Let
$$\varepsilon_j = y_j - \hat{\mathcal{E}}[y_j | X]$$
 for $j = 1, 2$. Show $\mathcal{E}[\varepsilon_j] = 0$.

$$E[x_j] = E[y_j - \hat{E}[y_j|x]] = E[E[y_j - \hat{E}[y_j|x]] = E[O] = O$$

Uses law of iterated expectations, as in Thm 6.3.1.

Note: $\hat{E}[y_j|x] = My_j + (x-Mx)\beta_x = My_j + (x-Mx)V_{xx}^{-1}V_{xy_j}$, and therefore $E_j = (y_j - My_j) - (x-Mx)V_{xx}^{-1}V_{xy_j}$.

For
$$\rho_{y:x} = corr(\epsilon_1, \epsilon_2) = \frac{cov(\epsilon_1, \epsilon_2)}{\cdots} = 0 \Rightarrow cov(\epsilon_1, \epsilon_2) = 0$$

$$cov(\epsilon_2, \epsilon_1) = \left[cov(\epsilon_1, \epsilon_2)\right]' = 0' = 0$$

Now show $cov(\xi_2, \xi_1) = cov(y_2, \xi_1)$.

$$cov(\xi_2, \xi_1) = cov(y_2 - \hat{\xi}[y_2|x], \xi_1) = cov(y_2, \xi_1) - cov(\hat{\xi}[y_2|x], \xi_1)$$

where the second term = 0, because predictions are orthogonal to errors, so their correlation (and thus covariance) is 0.

Together,
$$\rho_{yx}=0 \Rightarrow cov(\epsilon_2, \epsilon_1) = cov(y_2, \epsilon_1) = 0$$
 } No correlation between ϵ_1 and ϵ_2 , an no correlation between y_2 and ϵ_1 , or equivalently between y_1 and ϵ_2 .

Show $cov(\varepsilon_j, x-\mu_x) = 0$.

$$cov(\varepsilon_{j}, x-\mu_{x}) = cov\left((y_{j}-\mu_{y_{j}}) - (x-\mu_{x})v_{xx}^{-1}v_{xy_{j}}, x-\mu_{x}\right)$$

$$= cov\left((y_{j}-\mu_{y_{j}}), x-\mu_{x}\right) - cov\left((x-\mu_{x})v_{xx}^{-1}v_{xy_{j}}, x-\mu_{x}\right)$$

$$= v_{y_{j}x} - v_{xx} \cdot v_{xx}^{-1}v_{xy_{j}} = 0$$

There is no correlation between the errors and the mean-centered data.