

Repetition week 41

Delta method

$$\sqrt{n}(Y_n - \theta) \xrightarrow{D} N(0, \sigma^2) \Rightarrow \sqrt{n}(g(Y_n) - g(\theta)) \xrightarrow{D} N(0, \sigma^2 [g'(\theta)]^2)$$

$$g'(\theta) = 0$$

$$\sqrt{n}(Y_n - \theta) \xrightarrow{D} N(0, \sigma^2) \Rightarrow n(g(Y_n) - g(\theta)) \xrightarrow{D} \frac{\sigma^2}{2} [g''(\theta)] \chi_1^2$$

Sufficient statistics

A statistic $T(X)$ is a sufficient statistic for θ if the conditional distribution of the sample X given the value of $T(X)$ does not depend on θ .

A sufficient statistics for a parameter (-vector) θ is a statistic that in a certain sense, captures all the information about θ in the sample.

Theorem 6.2.2

If $p(\mathbf{x}|\theta)$ is the pdf/pmf of \mathbf{X} and $q(t|\theta)$ is the pdf/pmf of $T(\mathbf{X})$, then $T(\mathbf{X})$ is a sufficient statistics for θ if, for every \mathbf{x} in the sample space the ratio $\frac{p(\mathbf{x}|\theta)}{q(T(\mathbf{x})|\theta)}$ is a constant as a function of θ .

Theorem 6.2.6

Let $f(\mathbf{x}|\theta)$ be the joint pdf/pmf for a sample \mathbf{X} . $T(\mathbf{X})$ is a sufficient statistics for θ if and only if for all \mathbf{x} and all θ .

$$f(\mathbf{x}|\theta) = g(T(\mathbf{x})|\theta)h(\mathbf{x})$$