

Lec 2.

We observe "Data" $X = \langle 0, 0, 1, 0, 1, 0 \rangle$ realization from a random process

Assume a parametric model $f(x) = f(x; \theta) = \sin(ax)$

$$\tilde{\mathcal{F}} : \text{iid Binomial} = \{ p(x; \theta) : \theta \in \mathcal{H} \} = \{ \theta^x (1-\theta)^{1-x} : \theta \in (0,1) \}$$

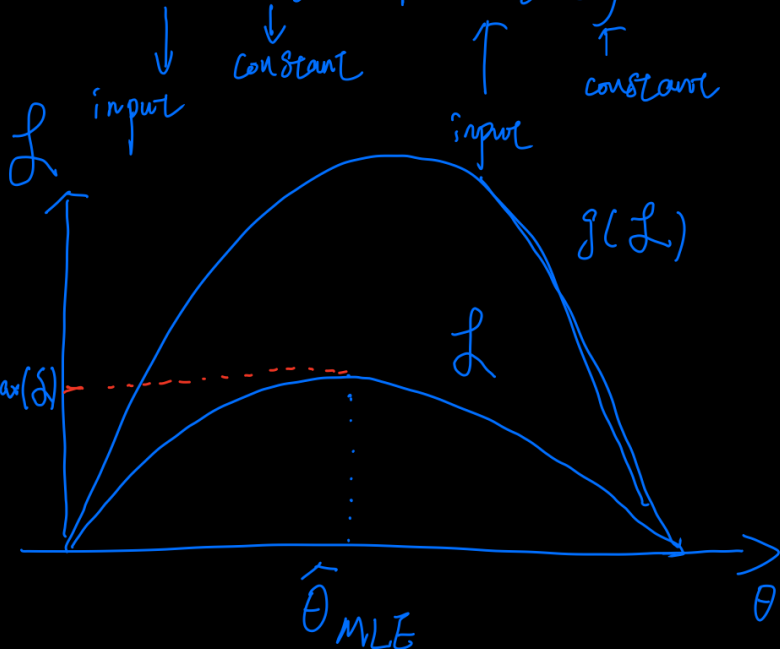
($p(x) = p(x; \theta)$ _{p.m.f.} constant θ needed to compute the prob)

We'd like to learn about θ (inference)

$$P(X; \theta) = P(\langle 0, 0, 1, 0, 1, 0 \rangle; \theta) \stackrel{\tilde{\mathcal{F}}}{=} \underbrace{(\theta^0 (1-\theta)^{1-0})}_{\substack{\text{6 data "0" 6 times}}} \cdot (\theta^0 (1-\theta)^{1-0}) \dots \underbrace{(\theta^1 (1-\theta)^{1-1})}_{(\theta^0 (1-\theta)^{1-0})} \\ \text{if } \theta = 0.5 \quad p(x; \theta) = 0.0156 \quad \theta = 0.25 \quad \dots \approx 0.0198 \quad \underline{= \theta^2 (1-\theta)^4}$$

"likelihood function" "probability"

$$\mathcal{L}(\theta; x) = P(x; \theta)$$



$$\int_{\mathcal{H}} \mathcal{L}(\theta; x) d\theta \stackrel{?}{=} 1 \quad \text{No}$$

$$P(x; \theta) \in (0,1)$$

$$\Rightarrow \mathcal{L}(\theta; x) \in (0,1)$$

$$\sum_x P(x; \theta) = 1$$

$$\int_x P(x; \theta) d\theta \stackrel{?}{=} 1$$

$$\hat{\theta}_{MLE} := \arg \max_{\theta \in (H)} \{L(\theta, x)\} = \arg \max_{\theta \in (H)} \{g(L(\theta, x))\}$$

MLE: maximum likelihood estimate

↑
argument
that
in the maximum

Consider g in
a strictly increasing
function

$$\ln \left(\prod_{i=1}^6 \theta^{x_i} (1-\theta)^{1-x_i} \right)$$

$$= \sum_{i=1}^6 \ln(\theta^{x_i} (1-\theta)^{1-x_i}) = \sum_{i=1}^6 [x_i \ln(\theta) + (1-x_i) \ln(1-\theta)] = \ln(\theta)^{\sum_{i=1}^6 x_i} + \ln(1-\theta)^{\sum_{i=1}^6 (1-x_i)}$$

$$\text{let } \bar{x} = \frac{1}{n} \sum_{i=1}^n x_i \quad \text{in our case } \bar{x} = \frac{1}{6} \sum_{i=1}^6 x_i \quad \& \bar{x} = \frac{2}{6} \sum_{i=1}^6 x_i$$

$$= 6 \left(\ln(\theta) \bar{x} + (1-\bar{x}) \ln(1-\theta) \right)$$

① find $L(\theta; x)$ ② Find $\frac{d}{d\theta} [L(\theta, x)] = l'(\theta, x)$ ③ $l'(\theta, x) = 0$ ④ solve $\hat{\theta}_{MLE}$

$$l'(\theta; x) = 6 \left(\frac{\bar{x}}{\theta} - \frac{1-\bar{x}}{1-\theta} \right) = 0 \Rightarrow \frac{\bar{x}}{\theta} = \frac{1-\bar{x}}{1-\theta} \quad \bar{x} = \theta \quad \hat{\theta}_{MLE} = \bar{x}$$

$$x = (0, 0, 1, 0, 1, 0) \Rightarrow \bar{x} = \frac{1}{6} \sum_{i=1}^6 x_i = \frac{0+0+1+0+1+0}{6} = \frac{2}{3} = \hat{\theta}_{MLE}$$

$$\underline{X_1, \dots, X_n | \theta \stackrel{iid}{\sim} \text{Bern}(\theta)} \Rightarrow \hat{\theta}_{MLE} = \bar{X}$$

MLE isn't the only point estimate strategy but it is common as it has nice properties

① $\hat{\theta}_{MLE} \xrightarrow[n \rightarrow \infty]{} \theta$ "consistency" i.e., it converges to the true value in n

(II) $\hat{\theta}_{MLE} \approx N(\theta, SE[\hat{\theta}_{MLE}]^2)$ Asymptotic Normality
standard error

(III) Among all consistent continuous, $\hat{\theta}_{MLE}$ has variable efficiency

Geometric case

$\tilde{F} = \text{iid Geometric}$ $X \sim \text{Geom}(\theta) = (1-\theta)^x \theta$ $L(\theta; x) = \prod_{i=1}^n (1-\theta)^{x_i} \theta$

take log $\Rightarrow L(\theta; x) = \sum_{i=1}^n \left[\ln \left[(1-\theta)^{x_i} \theta \right] \right] = n \ln(\theta) + \ln(1-\theta) \sum_{i=1}^n x_i = n \left(\ln(\theta) + \bar{x} \ln(1-\theta) \right)$

take derivative w.r.t θ

$L'(\theta; x) = n \left(-\frac{\bar{x}}{1-\theta} + \frac{1}{\theta} \right) \stackrel{\text{set}}{=} 0 \quad \theta = \frac{1}{1+\bar{x}}$

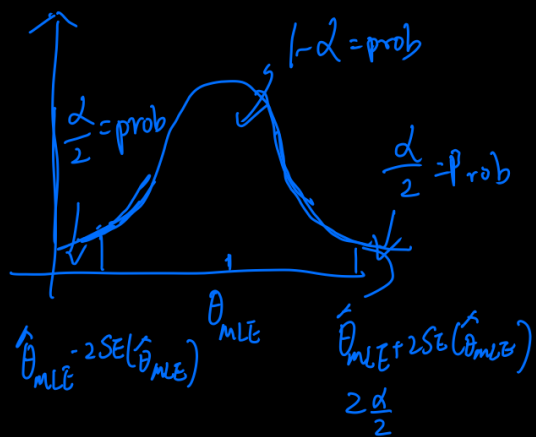
$\bar{x} = 99 \quad \theta_{MLE} = 0.01 \quad 1\%$
 $\bar{x} = 1 \quad \theta_{MLE} = 50\%$

if \tilde{F} is iid Bern $\hat{\theta}_{MLE} = \bar{X}$ $SE(\hat{\theta}_{MLE}) = SE[\bar{x}] = \sqrt{\frac{\theta(1-\theta)}{n}}$

$\hat{\theta}_{MLE} \stackrel{d}{\approx} N(\theta, SE(\hat{\theta}_{MLE})^2) \stackrel{\text{I}}{\approx} N\left(\hat{\theta}_{MLE}, SE\left[\hat{\theta}_{MLE}\right] \Big|_{\theta=\hat{\theta}_{MLE}}\right) \approx N\left(\bar{x}, \sqrt{\frac{\bar{x}(1-\bar{x})}{n}}\right)$
 Asymptotic

$\tilde{F} = \text{iid Geometric}$ $\hat{\theta}_{MLE} = \frac{1}{1+\bar{x}}$, $SE[\hat{\theta}_{MLE}] = SE\left[\frac{1}{1+\bar{x}}\right] = ?$

goal inference "____" is "provide a range of possible value of θ .



math 242, Econ 382
 Confidence interval (CI) _{$\theta, 1-\alpha$} = $\left[\hat{\theta}_{MLE} \pm Z_{\frac{\alpha}{2}} SE(\hat{\theta}_{MLE}) \right]$
 $1-\alpha$ is probability value

Third goal of Inference: Testing

also called Hypothesis Theory (half of math 242 ~~or~~ Econ 382 will teach)

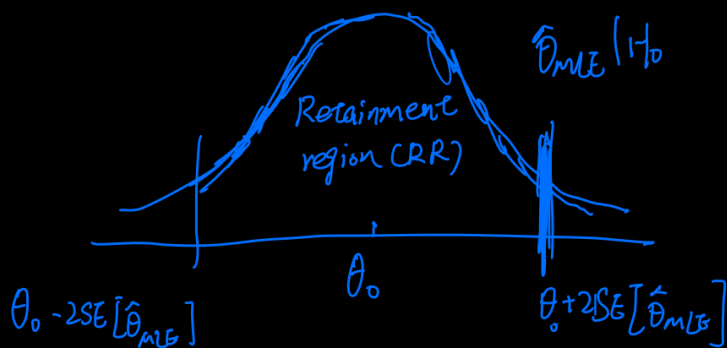
"null hypothesis" $H_0: \theta = \theta_0$ ^{some specific value}

"hypothesis" $H_a: \theta \neq \theta_0$

Assume my theory _____ and _____

tell me if I'm right or wrong

$$\Rightarrow \hat{\theta}_{MLE} \sim N(\theta_0, SE(\hat{\theta}_{MLE})^2)_{|\theta=\theta_0}$$



$$RR_{\theta_0, 1-\alpha} = \left[\theta_0 \pm Z_{\frac{\alpha}{2}} SE(\hat{\theta}_{MLE}) \right]_{|\theta=\theta_0}$$

if $\hat{\theta}_{MLE} \in RR \Rightarrow$ Retain H_0

if $\hat{\theta}_{MLE} \notin RR \Rightarrow$ Reject H_0 / Accept H_a

We have a strategy for all 3 inference goals we've done

"frequency inference"

which is a historical class