

## Lecture 18: The delta method

- If  $f$  is smooth from  $\mathbb{R}^p$  to  $\mathbb{R}^q$  and  $\sqrt{n}(X_n - x)$  converges in distribution to  $MVN_p(0, \Sigma)$  then

$$\sqrt{n}(f(X_n) - f(x)) \implies MVNN_q(0, D\Sigma D^T)$$

where  $D$  is the  $q \times p$  matrix with  $i, j$  entry

$$\frac{\partial f_i}{\partial x_j}$$

evaluated at  $x$ .

- I did an example:  $X_1, \dots, X_n$  iid with mean  $\mu$ , SD  $\sigma$ , and higher moments  $\mu_k = \mathbb{E}[(X - \mu)^k]$ .
- We studied *Coefficient of Variation* (CV) given by

$$\sigma/\mu$$

which is estimated by

$$\frac{\sqrt{\bar{X}^2 - \bar{X}^2}}{\bar{X}}.$$

- Define  $f(u, v) = \sqrt{v - u^2}/u$  to see that



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- The derivative matrix  $D$  is

$$\begin{bmatrix} -\frac{1}{\sigma} - \frac{CV}{\mu} & \frac{1}{2\mu\sigma} \end{bmatrix}$$

- I took the time to work out the variance covariance of

$$[X \quad X^2]^T$$

to get the matrix

$$\Sigma = \begin{bmatrix} \sigma^2 & \mu_3 + 2\mu\sigma^2 \\ \mu_3 + 2\mu\sigma^2 & \mu_4 - \sigma^4 + 4\mu\mu_3 + 4\mu^2\sigma^2 \end{bmatrix}$$

- After algebra

$$\tau^2 = CV^4 + \frac{\mu_4 - \sigma^4}{\sigma^4} \times \frac{CV^2}{4} + \frac{\mu_3}{\mu^3}$$

which is different by a sign from class.



# Lecture 18

- I talked about profiling for LD50.
- I talked about using  $\hat{\tau}$  for confidence intervals (and hypothesis tests).
- I talked about the duality between confidence sets and families of hypothesis tests.



## Coverage in the text

- Chapter 9.9, pp 131–133.
- Course slides “Convergence in Distribution”: 21-28.
- Also red slides 29-31.
- See “course notes” on web pages 68-71.

