W \subset V$  if and only if  $W^{\perp} \supset V^{\perp}$ 

Proof.

(a) Assuming  $W \subset V$ , then  $W^{\perp} \supset V^{\perp}$ .

Case 1. Let  $\mathbf{u} \in W$ , then  $\mathbf{u} \in V$ . Implies that  $\mathbf{u} \notin W^{\perp}$  and  $\mathbf{u} \notin V^{\perp}$  since

$$\langle \mathbf{u}, \mathbf{a} \rangle = 0$$
 and  $\langle \mathbf{u}, \mathbf{b} \rangle = 0$ 

 $\forall \mathbf{a} \in W^{\perp} \text{ and } \forall \mathbf{b} \in V^{\perp}.$ 

Stat 233 PS 1

Case 2. Now if  $\mathbf{u} \notin W$  but  $\mathbf{u} \in V$ , then  $\mathbf{u} \in W^{\perp}$  and  $\mathbf{u} \notin V^{\perp}$ . That is,

$$\langle \mathbf{u}, \mathbf{a} \rangle \neq 0$$
 and  $\langle \mathbf{u}, \mathbf{b} \rangle = 0$ 

 $\forall \mathbf{a} \in W^{\perp} \text{ and } \forall \mathbf{b} \in V^{\perp}.$ 

Case 3. Further, if  $\mathbf{u} \notin W$  and  $\mathbf{u} \notin V$ , then  $\mathbf{u} \in W^{\perp}$  and  $\mathbf{u} \in V^{\perp}$ , that is

$$\langle \mathbf{u}, \mathbf{a} \rangle \neq 0$$
 and  $\langle \mathbf{u}, \mathbf{b} \rangle \neq 0$ 

 $\forall \mathbf{a} \in W^{\perp} \text{ and } \forall \mathbf{b} \in V^{\perp}.$ 

To summarize,  $\mathbf{u} \in W^{\perp} \cap V^{\perp}$  in Case 3. And in Case 2,  $\mathbf{u} \in W^{\perp} \backslash V^{\perp}$ . Therefore,  $W^{\perp} \supset V^{\perp}$ .

(b) Assuming  $W^{\perp} \supset V^{\perp}$ , then  $W \subset V$ .

Let's prove this by contradiction, suppose not. That is, suppose  $W^{\perp} \supset V^{\perp}$  implies  $W \supset V$ .

Case 1. Let  $\mathbf{x} \in V^{\perp}$ , then  $\mathbf{x} \in W^{\perp}$ . It follows that,  $\mathbf{x} \notin V$  and  $\mathbf{x} \notin W$  since

$$\langle \mathbf{x}, \mathbf{a} \rangle = 0$$
 and  $\langle \mathbf{x}, \mathbf{b} \rangle = 0$ 

 $\forall \mathbf{a} \in V \text{ and } \forall \mathbf{b} \in W.$ 

Case 2. Now consider  $\mathbf{x} \notin V^{\perp}$  but  $\mathbf{x} \in W^{\perp}$ . Then  $\mathbf{x} \in V$  and  $\mathbf{x} \notin W$ . That is,

$$\langle \mathbf{x}, \mathbf{a} \rangle \neq 0$$
 and  $\langle \mathbf{x}, \mathbf{b} \rangle = 0$ 

 $\forall \mathbf{a} \in V \text{ and } \forall \mathbf{b} \in W.$ 

Case 3. Finally, if  $\mathbf{x} \notin V^{\perp}$  and  $\mathbf{x} \notin W^{\perp}$ . Then  $\mathbf{x} \in V$  and  $\mathbf{x} \in W$ . That is,

$$\langle \mathbf{x}, \mathbf{a} \rangle \neq 0$$
 and  $\langle \mathbf{x}, \mathbf{b} \rangle \neq 0$ 

 $\forall \mathbf{a} \in V \text{ and } \forall \mathbf{b} \in W.$ 

To summarize,  $\mathbf{x} \in V \cap W$  in Case 3, and  $\mathbf{x} \in V \backslash W$  in Case 2. Therefore  $W \subset V$ , which is a contradiction f.

5. Let  $\mathbf{Y} \sim N_n(\boldsymbol{\mu}, \boldsymbol{\Sigma})$ . Then  $(\mathbf{Y} - \boldsymbol{\mu})' \boldsymbol{\Sigma}^{-1} (\mathbf{Y} - \boldsymbol{\mu}) \sim \chi_n^2(\mathbf{0})$ .

*Proof.* If  $\mathbf{Z} = \mathbf{Y} - \boldsymbol{\mu}$ , then  $\mathbf{Z} \sim N_n(\mathbf{0}, \boldsymbol{\Sigma})$ . So that from Theorem 2.13(a),  $(\mathbf{Y} - \boldsymbol{\mu})' \boldsymbol{\Sigma}^{-1} (\mathbf{Y} - \boldsymbol{\mu}) \sim \chi_n^2 ((\boldsymbol{\mu} - \boldsymbol{\mu})' \boldsymbol{\Sigma}^{-1} (\boldsymbol{\mu} - \boldsymbol{\mu}))$ . Or simply,

$$(\mathbf{Y} - \boldsymbol{\mu})' \boldsymbol{\Sigma}^{-1} (\mathbf{Y} - \boldsymbol{\mu}) \sim \chi_n^2(\mathbf{0}).$$