

Stat 134 lec 36

Sec 6.3 Conditioning: density calc.

Recall $f_{Y|X=x}(y) = \frac{f(x,y)}{f_X(x)}$ conditional density

$$P(A) = \int_{x \in X} P(A|x=x) f_X(x) dx \quad \text{Integral Condition formula.}$$

warm up $X \sim \text{Unif}(0,1)$, $X = p$
 $I_1, \dots, I_n \stackrel{iid}{\sim} \text{Bern}(p)$

Find $P(\text{2nd toss H})$

$$\begin{aligned} P(\text{2nd toss H}) &= \int_0^1 P(\text{2nd H} | X=p) \cdot f_X(p) dp \\ &= \int_0^p p dp = \left(\frac{p^2}{2} \right) \end{aligned}$$

I have a p-cash but until I condition on $X=p$, p is a RV not a number

Bayesian Statistics

Last time we showed

$$P(1^{\text{st}} H, 2^{\text{nd}} H) = \frac{1}{5} \quad \text{and} \quad P(2^{\text{nd}} H | 1^{\text{st}} H) = \frac{2}{3}$$

$$\Rightarrow P(2^{\text{nd}} H | 1^{\text{st}} H) \neq P(2^{\text{nd}} H)$$

so our coin tosses are not independent unless we condition on $X=p$.

Here is how we compute the posterior density:

$$\Rightarrow f_{X|1^{\text{st}} H}(p) = \frac{P(1^{\text{st}} H | X=p) f_X(p)}{P(1^{\text{st}} H)} = \frac{\frac{1}{2} p}{\frac{1}{2}} = p$$

"posterior" "likelihood" "prior" "constant." "1/2"

Stat 134

Wednesday April 18 2018

1. Let $X \sim \text{Unif}(0, 1)$. Given $X=p$, let I_1, I_2, \dots, I_n be iid Bernoulli(p) trials. Given that the first two tosses are heads, the posterior density $f_{X|1\text{st H}, 2\text{nd H}}(p)$ is:

a $2p^2$

b $3p^2$

c $6p^2$

d none of the above

*two ways
to calculate:*

$$f_{X|1\text{st H}, 2\text{nd H}} = \frac{P(\text{2nd H} | 1\text{st H}, X=p) \cdot f_{X|1\text{st H}}^{(p)}}{P(\text{2nd H} | 1\text{st H})}$$

$\frac{p}{1}$ $\frac{2p}{1}$

$$f_{X|1\text{st H}, 2\text{nd H}} = \frac{P(\text{2nd H, 1st H} | X=p) f_X^{(x)}}{P(\text{1st H, 2nd H})}$$

$\frac{p^2}{1}$ $\frac{2/3}{1}$ $\frac{1}{1}$

$\frac{1}{3}$

Ex Find $P(3^{\text{rd}} H | 1^{\text{st}} H, 2^{\text{nd}} H)$

$$= \int_0^1 P(3^{\text{rd}} H | 1^{\text{st}} H, 2^{\text{nd}} H, X=p) \cdot f_{X|1^{\text{st}}, 2^{\text{nd}} H}(p) dp$$
$$= \int_0^1 3p^2 dp = \boxed{\frac{3}{4}}$$

$$P(1^{\text{st}} H, 2^{\text{nd}} H, 3^{\text{rd}} H) = \frac{1}{2} \cdot \frac{2}{3} \cdot \frac{3}{4} = \boxed{\frac{1}{4}}$$

What is

$$f_{X|1^{\text{st}} H, 2^{\text{nd}} H, 3^{\text{rd}} H} ? \quad - 4p^3$$

Review of Beta

$$\int_0^1 \underbrace{\frac{\Gamma(r+s)}{\Gamma(r)\Gamma(s)} p^{r-1} (1-p)^{s-1}}_{\text{beta}(r,s)} dp = 1$$

$$\Gamma(r) = (r-1)!$$

$$\text{beta}(1,1) = \text{Unif}(0,1)$$

Ex $X \sim \text{beta}(1,1), X = p$

S_n = the number of successes in n Bernoulli(p) trials.

Find $P(S_n=k)$:

$$P(S_n=k) = \int_0^1 P(S_n=k | X=p) f_X(p) dp$$

$$= \int_0^1 \binom{n}{k} p^{k+r-1} (1-p)^{n-k+s-1} 1 dp$$

$$\binom{n}{k} \cdot \frac{\Gamma(k+r) \Gamma(n-k+s)}{\Gamma(n+2)} = \frac{1}{n+1}$$

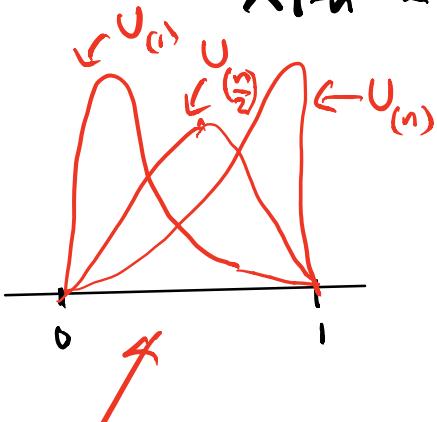
\Rightarrow Generalize

$$X \sim \text{beta}(r, s), X = p$$

$S_n = \# \text{ successes in } n \text{ Bern}(p) \text{ trials}$

Find posterior density $f_{X|S_n=k}(p) = \frac{\binom{n}{k} p^k (1-p)^{n-k} \Gamma(r+s)}{\Gamma(r) \Gamma(s)} p^k (1-p)^{n-k}$

$$f_{X|S_n=k}(p) = \frac{P(S_n=k|X=p) f_X(p)}{P(S_n=k)}$$



$$X \sim \text{beta}(r, s)$$

our prior density

is one of those,

which one is unbiased?

answ: $U\left(\frac{r}{r+s}\right)$ since $E(X) = \frac{r}{r+s}$

(i.e. $\text{Prob}(1^{\text{st}} \text{ H}) = \frac{r}{r+s}$)

$$= \frac{\Gamma(r+n+s)}{\Gamma(k+r) \Gamma(n-k+s)} p^{k+r-1} (1-p)^{n-k+s-1}$$

$$\text{beta}(k+r, n-k+s)$$

Here is some **extra material**, I didn't have time to cover in lecture but you are responsible for it:

Please Read.

In the above example the prior is $\text{beta}(r, s)$; the posterior is $\text{beta}(k+r, n-k+s)$ and the likelihood is binomial (n, p) where p is a value from the distribution of the prior.

When the prior and the posterior belong to the same family, we say that the prior and the likelihood are **conjugate**,

e.g. the beta and the binomial are conjugate

e.g. the gamma ^{Prior} and the Poisson ^{Likelihood} are conjugate.

When this happens it is easy to update the prior with the posterior, as we did in the i-clicker question

The summary on p 425 is useful.

Conditioning Formulae: Density Case

Multiplication rule: The joint density is the product of the marginal and the conditional

$$f(x, y) = f_X(x)f_Y(y | X = x)$$

Division rule: The conditional density of Y at y given $X = x$ is

$$f_Y(y | X = x) = \frac{f(x, y)}{f_X(x)}$$

Bayes' rule:

$$f_X(x | Y = y) = \frac{f_Y(y | X = x)f_X(x)}{f_Y(y)}$$

Conditional distribution of Y given $X = x$: Integrate the conditional density

$$P(Y \in B | X = x) = \int_B f_Y(y | X = x) dy \quad \begin{matrix} \text{I wrote conditional density} \\ f(y) \\ y|x=x \end{matrix}$$

Conditional expectation of $g(Y)$ given $X = x$: Integrate g against the conditional density:

$$E(g(Y) | X = x) = \int g(y)f_Y(y | X = x) dy$$

Average conditional probability:

$$P(B) = \int P(B | X = x)f_X(x) dx$$

$$f_Y(y) = \int f_Y(y | X = x)f_X(x) dx$$

Average conditional expectation:

$$E(Y) = \int E(Y | X = x)f_X(x) dx$$

We use the conditional density to compute the conditional expectation as you would expect.

Conditional Expectation

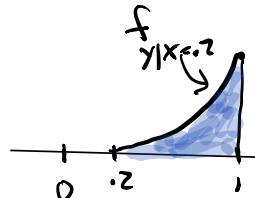
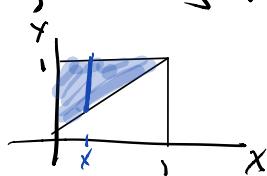
$$E(g(y) | x=z) = \int_{y \in Y} g(y) f_{Y|X=z}(y) dy$$

\Leftrightarrow Last time

$$f(x,y) = 90(y-x)^8, 0 < x < y < 1$$

we found

$$f_{Y|X=z}(y) = \frac{9}{(1-z)^9} (y-z)^8$$



Find $E(Y|X=.2)$

Soln

$$\begin{aligned} E(Y|X=.2) &= \int_{.2}^1 y f_{Y|X=.2}(y) dy \\ &= \int_{.2}^1 y \frac{9}{(.8)^9} (y-.2)^8 dy \end{aligned}$$

Substitute $u = y - .2, y = u + .2$

$$u = .8 \quad du = dy$$

$$= \frac{9}{(.8)^9} \int (u+.2) u^8 du$$

$$= \frac{9}{(.8)^9} \left[\int_0^{.8} u^9 du + .2 \int_0^{.8} u^8 du \right]$$

$$= \frac{9}{(.8)^9} \left[\frac{(.8)^{10}}{10} + \frac{.2(.8)^9}{9} \right]$$

$$= 9 \left[\frac{.8}{10} + \frac{.2}{9} \right] = \boxed{.92}$$