

Tutorial/Worksheet on “M0a. Probability Background”

CS5691 PRML Jul–Nov 2025

August 9, 2025

1. [PROBABILITY BASICS] Do all the tutorial/exercise problems in the slides on “M0a. Probability Background”. The specific slides in this slide deck include:

(a) Slides 10-11 on conditional probability and Bayes’ theorem examples.

Solution:

Slide 10:

$[Pr(Y = 1|X = 9), Pr(Y = 2|X = 9)]^T \propto [0 \ 2]^T$. Therefore,
 $[Pr(Y = 1|X = 9), Pr(Y = 2|X = 9)]^T = [\frac{0}{2} \ \frac{2}{2}]^T = [0 \ 1]^T$.

We can calculate $Pr(Y|X = 5)$ as above, or equivalently as:

$Pr(Y = 1|X = 5) = \frac{Pr(Y=1, X=5)}{Pr(X=5)} = \frac{(5/N)}{(9/N)} = \frac{5}{9}$, and

$Pr(Y = 2|X = 5) = 1 - Pr(Y = 1|X = 5) = \frac{4}{9}$.

Slide 11:

Handwritten solution for Slide 11:

Joint probability table for X and Y:

	X			
Y	R	G	B	
S	20	10	15	45
C	10	20	25	55
	30	30	40	

(a) $P(X)$:

$$P(X=R) = \frac{30}{100} = 0.3$$

$$P(X=G) = \frac{30}{100} = 0.3$$

$$P(X=B) = \frac{40}{100} = 0.4$$

$P(Y)$:

$$P(Y=S) = \frac{45}{100} = 0.45$$

$$P(Y=C) = \frac{55}{100} = 0.55$$

(b) $P(X/Y=C)$:

$$P(X=R/Y=C) = \frac{10}{55} = \frac{2}{11}$$

$$P(X=G/Y=C) = \frac{20}{55} = \frac{4}{11}$$

$$P(X=B/Y=C) = \frac{25}{55} = \frac{5}{11}$$

$P(Y/X=B)$:

$$P(Y=S/X=B) = \frac{15}{40} = \frac{3}{8}$$

$$P(Y=C/X=B) = \frac{25}{40} = \frac{5}{8}$$

(c) $P(Y=C/X=B)$:

$$P(Y=C/X=B) = \frac{P(X=B/Y=C) \cdot P(Y=C)}{P(X=B)}$$

$$= \frac{\frac{5}{11} \cdot \frac{55}{100}}{\frac{40}{100}}$$

$$= \frac{5}{8}$$

[SOURCE: From [CC]Notes (by Prof. Chandra)]

- (b) Slide 19 on conditional distribution and conditional expectation questions.

Solution: Discussed extensively in class. So explicit solutions not provided here.

- (c) Slides 32 and 35 on questions related to pmf (probability mass function) and pdf (probability density function) respectively.

Solution:

Slide 32:

(i) If $X \sim \text{Bernoulli}(\theta)$, then $E[X] = \sum_{x=0,1} x \Pr(X=x) = 0 \Pr(X=0) + 1 \Pr(X=1) = \Pr(X=1) = \theta$.

(ii) A Binomial-distributed rv can be expressed as a sum of n iid (independent and identically distributed) Bernoulli rvs as follows. For $i \in \{1, 2, \dots, n\}$, let $X_i \sim \text{Bernoulli}(p)$. Then $X = \sum_{i=1}^n X_i$ follows a Binomial(n, p) distribution. So,
 $E[X] = E[\sum_{i=1}^n X_i] = \sum_{i=1}^n E[X_i] = \sum_{i=1}^n p = np$.

We have used linearity of expectation in the second step, and expectation of a Bernoulli rv from above in the third step.

(iii) Answer is $1/\theta$. You can compute mean of Geometric rv X either directly using the expectation formula, or using the conditional expectation property. For the latter, let $Y = 1$ if first toss is Heads, and 0 otherwise (ow). Then, use the Law of Total Expectation, i.e., compute $E[E[X|Y]]$ to obtain $E[X]$. Specifically,
 $E[X] = E[E[X|Y]] = \theta E[X|Y=1] + (1-\theta) E[X|Y=0] = \theta \cdot 1 + (1-\theta) \cdot (1 + E[X])$
 $\Rightarrow E[X] = 1 + (1-\theta)E[X] \Rightarrow E[X] = 1/\theta$.

Slide 35: Mean and variance of uniform distribution can be derived from the formula directly.

- (d) Slide 43 on covariance between two rvs (random variables) X, Y whose joint distribution is given.

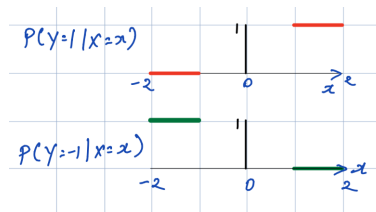
Solution: In this example, $X = 3 - Y$, and so they are negatively related and the covariance is negative also then.

2. [DERIVING POSTERIORIS]

- (a) Consider a continuous random variable X and a discrete random variable Y . Let

- $P_Y(Y=1) = 0.5$ and $P_Y(Y=-1) = 0.5$, and
- $(X|Y=1) \sim \text{Unif}(1, 2)$ and $(X|Y=-1) \sim \text{Unif}(-2, -1)$.

Draw the plots for $P(Y=1|X=x)$ and $P(Y=-1|X=x)$ given the above assumptions.

Solution:

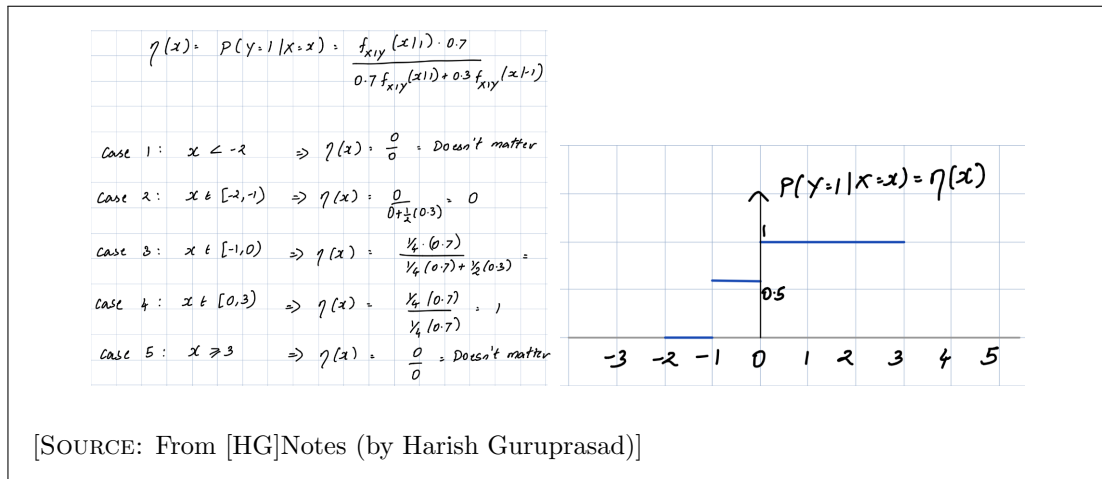
[SOURCE: From [HG]Notes (by Harish Guruprasad)]

- (b) Consider the following setting:

- $P_Y(Y=1) = 0.7$ and $P_Y(Y=-1) = 0.3$
- $(X|Y=1) \sim \text{Unif}(-1, 3)$ and $(X|Y=-1) \sim \text{Unif}(-2, 0)$

1. Derive $P(Y=1|X=x)$ for different possible values of x .
2. Plot the derived $P(Y=1|X=x)$.

Solution:



3. [BAYES' THEOREM ON EVENTS] It is estimated that 50% of emails are spam emails. Some software has been applied to filter these spam emails before they reach your inbox. A certain brand of software claims that it can detect 99% of spam emails, and the probability for a false positive (a non-spam email detected as spam) is 5%. Now if an email is detected as spam, then what is the probability that it is in fact a non-spam email?

Solution:

Define events

A = event that an email is detected as spam,
 B = event that an email is spam,
 B^c = event that an email is not spam.

We know $P(B) = P(B^c) = .5$, $P(A|B) = 0.99$, $P(A|B^c) = 0.05$.

Hence by the Bayes's formula we have

$$\begin{aligned}
 P(B^c|A) &= \frac{P(A|B^c)P(B^c)}{P(A|B)P(B) + P(A|B^c)P(B^c)} \\
 &= \frac{0.05 \times 0.5}{0.05 \times 0.5 + 0.99 \times 0.5} \\
 &= \frac{5}{104}
 \end{aligned}$$

[SOURCE: <https://www.studocu.com/row/document/united-international-university/data-structure-and-algorithm-ii-lab/ai-final-spring-2021/31515566>]

4. [RANDOM VARIABLE BASICS] A discrete random variable X is said to follow/have a Poisson distribution with parameter $\lambda > 0$ over the support $S = \{0, 1, 2, \dots\}$ if it has the following pmf:

$$P(X = x) = \frac{\lambda^x e^{-\lambda}}{x!}$$

In this question, you are required to verify if the above pmf of X is indeed a valid pmf by verifying the following properties:

1. $P(X = x) \geq 0 \quad \forall x \in S$, and
2. $\sum_{x \in S} P(X = x) = 1$.

In addition, derive $\mathbb{E}(X)$ and $Var(X)$.

[Hint: Use the Maclaurin's series of exponential function: $e^y = \sum_{n=0}^{\infty} \frac{y^n}{n!}$]

Solution:

Verifying that for probability mass function, $P(X = x) = \frac{\lambda^x e^{-\lambda}}{x!}$, follow the following properties:

- 1) $P(X = x) \geq 0 \quad \forall x \in S$

This statement says that for every element x in the support S , all the probabilities must be positive.

Proof:

Given parameter $\lambda > 0$

$\Rightarrow \lambda^x > 0$ As any power of positive number is positive

As $x \in S$ and $S = \{0, 1, 2, \dots\}$ So, $x \geq 0$

$\Rightarrow x! \geq 0$

As we know that any e is a constant with a positive value 2.71828.

$\Rightarrow e^{-\lambda} > 0$ As any power of positive number is positive

$\Rightarrow P(X = x) = \frac{\lambda^x e^{-\lambda}}{x!} \geq 0$ As multiplication and division of 2 positive numbers is positive

Hence, $P(X = x) \geq 0 \quad \forall x \in S$

$$2) \sum_{x \in S} P(X = x) = 1$$

This statement says that if we add up all the probabilities for all the possible values of x , in the support S , then that sum equals 1.

Proof:

Given $P(X = x) = \frac{\lambda^x e^{-\lambda}}{x!}$

$$\Rightarrow \sum_{x=0}^{\infty} P(X = x) = \sum_{x=0}^{\infty} \frac{\lambda^x e^{-\lambda}}{x!}$$

$$\Rightarrow e^{-\lambda} \sum_{x=0}^{\infty} \frac{\lambda^x}{x!}$$

Since We know that, $\sum_{x=0}^{\infty} \frac{\lambda^x}{x!} = e^{\lambda}$

$$\Rightarrow \sum_{x=0}^{\infty} P(X = x) = e^{-\lambda} e^{\lambda}$$

$$\Rightarrow \sum_{x=0}^{\infty} P(X = x) = 1$$

Hence proved.

Calculating Expectation,

$$E(X) = \sum_{x=0}^{\infty} x \frac{\lambda^x e^{-\lambda}}{x!}$$

$$= \sum_{x=1}^{\infty} x \frac{\lambda^x e^{-\lambda}}{x!}$$

$$= \lambda e^{-\lambda} \sum_{x=1}^{\infty} x \frac{\lambda^{x-1}}{x!}$$

$$= \lambda e^{-\lambda} \sum_{x-1=0}^{\infty} \frac{\lambda^{x-1}}{(x-1)!}$$

$$\text{As, } \sum_{n=0}^{\infty} \frac{y^n}{n!} = e^y$$

$$\text{Thus, } E(X) = \lambda e^{-\lambda} e^{\lambda} = \lambda$$

Calculating variance

$$\text{Var}(X) = \sigma^2 = E(X^2) - (E(X))^2$$

$$E(X^2) = \sum_{x=1}^{\infty} x^2 \frac{\lambda^x e^{-\lambda}}{x!}$$

$$\begin{aligned}
&= \lambda e^{-\lambda} \sum_{x=1}^{\infty} x^2 \frac{\lambda^{x-1}}{x!} \\
&= \lambda e^{-\lambda} \sum_{x=1}^{\infty} x \frac{\lambda^{x-1}}{(x-1)!} \\
&= \lambda e^{-\lambda} \left(\sum_{x=1}^{\infty} (x-1) \frac{\lambda^{x-1}}{(x-1)!} + \sum_{x=1}^{\infty} \frac{\lambda^{x-1}}{(x-1)!} \right) \\
&= \lambda e^{-\lambda} \left(\sum_{x=2}^{\infty} \frac{\lambda^{x-1}}{(x-2)!} + \sum_{x=1}^{\infty} \frac{\lambda^{x-1}}{(x-1)!} \right) \\
&= \lambda e^{-\lambda} \left(\lambda \sum_{x=2}^{\infty} \frac{\lambda^{x-2}}{(x-2)!} + \sum_{x=1}^{\infty} \frac{\lambda^{x-1}}{(x-1)!} \right)
\end{aligned}$$

Let $y = x - 1, z = x - 2$.

$$\text{then, } \implies \lambda e^{-\lambda} \left(\lambda \sum_{z=0}^{\infty} \frac{\lambda^z}{z!} + \sum_{y=0}^{\infty} \frac{\lambda^y}{y!} \right)$$

Using the Maclaurin series of exponential function, we've:

$$E[X^2] = \lambda e^{-\lambda} (\lambda e^{\lambda} + e^{\lambda})$$

$$= \lambda e^{-\lambda} e^{\lambda} (\lambda + 1) = \lambda^2 + \lambda$$

$$\text{Therefore, } \text{Var}(X) = E(X^2) - (E(X))^2 = (\lambda^2 + \lambda) - \lambda^2 = \lambda.$$

5. [RV BASICS CONTD.] A fair coin is tossed twice, and \mathbf{X} is defined as the number of heads that are observed. Find the range of \mathbf{X} (R_X) and the probability mass function (P_X).

Solution:

Here, our sample space is given by

$$S = \{HH, HT, TH, TT\}.$$

The number of heads will be 0, 1 or 2. Thus

$$R_X = \{0, 1, 2\}.$$

Since this is a finite (and thus a countable) set, the random variable X is a discrete random variable. Next, we need to find PMF of X . The PMF is defined as

$$P_X(k) = P(X = k) \text{ for } k = 0, 1, 2.$$

We have

$$P_X(0) = P(X = 0) = P(TT) = \frac{1}{4},$$

$$P_X(1) = P(X = 1) = P(\{HT, TH\}) = \frac{1}{4} + \frac{1}{4} = \frac{1}{2},$$

$$P_X(2) = P(X = 2) = P(HH) = \frac{1}{4}.$$

Figure 1:

[SOURCE: https://www.probabilitycourse.com/chapter3/3_1_3_pmf.php]

6. [CHANGE OF VARIABLES VS. THE LAW OF THE UNCONSCIOUS STATISTICIAN] Consider a continuous random variable X and a discrete random variable Y . Let

- $P_Y(Y = 1) = 0.5$ and $P_Y(Y = -1) = 0.5$, and
- $(X|Y = 1) \sim \text{Unif}(1, 2)$ and $(X|Y = -1) \sim \text{Unif}(-2, -1)$.

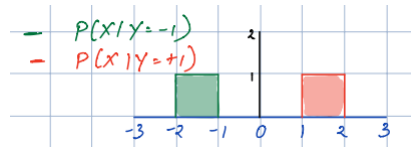
- What is the marginal distribution of X ? Specifically, plot the pdf of X denoted $f_X(x)$.
- Write down the pdf $f_X(x)$ of X .
- Let $Z = X^2$. What is the pdf of Z ? Use it to compute $E[Z]$. (Hint: To obtain pdf of Z , you could simply derive the cdf of Z and differentiate it (or you could also use the change-of-variables formula).)

- d. Now, use the pdf of X directly to compute $E[X^2]$ (using the law of the unconscious statistician). Does this give the same answer as the previous question? Which of these two methods do you prefer to compute the expectation of a function of a rv?

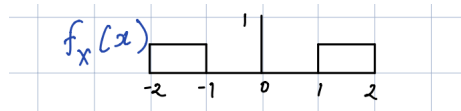
Solution:

a. Solution

Class conditionals $f_{X|Y}(x)$:



Marginal $f_X(x)$:



b. See Figure b.

$$f_X(x) = \begin{cases} 1/2 & \text{if } x \in [-2, -1] \\ 0 & \text{o.w.} \\ 1/2 & \text{if } x \in [1, 2] \end{cases}$$

[SOURCE: Parts a,b from [HG]Notes]

- c. Follow hint to obtain $f_Z(z)$; then use standard expectation formula, $E[Z] = \int_{-\infty}^{\infty} z f_Z(z) dz$. The cdf of $Z = X^2$ is given by:

$$\begin{aligned} F_Z(z) &= P(Z \leq z) \\ &= P(X^2 \leq z) \\ &= P(X \in [-\sqrt{z}, +\sqrt{z}]) \\ &= 2 P(X \in [0, \sqrt{z}]) \\ &= \begin{cases} 2 \frac{\sqrt{z}-1}{2} & \text{if } \sqrt{z} \in [1, 2] \\ 2 \frac{1}{2} & \text{if } \sqrt{z} > 2 \\ 0 & \text{o.w.} \end{cases} \\ &= \begin{cases} \sqrt{z} - 1 & \text{if } z \in [1, 4] \\ 1 & \text{if } z > 4 \\ 0 & \text{o.w.} \end{cases} \end{aligned}$$

Differentiating the above cdf gives:

$$f_Z(z) = F'_Z(z) = \begin{cases} \frac{1}{2\sqrt{z}} & \text{if } z \in [1, 4] \\ 0 & \text{o.w.} \end{cases}$$

Therefore, $E[Z] = \int_{z=1}^4 z \frac{1}{2\sqrt{z}} dz = \left[\frac{z^{3/2}}{3} \right]_{z=1}^{z=4} = (8 - 1)/3 \approx 2.33$.

Another way of deriving $f_Z(z)$ using change-of-variables formula: Let $Z = g(X) = X^2$. Assume $z \geq 0$ (otherwise, pdf is zero). Then, $dz = 2x dx \implies |dx/dz| = |1/(2x)| = 1/(2\sqrt{z})$. Since $g(\cdot)$ is a two-to-one function, we use the following change-of-variables formula:

$$\begin{aligned} f_Z(z) &= f_X(x) \left| \frac{dx}{dz} \right| \text{ (at } x = -\sqrt{z}) + f_X(x) \left| \frac{dx}{dz} \right| \text{ (at } x = +\sqrt{z}) \\ &= f_X(\sqrt{z}) \frac{1}{2\sqrt{z}} + f_X(-\sqrt{z}) \frac{1}{2\sqrt{z}} \\ &= \begin{cases} 2 \left(\frac{1}{2} \frac{1}{2\sqrt{z}} \right) & \text{if } \sqrt{z} \in [1, 2] \\ 0 & \text{o.w.} \end{cases} \\ &= \begin{cases} \frac{1}{2\sqrt{z}} & \text{if } z \in [1, 4] \\ 0 & \text{o.w.} \end{cases} \end{aligned}$$

d. By law of the unconscious statistician,

$$\begin{aligned} E[X^2] &= \int_{x=-\infty}^{\infty} f_X(x) x^2 dx \\ &= \int_{-2}^{-1} (x^2/2) dx + \int_1^2 (x^2/2) dx \\ &= [x^3/6]_{-2}^{-1} + [x^3/6]_1^2 \\ &= (-1 + 8 + 8 - 1)/6 \\ &= (8 - 1)/3 \approx 2.33 \end{aligned}$$

The two answers match as expected. **Going forward, you can use the law of the unconscious statistician for all problems where expectation of a function of a rv is needed, as it greatly simplifies calculations as illustrated in this question.**