INADMISSIBILITY OF THE SAMPLE VARIANCE

For any observations $X_1, ..., X_n$ with $n \geq 2$, assumed to be i.i.d. with finite variance σ^2 , let $s_X^2 = \frac{1}{n-1} \sum_{j=1}^n (X_j - \overline{X})^2$, the usual unbiased estimator of σ^2 . Y. G. Yatracos (2005) proved that s_X^2 is inadmissible as an estimator of σ^2 by providing an estimator which has smaller mean-squared error in all relevant cases. More precisely, he proved the following:

Theorem 1 (Yatracos). There is a constant c_n depending on n, namely $c_n = \frac{(n+2)(n-1)}{n(n+1)}$, such that for any $n \geq 2$ and for all $X_1, ..., X_n$ i.i.d. with $E(X_1^4) < +\infty$ and variance σ^2 with $\sigma > 0$, the mean-square error of $c_n s_X^2$ as an estimator of σ^2 is less than that of s_X^2 .

Proof. First let's find the mean-square error as a function of an unspecified c_n . We can assume without loss of generality that $\mu := EX_1 = 0$ because replacing all X_j by $X_j - \mu$ does not change either σ^2 or s_X^2 . Then let $\tau := \tau_X := E(X_1^4)/\sigma^4$. We have

(1)
$$E\left((c_n s_X^2 - \sigma^2)^2\right) = c_n^2 E(s_X^4) - 2c_n \sigma^4 + \sigma^4,$$

and

(2)
$$E(s_X^4) := E((s_X^2)^2) = \frac{1}{(n-1)^2} [n(n-1)T_{12} + nT_4]$$

where

$$T_{12} := E\left\{ (X_1 - \overline{X})^2 (X_2 - \overline{X})^2 \right\}$$

and

$$T_4 := E\left\{ (X_1 - \overline{X})^4 \right\}.$$

Next,

$$T_{12} = E[(X_1^2 - 2X_1\overline{X} + \overline{X}^2)(X_2^2 - 2X_2\overline{X} + \overline{X}^2)]$$

$$= \sigma^4 - 2E(X_1X_2^2\overline{X}) - 2E(X_2X_1^2\overline{X}) + 4E(X_1X_2\overline{X}^2)$$

$$-2E(X_1\overline{X}^3) - 2E(X_2\overline{X}^3) + E(X_1^2\overline{X}^2) + E(X_2^2\overline{X}^2) + E(\overline{X})^4$$

$$= \sigma^4 - 4E(X_1X_2^2\overline{X}) + 4E(X_1X_2\overline{X}^2) - 4E(X_1\overline{X}^3) + 2E(X_1^2\overline{X}^2) + E(\overline{X}^4)$$

$$= \sigma^4 - 4\frac{\sigma^4}{n} + 8\frac{\sigma^4}{n^2} - 4\left[\frac{\tau\sigma^4 + 3(n-1)\sigma^4}{n^3}\right]$$

$$+2\left[\frac{\tau\sigma^4 + (n-1)\sigma^4}{n^2}\right] + \frac{n\tau\sigma^4 + 3n(n-1)\sigma^4}{n^4},$$

where
$$3n(n-1) = \binom{n}{2} \binom{4}{2}$$
. Thus

$$T_{12} = \sigma^4 \left\{ \left[1 - \frac{4}{n} + \frac{8}{n^2} - 12 \frac{(n-1)}{n^3} + \frac{2(n-1)}{n^2} + \frac{3(n-1)}{n^3} \right] + \tau \left[-\frac{4}{n^3} + \frac{2}{n^2} + \frac{1}{n^3} \right] \right\}$$

$$= \sigma^4 \left\{ 1 - \frac{2}{n} - \frac{3}{n^2} + \frac{9}{n^3} + \tau \left[\frac{2}{n^2} - \frac{3}{n^3} \right] \right\}.$$
(3)

That finishes the representation of T_{12} . For another term of (2) we have

$$T_{4} = E(X_{1}^{4}) - 4E(X_{1}^{3}\overline{X}) + 6E(X_{1}^{2}\overline{X}^{2}) - 4E(X_{1}\overline{X}^{3}) + E(\overline{X}^{4})$$

$$= \sigma^{4} \left(\tau - 4\frac{\tau}{n} + 6\left\{\frac{\tau + n - 1}{n^{2}}\right\} - 4\left\{\frac{\tau + 3(n - 1)}{n^{3}}\right\} + \frac{\tau + 3(n - 1)}{n^{3}}\right)$$

$$= \sigma^{4} \left[\frac{6(n - 1)}{n^{2}} - \frac{9(n - 1)}{n^{3}} + \tau\left\{1 - \frac{4}{n} + \frac{6}{n^{2}} - \frac{4}{n^{3}} + \frac{1}{n^{3}}\right\}\right]$$

$$= \sigma^{4} \left[\tau\left\{1 - \frac{4}{n} + \frac{6}{n^{2}} - \frac{3}{n^{3}}\right\} + \frac{6}{n} - \frac{15}{n^{2}} + \frac{9}{n^{3}}\right].$$

Inserting the results of (3) and (4) into (2) gives

$$E[(c_n s_X^2 - \sigma^2)^2] = \sigma^4[c_n^2 A_n - 2c_n + 1]$$

where

$$A_n = \frac{n}{n-1} \left[1 - \frac{2}{n} - \frac{3}{n^2} + \frac{9}{n^3} + \tau \left(\frac{2}{n^2} - \frac{3}{n^3} \right) \right] + \frac{n}{(n-1)^2} \left[\frac{6}{n} - \frac{15}{n^2} + \frac{9}{n^3} + \tau \left(1 - \frac{4}{n} + \frac{6}{n^2} - \frac{3}{n^3} \right) \right].$$

This implies

(5)
$$E\left[(c_n s_X^2 - \sigma^2)^2 \right] = \sigma^4 \left[\frac{c_n^2}{(n-1)^2} (B_n + \tau C_n) - 2c_n + 1 \right]$$

where

$$B_n = n(n-1) - 2(n-1) - \frac{3(n-1)}{n} + \frac{9(n-1)}{n^2} + 6 - \frac{15}{n} + \frac{9}{n^2}$$

and

$$C_n = \frac{2(n-1)}{n} - \frac{3(n-1)}{n^2} + n - 4 + \frac{6}{n} - \frac{3}{n^2}.$$

Then

$$B_n = n^2 - 3n + 2 - 3 + 6 + \frac{1}{n}(3 + 9 - 15) + \frac{1}{n^2}(-9 + 9)$$

$$= n^2 - 3n + 5 - \frac{3}{n}$$

and

$$C_n = n - 4 + 2 + \frac{1}{n}(6 - 2 - 3) + \frac{1}{n^2}(3 + 11) = n - 2 + \frac{1}{n} + \frac{14}{n^2}.$$

Let

$$D_n := n^2 - 3n + 5 - \frac{3}{n} + \tau \left(n - 2 + \frac{1}{n} + \frac{14}{n^2} \right).$$

Thus the mean-square error (MSE) on the left in (1) or (5) equals

$$\sigma^4 \left[\frac{c_n^2}{(n-1)^2} D_n - 2c_n + 1 \right].$$

So, in order to get a smaller MSE for a given $c_n < 1$ than for $c_n = 1$, we ask whether

$$\frac{c_n^2}{(n-1)^2}D_n - 2c_n + 1 < \frac{D_n}{(n-1)^2} - 1,$$

the right side being the result if c_n on the left is replaced by 1. in other words whether

$$\frac{c_n^2 - 1}{(n-1)^2} D_n < 2c_n - 2,$$

or, reversing signs and dividing by $1-c_n$, whether

$$\frac{1+c_n}{(n-1)^2}D_n > 2.$$

We can also write

$$D_n = n^2 - 2n + 3 - \frac{2}{n} + \frac{14}{n^2} + (\tau - 1)\left(n - 2 + \frac{1}{n} + \frac{14}{n^2}\right).$$

Now $n-2+\frac{1}{n}+\frac{14}{n^2}>0$ for all $n\geq 2$ because $n^3-2n^2+n+14=n^2(n-2)+n+14>0$. Thus

$$D_n \ge E_n := n^2 - 2n + 3 - \frac{2}{n} + \frac{14}{n^2}$$

since $\tau \geq 1$. Now plugging in Yatracos's value $c_n = \frac{(n+2)(n-1)}{n(n+1)}$ we're asking whether

$$\left[1 + \frac{(n+2)(n-1)}{n(n+1)}\right] E_n > 2(n-1)^2,$$

or equivalently whether

$$[n(n+1) + (n+2)(n-1)]E_n > 2(n-1)^2 n(n+1),$$

or whether

$$(2n^2 + 2n - 2)E_n > 2(n^4 - n^3 - n^2 + n),$$

or whether

$$n^4 - n^3 + n(-2+3+2) + 14 - 2 - 3 + \frac{1}{n}(14+2) - \frac{14}{n^2} > n^4 - n^3 - n^2 + n.$$

This reduces to

$$n^2 + 2n + 9 + \frac{16}{n} - \frac{14}{n^2} > 0$$

which is clearly true for all all $n \geq 2$, proving the theorem.

Remarks. It do not know at this writing whether with Yatracos's c_n the resulting estimator is admissible.

The mean-square error of $c_n s_X^2$ depends on the distribution of X_1 only through the number τ . For each $\tau \geq 1$, it seems possible to determine a $c_n = c_n(\tau)$ which minimizes the mean-square error.

Notes. I appreciate that Y. Yatracos sent me an advance copy, in 2004 or 2005, of his work. In 18.466, Fall 2005, Songzi Du wrote a term paper in which he gave an exposition of Yatracos's paper, with proofs, which helped my understanding. Both Yatracos's paper, and Songzi Du's exposition, use a method of augmented samples. Whereas, I looked for a direct proof of Yatracos's theorem without augmented samples.

REFERENCES

Du, Songzi (2005). Mean-squared error reduction for unbiased estimators. Term paper, 18.466, Fall 2005.

Yatracos, Yannis G. (2005). Artificially augmented samples, and mean squared error reduction. *J. Amer. Statistic. Assoc.* **100**, pp. 1168–1175.