

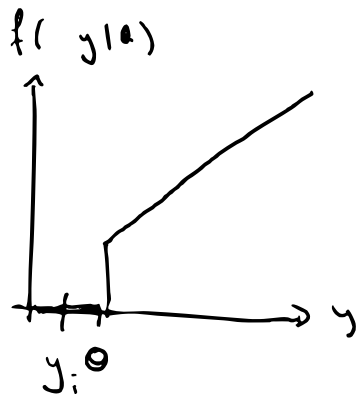
QUIZ 1

MLE

$$f(y; \theta) = \frac{2y}{1-\theta^2}$$

$$0 \leq \theta \leq y \leq 1$$

$$y_i \in \{0.7, 0.63, 0.92, 0.86, 0.43,$$



$$\bullet \theta \leq y_i = 0.21$$

$$\bullet \ln L$$

$$= \sum_{i=1}^n \ln \left(\frac{2y_i}{1-\theta^2} \right)$$

$$\hat{\theta}_{MLE} = 0.21$$

argmax θ $f(y|\theta)$

$$\frac{\partial \ln L}{\partial \theta} > 0$$

$$= \sum_{i=1}^n \ln(2y_i) - \ln(1-\theta^2)$$

$$= \underbrace{n \ln 2}_{\text{constant}} - \underbrace{\ln(1-\theta^2)}_{\text{decreasing}} + \sum_{i=1}^n \ln y_i$$

QUIZ 2

R, w

$$p \in \{1/5, 1/3, 1/2, 3/4\}$$

4: R w w w

1R 3w

MLE of p .

$X = \# \text{ Reds}$

~~$\mathbb{E} x_i = 0$~~

~~X~~ $\Rightarrow X \sim B(4, p)$

~~Unconstrained~~
~~MLE~~ $p = 1/4$

$P(X=1|p)$ for $p \in \{1/5, 1/3, 1/2, 3/4\}$

$$P(X=1 | p=1/5) = \underbrace{\binom{4}{1}}_4 \left(\frac{1}{5}\right)^1 \left(\frac{4}{5}\right)^3 = 4 \cdot \frac{1 \times 4^3}{5^4}$$

$$P(X=1 \mid p = \frac{1}{3}) = \binom{4}{1} \left(\frac{1}{3}\right)^1 \left(\frac{2}{3}\right)^3 .$$

$$P(X=1 \mid p = \frac{1}{2}) = \binom{4}{1} \left(\frac{1}{2}\right)^1 \left(\frac{1}{2}\right)^3 .$$

$$P(X=1 \mid p = \frac{3}{4}) = \binom{4}{1} \left(\frac{3}{4}\right)^1 \left(\frac{1}{4}\right)^3 .$$

$$P(X=1 \mid p = \frac{1}{5})$$

$$\hat{p}_{MLE} = \frac{1}{5} . \quad \square$$

~~Q.E.D.~~

Quiz 3

$$n = 6$$



$\{y_1, \dots, y_6\}$

~~the~~ $f(y|\theta) = \frac{1}{3}$

$$\hat{\theta} = y_{\max}$$

$$0 \leq y \leq 3$$

~~or~~

$$P(\hat{\theta} \in [\theta - 0.3, \theta + 0.3])$$

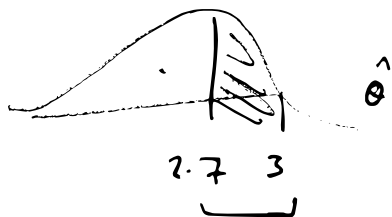
||

$$P(\hat{\theta} \leq 3) - P(\hat{\theta} \leq 2.7)$$

$$P(2.7 \leq \hat{\theta} \leq 3.3)$$

||

$$P(2.7 \leq \hat{\theta} \leq 3)$$



↖

$$1 - P(\hat{\theta} \leq 2.7)$$

$$P(Y_{\max} \leq 2.7) = P(Y_1 < 2.7 \wedge Y_2 < 2.7 \dots)$$

Random sample

\Leftrightarrow iid

Independence

$$P(A \cap B) = P(A)P(B)$$

$$= \prod_{i=1}^6 P(Y_i < 2.7)$$

Identically distributed

$$= P(Y_1 < 2.7)^6$$



$$Y_i \sim U[0, 3]$$

$$P(Y_1 < 2.7) = \int_0^{2.7} \frac{1}{3} dy$$

$$= \left[\frac{1}{3} y \right]_0^{2.7}$$

$$= \frac{2.7}{3} = \frac{9}{10}$$

$$P(\hat{\theta} \in [2.7, 3.3])$$

$$= 1 - \left(\frac{9}{10}\right)^6 \approx 0.47. \quad \square$$

Q112 4

5: {17, 42, 46, 39, 56}

$$E(Y|\hat{\theta}) = \frac{1}{n} \sum y_i$$

\uparrow
 $\frac{0}{2}$

\uparrow
50

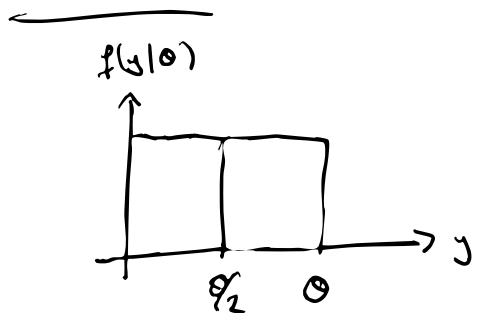
$$E(Y) = \int_0^{\theta} \frac{y}{\theta} dy$$

$$= \frac{1}{\theta} \left[\frac{1}{2} y^2 \right]_0^{\theta} =$$

$$\frac{1}{2} \theta = 50$$

$$\underline{\hat{\theta} = 100}$$

$$f(y|\theta) = \frac{1}{\theta} \quad 0 \leq y \leq \theta$$



$$\frac{1}{2} \frac{\theta^2}{\theta} - 0 = \frac{\theta}{2}$$

□

Quiz 5

$$X \sim N(0, \sigma^2)$$

$$\theta, \sigma^2$$

$$n = 9$$

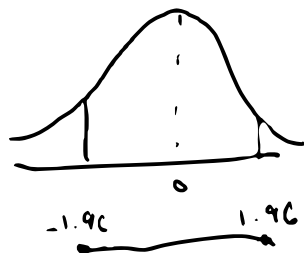
$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i = 1$$

$$S = \left[\frac{1}{n} \sum_{i=1}^n x_i^2 - \bar{x}^2 \right]^{1/2} = 3$$

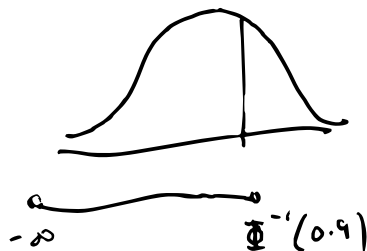
$$P(X \in C) = 90\%$$

'A CI'

'The CI'



'The CI'



CI \Leftrightarrow Test

'The CI' \Leftrightarrow "Most powerful" test.

$$P(\bar{X} \in C) = 95\%$$

$$\rightarrow \bar{x} = 1 \quad s = 3 \quad n = 9$$

$$T = \frac{1}{3/\sqrt{9}} = 1$$

$$[\bar{x} - s z, \bar{x} + s z]$$

$$H_0: \theta = 0$$

$$H_1: \theta \neq 0$$

$$\bar{x}$$

$$s/\sqrt{n}$$

$$P(\bar{X} \in (-\infty, 2.86]) = 0.95$$

\Leftrightarrow

$$P(\bar{X} > 2.86) = 0.05$$

$$P(\bar{X} > 2.86) = 0.05$$

$$P\left(\frac{\bar{X} - \bar{x}}{s/\sqrt{n}} > \frac{2.86 - \bar{x}}{s/\sqrt{n}}\right)$$

$\underbrace{\hspace{1.5cm}}_T \qquad \qquad \underbrace{\hspace{1.5cm}}_1$

$$P(\bar{X} > 2.86) = P\left(T > \frac{2.86 - 1}{1}\right)$$

$$= P(T > 1.86)$$

$$= 5\%$$

□

$$X \sim N(5, 7)$$

↓

$$\frac{X - 5}{\sqrt{7}} \sim N(0, 1)$$

$$P(-5.91 \leq \overset{\downarrow}{(\bar{X})} \leq 7.91)$$

$$= P\left(\frac{-5.91 - 1}{1} \leq T \leq \frac{7.91 - 1}{1}\right)$$

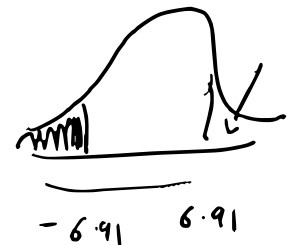
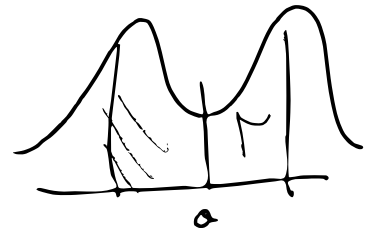
$$= P(-6.91 \leq T \leq 6.91)$$

$$= P(T \leq 6.91) - P(T \leq -6.91)$$

$$= P(T \leq 6.91) - [1 - P(T \leq 6.91)]$$

$$= \underline{2 \cdot P(T \leq 6.91) - 1} \neq 5\%$$

$$0: \{\overset{\downarrow}{\hat{\theta}} - \sim, \overset{\downarrow}{\hat{\theta}} + \sim\}$$



Quiz 6

$$y_1, \dots, y_n$$

$$\hat{\theta}_1 = y_1$$

$$\hat{\theta}_2 = n \cdot \min y_i$$

$$f(y|\theta) = \frac{1}{\theta} e^{-y/\theta} \quad \theta > 0 \quad y \geq 0$$

$$\exp\left(\frac{1}{\theta}\right)$$

Un Biased:

$$E \hat{\theta} = \theta$$

THINK OF AN ARCHERY
TARGET

UNBIASED:

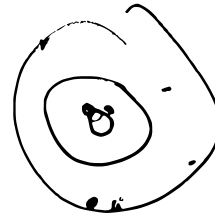
ARROWS HIT MIDDLE
ON AVERAGE



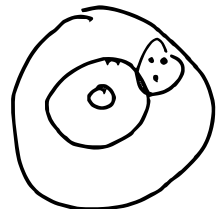
EFFICIENT:

ARROWS CLOSELY
PACKED TOGETHER

$$MSE = Bias^2 + Var$$



U



E

$$f(y|\theta) = \frac{1}{\theta} e^{-y/\theta}$$

$$F(y|\theta) = 1 - e^{-y/\theta}$$

$$f_{Y_{\min}}(y) = n \cdot \underbrace{f(y)} \cdot \underbrace{[1 - F(y)]^{n-1}}$$

$$f_{Y_{\min}}(y) = n \cdot \left[\frac{1}{\theta} e^{-y/\theta} \right]$$

$$\cdot \left[e^{-y/\theta} \right]^{n-1}$$

$$e^{A+B} = e^A e^B$$

$$\hat{\theta}_1 = Y_{01} \sim \exp\left(\frac{1}{\theta}\right)$$

$$\hat{\theta}_1 = n \cdot Y_{\min} \sim \exp\left(\frac{n}{\theta}\right) \cdot n$$

$$= \frac{n}{\theta} e^{-y/\theta} \cdot \underbrace{e^{(-y/\theta)(n-1)}}$$

$$= \frac{n}{\theta} e^{-y/\theta \cdot n}$$

$$= \left(\frac{n}{\theta}\right) e^{-\left(\frac{n}{\theta}\right)y}$$

$$\exp.(\lambda) = \lambda e^{-\lambda y}$$

$$\hat{\theta}_1 \sim \exp\left(\frac{1}{\theta}\right)$$

$$\hat{\theta}_2 \sim n \cdot \exp\left(\frac{n}{\theta}\right)$$

$$\mathbb{E}(\exp(\lambda)) = \frac{1}{\lambda}$$

$$f(y|\lambda) = \lambda e^{-\lambda y}$$

$y > 0$

$$\mathbb{E} \hat{\theta}_1 = \theta$$

$$\mathbb{E} \hat{\theta}_2 = \frac{\theta}{n} \cdot n = \theta$$

} Both
unbiased

$$\text{Var}(\exp(\lambda)) = \frac{1}{\lambda^2}$$

$$\text{Var}(\hat{\theta}_1) = \theta = \text{Var}(\hat{\theta}_2)$$

↑
equally efficient.

l.

QUIZ 7

• $X_4 \sim N(\mu, \sigma^2)$

$n=3$

\bar{X}
"

(X_4)

$\frac{1}{3}(X_1 + X_2 + X_3)$

• $\bar{X} \sim N(\mu, \frac{\sigma^2}{3})$

$P(\bar{X} - 2\sigma \leq X_4 \leq \bar{X} + 2\sigma)$

$\frac{1}{3}(X_1 + X_2 + X_3)$

$\sim N(3\mu, \sigma^2)$

$E(X_4 - \bar{X}) = 0$

$X_4 \perp \bar{X} \Rightarrow \text{cov}(\cdot) = 0$

$\text{Var}(X_4 - \bar{X}) = \text{Var}(X_4) + \text{Var}(\bar{X}) + \cancel{\text{cov}(X_4, \bar{X})}$

$= \sigma^2 + \frac{\sigma^2}{3} = \frac{4\sigma^2}{3}$

$$P(\bar{x} - 2\sigma \leq X_4 \leq \bar{x} + 2\sigma)$$

$$= P(\bar{x} - 2\sigma \leq \underline{X_4 - \bar{x}} \leq 2\sigma)$$

$$= P\left(-\frac{2\sigma}{(2\sigma/\sqrt{3})} \leq \frac{X_4 - \bar{x}}{2\sigma/\sqrt{3}} \leq \frac{2\sigma}{(2\sigma/\sqrt{3})}\right)$$

$$\text{Var}(X_4 - \bar{x}) = \frac{4\sigma^2}{3}$$

$$\sqrt{\text{Var}(\dots)} = \frac{2\sigma}{\sqrt{3}}$$

$$= P\left(-\sqrt{3} \leq \frac{X_4 - \bar{x}}{2\sigma/\sqrt{3}} \leq \sqrt{3}\right)$$

$$= P(-\sqrt{3} \leq Z \leq \sqrt{3})$$

$$X_4 - \bar{x} \sim N(0, \frac{4\sigma^2}{3})$$

$$\frac{X_4 - \bar{x}}{2\sigma/\sqrt{3}} \sim N(0, 1)$$

$$= P(Z \leq \sqrt{3}) - P(Z \leq -\sqrt{3})$$

$$= 2P(Z \leq \sqrt{3}) - 1 \quad \square$$

$$f(x) = \frac{x^{k-1} e^{-x/\theta}}{\Gamma(k) \theta^k}$$

$$k > 0, \theta > 0$$

$$\Gamma(k) = \int_0^{\infty} t^{k-1} e^{-t} dt$$

$$E X^{k-1} \quad X \sim \exp(1)$$

$$\hat{\theta}, \hat{k} = \underset{\theta, k}{\operatorname{argmax}} L(x|\theta, k)$$

$$L = \prod_{i=1}^n \frac{x_i^{k-1} e^{-x_i/\theta}}{\Gamma(k) \theta^k}$$

Independence

$$P(A \cap B) = P(A)P(B) \quad \{X_i\} \text{ iid}$$

Optimization

$$x_0 = \underset{x}{\operatorname{argmax}} f(x)$$

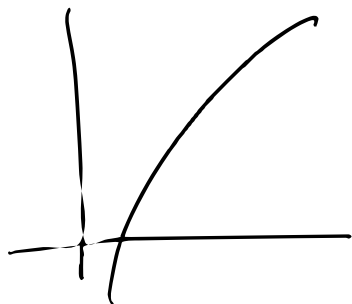
$$\forall x \quad g(f(x_0)) \geq g(f(x)) \Leftrightarrow$$

$$\Leftrightarrow f(x_0) \geq f(x) \quad \forall x$$

$$\Leftrightarrow x_0 = \underset{x}{\operatorname{argmax}} g(f(x))$$

g increasing

$$y = \ln x$$



$$\frac{\partial}{\partial x} \ln x = \frac{1}{x} > 0$$

$$\ln \left(\frac{1}{\Gamma(k)\theta^k} \right)^n = -n [\ln(\Gamma(k)\theta^k)]$$

$$L = \prod_{i=1}^n \frac{x_i^{k-1} e^{-x_i/\theta}}{(\Gamma(k)\theta^k)} = -n \ln(\Gamma(k)) - nk \ln(\theta)$$

$$e^{A+B} = e^A \cdot e^B$$

$$= \frac{1}{(\Gamma(k)\theta^k)^n} \left[x_1^{k-1} e^{-x_1/\theta} \cdot x_2^{k-1} e^{-x_2/\theta} \cdots \right]$$

$$= \frac{1}{(\Gamma(k)\theta^k)^n} \left[\prod_{i=1}^n x_i^{k-1} \cdot e^{-\frac{1}{\theta} \sum_{i=1}^n x_i} \right]$$

$$\begin{aligned} \log L(k, \theta) &= -nk \ln(\theta) - n \cdot \ln(\Gamma(k)) \\ &\quad + (k-1) \sum_i \ln(x_i) - \frac{1}{\theta} \sum x_i \end{aligned}$$

$$\bullet \quad \frac{1}{n} \sum x_i = E(x | \hat{\theta}, \hat{k})$$

$$X \sim \text{Gamma}(k, \theta)$$

$$E X = \theta k$$

$$\text{Var}(X) = \theta^2 k$$

$$\frac{1}{n} \sum x_i^2 = E(x^2 | \hat{\theta}, \hat{k})$$

$$\bullet \quad s^2 = \frac{1}{n-1} \sum (x_i - \bar{x})^2 = \text{Var}(x | \hat{\theta}, \hat{k})$$

$$\left\{ \begin{array}{l} s^2 = \hat{\theta}^2 \hat{k} \\ \bar{x} = \hat{\theta} \hat{k} \end{array} \right.$$

$$\bar{X} = \hat{\Theta}^1 k^1$$

$$S^2 = \hat{\Theta}^{12} k^1$$

$$\frac{S^2}{\bar{X}} = \frac{\hat{\Theta}^{12} k^1}{\hat{\Theta}^1 k^1} = \hat{\Theta}^1 \quad \square$$

Better
est.

Most
assumptions

MLE

↓

OLS

↓

kernel
reg.

$$\bar{X} \approx \frac{S^2}{\bar{X}} k^1$$

$$\frac{\bar{X}^2}{S^2} = k^1$$

$$\square \quad y_i = x_i^1 \beta + \varepsilon_i$$

$$\underline{\underline{E x_i \varepsilon_i = 0}}$$

$$E x = \Theta k$$

$$Var(x) = \Theta^2 k$$

$$y_i = f(x_i) + \varepsilon_i$$

f continuous.