

Basic formulars:

$$g_n(\delta) = \bar{g} = \frac{1}{n} \sum_{i=1}^n g_i$$

GMM estimator:

$$\tilde{\delta}(\hat{W}) = (S'_{xz} \hat{W} S_{xz})^{-1} S'_{xz} \hat{W} s_{xy}$$

sample error:

$$\tilde{\delta} - \delta = (S'_{xz} \hat{W} S_{xz})^{-1} S'_{xz} \hat{W} \bar{g}$$

sample moment:

$$g_n(\tilde{\delta}) = s_{xy} - S_{xz} \tilde{\delta}$$

In **efficient GMM**: $\hat{W} = \hat{S}^{-1}$. By the definition, $\hat{W}$ is a $K \times K$ symmetric and positive definite matrix.

First-order Taylor expansion:

$$\begin{aligned} g_n(\tilde{\delta}) &\approx g_n(\delta) + \left. \frac{\partial g_n(\tilde{\delta})}{\partial \tilde{\delta}} \right|_{\tilde{\delta}=\delta} (\tilde{\delta} - \delta) \\ &= \bar{g} - S_{xz}(\tilde{\delta} - \delta) \\ &= \bar{g} - S_{xz}(S'_{xz} \hat{W} S_{xz})^{-1} S'_{xz} \hat{W} \bar{g} \end{aligned}$$

the weighted form projection matrix:

$$P = S_{xz}(S'_{xz} \hat{W} S_{xz})^{-1} S'_{xz} \hat{W}$$

$Rank(P) = L$, since S_{xz} is full column rank(=L).

$$g_n(\tilde{\delta}) = (I_K - P)\bar{g} = M g_n(\delta)$$

$$Rank(M) = tr(M) = tr(I_K) - tr(P) = K - L$$

Conclusion:

$$J(\delta, S^{-1}) = n \cdot \bar{g}' S^{-1} \bar{g} = (\sqrt{n} \bar{g})' S^{-1} (\sqrt{n} \bar{g}) \sim \chi^2(K)$$

$$J(\tilde{\delta}(\hat{S}^{-1}), \hat{S}^{-1}) = n \cdot g_n(\tilde{\delta}(\hat{S}^{-1}))' \hat{S}^{-1} g_n(\tilde{\delta}(\hat{S}^{-1})) \xrightarrow{d} \chi^2(K - L)$$

The reason for the loss of degrees of freedom is: When constructing $g_n(\tilde{\delta})$, we used $\tilde{\delta}$ to replace the true δ , which is equivalent to imposing L constraints on the sample moments (by projecting $\bar{g}$ onto the column space of S_{xz} through the projection matrix P), which can also explain why we use the degrees of freedom $n - K$ in OLS estimation.