In our first example of using a "pivotal quantity" to derive confidence intervals this day, I wasn't very clear about exactly how you derive the formulas for the endpoints. This sheet should clarify the algebra involved.

We considered the case of sampling from a  $N(\mu, \sigma^2)$  population to estimate the target parameter  $\mu$  (the population mean). From our work in section 7.2 of the text, we know that with *n* independent samples, the quantity

$$Q = \frac{\overline{Y} - \mu}{s/\sqrt{n}},$$

(where  $s = \frac{1}{n-1} \sum_{i=1}^{n} (Y_i - \overline{Y})^2$ , the sample standard deviation) satisfies

- Q contains  $\mu$  as the only unknown, and
- Q has a t(n-1) distribution.

Hence we can use Q as a pivotal quantity to derive confidence intervals. Here are complete derivations of each of the three cases (being more careful about signs than I was in class):

Two-sided,  $(1 - \alpha) \times 100\%$  confidence interval

Let  $t_{\alpha/2}$  be the  $\alpha/2$ -percentile point for the *t*-distribution: that is  $P(T > t_{\alpha/2}) = \alpha/2$ . Then by symmetry:

$$\begin{aligned} 1 - \alpha &= P\left(-t_{\alpha/2}(n-1) < \frac{\overline{Y} - \mu}{s/\sqrt{n}} < t_{\alpha/2}(n-1)\right) \\ &= P\left(-t_{\alpha/2}(n-1)\frac{s}{\sqrt{n}} < \overline{Y} - \mu < t_{\alpha/2}(n-1)\frac{s}{\sqrt{n}}\right) \\ &= P\left(-\overline{Y} - t_{\alpha/2}(n-1)\frac{s}{\sqrt{n}} < -\mu < -\overline{Y} + t_{\alpha/2}(n-1)\frac{s}{\sqrt{n}}\right) \\ &= P\left(\overline{Y} + t_{\alpha/2}(n-1)\frac{s}{\sqrt{n}} > \mu > \overline{Y} - t_{\alpha/2}(n-1)\frac{s}{\sqrt{n}}\right) \\ &= P\left(\overline{Y} - t_{\alpha/2}(n-1)\frac{s}{\sqrt{n}} < \mu < \overline{Y} + t_{\alpha/2}(n-1)\frac{s}{\sqrt{n}}\right) \end{aligned}$$

(Note: at the next-to-last step, we multiply through by -1, which reverses all inequalities, then we switch back to write the terms increasing left to right as usual).

(turn over)

One-sided,  $(1 - \alpha) \times 100\%$  lower confidence bound

*Note:* The formula I gave for this in class *was correct*, but the way I suggested to derive it wasn't (d'oh!). We start from the statement (with  $t_{\alpha}$  instead of  $t_{\alpha/2}$ ):

$$1 - \alpha = P\left(\frac{\overline{Y} - \mu}{s/\sqrt{n}} < t_{\alpha}(n-1)\right)$$
$$= P\left(\overline{Y} - \mu < t_{\alpha}(n-1)\frac{s}{\sqrt{n}}\right)$$
$$= P\left(-\mu < -\overline{Y} + t_{\alpha}(n-1)\frac{s}{\sqrt{n}}\right)$$
$$= P\left(\mu > \overline{Y} - t_{\alpha}(n-1)\frac{s}{\sqrt{n}}\right)$$

One-sided,  $(1 - \alpha) \times 100\%$  upper confidence bound

Starting from

$$1 - \alpha = P\left(-t_{\alpha}(n-1) < \frac{\overline{Y} - \mu}{s/\sqrt{n}}\right)$$

and proceeding as for the lower confidence bound,

$$1 - \alpha = P\left(\mu < \overline{Y} + t_{\alpha}(n-1)\frac{s}{\sqrt{n}}\right)$$