## Estimating a mean

Probably the most widely used formula for a confidence interval is

$$
\bar{x} \pm 2 \sqrt{s^{2} / n}
$$

- $\overline{\mathrm{x}}$ is the sample mean
- $s^{2}$ is the sample variance
- n is the sample size
- 2 is PJD's approximation to 1.96

Strictly, you should use a value $\mathbf{c}_{\mathbf{n}}$ which depends on $\mathbf{n}$, but is approximately 2 for reasonably large $\mathbf{n}$, for example:

| n | $\mathrm{c}_{\mathrm{n}}$ |
| ---: | ---: |
| 5 | 2.78 |
| 10 | 2.26 |
| 20 | 2.09 |
| 50 | 2.01 |
| $\infty$ | 1.96 |

Where does the formula for the confidence interval come from?

This diagram shows how the distribution of the sample mean changes with the sample size:

Histogram of one


Histogram of four


Histogram of two


Histogram of eight


And this diagram shows how the variance of the sample mean changes with the sample size:


## Conclusions

- sample means are approximately Normally distributed (symmetric, bell-shaped histogram)
- larger samples lead to more precise estimates
- the variance of an estimate is inversely proportional to the sample size, $n$
- the standard error of an estimate is therefore inversely proportional to $\sqrt{ } \mathbf{n}$
- Murphy's law of diminishing returns - doubling the sample size does not double the precision of your estimate

