ST501 Group Project

Introduction

Part 1 Convergences

 \mathbf{A}

$$f_Y(y) = \frac{1}{2b}e^{-\left(\frac{|y-\mu|}{b}\right)}$$

Given: $\mu = 0$ and b = 5

$$E(Y) = \int_{-\infty}^{\infty} y * f(y) dy$$

$$= \int_{-\infty}^{0} y * \frac{1}{10} e^{\frac{y}{5}} + \int_{0}^{\infty} y * \frac{1}{10} e^{\frac{-y}{5}}$$

$$= \frac{1}{10} \left[\int_{-\infty}^{0} y e^{\frac{y}{5}} + \int_{0}^{\infty} y * e^{\frac{-y}{5}} \right]$$

$$= \frac{1}{10} \left[\left[(5y - 25) e^{\frac{y}{5}} \right]_{-\infty}^{0} + \left[(-5y - 25) e^{\frac{-y}{5}} \right]_{0}^{\infty} \right]$$

$$= \frac{1}{10} \left[-25 + 25 \right]$$

$$E(Y) = 0 \blacksquare$$

$$E(Y^{2}) = \int_{-\infty}^{\infty} y^{2} * f(y) dy$$

$$= \frac{1}{10} \left[\int_{-\infty}^{0} y^{2} e^{\frac{y}{5}} + \int_{0}^{\infty} y^{2} * e^{\frac{-y}{5}} \right]$$

$$= \frac{1}{10} * \left[250 + 250 \right]$$

$$E(Y^{2}) = 50 \blacksquare$$

Therefor $E(Y^2)$ exists.

Thus:

 $E(Y)=0,\, E(Y^2)=50$ then Var(Y)=50 by the variance computing formula.

$$L = \frac{1}{n} \sum_{i=1}^{n} Y_i^2$$

By the Law of Large Numbers:

$$\frac{1}{n} \sum_{i=1}^{n} Y_i \quad \underset{\rightarrow}{p} \ E(Y_i)$$

By generalisation of the Law of Large Numbers:

$$L = \frac{1}{n} \sum_{i=1}^{n} Y_i^2 \quad \underline{p} \ E(Y_i^2) = 50$$

Therefore:

$$L \xrightarrow{p} 50 \blacksquare$$

 \mathbf{B}

By the continuity theorem, $K = \sqrt{L} \quad p \sqrt{50}$

Thus: $K \xrightarrow{p} \sqrt{50}$.

 \mathbf{C}

Derrive the CDF of the double exponential distribution.

$$f_Y(y) = \frac{1}{2b} * e^{-(\frac{|y-\mu|}{b})}$$

This yields two cases for the CDF:

$$F_Y(y) = \begin{cases} \int_{-\infty}^t f_Y(Y)dt & \text{for } y \le \mu \\ \int_t^\infty f_Y(Y)dt & \text{for } y \ge \mu \end{cases}$$

Solving for these leaves us with the following:

For $y \leq \mu$:

$$\int_{-\infty}^{y} f_Y(Y)dt$$
$$\int_{-\infty}^{y} \frac{1}{2b} * e^{-(\frac{t-\mu}{b})} dt$$

We know that in this case we are solving for when the exponent is positive.

$$\int_{-\infty}^{y} \frac{1}{2b} * e^{\left(\frac{|t-\mu|}{b}\right)} dt$$

$$\frac{1}{2b} * e^{(\frac{t-\mu}{b})} * b|_{-\infty}^{y} = \frac{1}{2} e^{(\frac{t-\mu}{b})} - 0 = \frac{1}{2} e^{(\frac{t-\mu}{b})}$$

For the case where $y \ge \mu$

$$\int_{-\infty}^{y} f_Y(Y)dt$$

$$\int_{-\infty}^{y} \frac{1}{2b} * e^{-(\frac{t-\mu}{b})} dt$$

This must be further split into:

$$\int_{-\infty}^{\mu} \frac{1}{2b} * e^{-(\frac{t-\mu}{b})} dt + \int_{\mu}^{y} \frac{1}{2b} * e^{-(\frac{t-\mu}{b})} dt$$
$$\frac{1}{2b} * e^{(\frac{-t+\mu}{b})} * -b|_{-\infty}^{\mu} + \frac{1}{2b} * e^{(\frac{-t+\mu}{b})} * -b|_{\mu}^{y}$$
$$\frac{1}{2} - \frac{1}{2} e^{\frac{\mu-y}{b}} + \frac{1}{2}$$

Thus for this case:

$$1 - \frac{1}{2}e^{\frac{\mu - y}{b}}$$

In conclusion the CDF for the double exponential distribution is:

$$F_Y(y) = \begin{cases} \frac{1}{2}e^{(\frac{y-\mu}{b})} & \text{for } y \le \mu\\ 1 - \frac{1}{2}e^{\frac{\mu-y}{b}} & \text{for } y \ge \mu \end{cases}$$

 \mathbf{D}

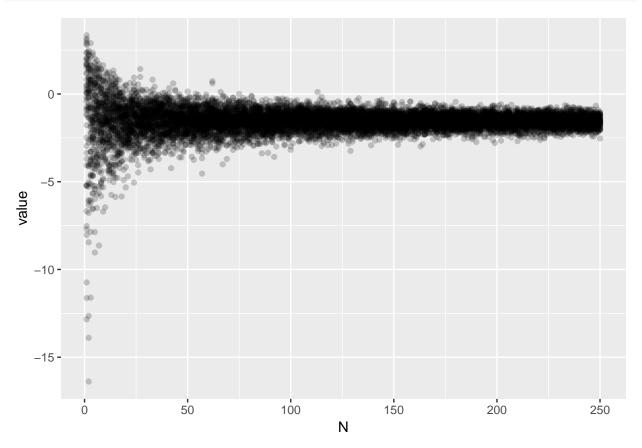
```
rdoublex <- function(u, mu = 0, b = 5){
   if( u > 0){
        mu + b*log(2*u)
   }else {
        mu - b*log(2-2*u)
   }
}

output_matrix <- matrix(0, nrow= 250, ncol = 50)

for( j in 1:50){
   for(i in 1:250){
        random_value <- runif(i, 0, 1)
        output_matrix[i, j]<- mean(vapply(random_value, FUN = rdoublex, double(1)))
   }
}

output_matrix %>%
   as.data.frame() %>%
```

```
mutate(N = 1:250) %>%
gather(replication, value, -N) %>%
ggplot(aes(N, value))+
geom_point(alpha=1/5)
```



 \mathbf{E}