

get

$$\begin{pmatrix} \theta \\ 0 \end{pmatrix} = \begin{pmatrix} X \\ H \end{pmatrix} \beta = \left( \begin{array}{ccccc} 1 & 1 & 0 & 1 & 0 \\ 1 & 1 & 0 & 0 & 1 \\ 1 & 0 & 1 & 1 & 0 \\ 1 & 0 & 1 & 0 & 1 \\ \hline 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 \end{array} \right) \begin{pmatrix} \mu \\ \alpha_1 \\ \alpha_2 \\ \tau_1 \\ \tau_2 \end{pmatrix},$$



where the augmented matrix now has five linearly independent columns. Thus given  $\theta$ ,  $\beta$  is now unique.

$$X(\beta - \hat{\beta}_1) = X(\beta - \hat{\beta}_2)$$

### EXERCISES 3h

1. Suppose that  $X$  does not have full rank, and let  $\hat{\beta}_i$  ( $i = 1, 2$ ) be any two solutions of the normal equations. Show directly that

$$\|Y - X\hat{\beta}_1\|^2 = \|Y - X\hat{\beta}_2\|^2. \quad \text{both sides.}$$

2. If the columns of  $X$  are linearly dependent, prove that there is no matrix  $C$  such that  $CY$  is an unbiased estimate of  $\beta$ .

*+ Unbiased est of  $\beta$ , ask for  $\beta + N(\lambda)$ .  $X(\hat{\beta}_1 - \hat{\beta}_2) = 0$*

#### 3.9.2 Estimable Functions

Since  $\hat{\beta}$  is not unique,  $\beta$  is not estimable. The question then arises: What can we estimate? Since each element  $\theta_i$  of  $\theta$  ( $= X\beta$ ) is estimated by the  $i$ th element of  $\hat{\theta} = PY$ , then every linear combination of the  $\theta_i$ , say  $b'\theta$ , is also estimable. This means that the  $\theta_i$  form a linear subspace of estimable functions, where  $\theta_i = x'_i \beta$ ,  $x'_i$  being the  $i$ th row of  $X$ . Usually, we define estimable functions formally as follows.

**Definition 3.1** The parametric function  $a'\beta$  is said to be estimable if it has a linear unbiased estimate,  $b'Y$ , say.

We note that if  $a'\beta$  is estimable, then  $a'\beta = E[b'Y] = b'\theta = b'X\beta$  identically in  $\beta$ , so that  $a' = b'X$  or  $a = X'b$  (A.11.1). Hence  $a'\beta$  is estimable if and only if  $a \in C(X')$ .

**EXAMPLE 3.8** If  $a'\beta$  is estimable, and  $\hat{\beta}$  is any solution of the normal equations, then  $a'\hat{\beta}$  is unique. To show this we first note that  $a = X'b$  for some  $b$ , so that  $a'\beta = b'X\beta = b'\theta$ . Similarly,  $a'\hat{\beta} = b'X\hat{\beta} = b'\hat{\theta}$ , which is unique. Furthermore, by Theorem 3.2,  $b'\hat{\theta}$  is the BLUE of  $b'\theta$ , so that  $a'\hat{\beta}$  is the BLUE of  $a'\beta$ .  $\square$

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