

EC402 and FM437: additional handout.

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December 5, 2008

This is a (very) detailed proof of the unbiasedness of the s^2 estimator of σ_ϵ

Assume the model is:

$$Y = X\beta + \epsilon$$

(Where X is $N \times K$.) Then under $A1, A2, A3, A4GM$ ($A4GM$ is crucial here so let me spell it out: $Var(\epsilon|X) = \sigma_\epsilon^2 I_n$), the following holds:

$$E\left(\sum_i (\hat{\epsilon}_i^2) | X\right) = (N - K)\sigma_\epsilon^2$$

A proof goes like this:

$$E\left(\sum_i (\hat{\epsilon}_i^2) | X\right) = E(\hat{\epsilon}'\hat{\epsilon} | X) = E((M_X\epsilon)'(M_X\epsilon) | X) \quad (1)$$

$$\begin{aligned} &= E(\epsilon' M_X M_X \epsilon | X) \\ &= E(\epsilon' M_X \epsilon | X) \end{aligned} \quad (2)$$

In (1), I use the identity $M_X\epsilon = \hat{\epsilon}$; in (2) I use the idempotence of M_X .

Here is the property that I am going to need next:

Lemma 0.1. $\forall (k, n) \in \mathbb{N}^2$, if A is a $(N \times K)$ matrix and B is a $(K \times N)$ matrix, then $tr(AB) = tr(BA)$.

Proof. $C = AB \Rightarrow c_{ij} = \sum_{k=1}^K (a_{ik}b_{kj})$, hence $tr(AB) = \sum_{n=1}^N c_{nn} = \sum_{n=1}^N \sum_{k=1}^K (a_{nk}b_{kn}) = \sum_{k=1}^K \sum_{n=1}^N (b_{kn}a_{nk}) = \sum_{k=1}^K d_{kk}$ where $D = BA$, so $tr(AB) = tr(BA)$ \square

Back to the main proof:

$$\begin{aligned}
E\left(\sum_i(\hat{\epsilon}_i^2)|X\right) &= E\left(\text{tr}(\boldsymbol{\epsilon}'M_X\boldsymbol{\epsilon})|X\right) = E\left(\text{tr}(M_X\boldsymbol{\epsilon}\boldsymbol{\epsilon}')|X\right) \\
&= \text{tr}\left(E\left(M_X\boldsymbol{\epsilon}\boldsymbol{\epsilon}'|X\right)\right) \\
&= \text{tr}\left(M_X E\left(\boldsymbol{\epsilon}\boldsymbol{\epsilon}'|X\right)\right) & (3) \\
&= \text{tr}\left(M_X(\sigma_\epsilon^2 I_n)\right) & (4) \\
&= \sigma_\epsilon^2 \text{tr}(M_X) \\
&= (N - K)\sigma_\epsilon^2 & (5)
\end{aligned}$$

In (3), I use the linearity of conditional expectations (remember that the only stochastic thing in M_X is X , which is conditioned on in the expected value). In (4), I use the homoskedasticity assumption. In (5), I use the following lemma:

Lemma 0.2. *If P_X ($n \times n$) is an orthogonal projection matrix on a vector space of dimension k (with basis given by the k columns of the $(n \times k)$ matrix X , so that $P_X = X(X'X)^{-1}X'$), then $\text{tr}(P_X) = k$.*

Proof. $\text{tr}(P_X) = \text{tr}(X(X'X)^{-1}X') = \text{tr}(X'X(X'X)^{-1}) = \text{tr}(I_k)$ (I used Lemma (0.1) to commute within the trace operator.) \square

Corollary 0.3. *If M_X ($n \times n$) is the annihilator matrix of X , so that $M_X = I_n - P_X$, then $\text{tr}(M_X) = \text{tr}(I_n) - \text{tr}(P_X) = n - k$.*

Note that another way to think about Corollary (0.3) is to understand that M_X is also the orthogonal projection matrix on the orthogonal to the vector space spanned by the columns of X (i.e. the set of all vectors that are orthogonal to the columns of X), which is a vectorial space of dimension $n - k$: then Corollary (0.3) is a direct application of Lemma (0.2).

A useful reference on the geometry of OLS, if you are curious, would be chapter 2 of Davidson and McKinnon's "Econometric Theory and Methods".