

Stat 134 lec 34

Last time

Sec 6.2

X discrete.

Conditional expectation

$$E(Y|X=x) = \sum_{y \in Y} y \cdot P(Y|X=x)$$

$E(Y|X)$ is a RV (function of X)

law of iterated expectation

$$E(Y) = E(E(Y|X))$$

Today

① Sec 6.2 Properties of conditional expectation.

② Sec 6.3

a) Conditional densities,

b) multiplication rule

c) rule of average conditional probabilities

①

Sec 6.2Properties of conditional expectation

For a fixed conditioning event A, conditional expectation has all the familiar properties of expectation

$$E(X+Y) = E(X) + E(Y)$$

$$E(X+Y|A) = E(X|A) + E(Y|A)$$

since $X+Y|A$ is
a restriction of
 $X+Y$ to a smaller
outcome space.

So suppose A is the event $X=x$

What is $E(X+Y|X=x)$?

$$= E(X|X=x) + E(Y|X=x)$$

$$= x + E(Y|X=x)$$

so we write, equality of RV.

$$E(X+Y|X) = X + E(Y|X)$$

Properties

- ① $E(X) = E(E(X|Y))$ equality of numbers
 - ② $E(aY+b|X) = aE(Y|X) + b$
 - ③ $E(Y+Z|X) = E(Y|X) + E(Z|X)$
 - ④ $E(g(X)|X) = g(X)$
 - ⑤ $E(g(X)Y|X) = g(X)E(Y|X)$
- equality of RV

(2)

(a)

sec 6.3 Conditional Density:

Let A be an event.

If X is a discrete RV.

$$P(A|X=x) = \frac{P(A, X=x)}{P(X=x)} \text{ by Bayes' rule.}$$

If X is continuous,

$$P(A|X \in dx) = \frac{P(A, X \in dx)}{P(X \in dx)}$$

We define

$$P(A|X=x) = \lim_{dx \rightarrow 0} \frac{P(A, X \in dx)}{P(X \in dx)}$$

If Y is continuous

$$P(Y \in dy | X=x) = \lim_{dx \rightarrow 0} \frac{P(Y \in dy, X \in dx)}{P(X \in dx)}$$

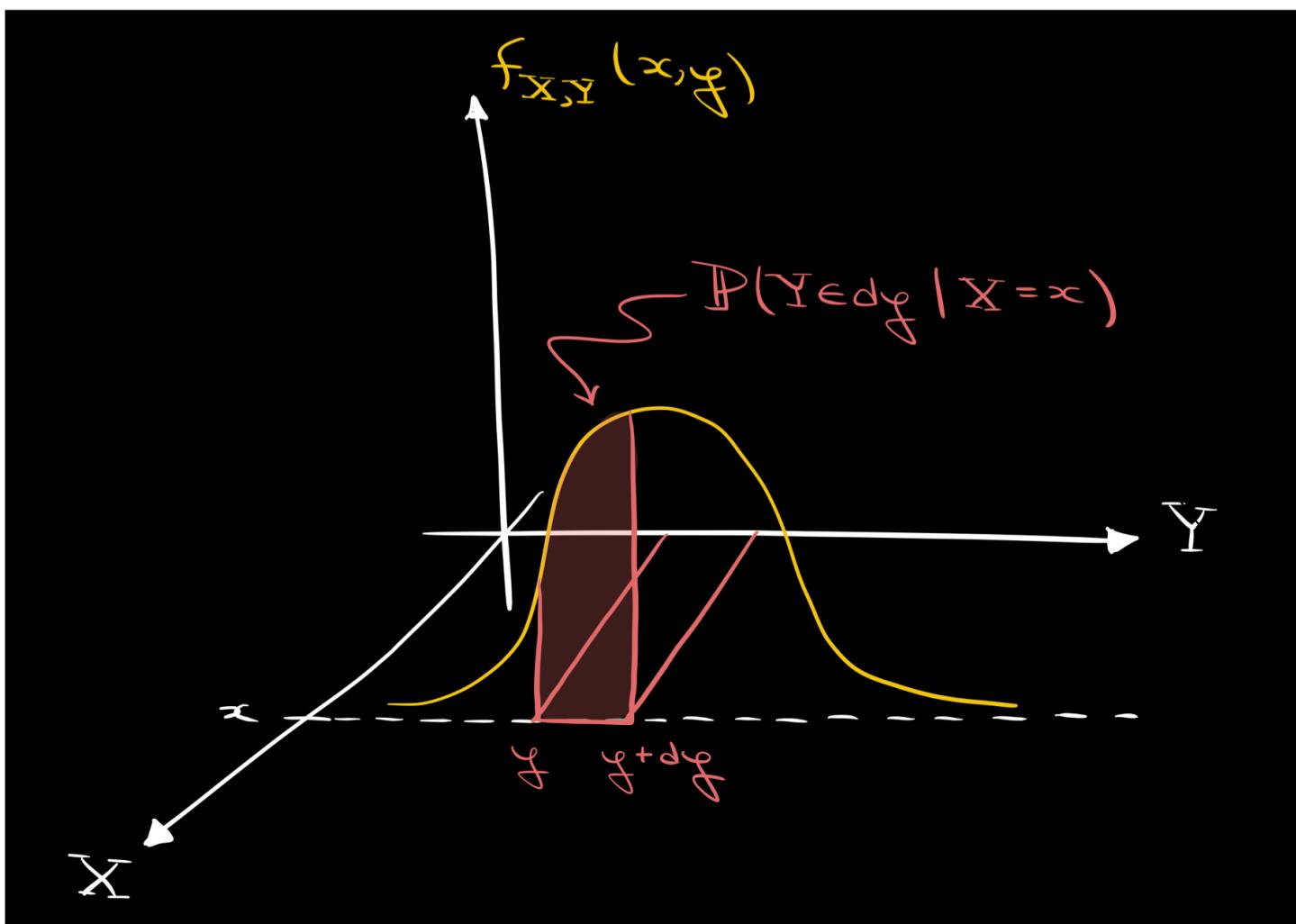
$$\begin{aligned} f_{Y|X=x}(y) dy &= \lim_{dx \rightarrow 0} \frac{f_{X,Y}(x,y) dx dy}{f_X(x) dx} \\ &= \frac{f_{X,Y}(x,y) dy}{f_X(x) dx} \end{aligned}$$

$$\Rightarrow \boxed{f_{Y|X=x}(y) = \frac{f_{X,Y}(x,y)}{f_X(x)}}$$

conditional density
of Y given $X=x$

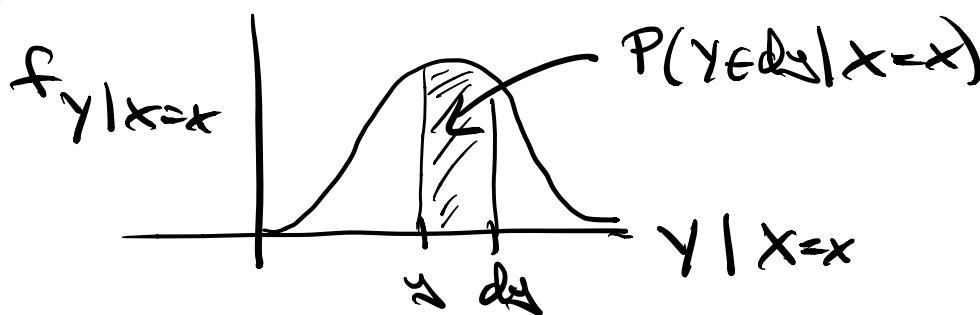
constant (not a function of y).

Picture (Courtesy of John Michael Laurel
from our class!)



$f(y)$ consists of a slice of $f(x,y)$
 $y|x=x$ through $X=x$.

I will draw the distribution of $y|x=x$ as



$$\text{Ex} \quad f(x,y) = K(y-x)^8, \quad 0 < x < y < 1$$

a) find K

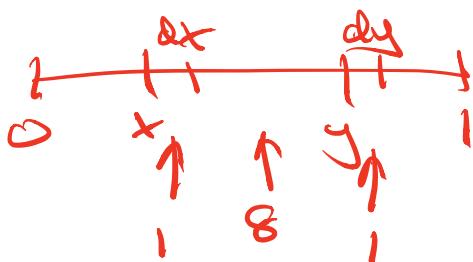
b) find the marginal distribution of x.

Soln a :

Recognize $f(x,y)$ as joint density of two order statistics,

$$X = U_{(1)}$$

$$Y = U_{(10)}$$



$$\Rightarrow f(x,y) = \binom{10}{1,8,1} (y-x)^8$$

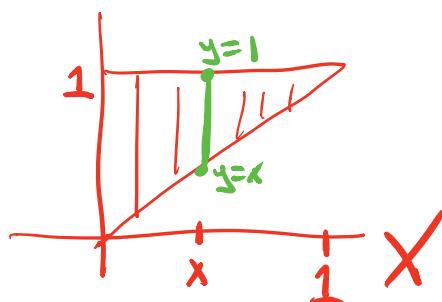
$$\Rightarrow K = \frac{10!}{1!8!1!} = 90.$$

Soln b : method 1 (integrate out y)

$$f_X(x) = \int_{y=x}^{y=1} f(x,y) dy$$

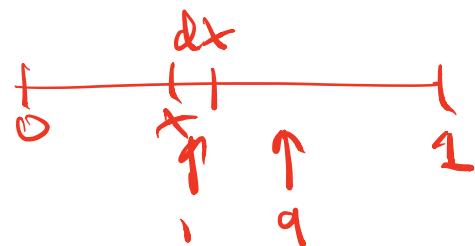
$$= \int_{y=x}^{y=1} 90(y-x)^8 dy \quad \begin{aligned} \text{let } u &= y-x \\ du &= dy \end{aligned}$$

$$= \int_{u=0}^{u=1-x} 90u^8 du = \frac{90u^9}{9} \Big|_0^{1-x} = \boxed{10(1-x)^9, \quad 0 < x < 1}$$



(or) method 2 (Find $P(X \in dx)$)

$$P(X \in dx) = f_X(x) dx$$

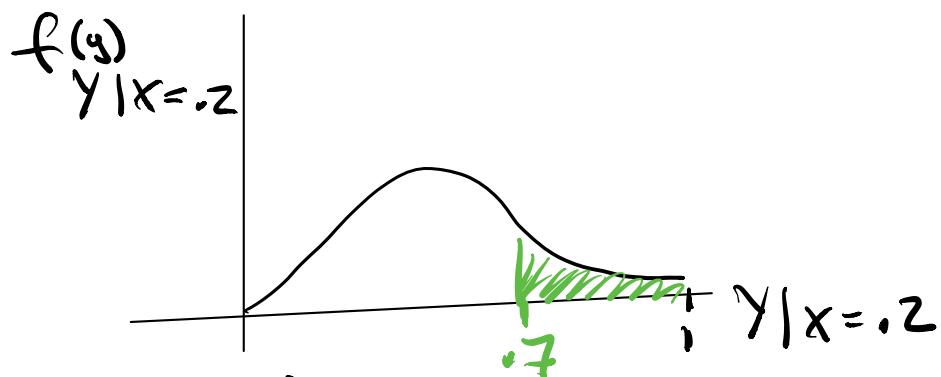


$$\Rightarrow f_X(x) = \binom{10}{1,9} (1-x)^9 \\ = \boxed{\frac{10}{10} (1-x)^9, \quad 0 < x < 1}$$

c) Find $P(Y > .7 | X=.2)$

Soln

Method 1 (integrate the conditional density)



$$P(Y > .7 | X=.2) = \int_{y=.7}^{y=1} f_{Y|X=.2}(y) dy$$

$$f_{Y|X=.2}(y) = \frac{f(.2, y)}{f(.2)} = \frac{90(y-.2)^8}{10(1-.2)^9} \Big|_{x=.2} \\ = \frac{90(y-.2)^8}{10(.8)^9}$$

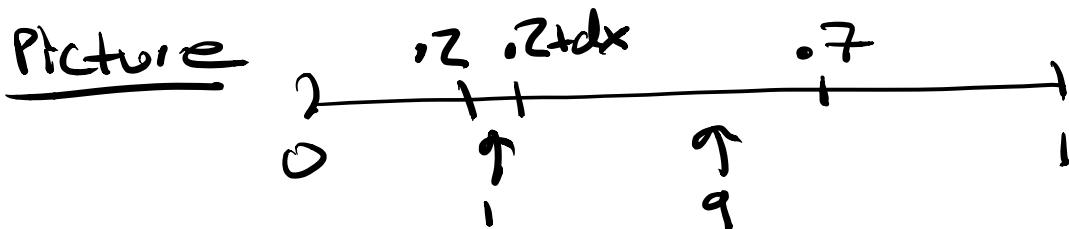
$$\begin{aligned}
 P(Y > .7 | X = .2) &= \frac{9}{(.8)^9} \int_{y=.7}^{y=1} (y - .2)^8 dy \\
 &= \frac{9}{(.8)^9} \int_{U=.5}^{U=.8} U^8 dU = \frac{9}{(.8)^9} \frac{U^9}{9} \Big|_{.5}^{.8} \\
 &= \frac{9}{(.8)^9} \left(\frac{.8^9}{9} - \frac{.5^9}{9} \right) \\
 &= \boxed{1 - \left(\frac{.5}{.8} \right)^9}
 \end{aligned}$$

Method 2 (use fact that $X = U_{(1)}$ and $Y = U_{(10)}$)

$$P(Y > .7 | X = .2) = 1 - P(Y < .7 | X = .2)$$

By Bayes' rule,

$$P(Y < .7 | X = .2) = \frac{P(Y < .7, X = .2)}{P(X = .2)}$$



$P(Y < .7, X = .2)$ is the chance that the remaining 9 darts land between .2 and .7.

$$\text{this is } (.7 - .2)^9 = (.5)^9$$

$P(X = .2)$ is the chance that the remaining 9 darts land between .2 and 1. This is $(1 - .2)^9 = (.8)^9$

Hence,

$$P(Y > .7 | X = .2) = \boxed{1 - \left(\frac{.5}{.8} \right)^9}$$

b) Multiplication rule

discrete case :

$$P(X=x, Y=y) = P(Y=y|X=x) \cdot P(X=x)$$

continuous case :

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

$$f_{x,y} = f_{Y|X=x} f_X(x)$$

Ex $X \sim \text{Gamma}(2, \lambda)$

given $X=x$, $Y \sim U(0, x)$ (i.e. $Y|X=x \sim U(0, x)$)
Find $f(x, y)$

Soln

$$f_{x,y} = f_{Y|X=x} f_X(x)$$

where

$$f_{Y|X=x}(y) = \frac{1}{x}, \quad 0 < y < x < \infty$$

$$f_X(x) = x^2 e^{-\lambda x}$$

$$\Rightarrow f(x, y) = \frac{1}{x} x^2 e^{-\lambda x} = \boxed{\lambda^2 e^{-\lambda x}}$$

(c) The rule of averages conditional probabilities

discrete case (X discrete):

$$P(A) = \sum_{x \in X} P(A|X=x) P(X=x)$$

ex let A = event that 2nd card in deck is K of hearts.
 X = 1st card in deck

There are 52 possible values of X each with probability $\frac{1}{52}$.

average of all possible conditionals.

$$P(A) = P(A|X=K \text{ of hearts})P(X=K \text{ of hearts}) +$$

\vdots " " $\frac{1}{52}$

$$P(A|X=5 \text{ of spades})P(X=5 \text{ of spades}) + \dots$$

$$\frac{1}{51} \quad \frac{1}{52}$$

$$= 51 \left(\frac{1}{51} \cdot \frac{1}{52} \right) = \frac{1}{52}$$

continuous case (X continuous):

$$P(A) = \int_{x \in X} P(A|X=x) f_X(x) dx$$

integral conditioning formula

$\stackrel{\text{def}}{=} X \sim \text{Unif}(0,1)$

given $X = p$, let $I_1, I_2 \stackrel{\text{iid}}{\sim} \text{Ber}(p)$

Find the probability the first toss is heads
(i.e. $I_1 = 1$).

Soln

$A = \text{event } I_1 = 1$

$$P(A|X=p) = p$$

$$f_X(p) = \frac{1}{p=1}$$

$$P(A) = \int_{p=0}^1 P(A|X=p) f_X(p) dp$$

$$= \int_0^1 p dp = \left. \frac{p^2}{2} \right|_0^1 = \left(\frac{1}{2} \right)$$

So not knowing anything about the coin, the chance it will land heads is $\frac{1}{2}$. If $I_1 = 1$ then we will see that the chance $I_2 = 1$ is $\frac{2}{3}$.

$$(\text{i.e. } P(I_2 = 1 | I_1 = 1) = \frac{2}{3})$$

To see this we need to know about Bayesian statistics.

Bayesian Statistic

Posterior \propto likelihood \cdot prior.
 proportional.

$$\text{ex } X \sim \text{Unif}(0,1)$$

given $X = p$, let $I_1, I_2 \stackrel{\text{iid}}{\sim} \text{Ber}(p)$

likelihood : $P(I_1=1 | X=p) = p$ given that we have a p -coin, the chance it lands head is p .

Posterior : $f(X=p | I_1=1)$ ← computed below

Prior : $f_X(p) = 1$ ← $X \sim U(0,1)$ says p is equally likely to be any value $0 < p < 1$.

$$f(X=p | I_1=1) = \frac{P(I_1=1 | X=p) \cdot f_X(p)}{P(I_1=1)} \quad \begin{matrix} \leftarrow \text{likelihood} \\ \leftarrow \text{prior} \end{matrix}$$

$\leftarrow \text{constant}$

Since $f(X=p, I_1=1) = f(X=p | I_1=1) P(I_1=1)$

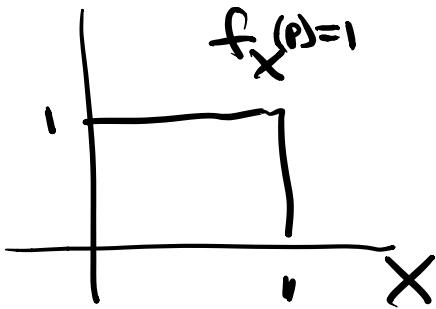
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$$f(I_1=1, X=p) = P(I_1=1 | X=p) f_X(p)$$

Hence, $f(X=p | I_1=1) = \frac{P \cdot 1}{\sum} = \boxed{2P}$

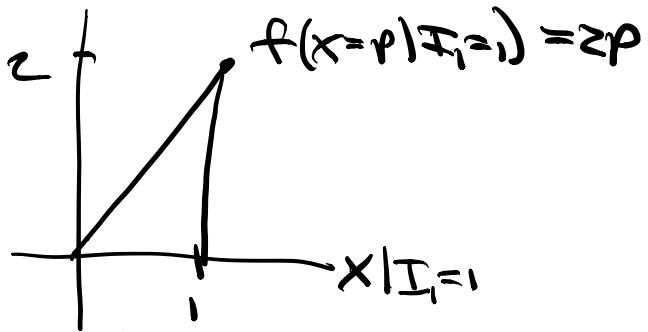
$\begin{matrix} \leftarrow \text{likelihood} \\ \leftarrow \text{prior} \\ \uparrow \text{posterior} \\ \uparrow \text{constant} \end{matrix}$

Prior



↑
no idea
What is
Probability
coin lands heads,

Posterior



↑ based on our
1st test this is
the updated
distribution of
the probability the
coin lands heads,

We will make this posterior our new prior
when we flip the coin a second time.

To be continued next time.