

1. (a) Consider each of the following sets of vectors in  $R^4$  where  $a$ ,  $b$  and  $c$  represent any real numbers.

All vectors of the form:

$$(i) \begin{bmatrix} a \\ b \\ a+b \\ 1 \end{bmatrix} \quad (ii) \begin{bmatrix} a \\ b \\ a+b \\ 0 \end{bmatrix} \quad (iii) \begin{bmatrix} a \\ b \\ c \\ a+b-2c \end{bmatrix}$$

In each case, determine whether the set of vectors represents a subspace of  $R^4$  (you need to justify your answer).

- (b) Consider the subspace of  $R^4$  defined as all vectors of the form

$$\begin{bmatrix} a \\ a+b \\ a-b \\ -2b \end{bmatrix}$$

What is the dimension of this subspace? Give a set of vectors that form a basis for this subspace.

- (c) Consider a random vector  $X \sim N_3(\mu_X, \Sigma_X)$ :

$$X = \begin{bmatrix} X_1 \\ X_2 \\ X_3 \end{bmatrix}, \quad \mu_X = \begin{bmatrix} \mu_1 \\ \mu_2 \\ \mu_3 \end{bmatrix}, \quad \Sigma_X = \begin{bmatrix} 2 & 1 & 2 \\ 1 & 2 & -1 \\ 2 & -1 & 3 \end{bmatrix}$$

- i. Find a matrix  $A$  such that if we let  $W = AX$  then:

$$\mu_W = \begin{bmatrix} \mu_1 - \mu_2 \\ \mu_2 + 2\mu_3 \\ 3\mu_1 - 2\mu_2 - \mu_3 \end{bmatrix}$$

- ii. Find the covariance matrix for  $W$ , i.e.  $\Sigma_W$ .

- (d) The Best Linear Unbiased Estimate (BLUE) of a parameter is defined as the estimator that has the smallest variance among the set of all unbiased linear estimators. Note a linear estimator is simply any estimator which is a linear combination of the data, i.e. can be written as  $\mathbf{a}^t \mathbf{Y}$ .

Given the linear model,

$$\mathbf{Y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\epsilon} \quad \text{where} \quad \boldsymbol{\beta} \sim N(\mathbf{0}, \sigma^2 \mathbf{I}_n)$$

and the least squares estimates of  $\boldsymbol{\beta}$

$$\hat{\boldsymbol{\beta}} = (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t \mathbf{Y},$$

consider estimating a linear combination of the parameters  $\mathbf{a}^t \boldsymbol{\beta}$ . We can show that among all possible linear unbiased estimators of  $\mathbf{a}^t \boldsymbol{\beta}$ ,  $\mathbf{a}^t \hat{\boldsymbol{\beta}}$  has the smallest variance using the following steps.

- i. First, consider any linear estimator  $\mathbf{b}^t\mathbf{Y}$  and show:
  - A.  $\mathbf{b}^t\mathbf{Y}$  is unbiased for  $\mathbf{a}^t\boldsymbol{\beta}$  if  $\mathbf{b}^t\mathbf{X} = \mathbf{a}^t$ .
  - B.  $\text{Var}(\mathbf{b}^t\mathbf{Y}) = \sigma^2\mathbf{b}^t\mathbf{b}$ .
- ii. Now consider the estimator based on the least squares estimates  $\mathbf{a}^t\hat{\boldsymbol{\beta}}$ . Show the following:
  - A.  $\mathbf{a}^t\hat{\boldsymbol{\beta}}$  is unbiased for  $\mathbf{a}^t\boldsymbol{\beta}$ ,
  - B.  $\text{Var}(\mathbf{a}^t\hat{\boldsymbol{\beta}}) = \sigma^2\mathbf{a}^t(\mathbf{X}^t\mathbf{X})^{-1}\mathbf{a}$
- iii. Now set  $\mathbf{b}^* = \mathbf{X}(\mathbf{X}^t\mathbf{X})^{-1}\mathbf{a}$ . For any  $\mathbf{b}$

$$(\mathbf{b} - \mathbf{b}^*)^t(\mathbf{b} - \mathbf{b}^*) \geq 0$$

and the equality holds only if  $\mathbf{b} = \mathbf{b}^*$ . Show that

$$\sigma^2(\mathbf{b} - \mathbf{b}^*)^t(\mathbf{b} - \mathbf{b}^*) = \text{Var}(\mathbf{b}^t\mathbf{Y}) - \text{Var}(\mathbf{b}^{*t}\mathbf{Y})$$

which means  $\text{Var}(\mathbf{b}^t\mathbf{Y}) \geq \text{Var}(\mathbf{b}^{*t}\mathbf{Y})$  and the equality holds only if  $\mathbf{b} = \mathbf{b}^*$ .

Hint: expand the left hand side and simplify using the results in (a) and (b).