

Recap:

pivotal quantities $g(\underline{x}, \theta)$ whose distributions are free of the parameter.

Ex: $x_1, \dots, x_n \stackrel{iid}{\sim} N(\mu, \sigma^2)$, μ, σ^2 both unknown,

Goal: Find 95% CI for σ^2 .

$$\boxed{\frac{(n-1)S^2}{\sigma^2}} \sim \chi_{n-1}^2$$

↑
pivotal quantity. This can be used to find 95% CI for σ^2

$$P\left(\chi_{n-1, 0.025}^2 \leq \frac{(n-1)S^2}{\sigma^2} \leq \chi_{n-1, 0.975}^2\right) = 0.95$$

$$\Leftrightarrow P\left(\frac{(n-1)S^2}{\chi_{n-1, 0.975}^2} \leq \sigma^2 \leq \frac{(n-1)S^2}{\chi_{n-1, 0.025}^2}\right) = 0.95$$

95% CI for σ^2 is $\left[\frac{(n-1)S^2}{\chi_{n-1, 0.975}^2}, \frac{(n-1)S^2}{\chi_{n-1, 0.025}^2} \right]$

Poisson Interval estimator:

$x_1, \dots, x_n \stackrel{iid}{\sim} \text{Pois}(\lambda)$, need to find $100(1-\alpha)\%$ CI for λ .

Note that $Y = \sum_{i=1}^n x_i \sim \text{Pois}(n\lambda)$. Let,

$$S = \left\{ \lambda : F_\lambda(Y) \leq 1 - \frac{\alpha}{2}, F_\lambda(Y) \geq \frac{\alpha}{2} \right\}$$

where $F_\lambda(y)$ is the CDF evaluated at y of $\sum_{i=1}^n x_i$. Then S is a $100(1-\alpha)\%$ confidence set for λ .

$$F_\lambda(y) = \sum_{k=0}^y e^{-n\lambda} \frac{(n\lambda)^k}{k!} = P(X_{2(y+1)}^\vee > 2n\lambda)$$

This can be proved by using the fact that $X_{2(y+1)}^\vee \stackrel{d}{=} \text{Gamma}(y+1, \frac{1}{2})$ and by using integration by parts.

$$P(X_{2(y+1)}^\vee > 2n\lambda) = F_\lambda(y) \geq \frac{\alpha}{2}$$

this tells us that $\lambda < \frac{X_{2(y+1)}^\vee, 1-\alpha/2}{2n}$

Also use the other equation

$$F_\lambda(y) \leq 1 - \alpha/2 \quad \text{to ensure that}$$

$$\lambda > \frac{X_{2(y+1)}^\vee, \alpha/2}{2n}$$

$$\Leftrightarrow S = \left\{ \lambda : \frac{X_{2(y+1)}^\vee, \alpha/2}{2n} < \lambda < \frac{X_{2(y+1)}^\vee, 1-\alpha/2}{2n} \right\}$$

Bayesian Credible intervals:

In frequentist θ is a fixed but unknown quantity. Confidence interval is a random quantity as all x_i 's are random.

Ex: ~~x_1, \dots, x_n~~ $x_1, x_2 \stackrel{iid}{\sim} U\left[\theta - \frac{1}{2}, \theta + \frac{1}{2}\right]$

Since $x_i - \theta \sim U\left[-\frac{1}{2}, \frac{1}{2}\right]$, this is a location family.

~~$\bar{x} - \theta$~~ $\bar{x} - \theta$ is a pivotal quantity.

$$P(-c \leq \bar{x} - \theta \leq c) = 1 - \alpha \quad \text{--- (1)}$$

where, ~~c~~ c is chosen from the distribution of $\bar{x} - \theta$. Note that, this distribution is free of θ .

① is equivalent to

$$P(\bar{x} - c \leq \theta \leq \bar{x} + c) = 1 - \alpha \quad \text{--- (2)}$$

\Rightarrow $100(1-\alpha)\%$ CI for θ is $[\bar{x} - c, \bar{x} + c]$.

Suppose you observe $x_1 = 1, x_2 = 2$. Then θ has to equal to 1.5.

\Rightarrow $[\bar{x} - c, \bar{x} + c] = [1.5 - c, 1.5 + c]$ must contain θ .

As a Bayesian ~~we start~~ we assume the parameter θ is random and

② calculate posterior distribution of $\theta | x_1, \dots, x_n$ from the prior distribution $\pi(\theta)$.

$$\pi(\theta | x_1, \dots, x_n) \propto \pi(\theta) \prod_{i=1}^n f(x_i | \theta)$$

We define \odot credible intervals of θ .

Definition: For $0 < \alpha < 1$, a $100(1-\alpha)\%$ credible set

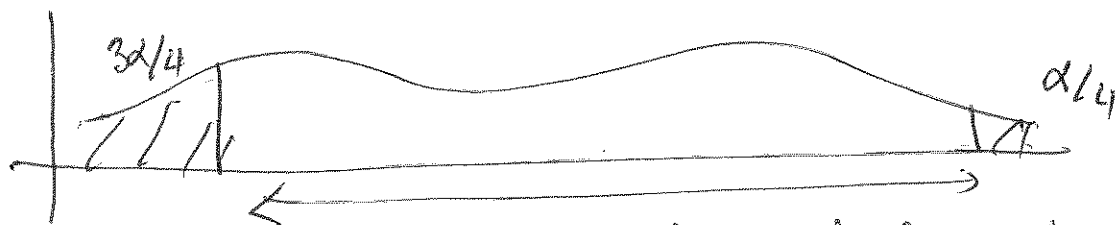
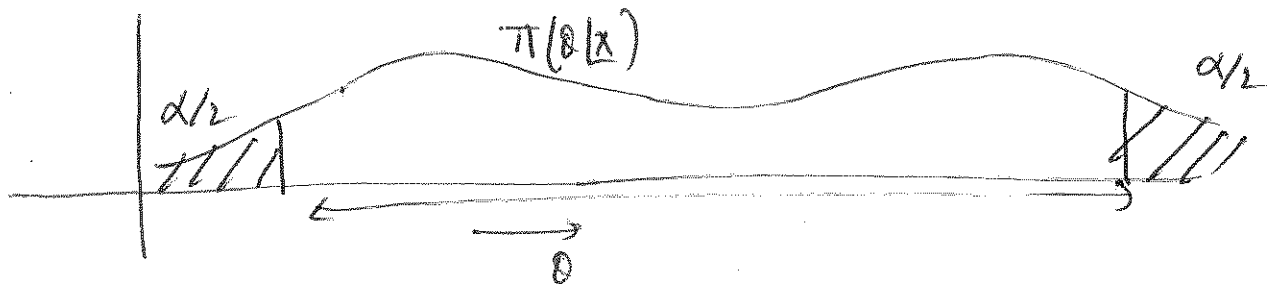
C is given by a set s.t. $P(\theta \in C | \underline{x}) = 1 - \alpha$.

If the credible set is an interval, then it is called a credible interval.

If you have a posterior dist. $\pi(\theta | \underline{x})$, you are interested in finding an interval

(a, b) s.t. $\int_a^b \pi(\theta | \underline{x}) d\theta = 1 - \alpha$.

There are many different ways ~~to find~~ to find $100(1-\alpha)\%$ credible intervals.

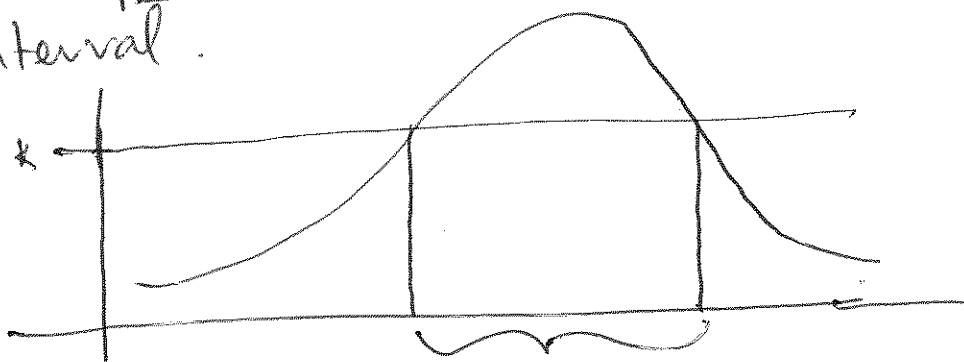


Bayesians prefer to construct highest posterior density credible intervals.

For a posterior density $\pi(\theta|x)$, the highest posterior density credible set is given by

$$C = \{ \theta : \pi(\theta|x) \geq k \}$$

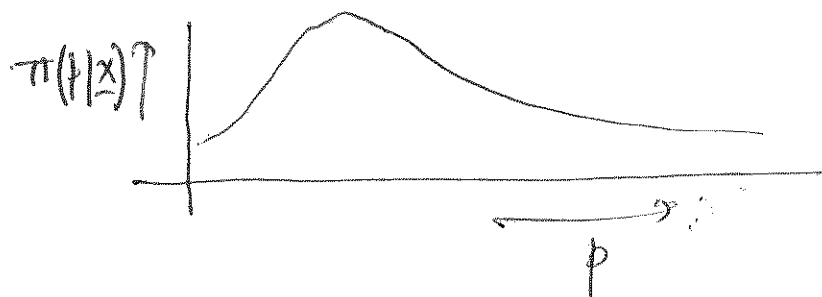
the $\theta|x$ is unimodal C is actually an interval.



This is called the ~~HPD~~ highest posterior density (HPD) interval.

Example: $X_1, \dots, X_n \stackrel{iid}{\sim} \text{Ber}(p)$, $p \sim \text{Beta}(a, b)$

$$p | X_1, \dots, X_n \sim \text{Beta}\left(a + \sum_{i=1}^n X_i, b + n - \sum_{i=1}^n X_i\right)$$



one can find HPD $100(1-\alpha)\%$ interval easily.

Look at the set $C = \{p: \pi(p|x) \geq k\}$

C is an interval.

Now find k by equating

$$\pi(C | x) = 1 - \alpha.$$

When the posterior is not in a closed form
use MCMC, draw M MCMC samples of θ .

Let these MCMC samples of θ be

$$\theta^{(1)}, \dots, \theta^{(M)}$$

Approximate posterior dist. $\pi(\theta|x)$ by a ~~dist.~~
discrete dist. that has masses at $\theta^{(1)}, \dots, \theta^{(M)}$
each with prob. $\frac{1}{M}$.

2:00 PM : Monday.

Bayesian ~~confidence~~ credible intervals

HPD credible interval.

$$\{\theta : \pi(\theta | X) > \kappa\} = \text{HPD credible set.}$$

When $\pi(\theta | X)$ is unimodal, this set will ~~give~~
result in an interval.

~~Let~~ $\theta^{(1)}, \dots, \theta^{(M)}$ be the M MCMC
post burn-in samples.

Construct an empirical dist. ~~using~~ using $\theta^{(1)}, \dots, \theta^{(M)}$,
i.e. a dist. which puts mass $\frac{1}{M}$ to any $\theta^{(i)}$.

Now one can create HPD credible interval in
the following way.

Order $\theta^{(1)}, \dots, \theta^{(M)}$ in an increasing
way

$$\theta_n^{(1)} < \dots < \theta_n^{(M)}$$

look for i s.t. $\frac{i}{M} = \alpha/2$

Also look for j s.t. $\frac{j}{M} = 1 - \alpha/2$.

then $[\theta_n^{(i)}, \theta_n^{(j)}]$ is the HPD credible interval.

Method of evaluating confidence intervals

Let's fix the coverage and look at the length of the confidence intervals.

~~Then~~ $x_1, \dots, x_n \stackrel{iid}{\sim} N(\mu, \sigma^2)$, σ^2 known

$$\left\{ \mu: \bar{x} + \frac{\sigma}{\sqrt{n}} a \leq \mu \leq \bar{x} + \frac{\sigma}{\sqrt{n}} b \right\}$$

$$a = -b = z_{\alpha/2}$$

Thm: If $f(x)$ is unimodal pdf. If the interval $[a, b]$ satisfies

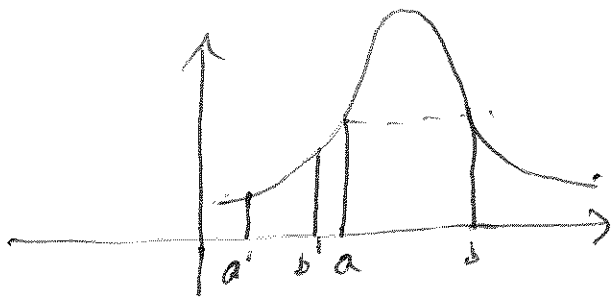
$$(1) \int_a^b f(x) dx = 1 - \alpha \quad (2) f(a) = f(b) > 0 \quad (3) a \leq x^* \leq b,$$

where x^* is the mode of $f(x)$. Then $[a, b]$ is the shortest interval that satisfies (1).

Pf: Let $[a', b']$ be any other interval satisfying $b' - a' < b - a$. We will prove $\int_{a'}^{b'} f(x) dx < 1 - \alpha$.

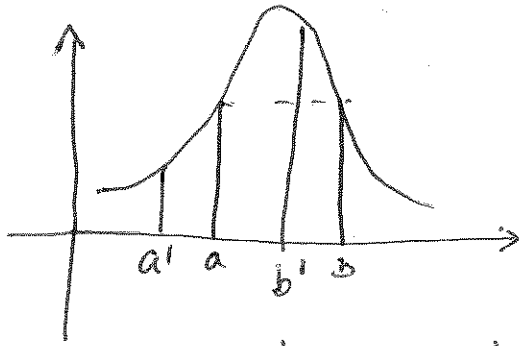
We will prove the result for $a' \leq a$. The proof will be similar for $a' > a$. Also for $a \leq a'$, consider two cases:

Case 1: $b' \leq a$, i.e. $a' \leq b' \leq a < b$



$$\begin{aligned} \int_{a'}^{b'} f(x) dx &\leq f(a) (b' - a') \\ &< f(a) (b - a) < \int_a^b f(x) dx \\ &= 1 - \alpha \end{aligned}$$

Case 2!



$b' > a$ and $a' \leq a \leq b' \leq b$

$$\int_{a'}^{b'} f(x) dx = \int_a^b f(x) dx + \left[\int_{a'}^a f(x) dx - \int_{b'}^b f(x) dx \right]$$

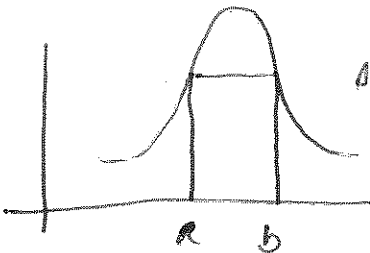
$$\int_{a'}^a f(x) dx \leq f(a)(a-a'), \quad \int_{b'}^b f(x) dx \geq f(b)(b-b')$$

$$\Rightarrow \int_{a'}^{b'} f(x) dx \leq \int_a^b f(x) dx + f(a)(a-a') - f(b)(b-b')$$

$$= \int_a^b f(x) dx + f(a) [(a-a') - (b-b')] \quad [as \underline{f(a)=f(b)}]$$

$$= \int_a^b f(x) dx + f(a) \underbrace{[(b'-a') - (b-a)]}_{< 0}$$

$$< \int_a^b f(x) dx = 1 - \alpha.$$



s.t. $f(a) = f(b)$.

This is an important result for the location families where the length of the interval $= c(b-a)$.

look at the interval μ for $x_1, \dots, x_n \stackrel{iid}{\sim} N(\mu, \sigma^2)$

$$\left\{ \mu: \bar{x} + \frac{\sigma}{\sqrt{n}} a \leq \mu \leq \bar{x} + \frac{\sigma}{\sqrt{n}} b \right\}$$

the length of the interval is $(b-a) \frac{\sigma}{\sqrt{n}}$.

So, this length can be minimized if one chooses $f(b) = f(a)$.

This problem is equivalent to a constrained optimization problem.

minimize $(b-a)$ subject to $\int_a^b f(x) = 1-\alpha$.

Thus b has to be a

function of a . Take derivative w.r.t. a .

$$\frac{db}{da} - 1 = 0, \quad \frac{db}{da} f(b) - f(a) = 0$$

$$\Rightarrow f(b) = f(a).$$

Example: Suppose $X \sim \text{Gamma}(\kappa, \beta)$. The quantity

$Y = \frac{X}{\beta} \sim \text{Gamma}(\kappa, 1)$. Y is a pivot.

A $100(1-\alpha)\%$ CI for β can be obtained in the following way:

$$P\left(a \leq \frac{Y}{\beta} \leq b\right) = 1-\alpha \Leftrightarrow P\left(\frac{Y}{b} \leq \beta \leq \frac{Y}{a}\right) = 1-\alpha$$

$\Rightarrow \left[\frac{Y}{b}, \frac{Y}{a}\right]$ is a $100(1-\alpha)\%$ CI for β .

~~Any~~ This interval obtained from the pivotal

quantity is known as a pivotal interval.

~~Here~~ Here any pivotal interval using the pivot $\frac{Y}{\beta}$ should be of the form $[\frac{Y}{b}, \frac{Y}{a}]$

Qn: Find a, b that gives minimum length of such an interval with coverage $1-\alpha$.

$$\text{length} = Y \left[\frac{1}{a} - \frac{1}{b} \right]$$

Thus the optimization problem is $\min_{a, b} \frac{1}{a} - \frac{1}{b}$ s.t. $\int_a^b f(x) dx = 1-\alpha$.

$$\bullet \frac{1}{a^2} + \frac{1}{b^2} \frac{db}{da} = 0 \quad \text{and} \quad f(b) \frac{db}{da} - f(a) = 0$$

$$\Rightarrow \frac{db}{da} = \frac{b^2}{a^2} \Rightarrow f(b) \frac{b^2}{a^2} = f(a) \Rightarrow b^2 f(b) = a^2 f(a)$$

Thus we have to find a, b in the density Gamma($k, 1$) s.t. $a^2 f(a) = b^2 f(b)$.

This is the criterion to find the shortest interval among all pivotal intervals using the pivot $\frac{Y}{\beta}$.

Is there any connection between optimal tests and optimal intervals.

Note that probability of true coverage for an interval $C(\underline{x})$ is given by $P_{\theta}(\theta \in C(\underline{x}))$

The probability of false coverage is given by

$$P_{\theta}(\theta' \in C(\underline{x})), \theta' \neq \theta, \text{ if } C(\underline{x}) = [L(\underline{x}), U(\underline{x})]$$

$$P_{\theta}(\theta' \in C(\underline{x})), \theta < \theta', \text{ if } C(\underline{x}) = (-\infty, U(\underline{x}))]$$

$$P_{\theta}(\theta' \in C(\underline{x})), \theta > \theta', \text{ if } C(\underline{x}) = [L(\underline{x}), \infty)$$

Note that ~~coverage~~ the false coverage will be big if we unnecessarily cover unimportant θ' .

Thus we want an interval (or a set) which minimizes the prob. of false coverage.

UMA confidence set: A $(1-\alpha)$ confidence set that minimizes prob. of false coverage over a class of $(1-\alpha)$ confidence sets is called a uniformly most accurate (UMA) confidence set.

Thm: $X_1, \dots, X_n \stackrel{iid}{\sim} f_{\theta}(x)$, where θ is real valued.

For each $\theta_0 \in \Theta$, let $A^{\alpha}(\theta_0)$ be the UMP level α acceptance region of a test $H_0: \theta = \theta_0$ vs.

$H_1: \theta > \theta_0$, let $C^*(\underline{x})$ be the $(1-\alpha)$ confidence set formed by inverting the UMP acceptance regions.

Then for any $(1-\alpha)$ confidence set $C(\underline{x})$,

$$P_{\theta}(\theta' \in C^*(\underline{x})) \leq P_{\theta}(\theta' \in C(\underline{x})) \quad \forall \theta' < \theta.$$

$X_1, \dots, X_n \stackrel{iid}{\sim} N(\mu, \sigma^2)$,

$C(\underline{x}) = \left\{ \mu: \mu > \bar{x} - z_{\alpha} \frac{\sigma}{\sqrt{n}} \right\}$ is the $100(1-\alpha)\%$ CI for μ obtained by inverting the UMP test for $H_0: \mu = \mu_0$ vs. $H_1: \mu > \mu_0$.

$C(\underline{x})$ has the lowest prob. of false coverage.

⊗ Is there any optimality guarantee with UMPU?

Definition: A $(1-\alpha)$ confidence set $C(\underline{x})$ is called unbiased if $P_{\theta'}(\theta' \in C(\underline{x})) \leq 1-\alpha \quad \forall \theta' \neq \theta$.

$$P_{\theta}(\theta \in C(\underline{x})) = 1-\alpha.$$

~~the~~ ~~are~~ ~~obtained~~ By inverting acceptance region of a UMP unbiased (UMPU) test one obtains an unbiased interval.

$H_0: \theta = \theta_0$ vs. $H_1: \theta \neq \theta_0$

Thm: $X \sim f_{\theta}$, θ is a real valued parameter.

Let $C(\underline{x}) = [L(\underline{x}), U(\underline{x})]$ be the confidence interval for θ . If $L(\underline{x}), U(\underline{x})$ are both increasing functions of \underline{x} , then for any θ'

$$E_{\theta'}[\text{length}(C(\underline{x}))] = \int_{\theta \neq \theta'} P_{\theta'}(\theta \in C(\underline{x})) d\theta.$$

$$E_{\theta'}[\text{length}(C(x))] = \int_X \text{length}[C(x)] f_{\theta'}(x) dx$$

$$= \int_X [\cancel{U(x)} - L(x)] f_{\theta'}(x) dx$$

$$= \int_X \left[\int_{L(x)}^{U(x)} d\theta \right] f_{\theta'}(x) dx$$

$$= \int_{\theta \in \mathbb{H}} \left[\int_{U^{-1}(\theta)} f_{\theta'}(x) dx \right] d\theta$$

$$= \int_{\theta \in \mathbb{H}} P_{\theta'}(U^{-1}(\theta) \leq x \leq L^{-1}(\theta)) d\theta$$

$$= \int_{\theta \in \mathbb{H}} P_{\theta'}(L(x) \leq \theta \leq U(x)) d\theta$$

$$= \int_{\theta \in \mathbb{H}} P_{\theta'}(\theta \in C(x)) d\theta$$

back ①