Stat 257.01 Fall 2004 Assignment #7 Solutions

(6.8) There was a misprint in this problem in earlier editions of the text. The earlier editions had  $\overline{y} = 169$  while the corrected editions have  $\overline{y} = 2.6$ . If you use  $\overline{y} = 169$ , though, then the problem does have a "solution." However, it should be noted that the only way that  $\overline{y}$  can equal  $\sum y_i^2$  is if  $y_i = 0$  for all *i*, or there is a single data point  $y_1 = 1$ .

 $\overline{\overline{y} = 169}$ : We begin by finding r, the estimator of the population ratio R. Recall that  $r = \overline{y}/\overline{x}$  so that  $\overline{\overline{y}} = 160.0$ 

$$r = \frac{\overline{y}}{\overline{x}} = \frac{169.0}{9.2} \approx 18.4.$$

From the computational formula for  $\hat{V}(r)$ , and a little algebra, we find that

$$\begin{split} \hat{V}(r) &= \left(\frac{N-n}{nN}\right) \left(\frac{1}{\mu_x^2}\right) (s_y^2 + r^2 s_x^2 - 2r\hat{\rho}s_x s_y) \\ &= \left(\frac{N-n}{nN}\right) \left(\frac{1}{\mu_x^2}\right) \left(\frac{1}{n-1} (\sum y_i^2 - 2r\sum x_i y_i + r^2 \sum x_i^2)\right). \end{split}$$

Using  $\overline{x}$  to approximate  $\mu_x$ , and plugging in the given data, we conclude that

$$\hat{V}(r) \approx \left(\frac{275 - 25}{25 \cdot 275}\right) \left(\frac{1}{9.2^2}\right) \left(\frac{1}{25 - 1}(169.0 - 2 \cdot 18.4 \cdot 522 + (18.4)^2 \cdot 2240)\right) \approx 13.2.$$

Thus, a reasonable bound on the error of estimation is  $2\sqrt{\hat{V}(r)} \approx 7.3$ . In other words, an approximate 95% confidence interval for R is  $18.4 \pm 13.2$ .

If  $\overline{y} = 2.6$ : so that  $\overline{y} = 2.6$ :  $\overline{y}/\overline{x} = 2.6$ . We begin by finding r, the estimator of the population ratio R. Recall that  $r = \overline{y}/\overline{x}$ 

$$r = \frac{\overline{y}}{\overline{x}} = \frac{2.6}{9.2} \approx 0.283$$

From the computational formula for  $\hat{V}(r)$ , and a little algebra, we find that

$$\begin{split} \hat{V}(r) &= \left(\frac{N-n}{nN}\right) \left(\frac{1}{\mu_x^2}\right) \left(s_y^2 + r^2 s_x^2 - 2r\hat{\rho}s_x s_y\right) \\ &= \left(\frac{N-n}{nN}\right) \left(\frac{1}{\mu_x^2}\right) \left(\frac{1}{n-1} \left(\sum y_i^2 - 2r\sum x_i y_i + r^2 \sum x_i^2\right)\right). \end{split}$$

Using  $\overline{x}$  to approximate  $\mu_x$ , and plugging in the given data, we conclude that

$$\hat{V}(r) \approx \left(\frac{275 - 25}{25 \cdot 275}\right) \left(\frac{1}{9.2^2}\right) \left(\frac{1}{25 - 1}(169.0 - 2 \cdot 0.283 \cdot 522 + (0.283)^2 \cdot 2240)\right) \approx 0.000948$$

Thus, a reasonable bound on the error of estimation is  $2\sqrt{\hat{V}(r)} \approx 0.0616$ . In other words, an approximate 95% confidence interval for R is  $0.283 \pm 0.0616$ .

(6.11) Again, we begin by finding r, the estimator of the population ratio R. Recall that  $r = \overline{y}/\overline{x}$  so that

$$r = \frac{\overline{y}}{\overline{x}} = \frac{\sum y_i}{\sum x_i} = \frac{11458}{10103} \approx 1.134.$$

Thus, we find that

$$\hat{\mu}_y = r\mu_x \approx (1.134)(880) \approx 997.92.$$

As for a bound on the error of estimation, we find that  $\sum (y_i - rx_i)^2 \approx 46939.53$  so that  $s_r^2 \approx 94.067$ . Hence,

$$B = 2\sqrt{\hat{V}(\hat{\mu}_y)} = 2\sqrt{\left(\frac{N-n}{nN}\right)s_r^2} \approx 2\sqrt{\left(\frac{500-12}{12\cdot 500}\right)(94.067)} \approx 53.65.$$

(6.23) Suppose that x denotes income (in 1982 constant billions) in 1980 and that y denotes income (in 1982 constant billions) in 1989. From the data given in the problem, we find that n = 6, N = 19, and  $\tau_x = 674$ . Furthermore,

$$\sum x_i = 245, \quad \sum x_i^2 = 11991, \quad \sum y_i = 327, \quad \sum y_i^2 = 22131, \quad \sum x_i y_i = 16196$$

so that

$$r = \frac{\sum y_i}{\sum x_i} = \frac{327}{245} \approx 1.3347.$$

Finally, let  $\tau_y$  denote the 1989 total income.

(a) We find  $\hat{\tau}_y$ , a ratio estimator of  $\tau_y$ . Hence,

$$\hat{\tau}_y = r\tau_x \approx 1.3347 \cdot 674 \approx 899.58$$

As for the estimated variance of  $\hat{\tau}_y$ , we have

$$\begin{split} \hat{V}(\hat{\tau}_y) &= \left(\frac{\tau_x^2}{\mu_x^2}\right) \left(\frac{N-n}{nN}\right) \left(\frac{1}{n-1}\right) \sum (y_i - rx_i)^2 \\ &= N^2 \left(\frac{N-n}{nN}\right) \left(\frac{1}{n-1} (\sum y_i^2 - 2r \sum x_i y_i + r^2 \sum x_i^2)\right) \\ &\approx 19^2 \left(\frac{19-6}{6\cdot 19}\right) \left(\frac{1}{6-1} (22131 - 2\cdot 1.3347 \cdot 16196 + 1.3347^2 \cdot 11991)\right) \\ &\approx 2127.94 \end{split}$$

so that a bound on the error of estimation is given by  $2\sqrt{\hat{V}(\hat{\tau}_y)} \approx 2\sqrt{2127.94} \approx 92.26$ .

(b) We find  $\hat{\tau}_{yL}$ , a regression estimator of  $\tau_y$ . Since

$$\hat{\tau}_{yL} = N\hat{\mu}_{yL} = N(\overline{y} + b(\mu_x - \overline{x}))$$

where

$$b = \frac{\sum (y_i - \overline{y})(x_i - \overline{x})}{\sum (x_i - \overline{x})^2} = \frac{\sum x_i y_i - n\overline{x} \overline{y}}{\sum x_i^2 - n\overline{x}^2},$$

we conclude that

$$b \approx \frac{16196 - 5 \cdot (245/6) \cdot (327/6)}{11991 - 5 \cdot (245/6)^2} \approx 1.387$$

and

$$\hat{\tau}_{yL} \approx 19 \cdot ((327/6) + 1.387((674/19) - (245/6))) \approx 894.24.$$

As for the estimated variance of  $\hat{\tau}_{yL}$ , we have

$$\begin{split} \hat{V}(\hat{\tau}_{yL}) &= N^2 \left(\frac{N-n}{nN}\right) \left(\frac{1}{n-2}\right) \left[\sum (y_i - \overline{y})^2 - b^2 \sum (x_i - \overline{x})^2\right] \\ &= N^2 \left(\frac{N-n}{nN}\right) \left(\frac{1}{n-2}\right) \left[\sum y_i^2 - b^2 \sum x_i^2 - n(\overline{y}^2 - b^2 \overline{x}^2)\right] \\ &\approx 19^2 \left(\frac{19-6}{6\cdot 19}\right) \left(\frac{1}{6-2}\right) \left[22131 - 1.387^2 \cdot 11991 - 6((327/6)^2 - 1.387^2(245/6)^2)\right] \\ &\approx 5015.04 \end{split}$$

so that a bound on the error of estimation is given by  $2\sqrt{\hat{V}(\hat{\tau}_{yL})} \approx 2\sqrt{5015.04} \approx 141.63$ .

(c) We find  $\hat{\tau}_{yD}$ , a difference estimator of  $\tau_y$ . Let  $d_i$  be the differences between the 1989 incomes and the 1980 incomes, so that  $d_i = \{5, 28, 12, 10, 1, 26\}$ . We also find

$$\sum d_i = 82, \quad \sum d_i^2 = 1730.$$

Hence,

$$\hat{\tau}_{yD} = N\hat{\mu}_{yD} = N(\mu_x + \overline{d}) = 19(674/19 + 82/6) \approx 933.67.$$

As for the estimated variance of  $\hat{\tau}_{yD}$ , we have

$$\hat{V}(\hat{\tau}_{yD}) = N^2 \left(\frac{N-n}{nN}\right) \left(\frac{1}{n-1}\right) \sum (d_i - \overline{d})^2 = N^2 \left(\frac{N-n}{nN}\right) \left(\frac{1}{n-1}\right) \left(\sum d_i^2 - n\overline{d}^2\right)$$
$$\approx 19^2 \left(\frac{19-6}{6\cdot 19}\right) \left(\frac{1}{6-1}\right) \left(1730 - 6\cdot (82/6)^2\right)$$
$$\approx 5016.84$$

so that a bound on the error of estimation is given by  $2\sqrt{\hat{V}(\hat{\tau}_{yD})} \approx 2\sqrt{5016.84} \approx 141.66$ .

(d) The bound on the error of estimation for the ratio estimator is somewhat smaller than the bounds for the other two estimators, making it somewhat preferable.