

# Delta Method and the Bootstrap

ST551 Lecture 27

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# Announcements

Lectures this week:

- Today lecture: Delta method and Bootstrap
- Weds lecture: Randomization & Permutation
- Friday lecture: **Cancelled** - Office hours instead

**Formula Sheet** The final is closed book, no note sheet. I am willing to provide some of the harder (less common) formulae.

**Lab:** No set material, I'll encourage Chuan to lead a formula strategy session.

# Delta Method

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# Delta Method

If the sampling distribution of a statistic converges to a Normal distribution, the **Delta method**, provides a way to approximate the sampling distribution of a function of a statistic.

## *Univariate* Delta Method

If

$$\sqrt{n}(\hat{\theta} - \theta) \rightarrow_D N(0, \sigma^2)$$

then

$$\sqrt{n}(g(\hat{\theta}) - g(\theta)) \rightarrow_D N(0, \sigma^2[g'(\theta)]^2)$$

(As long as  $g'(\theta)$  exists and is non-zero valued.)

## Another way of saying it

If we know,

$$\hat{\theta} \sim N(\theta, \sigma^2)$$

then,

$$g(\hat{\theta}) \sim N(g(\theta), \sigma^2 [g'(\theta)]^2)$$

The approximation can be pretty rough. I.e. just because the sample is large enough that the original statistic is reasonably Normal, doesn't mean the transformed statistic will be.

## Example: Log Odds

Let  $Y_1, \dots, Y_n \sim \text{Bernoulli}(p)$ , and  $X = \sum_{i=1}^n Y_i$ .

We know  $\hat{p} = \frac{X}{n} \sim N(p, \frac{p(1-p)}{n})$ .

We might estimate the log odds with:

$$\log\left(\frac{\hat{p}}{1 - \hat{p}}\right)$$

**What is the asymptotic distribution of the estimated log odds?**

## Example: Log Odds cont.

$$g(p) = \log\left(\frac{p}{1-p}\right) = \log(p) - \log(1-p)$$

## Other comments on delta method

Derived using a Taylor expansion of  $g(\hat{\theta})$  around  $g(\theta)$

There is also a multivariate version (useful if you need some function of two statistics, e.g. ratio of sample means)



# Bootstrap

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A method to approximate the sampling distribution of a statistic

## Idea:

- Recall, one way to approximate the sampling distribution of a statistic was by **simulation**, but you have to assume a population distribution.
- The bootstrap uses the *empirical distribution function* as an estimate for the population distribution, i.e. relies on

$$\hat{F}(y) \approx F(y)$$

## Example - Sampling distribution of Median by simulation

Assume a population distribution, i.e.  $Y \sim N(\mu, \sigma^2)$

**Repeat for**  $k = 1, \dots, B$

1. Sample  $n$  observations from  $N(\mu, \sigma^2)$
2. Find sample median,  $m^{(k)}$

Then the simulated sample medians,  $m^{(k)}, k = 1, \dots, B$  approximate the sampling distribution of the sample median.

## Example - Sampling distribution of Median by bootstrap

Estimate the population distribution from the sample, i.e.  $\hat{F}(y)$

**Repeat for**  $k = 1, \dots, B$

1. Sample  $n$  observations from a population with c.d.f  $\hat{F}(y)$
2. Find sample median,  $m^{(k)}$

Then the bootstrapped sample medians,  $m^{(k)}, k = 1, \dots, B$  approximate the sampling distribution of the sample median.

## Sampling from a c.d.f

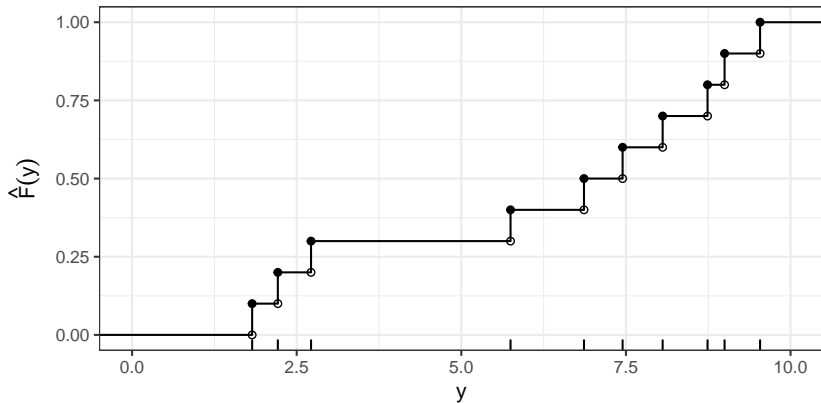
You can sample from any c.d.f by sampling from a Uniform(0, 1), then transforming with the inverse c.d.f.

I.e. sample  $u_1, \dots, u_n$  i.i.d from Uniform(0,1), then

$$y_i = F^{-1}(u_i) \quad i = 1, \dots, n$$

are distributed with c.d.f  $F(y)$ .

## In the empirical case



Sampling from the ECDF is equivalent to sampling with replacement from the original sample.

## Example - Sampling distribution of Median by bootstrap

**Repeat for**  $k = 1, \dots, B$

1. Sample  $n$  observations with replacement from  $Y_1, \dots, Y_n$
2. Find sample median,  $m^{(k)}$

Then the bootstrapped sample medians,  $m^{(k)}$ ,  $k = 1, \dots, B$  approximate the sampling distribution of the sample median.

A little more subtly:

$$\hat{m} - m \sim \tilde{m} - \hat{m}$$

## Example

Sample values: 1.8, 2.2, 2.7, 5.7, 6.9, 7.4, 8.1, 8.7, 9 and 9.5

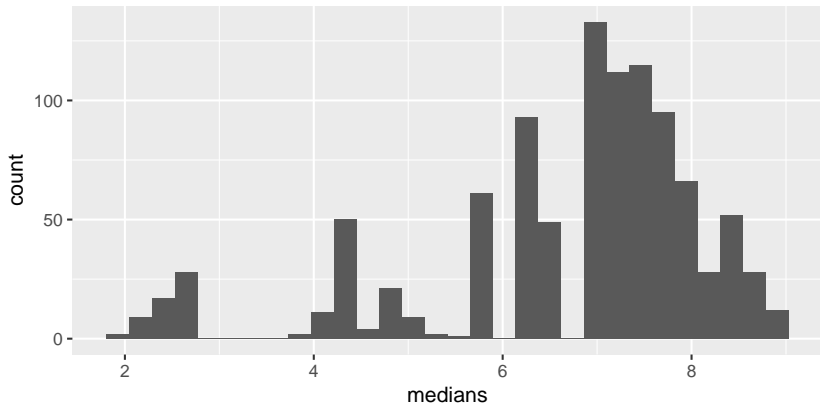
Sample median: 7.1562828

A bootstrap resample: 1.8, 2.7, 2.7, 5.7, 6.9, 7.4, 8.1, 8.1, 8.7 and 9.5

Sample median: 7.1562828



# Many resamples



# Bootstrap confidence intervals

Many methods..

A common one:

- **Quantile:**  $100(\alpha/2)$  largest resampled statistic value, and  $100(1 - \alpha/2)$  largest resampled statistic value

## Comments on the bootstrap

Relies on  $\hat{F}(y)$  being a good estimate of the  $F(y)$ , doesn't necessarily solve small sample problems.

Resampling should generally mimic original study design. E.g. If pairs of observations are sampled from a population, pairs should be resampled