

Suppose we are doing a binomial experiment with estimator

$$\tilde{\theta} = \frac{Y}{n}$$

And we want to calculate the probability

$$P(|\tilde{\theta} - \theta| \leq d) \quad \text{for some } d$$

$\hookrightarrow$  true mean (can be unknown)

Here, we are essentially asking: what is the probability that the difference between the estimator's output and the true mean (which is unknown) is within some threshold  $d$ ?

- Suppose we collect 100 samples, all with size  $n$ . If  $p$  is the probability value calculated above, then in 100 $p$  samples, the point estimate provided by the estimator  $\tilde{\theta}$  will be within  $d$  units of the true mean.

In most of these problems, this is done by approximating the binomial distribution using the normal/Gaussian distribution.

Recall that in the Gaussian distribution, variables  $Y$  are standardized to z-scores using the formula

$$z = \frac{Y - \mu}{\sigma} \sim \mathcal{N}(0, 1)$$

Here,  $Y \rightarrow \tilde{\theta}$  ;  $\mu \rightarrow \theta$  (sample mean)

To complete the standardization, we need to compute the standard deviation (denominator).

For binomial distributions:

$$sd(X) = \sqrt{\text{Var}(Y)} = \sqrt{n\theta(1-\theta)}$$

However, we are calculating the standard deviation of the *estimator*, which is equal to  $Y/n$ . Then:

$$\begin{aligned} sd(\tilde{\theta}) &= sd(Y/n) = \text{Var}(Y/n)^{1/2} = (n^{-2} \text{Var}(Y))^{1/2} \\ &= (n^{-2} n\theta(1-\theta))^{1/2} \\ &= \sqrt{\frac{\theta(1-\theta)}{n}} = \sigma \end{aligned}$$

$$\text{Then: } P(|\tilde{\theta} - \theta| \leq d) = P\left(-\frac{d}{\sigma} \leq \frac{\tilde{\theta} - \theta}{\sigma} \leq \frac{d}{\sigma}\right)$$

As such, an asymptotic pivotal quantile for binomial distributions is

$$\frac{\tilde{\theta} - \theta}{\sqrt{\frac{\theta(1-\theta)}{n}}}$$

## Confidence Intervals

General formula:  $\pm z \cdot \text{sd}(\tilde{\theta})$

Ex. for binomial, a 95% CI is  $\pm 1.96 \sqrt{\frac{\theta(1-\theta)}{n}}$

Proof: An asymptotic pivotal quantity for  $\tilde{\theta}$  is

$$\frac{\tilde{\theta} - \theta}{\sqrt{\frac{\theta(1-\theta)}{n}}}$$

which follows  $G(0,1)$  as  $n \rightarrow \infty$ .

As such, we can set

$$P\left(-d \leq \frac{\tilde{\theta} - \theta}{\sqrt{\frac{\theta(1-\theta)}{n}}} \leq d\right) \approx 0.95 \quad \text{for some } d$$

Then, we can set  $d = 1.96$ , since  $P(-1.96 \leq z \leq 1.96) \approx 0.95$  for any  $z$  on the Gaussian dist.

$$\Rightarrow P\left(-1.96 \leq \frac{\tilde{\theta} - \theta}{\sqrt{\frac{\theta(1-\theta)}{n}}} \leq 1.96\right) \approx 0.95$$

$$\Rightarrow P\left(-1.96 \sqrt{\frac{\theta(1-\theta)}{n}} \leq \tilde{\theta} - \theta \leq 1.96 \sqrt{\frac{\theta(1-\theta)}{n}}\right) \approx 0.95$$

$$\Rightarrow P\left(-1.96 \sqrt{\frac{\theta(1-\theta)}{n}} - \tilde{\theta} \leq \theta \leq 1.96 \sqrt{\frac{\theta(1-\theta)}{n}} - \tilde{\theta}\right) \approx 0.95$$

$$\Rightarrow P\left(\tilde{\theta} - 1.96 \sqrt{\frac{\theta(1-\theta)}{n}} \leq \theta \leq \tilde{\theta} + 1.96 \sqrt{\frac{\theta(1-\theta)}{n}}\right) \approx 0.95$$

$$\Rightarrow P(\theta \in [\tilde{\theta} \pm 1.96 \sqrt{\frac{\tilde{\theta}(1-\tilde{\theta})}{n}}]) \approx 0.95 \quad \star \quad \text{if on test, end proof here}$$

As such, the probability that the true mean is within  $1.96 \cdot \text{sqrt}(\dots)$  of the estimator is 0.95.

Thus, that is a 95% confidence interval.

PCs:

If  $\sigma$  is known:  $\frac{\tilde{\theta} - \theta}{\sigma/\sqrt{n}} \sim G(0, 1)$   $\sigma$  is population sd

If  $\sigma$  is not known:  $\frac{\tilde{\theta} - \theta}{\sigma} \sim G(0, 1)$

Then get estimator for  $\sigma$