

## A simple proof of the FWL Theorem

Here is a simple proof of the FWL Theorem (established in class). Before proceeding be sure that you understand the basic properties of  $M$  and  $X$ .

Start with:

$$Y = (P + M)Y = PY + MY = X_1\hat{\beta}_1 + X_2\hat{\beta}_2 + MY.$$

Multiply both sides by  $X_1'M_2$  to get:

$$X_1'M_2Y = X_1'M_2X_1\hat{\beta}_1 + X_1'M_2X_2\hat{\beta}_2 + X_1'M_2MY$$

Now

- $M_2X_2 = 0$  by property of the  $M_2$  matrix, so the second term of the right hand side vanishes.
- $X_1'M_2MY = X_1^{*'}e$ , with  $X_1^{*'} \equiv X_1'M_2$  and  $e \equiv MY$ .

Now  $e$  are errors of the full regression of  $Y$  on  $X_1$  and  $X_2$ . Note

$$X'e = X'Mu = 0$$

since  $X'M = 0$ . This implies that all columns of  $X$  are uncorrelated with  $e$ .  $X_1^{*'}$  are residuals of regressing  $X_1$  on  $X_2$ , that is, they are the part of  $X_1$  not linearly explained by  $X_2$ . Consequently, by construction  $X_1^{*'}$  is correlated with  $X_1$  (though not with  $X_2$ ). Then  $X_1^{*'}e = 0$  since  $e$  is uncorrelated with  $X_1^{*'}$ . Hence the third term of the right hand side vanishes.

This leaves:  $X_1'M_2Y = X_1'M_2X_1\hat{\beta}_1$ . So:

$$\hat{\beta}_1 = (X_1'M_1X_1)^{-1} X_1'M_1Y = \tilde{\beta}_1$$