

Asymptotic distribution of the MLE

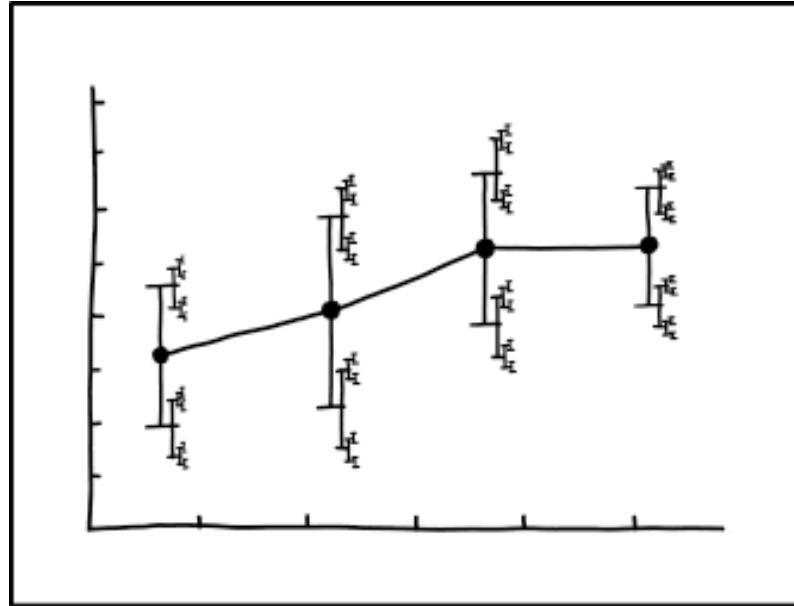
Hw 2 due date changed to next Thursday (9/15)

Warm-up

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

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

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



Eventually:

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

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

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
 $\forall \varepsilon > 0, P(|X_n - X| > \varepsilon) \rightarrow 0$ as $n \rightarrow \infty$

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

Key results:

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

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

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

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

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 $\gamma_1, \dots, \gamma_n$ are iid w/ probability function $f(\gamma_i; \theta)$, $\theta \in \mathbb{R}$

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

$$\hat{\Sigma}_1(\theta) = -\mathbb{E}\left[\frac{\partial^2}{\partial \theta^2} \log f(\gamma; \theta)\right]$$

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

- Proof sketch:
- 1) $\sqrt{n}(\hat{\theta} - \theta) \approx \frac{-\frac{1}{\sqrt{n}} \ell'(\theta)}{\frac{1}{n} \ell''(\theta)}$
 - 2) $\frac{1}{n} \ell''(\theta) \xrightarrow{P} -\hat{\Sigma}_1(\theta)$
 - 3) $\frac{1}{\sqrt{n}} \ell'(\theta) \xrightarrow{d} N(0, \hat{\Sigma}_1(\theta))$
 - 4) Slutsky's theorem: $\sqrt{n}(\hat{\theta} - \theta) \xrightarrow{d} \frac{1}{\hat{\Sigma}_1(\theta)} N(0, \hat{\Sigma}_1(\theta))$

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

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

$$0 = l'(\hat{\theta}) \approx l'(\theta) + (\hat{\theta} - \theta) l''(\theta) \quad (\text{First-order Taylor expansion})$$

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

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

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

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

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

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

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

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

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

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

$$y_1, \dots, y_n \text{ are iid}, \Rightarrow \frac{1}{\sqrt{n}} \cdot n \cdot \mathbb{E} \left[\frac{\partial}{\partial \theta} \log f(y_i; \theta) \right] \\ = \sqrt{n} \mathbb{E} \left[\frac{\partial}{\partial \theta} \log f(y_i; \theta) \right]$$

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

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

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

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

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

$$= \text{Var}\left(\frac{\partial}{\partial \theta} \log f(y_i; \theta)\right)$$