

* midterm 1 : Tue 10/18

lecture to follow

2.6 conditioning

Given $X: \Omega \rightarrow \mathbb{R}$ a r.v.,
the PMF was

$$P_X(x) = P(X=x)$$

we can do this with any Prob. law,
not just $P(\cdot)$. Do this for
 $P(\cdot | A)$.

Defn

Let $X: \Omega \rightarrow \mathbb{R}$ be an r.v.,

$A \subseteq \Omega$ with $P(A) > 0$.

The conditional PMF of X
given A is

$$P_{X|A}(x) = P(X=x | A)$$

note

$$P_{X|A}(x) = \frac{P(X=x, A)}{P(A)}$$

note the events

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$$\{X=x\} \cap A \quad \text{for } x \in \text{range}(X)$$

are disjoint, with union A .

$$\therefore P(A) = \sum_x P(\{X=x\} \cap A)$$

so

$$\sum_x P_{X|A}(x) = \frac{\sum_x P(\{X=x\} \cap A)}{P(A)}$$

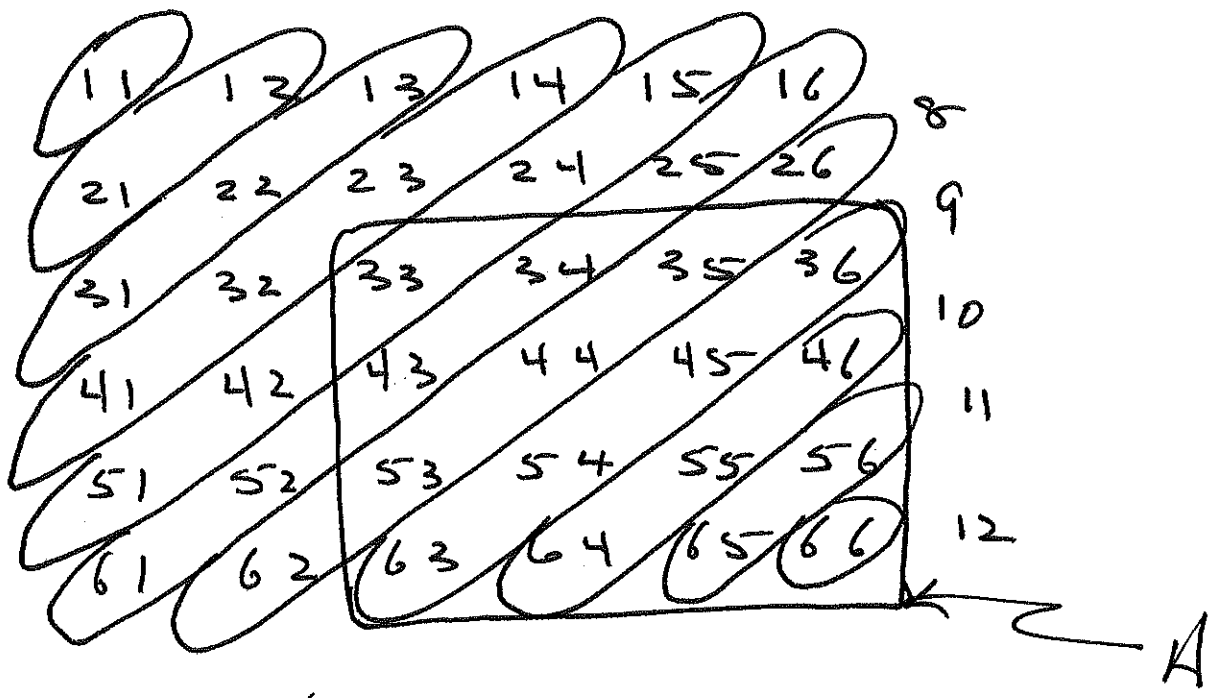
$$= \frac{P(A)}{P(A)} = 1,$$

as expected (and required) of
a PMF.

Ex.

Roll 2 independent, fair 6-sided dice,
and let X be sum of top faces.

$X =$ 2 3 4 5 6 7



$$P_X(k) = \begin{cases} 1/36 & \text{if } k = 2, 12 \\ 2/36 & k = 3, 11 \\ 3/36 & k = 4, 10 \\ 4/36 & k = 5, 9 \\ 5/36 & k = 6, 8 \\ 6/36 & k = 7 \\ 0 & \text{otherwise} \end{cases}$$

$$\text{Let } A = \{ \min \geq 3 \}. P(A) = \frac{16}{36}$$

$$P_{X|A}(k) = \begin{cases} 1/16 & k = 6, 12 \\ 2/16 & k = 7, 11 \\ 3/16 & k = 8, 10 \\ 4/16 & k = 9 \\ 0 & \text{otherwise} \end{cases}$$

Ex.

Flip a weighted coin, $P(H) = p$, until 1st head. Let $X = \# \text{ flips}$. Then X is geometric with parameter p .

$$P_X(k) = \begin{cases} (1-p)^{k-1} p & \text{if } k = 1, 2, 3, \dots \\ 0 & \text{otherwise} \end{cases}$$

Let A be the event that the 1st head occurs with n flips.

$$P(A) = \sum_{k=1}^n P_X(k)$$

$$= \sum_{k=1}^n (1-p)^{k-1} \cdot p$$

$$= p \cdot \sum_{k=0}^{n-1} (1-p)^k$$

$$= \cancel{p} \cdot \frac{1 - (1-p)^n}{\cancel{1 - (1-p)}} = 1 - (1-p)^n$$

Another way:

$$P(A) = 1 - P(A^c)$$

$$= 1 - (1-p)^n$$

Thus

$$P_{X|A}^{(k)} = \begin{cases} \frac{(1-p)^{k-1} \cdot p}{1-(1-p)^n} & k = 1, 2, \dots, n \\ 0 & \text{otherwise} \end{cases}$$

Now let Y be another r.v. such that

$$P_Y(y) = P(Y=y) > 0$$

for all $y \in \text{range}(Y)$, we can now condition on events of form

$$A = \{Y=y\}$$

Defn

The conditional PMF of X given Y is

$$P_{X|Y}(x|y) = P(X=x | Y=y) \\ = \frac{P(X=x, Y=y)}{P(Y=y)}$$

Think of y as a fixed constant and x as variable.

note:

$$P_{X|Y}(x|y) = \frac{P_{X,Y}(x,y)}{P_Y(y)}$$

Thus

$$P_{X,Y}(x,y) = P_Y(y) \cdot P_{X|Y}(x|y)$$

Also

$$P_{X,Y}(x,y) = P_X(x) \cdot P_{Y|X}(y|x)$$

Ex.

A professor answers questions wrongly with prob. $\frac{1}{4}$, independent of other questions. The # of questions asked is 0, 1, 2 with eq. Prob. $\frac{1}{3}$.

let

$X = \# \text{ questions asked}$

and

$Y = \# \text{ questions answered wrongly.}$

find $P_{X,Y}(x,y)$. we have

$$P_X(x) = \begin{cases} \frac{1}{3} & x = 0, 1, 2 \\ 0 & \text{otherwise} \end{cases}$$

$$P_{Y|X}(y|x) = \begin{cases} 1 & x=0, y=0 \\ \begin{bmatrix} 3/4 \\ 1/4 \end{bmatrix} & x=1, y=0 \\ \begin{bmatrix} 9/16 \\ 3/8 \\ 1/16 \end{bmatrix} & x=2, y=0 \\ \begin{bmatrix} 3/8 \\ 1/16 \end{bmatrix} & x=2, y=1 \\ 0 & x=2, y=2 \\ 0 & \text{otherwise} \end{cases}$$

Therefore

$$P_{X,Y}(x,y) = \begin{cases} 1/3 & x=0, y=0 \\ 1/4 & 1 \quad 0 \\ 1/12 & 1 \quad 1 \\ 3/16 & 2 \quad 0 \\ 1/8 & 2 \quad 1 \\ 1/48 & 2 \quad 2 \\ 0 & \text{otherwise} \end{cases}$$

find marginal $P_Y(y)$.

$$P_Y(y) = \sum_x P_{X,Y}(x,y)$$

$$= \sum_x P_X(x) \cdot P_{Y|X}(y|x)$$

Thus

$$P_Y(y) = \begin{cases} 37/48 & y=0 \\ 10/48 & y=1 \\ 1/48 & y=2 \\ 0 & \text{otherwise} \end{cases}$$

Exercise find $P_{X|Y}(x|y)$

conditional Expectation

let X be a r.v., $A \subseteq \Omega$ with $P(A) > 0$. Defn:

$$E[X|A] = \sum_x x \cdot P_{X|A}(x)$$

also

$$E[g(X)|A] = \sum_x g(x) \cdot P_{X|A}(x)$$

we can condition on $A = \{Y=y\}$

provided $P(Y=y) > 0$.

$$E[X|Y=y] = \sum_x x \cdot P_{X|Y}(x|y)$$

Total expectation Theorem

Given a partition $A_1, A_2, \dots, A_n$ of Ω , then

$$E[X] = \sum_{i=1}^n P(A_i) \cdot E[X|A_i]$$

Proof

by total Prob. thm:

$$P_X(x) = \sum_{i=1}^n P(A_i) \cdot P_{X|A_i}(x)$$

mult. by x , sum over $x \in \text{range}(X)$

$$E[X] = \sum_x x \cdot P_X(x)$$

$$= \sum_x x \cdot \sum_{i=1}^n P(A_i) \cdot P_{X|A_i}(x)$$

$$= \sum_{i=1}^n P(A_i) \cdot \sum_x x \cdot P_{X|A_i}(x)$$

$$= \sum_{i=1}^n P(A_i) \cdot E[X|A_i]$$



Mean and Variance of Geometric

recall geom. With Param. p

$$P_X(k) = (1-p)^{k-1} \cdot p \quad (k=1,2,\dots)$$