

Asymptotic distribution of the MLE

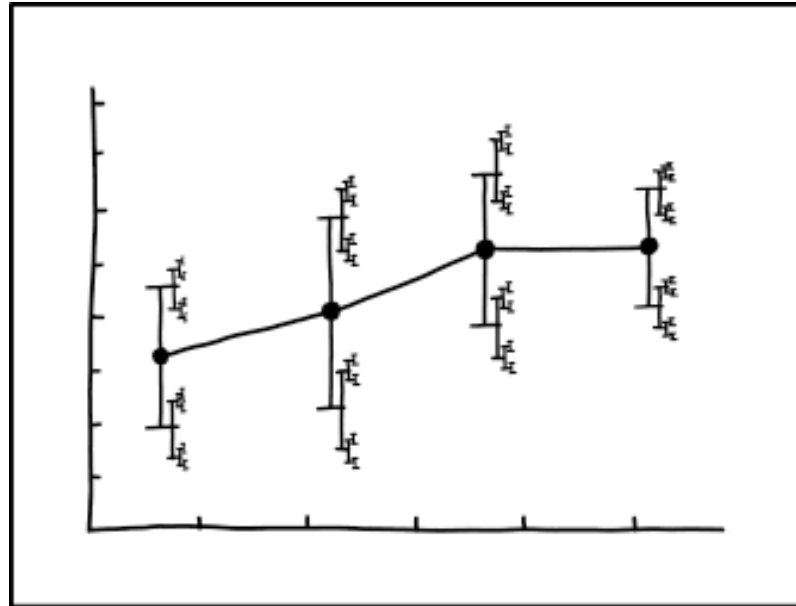
Hw 2 due date changed to next Thursday (9/15)

Warm-up

$$\frac{\bar{X} - \mu}{\hat{\sigma} / \sqrt{n}}$$

$$\frac{\bar{X} - \mu}{\textcircled{S / \sqrt{n}}}$$

$$\frac{\bar{X} - \mu}{S / \sqrt{n}} \approx N(0, 1)$$



I DON'T KNOW HOW TO PROPAGATE
ERROR CORRECTLY, SO I JUST PUT
ERROR BARS ON ALL MY ERROR BARS.

Eventually:

$$\hat{\beta} \approx N(\beta, \hat{\tau}^{-1}(\beta))$$

Why don't we estimate the standard error of our standard error?

Preliminaries

Let $X_1, X_2, X_3, \dots$ be a sequence of random variables, and let X be another random variable. Let F_n be the CDF of X_n , and F the CDF of X .

Def: (Convergence in probability) $X_n \xrightarrow{P} X$ if
 $\forall \varepsilon > 0, \quad P(|X_n - X| > \varepsilon) \rightarrow 0 \quad \text{as } n \rightarrow \infty$

Def: (convergence in distribution) $X_n \xrightarrow{d} X$ if,
 $\forall t$ where F is continuous,
 $F_n(t) \rightarrow F(t) \quad \text{as } n \rightarrow \infty$

Key results:

1) weak law of large numbers (WLLN): If $X_1, \dots, X_n$, then

$$\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i \xrightarrow{P} \mathbb{E}[X_1]$$

2) Central limit theorem (CLT): Let $X_1, \dots, X_n$ be iid with mean μ and variance σ^2 . Then

$$\frac{\bar{X}_n - \mu}{\sigma/\sqrt{n}} \xrightarrow{d} Z, \quad Z \sim N(0, 1)$$

3) Slutsky's theorem: If $X_n \xrightarrow{d} X$, $Y_n \xrightarrow{P} c \in \mathbb{R}$,
then $X_n Y_n \xrightarrow{d} cX$

Fact: $\hat{\sigma} \xrightarrow{P} \sigma$

Asymptotic distribution of the MLE

Suppose that $Y_1, \dots, Y_n$ are iid w/ probability function $f(Y_i; \theta)$, $\theta \in \mathbb{R}$

Let $l(\theta) = \sum_{i=1}^n \log f(Y_i; \theta)$, $\hat{\theta} = \text{MLE of } \theta$

$$\mathcal{I}_1(\theta) = -\mathbb{E} \left[\frac{\partial^2}{\partial \theta^2} \log f(Y; \theta) \right]$$

Theorem: $\sqrt{n}(\hat{\theta} - \theta) \xrightarrow{d} N(0, \mathcal{I}_1^{-1}(\theta))$

Proof sketch: 1) $\sqrt{n}(\hat{\theta} - \theta) \approx \frac{-\frac{1}{\sqrt{n}} l'(\theta)}{\frac{1}{n} l''(\theta)}$

2) $\frac{1}{n} l''(\theta) \xrightarrow{P} -\mathcal{I}_1(\theta)$

3) $\frac{1}{\sqrt{n}} l'(\theta) \xrightarrow{d} N(0, \mathcal{I}_1(\theta))$

4) Slutsky's theorem: $\sqrt{n}(\hat{\theta} - \theta) \xrightarrow{d} \underbrace{\frac{1}{\mathcal{I}_1(\theta)}}_{N(0, \mathcal{I}_1(\theta))} N(0, \mathcal{I}_1(\theta))$

Proof: $\hat{\theta} = \text{MLE} \Rightarrow l'(\hat{\theta}) = 0$

if $\hat{\theta} \approx \theta$ ("close enough")

$$0 = l'(\hat{\theta}) \approx l'(\theta) + (\hat{\theta} - \theta) l''(\theta)$$

(First-order Taylor expansion)

$$\Rightarrow \hat{\theta} - \theta \approx - \frac{l'(\theta)}{l''(\theta)}$$

$$\Rightarrow \sqrt{n}(\hat{\theta} - \theta) \approx \sqrt{n} \left(- \frac{l'(\theta)}{l''(\theta)} \right)$$

$$= - \frac{\frac{1}{\sqrt{n}} l'(\theta)}{\frac{1}{n} l''(\theta)}$$

Denominator

$$\frac{1}{n} \ell''(\theta) = \frac{1}{n} \sum_{i=1}^n \ell_i''(\theta), \quad \ell_i'' = \frac{\partial^2}{\partial \theta^2} \log f(y_i; \theta)$$

By
WLLN

$$), \quad \frac{1}{n} \sum_{i=1}^n \ell_i''(\theta) \xrightarrow{P} \mathbb{E}[\ell_i''(\theta)]$$

and we know that $\mathcal{I}_1(\theta) = -\mathbb{E}\left[\frac{\partial^2}{\partial \theta^2} \log f(y_i; \theta)\right]$

$$\Rightarrow \frac{1}{n} \sum_{i=1}^n \ell_i''(\theta) \xrightarrow{P} -\mathcal{I}_1(\theta)$$

$$\Rightarrow \frac{1}{n} \ell''(\theta) \xrightarrow{P} -\mathcal{I}_1(\theta)$$

Numerator : $\frac{1}{\sqrt{n}} l'(\theta)$

$$\mathbb{E} \left[\frac{1}{\sqrt{n}} l'(\theta) \right] = \frac{1}{\sqrt{n}} \sum_{i=1}^n \mathbb{E} \left[\frac{\partial}{\partial \theta} \log f(Y_i; \theta) \right]$$

$Y_1, \dots, Y_n$ are iid, $\Rightarrow \frac{1}{\sqrt{n}} \cdot n \cdot \mathbb{E} \left[\frac{\partial}{\partial \theta} \log f(Y_i; \theta) \right]$

$$= \sqrt{n} \mathbb{E} \left[\frac{\partial}{\partial \theta} \log f(Y_i; \theta) \right]$$

Claim: $\mathbb{E} \left[\frac{\partial}{\partial \theta} \log f(Y_i; \theta) \right] = 0$

PF: $\mathbb{E} \left[\frac{\partial}{\partial \theta} \log f(Y_i; \theta) \right] = \int \frac{\partial}{\partial \theta} \log f(y; \theta) f(y; \theta) d\theta$

$$= \int \frac{1}{\cancel{f(y; \theta)}} \cdot \frac{\partial}{\partial \theta} f(y; \theta) \cdot \cancel{f(y; \theta)} d\theta = \int \frac{\partial}{\partial \theta} f(y; \theta) d\theta$$
$$= \frac{\partial}{\partial \theta} \int f(y; \theta) d\theta = 0 \quad //$$

Numerator :
$$\text{Var} \left(\frac{1}{\sqrt{n}} \ell'(\theta) \right) = \frac{1}{n} \cdot \sum_{i=1}^n \text{Var} \left(\frac{\partial}{\partial \theta} \log f(Y_i; \theta) \right)$$
$$= \text{Var} \left(\frac{\partial}{\partial \theta} \log f(Y_i; \theta) \right)$$