

Show  $E[(x_0\hat{\delta} - x\beta)'(x_0\hat{\delta} - x\beta)] = \sigma^2 \text{tr}(M_0) + \beta'x'(1-M_0)x\beta$ ,

where  $M_0$  is p.p.m. onto  $C(x_0)$  and  $\hat{\delta}$  is L.S. estimator in simple model.

Here:  $Y = X\beta + \varepsilon$ ,  $\varepsilon \sim N(0, \sigma^2 I_n) \Rightarrow Y \sim N(X\beta, \sigma^2 I_n)$ ,  $E[Y] = X\beta$   
 $\text{Cov}(Y) = \sigma^2 I_n$

Since  $\hat{\delta}$  is LS,  $x_0\hat{\delta} = M_0 Y$ .  $E[x_0\hat{\delta}] = E[M_0 Y] = M_0 \cdot E[Y] = M_0 X\beta$

The left side, above, can be written with terms swapped:  $E[(x\beta - x_0\hat{\delta})'(x\beta - x_0\hat{\delta})] = *$

Now call  $z = x\beta - x_0\hat{\delta}$ , so that  $*$  =  $E[z'z]$ .

Using the variance identity:  $E[z'z] = \text{Cov}(z) + E[z]^2$ .

First, simplify  $z = x\beta - x_0\hat{\delta} = x\beta - M_0 Y$

Then find  $E[z] = E[x\beta - M_0 Y] = x\beta - M_0 \cdot E[Y] = x\beta - M_0 X\beta = (1-M_0)x\beta$

and  $\text{Cov}(z) = \text{Cov}(x\beta - M_0 Y) = M_0 \cdot \text{Cov}(Y) \cdot M_0' = M_0 \cdot \sigma^2 I_n \cdot M_0' = \sigma^2 I_n \cdot M_0 \cdot M_0' = \sigma^2 I_n \cdot M_0$   
 idempotent  $\uparrow$

Therefore  $E[z'z] = \text{Cov}(z) + E[z]^2$

$= \sigma^2 I_n \cdot M_0 + [(1-M_0)x\beta]'[(1-M_0)x\beta]$

$= \sigma^2 \cdot M_0 + \beta'x'(1-M_0)'(1-M_0)x\beta$

$= \sigma^2 \cdot \text{tr}(M_0) + \beta'x'(1-M_0)x\beta \quad \leftarrow \text{idempotent}$