

# Homework 5

MSSC 6010- Computational Probability

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**Question 1.** (4.4.42) From book. The Laplace distribution, also known as a double exponential has a **pdf** given by

$$f(x) = \frac{\lambda}{2} e^{-\lambda|x-\mu|}, \text{ where } -\infty < x < \infty, -\infty < \mu < \infty, \lambda > 0$$

- (a) Find the theoretical mean and variance of a Laplace distribution. (Hint: Integrals of absolute values should be done as a positive and a negative part, in this case, with limits from  $-\infty$  to  $\mu$  and from  $\mu$  to  $\infty$ .)

First, let's just integrate  $f(x)$ , since we will also need it later.

$$\begin{aligned} \int_{-\infty}^{\infty} \frac{\lambda}{2} e^{-\lambda|x-\mu|} dx &= \frac{\lambda}{2} \left( \int_{-\infty}^{\mu} e^{-\lambda(\mu-x)} dx + \int_{\mu}^{\infty} e^{-\lambda(x-\mu)} dx \right) \\ &= \frac{\lambda}{2} \left( \frac{1}{\lambda} e^{-\lambda(\mu-x)} \Big|_{-\infty}^{\mu} + \frac{-1}{\lambda} e^{-\lambda(x-\mu)} \Big|_{\mu}^{\infty} \right) \\ &= \frac{\lambda}{2} \left( \frac{1}{\lambda} e^0 + \frac{1}{\lambda} e^0 \right) \\ &= 1 \end{aligned}$$

Now

$$\begin{aligned}
E[X] &= \int_{-\infty}^{\infty} x \frac{\lambda}{2} e^{-\lambda|x-\mu|} dx \text{ let } y = x - \mu \\
&= \frac{\lambda}{2} \left( \int_{-\infty}^{\infty} y e^{-\lambda|y|} dy + \mu \int_{-\infty}^{\infty} e^{-\lambda|y|} dy \right) \text{ from above, the second integral is 1} \\
&= \frac{\lambda}{2} \left( \int_{-\infty}^0 y e^{\lambda(y)} dy + \int_0^{\infty} y e^{-\lambda(y)} dy \right) + \mu \\
&= \frac{\lambda}{2} \left[ \frac{y e^{\lambda y}}{\lambda} \Big|_{-\infty}^0 - \int_{-\infty}^0 \frac{1}{\lambda} e^{\lambda y} dy + \frac{y e^{-\lambda y}}{-\lambda} \Big|_0^{\infty} - \int_0^{\infty} \frac{1}{-\lambda} e^{-\lambda y} dy \right] + \mu \\
&= \frac{\lambda}{2} \left[ - \int_{-\infty}^0 \frac{1}{\lambda} e^{\lambda y} dy + \int_0^{\infty} \frac{1}{\lambda} e^{-\lambda y} dy \right] + \mu \\
&= \frac{\lambda}{2} \left[ -\frac{1}{\lambda^2} e^0 + \frac{1}{\lambda^2} e^0 \right] + \mu \\
&= \mu
\end{aligned}$$

Now, Variance

$$\begin{aligned}
E[X^2] &= \int_{-\infty}^{\infty} x^2 \frac{\lambda}{2} e^{-\lambda|x-\mu|} dx \\
&= \int_{-\infty}^{\mu} x^2 \frac{\lambda}{2} e^{-\lambda(\mu-x)} dx + \int_{\mu}^{\infty} x^2 \frac{\lambda}{2} e^{-\lambda(x-\mu)} dx \\
&= \frac{\lambda}{2} \left[ x^2 \frac{1}{\lambda} e^{-\lambda(x-\mu)} \Big|_{-\infty}^{\mu} - \int_{-\infty}^{\mu} 2x \frac{1}{\lambda} e^{-\lambda(\mu-x)} dx + \int_{\mu}^{\infty} x^2 e^{-\lambda(x-\mu)} dx \right] \\
&= \frac{\lambda}{2} \left[ \frac{\mu^2}{\lambda} - \left( 2x \frac{1}{\lambda^2} e^{-\lambda(\mu-x)} \Big|_{-\infty}^{\mu} - \int_{-\infty}^{\mu} 2 \frac{1}{\lambda^2} e^{-\lambda(\mu-x)} dx \right) + \int_{\mu}^{\infty} x^2 e^{-\lambda(x-\mu)} dx \right] \\
&= \frac{\lambda}{2} \left[ \frac{\mu^2}{\lambda} - \left( 2\mu \frac{1}{\lambda^2} e^0 - \frac{2}{\lambda^3} e^{-\lambda(\mu-x)} \Big|_{-\infty}^{\mu} \right) + \int_{\mu}^{\infty} x^2 e^{-\lambda(x-\mu)} dx \right] \\
&= \frac{\lambda}{2} \left[ \frac{\mu^2}{\lambda} - \left( 2\mu \frac{1}{\lambda^2} - \frac{2}{\lambda^3} e^0 \right) + \int_{\mu}^{\infty} x^2 e^{-\lambda(x-\mu)} dx \right] \\
&= \frac{\mu^2}{2} - \frac{\mu}{\lambda} + \frac{1}{\lambda^2} + \frac{\lambda}{2} \int_{\mu}^{\infty} x^2 e^{-\lambda(x-\mu)} dx \\
&= \frac{\mu^2}{2} - \frac{\mu}{\lambda} + \frac{1}{\lambda^2} + \frac{\lambda}{2} \left[ x^2 \frac{1}{-\lambda} e^{-\lambda(x-\mu)} \Big|_{\mu}^{\infty} - \int_{\mu}^{\infty} 2x \frac{1}{-\lambda} e^{-\lambda(x-\mu)} dx \right] \\
&= \frac{\mu^2}{2} - \frac{\mu}{\lambda} + \frac{1}{\lambda^2} + \frac{\lambda}{2} \left[ -\mu^2 \frac{1}{-\lambda} e^0 - \left( 2x \frac{1}{\lambda^2} e^{-\lambda(x-\mu)} \Big|_{\mu}^{\infty} - \int_{\mu}^{\infty} 2 \frac{1}{\lambda^2} e^{-\lambda(x-\mu)} dx \right) \right] \\
&= \frac{\mu^2}{2} - \frac{\mu}{\lambda} + \frac{1}{\lambda^2} + \frac{\lambda}{2} \left[ \frac{\mu^2}{\lambda} - \left( \frac{-2\mu}{\lambda^2} e^0 - \left( \frac{2}{-\lambda^3} e^{-\lambda(x-\mu)} \Big|_{\mu}^{\infty} \right) \right) \right] \\
&= \frac{\mu^2}{2} - \frac{\mu}{\lambda} + \frac{1}{\lambda^2} + \frac{\lambda}{2} \left[ \frac{\mu^2}{\lambda} - \left( \frac{-2\mu}{\lambda^2} - \left( \frac{2}{\lambda^3} e^0 \right) \right) \right] \\
&= \frac{\mu^2}{2} - \frac{\mu}{\lambda} + \frac{1}{\lambda^2} + \frac{\lambda}{2} \left[ \frac{\mu^2}{\lambda} + \frac{2\mu}{\lambda^2} + \frac{2}{\lambda^3} \right] \\
&= \frac{\mu^2}{2} - \frac{\mu}{\lambda} + \frac{1}{\lambda^2} + \frac{\mu^2}{2} + \frac{\mu}{\lambda} + \frac{1}{\lambda^2} \\
&= \mu^2 + \frac{2}{\lambda^2}
\end{aligned}$$

Finally

$$\begin{aligned}
Var(X) &= E[X^2] - E[X]^2 \\
&= \mu^2 + \frac{2}{\lambda^2} - \mu^2 \\
&= \frac{2}{\lambda^2}
\end{aligned}$$

- (b) Let  $X_1$ , and  $X_2$ , be independent exponential random variables, each with parameter  $\lambda$ . The distribution of  $Y = X_1 - X_2$  is a Laplace distribution with a mean of zero and a standard deviation of  $\frac{\sqrt{2}}{\lambda}$ . Set the seed equal to 3, and generate 25000  $X_1$  values from an  $Exp(\lambda = \frac{1}{2})$  and 25000  $X_2$  values from another  $Exp(\lambda = \frac{1}{2})$  distribution. Use these values to create the simulated distribution of  $Y = X_1 - X_2$

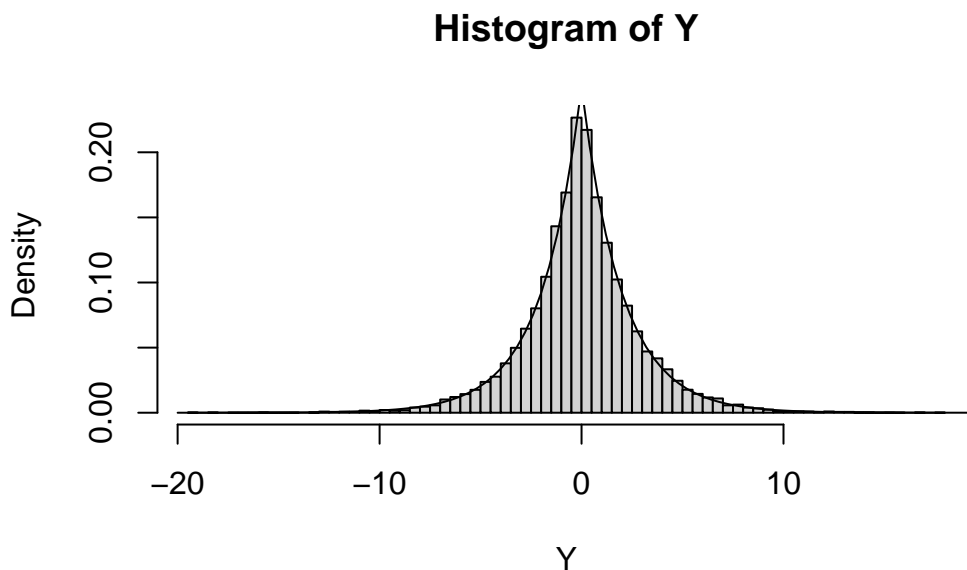
(i) Superimpose a Laplace distribution over a density histogram of the  $Y$  values (Hint: the R function `curve()` can be used to superimpose the Laplace distribution over the density histogram.)

(ii) Is the mean of  $Y$  within 0.02 of the theoretical mean?

(iii) Is the variance of  $Y$  within 2% of the theoretical answer?

```
# (b)
set.seed(3)
n <- 25000
X_1 <- rexp(n, 1/2)
X_2 <- rexp(n, 1/2)
Y <- X_1 - X_2
```

```
# (i)
hist(Y, freq= FALSE, breaks = 80)
curve(0.5/2*exp(-0.5*abs(x)), from=-20, to=20, add=TRUE)
```



```
library(dplyr)
```

```
# ii
theoretical_mean <- 0
between(theoretical_mean,
        theoretical_mean-0.02,theoretical_mean+0.2)
```

```
[1] TRUE
```

As we can see, the mean is within 0.02 of the theoretical.

```
# iii
theoretical_var <- 2/0.5**2
between(theoretical_var, theoretical_var-0.02*theoretical_var,
        theoretical_var+0.02*theoretical_var)
```

```
[1] TRUE
```

Also, the variance is within 2% of the theoretical answer.

**Question 2.** (4.4.24) Give a general expression to calculate the quantiles of a Weibull random variable.

$$f(X|\alpha, \beta) = \begin{cases} \alpha\beta^{-\alpha}x^{\alpha-1}e^{-(x/\beta)^\alpha} & \text{if } x \geq 0 \\ 0 & \text{if } x < 0 \end{cases}$$

The  $p^{th}$  percentile is the value of  $x_p$  such that

$$\int_{-\infty}^{x_p} f(x)dx = \frac{p}{100} = \int_0^{x_p} \frac{\alpha}{\beta^\alpha} x^{\alpha-1} e^{-(x/\beta)^\alpha} dx$$

Then we can use a u-sub to solve the integral

$$\begin{aligned} \Rightarrow \int_0^{(x_p/\beta)^\alpha} e^{-u} du &= \frac{p}{100} \\ -e^{-u} \Big|_0^{(x_p/\beta)^\alpha} &= \\ -e^{-(x_p/\beta)^\alpha} + 1 &= \\ 1 - \frac{p}{100} &= e^{-(x_p/\beta)^\alpha} \\ -\ln(1 - \frac{p}{100}) &= (x_p/\beta)^\alpha \\ \beta(-\ln(1 - \frac{p}{100}))^{\frac{1}{\alpha}} &= x_p \end{aligned}$$

**Question 3.** (4.4.29) Let  $X$  be a random variable with probability density function

$$f(x) = 3\left(\frac{1}{x}\right)^4, x \geq 1$$

- (a) Find the cumulative density function .

$$F(X) = \int_1^x 3t^{-4} dt = -x^3 - (-1)^3 = -x^3 + 1$$

- (b) Fix the seed at 98 (`set.seed(98)`) and generate a random sample of size  $n = 100,000$  from  $X$ 's distribution. Compute the mean, variance, and coefficient of skewness for the random sample.

```
set.seed(98)
n <- 100000
t <- runif(n)
X <- (1-t)**(-1/3)
mean(X)
```

```
[1] 1.50013
```

```
var(X)
```

```
[1] 0.7145073
```

```
X_star <- X-mean(X)
mean(X_star**3/(sqrt(var(X))*3))
```

```
[1] 10.63284
```

- (c) Obtain the theoretical mean, variance and coefficient of skewness of  $X$ .

First, mean

$$\begin{aligned} E[X] &= \int_1^{\infty} x 3x^{-4} dx \\ &= \frac{3}{2} \end{aligned}$$

Variance

$$\begin{aligned}
\text{Var}(X) &= E[X^2] - E[X]^2 \\
&= \int_1^{\infty} x^2 3x^{-4} dx - \frac{9}{4} \\
&= 3 - \frac{9}{4} \\
&= \frac{3}{4}
\end{aligned}$$

Skewness

$$\begin{aligned}
\gamma_1 &= \frac{E[(X - \mu)^3]}{\sigma^3} \\
&= \int_1^{\infty} (x - \mu)^3 3x^{-4} dx \\
&= \int_1^{\infty} (3x^{-1} - 9x^{-2}\mu + 9x^{-3}\mu^2 - 3x^{-4}\mu^3) dx \\
&= \infty, \text{ since } \lim_{x \rightarrow \infty} \ln|x| = \infty
\end{aligned}$$

- (d) How close are the estimates in (b) to the theoretical values in (c)?

```
cat("Mean is ", (mean(X)-1.5)/1.5*100, "% off of theoretical\n")
```

Mean is 0.008686381 % off of theoretical

```
cat("Variance is ", (var(X)-0.75)/0.75*100,"% off of theoretical")
```

Variance is -4.732365 % off of theoretical

While skewness is not even close to theoretical.

**Question 4.** (4.4.32) Let  $X$  be a random variable with probability density function

$$f(x) = (\theta + 1)(1 - x)^\theta, 0 \leq x \leq 1, \theta \geq 0$$

- (a) Verify that the area under  $f(x)$  is 1.

$$\begin{aligned}
\int_{-\infty}^{\infty} (\theta + 1)(1 - x)^\theta dx &= (\theta + 1) \int_0^1 (1 - x)^\theta dx \\
&= -((1 - 1)^{\theta+1} - 1^{\theta+1}) \\
&= 1
\end{aligned}$$

- (b) Find the cumulative density function.

$$\begin{aligned} F(X) &= \int_0^x (\theta + 1)(1 - t)^\theta dt \\ &= -(1 - t)^{\theta+1} \Big|_0^x \\ &= -(1 - x)^{\theta+1} + 1 \end{aligned}$$

- (c) What is  $\mathbb{P}(X \leq .25 | \theta = 2)$ ?

$$\begin{aligned} \mathbb{P}(X \leq 0.25 | \theta = 2) &= F(0.25) \\ &= -(1 - 0.25)^{2+1} + 1 \\ &= 0.578125 \end{aligned}$$

- (d) Fix the seed at 80, and generate 100,000 realizations of  $X$  with  $\theta = 2$ . What are the mean and variance of the random sample?

```
set.seed(80)
n = 100000
t = runif(n)
x = 1 - (-t+1)**(1/3)
cat("mean = ", mean(x), "\n")
```

```
mean = 0.2500348
```

```
cat("variance = ", var(x))
```

```
variance = 0.03769803
```

- (e) Calculate the theoretical mean and variance of  $X$  when  $\theta = 2$

mean when  $\theta = 2$

$$\begin{aligned} E[X] &= \int_0^1 x3(1 - x)^2 dx \\ &= 3 \int_0^1 x - 2x^2 + x^3 dx \\ &= 3\left(\frac{1}{2} - \frac{2}{3} + \frac{1}{4}\right) \\ &= \frac{1}{4} \end{aligned}$$



variance when  $\theta = 2$

$$\begin{aligned}
 \text{Var}(x) &= E[X^2] - E[X]^2 \\
 &= \int_0^1 x^2 3(1-x)^2 dx - \left(\frac{1}{4}\right)^2 \\
 &= 3 \int_0^1 x^2 - 2x^3 + x^4 dx - \left(\frac{1}{4}\right)^2 \\
 &= 3\left(\frac{1}{3} - \frac{2}{4} + \frac{1}{5}\right) - \left(\frac{1}{4}\right)^2 \\
 &= 0.0375
 \end{aligned}$$

- (f) How close are the estimates in (d) to the theoretical values in (e)?

```
cat("Mean is ", (mean(x)-1/4)/(1/4)*100, "% off of theoretical\n")
```

Mean is 0.01390041 % off of theoretical

```
cat("Variance is ", (var(X)-0.0375)/0.0375*100,"% off of theoretical")
```

Variance is 1805.353 % off of theoretical

**Question 5.** (4.4.34) A copper wire manufacturer produces conductor cables. These cables are of practical use if their resistance lies between 0.10 and 0.13 ohms per meter. The resistance of the cable follows a normal distribution, where 50% of the cables have a resistance under 0.11 ohms and 10% have a resistance over 0.13 ohms.

- (a) Determine the mean and the standard deviation for cable resistance.

Since the median is 0.11, and it follows a normal distribution, then the mean also has to be 0.11.

Then for standard deviation, since 10% of the cables have a resistance of 0.13 ohms, we have

$$\begin{aligned}
 \mathbb{P}(X \leq 0.13) &= 0.9 \\
 \Rightarrow \mathbb{P}\left(Z = \frac{X - 0.11}{\sigma} \leq \frac{0.13 - 0.11}{\sigma}\right) &= 0.9
 \end{aligned}$$

Because  $\mathbb{P}(Z \leq 1.28) = 0.9$ , set

$$\begin{aligned}
 \frac{0.13 - 0.11}{\sigma} &= 1.28 \\
 \Rightarrow \sigma &= \frac{0.13 - 0.11}{1.28} \\
 &= 0.015625
 \end{aligned}$$

- (b) Find the probability that a randomly chosen cable can be used.

$$\mathbb{P}(0.10 \leq X \leq 0.13) = \mathbb{P}(X \leq 0.13) - \mathbb{P}(X \leq 0.10)$$

```
p<-pnorm(0.13,0.11,0.015625)-pnorm(0.10,0.11,0.015625)
cat("Probability that it can be used is ", p)
```

Probability that it can be used is 0.6386411

- (c) Find the probability that at least 3 out of 5 randomly chosen cables can be used.

We can use the binomial distribution to help us calculate this

$$P(X \geq 3) = \sum_{i=3}^5 \binom{5}{i} \pi^i (1 - \pi)^{5-i}$$

```
cat("Probability that at least 3 can be used is ",sum(dbinom(3:5,5,p)))
```

Probability that at least 3 can be used is 0.7469351