

Lecture 9

Conditional Distributions

Recall that Prob. of A _{event} given another event B is

$$\underbrace{P(A|B)}_{\text{Conditional probability}} = \frac{P(A \cap B)}{P(B)} = \frac{P(A, B)}{P(B)}$$

(CDF)
CONDITIONAL DISTRIBUTION of a random variable X given an event M

$$F_x(x|M) = P(X \leq x|M) = \frac{P(X \leq x, M)}{P(M)}$$

Conditional Density

$$f_x(x|M) = \frac{d F_x(x|M)}{dx}$$

Eg 1: We are interested in finding the Conditional distribution of a r.v. X given that $X \leq a$

$$F_X(x | X \leq a) = P(X \leq x | X \leq a)$$

$$= \frac{P(X \leq x, X \leq a)}{P(X \leq a)} \quad (*)$$

To further simplify (*), consider two cases



$$\Rightarrow \{X \leq x, X \leq a\}$$

$$= \{X \leq x\}$$

$$\Rightarrow \{X \leq x, X \leq a \leq x\}$$

$$= \{X \leq a\}$$

Hence,

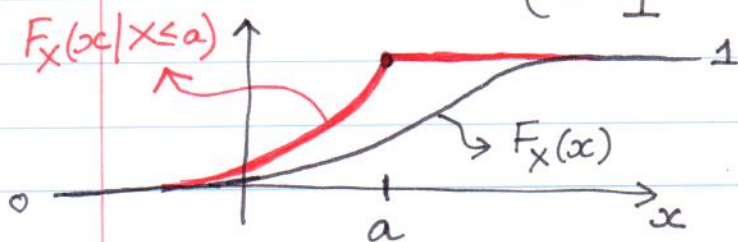
$$F_X(x | X \leq a) = \frac{P(X \leq x)}{P(X \leq a)}$$

$$\Rightarrow P(X \leq x, X \leq a) = P(X \leq a)$$

$$\Rightarrow F_X(x | X \leq a) = \frac{P(X \leq a)}{P(X \leq a)} = 1$$

$$= \frac{F_X(x)}{F_X(a)}$$

$$F_X(x | X \leq a) = \begin{cases} \frac{F_X(x)}{F_X(a)} & \text{if } x < a \\ 1 & \text{if } x \geq a \end{cases}$$



Eg. 2 $F_X(x | a \leq X \leq b) \stackrel{?}{=}$

$$F_X(x | a \leq X \leq b) = P(X \leq x | a \leq X \leq b)$$

$$= \frac{P(X \leq x, a \leq X \leq b)}{P(a \leq X \leq b)}$$

What is $P(a \leq X \leq b) \stackrel{?}{=} F_X(b) - F_X(a)$.

What about $P(X \leq x, a \leq X \leq b) \stackrel{?}{=} (*)$

Three
Cases

$$b \leq x$$

$$\Rightarrow (*) = P(a \leq X \leq b)$$

$$a \leq x < b$$

$$\begin{aligned} (*) &= P(a \leq X \leq x) \\ &= F_X(x) - F_X(a) \end{aligned}$$

$$(*) = 0$$

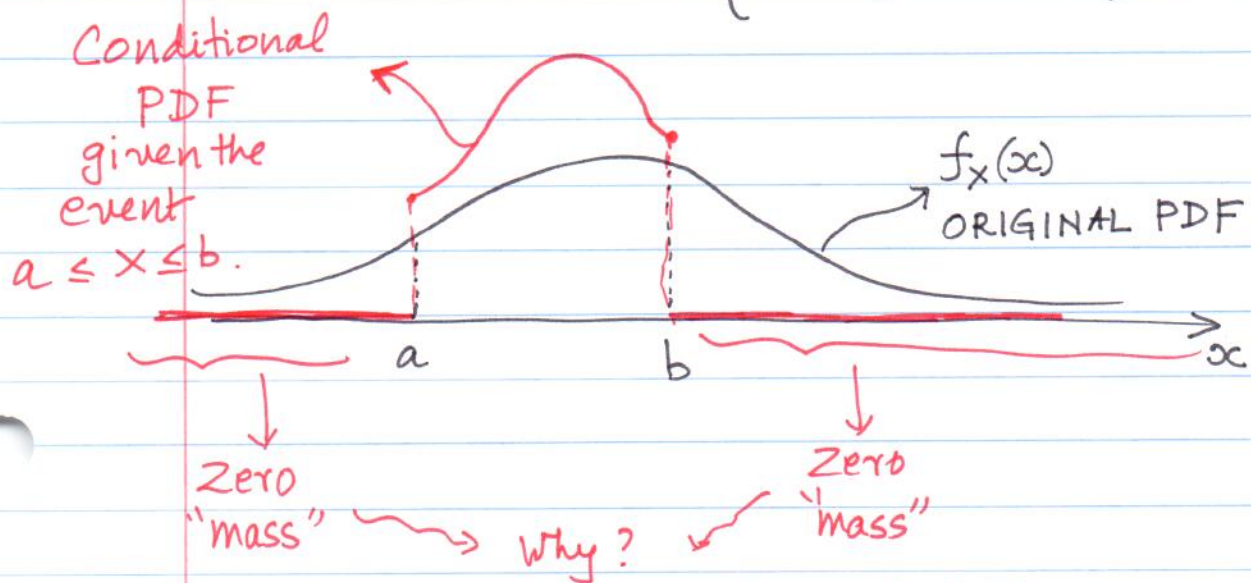
$$x < a$$

$$\Rightarrow F_X(x | a \leq X \leq b) = \begin{cases} 0 & x < a \\ \frac{F_X(x) - F_X(a)}{F_X(b) - F_X(a)} & a \leq x < b \\ 1 & b \leq x \end{cases}$$

(4)

What about the conditional PDF?

$$f_X(x | a \leq X \leq b) = \begin{cases} \frac{f_X(x)}{F_X(b) - F_X(a)} & a \leq x < b \\ 0 & \text{otherwise} \end{cases}$$



Since we are given that the event $a \leq X \leq b$ has occurred.

Note: Recall, in the past, we showed the (5)
"memoryless" property of exponential distribution

Eg 3: Memoryless Property of a Geometric R.v.

Geometric R.v. $P(X=k) = (1-p) \cdot p^{k-1}$ for $k=1, 2, 3, \dots$

Memoryless

Property: $P(X > m+n | X > m) = P(X > n)$

Check yourself....

Total Probability & Bayes' Theorem

$$P(X \leq x) = P(X \leq x | A_1) P(A_1) + \dots + P(X \leq x | A_n) P(A_n)$$

for A_1, A_2, \dots, A_n being a partition of Sample Space.

or

$$F_X(x) = F_X(x | A_1) \cdot P(A_1) + F_X(x | A_2) P(A_2) + \dots + F_X(x | A_n) \cdot P(A_n)$$

and, also

$$f_X(x) = f_X(x | A_1) P(A_1) + \dots + f_X(x | A_n) P(A_n).$$

(6)

event.
↑
~~~~~

$$P(\underbrace{A}_{\text{event}} | X \leq x) = \frac{P(A, X \leq x)}{P(X \leq x)}$$

$$= \frac{P(A) \cdot P(X \leq x | A)}{P(X \leq x)}$$

$$P(\underbrace{A}_{\text{Event}} | X \leq x) = \frac{F_X(x|A) \cdot P(A)}{F_X(x)}$$

Event.

What about

$$P(A | X = x) = ?$$

{

if  $X$  is  
continuous  
valued, then  
 $P(X=x)=0$

}

We define it as follows:

$$P(A | X = x) = \lim_{\Delta x \rightarrow 0} P(A | x < X \leq x + \Delta x)$$

Let's find  $P(A | x < X \leq x + \Delta x)$ 

$$= \frac{P(A, x < X \leq x + \Delta x)}{F_X(x + \Delta x) - F_X(x)}$$

$$P(A | \underbrace{x < X \leq x + \Delta x}_{\Delta x}) = P(A) \cdot \left\{ \frac{F_X(x + \Delta x | A) - F_X(x | A)}{F_X(x + \Delta x) - F_X(x)} \right\}$$

take lim  
 $\Delta x \rightarrow 0$

$$\Rightarrow \frac{f_X(x|A) \cdot P(A)}{f_X(x)}$$

or

$$P(A | X=x) = \frac{f_X(x|A) \cdot P(A)}{f_X(x)}$$

$$\Rightarrow \cancel{f_X(x|A)} \equiv \cancel{P(A)}$$

$$P(A) \cdot f_X(x|A) = P(A|X=x) f_X(x)$$

$$\Rightarrow P(A) \cdot \underbrace{\int_{-\infty}^{\infty} f_X(x|A) dx}_{=1} = \int_{-\infty}^{\infty} P(A|X=x) f_X(x) dx$$

$$\Rightarrow P(A) = \int_{-\infty}^{\infty} P(A|X=x) f_X(x) dx$$

CONTINUOUS VERSION of TOTAL  
PROBABILITY THEOREM.

from here,

CONTINUOUS VERSION  
OF BAYES' THEOREM

$$f_X(x|A) = \left( \frac{P(A|X=x)}{P(A)} \right) f_X(x)$$

$$f_X(x|A) = \left( \frac{P(A|X=x)}{\int_{-\infty}^{\infty} P(A|X=x) f_X(x) dx} \right) \cdot f_X(x)$$



## Applications of Conditional distribution (8)

Eg 1 Prob. of heads in a coin-tossing Experiment is a r.v.  $P$  with density  $f_P(p)$ . What is Prob. of Heads?

$$\text{Prob}\{H \mid P=p\} = p$$

$$\begin{aligned}\Rightarrow \text{Prob}\{H\} &= \int_{p=0}^1 \text{Prob}\{H \mid P=p\} f_P(p) dp \\ &= \int_0^1 p f_P(p) dp.\end{aligned}$$

Eg 2 Let  $P(H) = p$  represent the prob. of obtaining a head in a toss.

- $p$  can take any value in  $(0,1)$
- we can treat  $P$  as a random variable.
- Without any information about the coin, we assume the a priori pdf to be uniform. of  $P$

$$\text{i.e. } f_P(p) \sim \text{uniform}(0,1)$$

- We conduct an experiment → toss <sup>this</sup> coin ~~a~~  $n$  times & observe  $k$  heads

Q → How should we update  $f_P(p)$  ??



Let  $A = "k \text{ heads in } n \text{ tosses}"$  (ie the <sup>⑨</sup> Event that occurred)  
 for a specific value of  $P=p$ ,

$$P(A | P=p) = p^k (1-p)^{n-k}$$

$$\Rightarrow P(A) = \int_0^1 P(A | P=p) \cdot f_p(p) dp$$

Google:  
Euler's Integral

WHY??

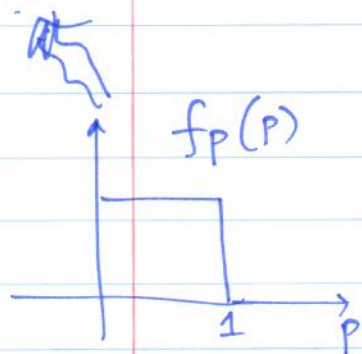
or  
Beta Function

$$= \int_0^1 p^k (1-p)^{n-k} dp = \frac{(n-k)! k!}{(n+1)!}$$

We are interested in a-posteriori pdf...

ie  $f_p(p | A) = \frac{P(A | P=p) \cdot f_p(p)}{P(A)}$

a-priori

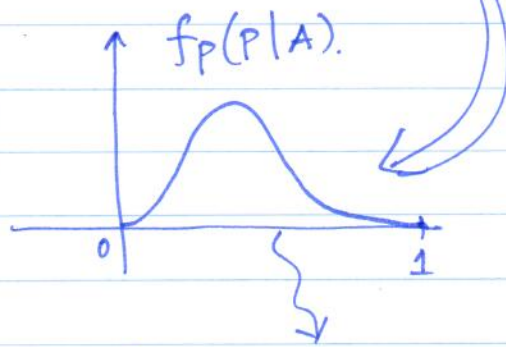


"event A occurred"

or

"after observing some data"

$$= \frac{(n+1)!}{(n-k)! k!} p^k (1-p)^{n-k}, \quad 0 < p \leq 1$$



a posteriori

We can use the a posteriori PDF to predict future outcomes.

Eg →  $B =$  "head occurring in  $(n+1)^{th}$  toss given that  $k$  heads occurred in first  $n$  tosses"  $\mathcal{B}$

$$P(B) = \int_0^1 \underbrace{P(B|P=p)}_{\downarrow P} \cdot \underbrace{f_p(p|A)}_{\downarrow \frac{(n+1)!}{(n-k)! k!} p^k (1-p)^{n-k}} dp$$

$$P(B) = \frac{k+1}{n+2}$$

Eg  $n=10, k=6$

$$P(B) = 7/12 = 0.58 \rightsquigarrow \text{Prediction using a-posteriori PDF}$$

vs

$$0.5 \rightsquigarrow \text{Prediction via (or non-informative) a-priori PDF}$$