

Problem Set 4 *

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*Problem list 3.10.6, 3.10.16, 3.12.6, 3.12.8

Problem 3.10.6. Let Y_1, Y_2, \dots, Y_n be a random sample from the exponential pdf $f_y(y) = e^{-y}, y \geq 0$. What is the smallest n for which $P(Y_{\min} < 0.2) > 0.9$?

Solution The evaluated $n > 11.513$, since x must be an integer (I'm assuming since these are trials) $n \geq 12$.

$$\begin{aligned}
 P(Y_{\min} < 0.2) &= \int_0^{0.2} f_{Y_{\min}}(y) \\
 &= \int_0^{0.2} n[1 - F_Y(y)]^{n-1} f_Y(y) \quad \text{Theorem 3.10.1.b (pg193)} \\
 &= \int_0^{0.2} n[1 - (1 - e^{-y})]^{n-1} (e^{-y}) \\
 &= \int_0^{0.2} n(e^{-y})^{n-1} (e^{-y}) \\
 &= \int_0^{0.2} n(e^{-ny}) \\
 &= -e^{-ny} \Big|_0^{0.2} \\
 &= -e^{-(0.2)n} - (-1) > 0.9 \\
 &= \log(e^{-0.2n}) < \log(0.1) \\
 n &> \frac{\log(0.1)}{-0.2} \approx 11.513
 \end{aligned}$$

Problem 3.10.16. Suppose a device has three independent components, all of whose lifetimes (in months) are modeled by the exponential pdf, $f_y(y) = e^{-y}, y > 0$. What is the probability that all three components will fail within two months of one another?

Solution¹ Range = $Y_{max} - Y_{min} = Y_3' - Y_1'$. $P(R < r) = 0.646$

The *memoryless property of the exponential distribution*:

$$P(X \geq s + t | X \geq s) = P(X \geq t)$$

This implies that the level of Y_1' is inconsequential. Thus, we can assume that $Y_1' = 0$. In which case we are really only interested in $P(Y_{max}' < r)$. Thus, we can apply Theorem 3.10.1.a (pg 193) with: $n=3$, $f_Y(y) = e^{-y}$, and $F_Y(y) = \int_0^y f_Y(y)dy = 1 - e^{-y}$.

$$P(Y_{max} < m) = \int_{-\infty}^m n[F_Y(y)]^{n-1} f_Y(y)$$

$$P(Y_3' < 2) = \int_0^2 3[1 - e^{-y}]^2 e^{-y} \quad \text{Enter WolframAlpha}$$

$$\approx 0.646.$$

Problem 3.12.6. Find $M_Y(t)$ if Y has the pdf:

$$f_Y(y) = \begin{cases} y, & 0 \leq y \leq 1 \\ 2 - y, & 1 \leq y \leq 2 \\ 0, & \text{elsewhere} \end{cases}$$

Solution $M_Y(t) = \frac{1-e^{-t}}{t^2}$

¹To start, I want to note that understanding of the "memoryless property of the exponential distribution" was critical to even approaching success in this problem. I studied the proof in this pdf, <http://www.cs.cmu.edu/afs/cs/academic/class/15750-s19/OldScribeNotes/lecture11.pdf> (pg 2). I also used wolframalpha to do some of the calculations that were a to complex for my patience level.

Since X is continuous, the second part of Definition 3.12.1 (pg 206) applies:

$$\begin{aligned}
M_Y(t) &= E(e^{tW}) = \int_{-\infty}^{\infty} e^{tw} f_W(w) dw \\
&= \int_{-\infty}^0 e^{ty} 0 + \int_0^1 e^{ty} y + \int_1^2 e^{ty} (2-y) + \int_2^{\infty} e^{ty} 0 \\
&= \int_0^1 e^{ty} y + \int_1^2 e^{ty} (2-y) && \text{I used WolframAlpha here.} \\
&= \left(\frac{1}{t} y - \frac{1}{t^2} \right) e^{ty} \Big|_0^1 + \frac{2}{t} e^{ty} \Big|_1^2 - \left(\frac{1}{t} y - \frac{1}{t^2} \right) e^{ty} \Big|_1^2 \\
&= \left(\frac{1}{t} - \frac{1}{t^2} \right) e^t + \left(\frac{1}{t^2} \right) + \frac{2}{t} e^{2t} - \frac{2}{t} e^t - \left(\frac{2}{t} - \frac{1}{t^2} \right) e^{2t} + \left(\frac{1}{t} - \frac{1}{t^2} \right) e^t \\
&= \frac{1}{t^2} + \frac{1}{t^2} e^{2t} - \frac{2}{t^2} e^t \\
&= \frac{1}{t^2} (e^t - 1)
\end{aligned}$$

Problem 3.12.8. Let Y be a continuous random variable with $f_Y(y) = ye^{-y}, 0 \leq y$. Show that $M_Y(t) = \frac{1}{(1-t)^2}$

Solution Since X is continuous, the second part of Definition 3.12.1 (pg 206) applies:

$$\begin{aligned}
M_Y(t) &= E(e^{tW}) = \int_{-\infty}^{\infty} e^{tw} f_W(w) dw \\
&= \int_0^{\infty} e^{ty} y e^{-y} \\
&= \int_0^{\infty} e^{ty-y} y \\
&= \frac{e^{(t-1)y}((t-1)y-1)}{(t-1)^2} \Big|_0^{\infty} \\
&= \lim_{y \rightarrow \infty} \frac{e^{(t-1)y}((t-1)y-1)}{(t-1)^2} - \lim_{y \rightarrow 0} \frac{e^{(t-1)y}((t-1)y-1)}{(t-1)^2} \\
&= \lim_{y \rightarrow \infty} \frac{e^{(t-1)y}((t-1)y-1)}{(t-1)^2} - \frac{e^{(t-1)0}((t-1)0-1)}{(t-1)^2} \\
&= \lim_{y \rightarrow \infty} \frac{e^U(U-1)}{(t-1)^2} - \frac{(-1)}{(t-1)^2} \quad \text{Distribute, let } U = (yt-y) \\
&= \frac{(\lim_{y \rightarrow \infty} e^{yt-y})(\lim_{y \rightarrow \infty} (yt-y-1))}{(t-1)^2} - \frac{(-1)}{(t-1)^2} \quad \text{Theorem 20.4 (Ross pg 156)}
\end{aligned}$$

It will suffice to show that if $\lim_{y \rightarrow \infty} e^{yt-y}$ converges to 0 then $M_Y(t) \rightarrow \frac{1}{(t-1)^2}$.

Since $0 * (\lim_{y \rightarrow \infty} (yt-y-1)) = 0$ and $\frac{0}{(t-1)^2} - \frac{(-1)}{(t-1)^2} = \frac{1}{(t-1)^2}$.

Proof. Let $\epsilon > 0$ and $\delta = \frac{\log(\epsilon)}{(t-1)}$. Note that, $\forall t$ such that $0 < t < 1$, $\delta > 0$.

So if $0 < |x| < \delta$ then $|x| < \frac{\log(\epsilon)}{(t-1)} \implies (t-1)|x| = \log(\epsilon) \implies e^{xt-x} < \epsilon$

Thus, by Theorem 20.6 (Ross pg 159²), $\lim_{x \rightarrow \infty} e^{xt-x} = 0$. \square

²Ross, Kenneth: Elementary Analysis the Theory of Calculus. Undergraduate Texts in Mathematics, 2nd edn. Springer, New York/Heidelberg/Berlin (2013)

Problem Assignment. The Inverse Transform Sampling procedure requires $F_Y^{-1}(u)$ where $F_Y^{-1}(u)$ is the inverse of the cumulative distribution function, $F_Y(y)$. We saw in problem 3.10.6 that $F_Y(y) = (1 - e^{-y})$. Solving for $x = -\log(1 - u)$ is our inverse CDF function. Now to be honest with you, I couldn't really follow your R code so I wrote my own. Now I'm not really sure what you are looking for to by way of the solution. But I've included my code below, and I am getting very close to 0.9 so I believe it is correct.

```
# inverse transform sampling

# This is the number of times you run the process,
positions <- 1000

# This will be a vector of minimums
samples <- c()

# You add one number to sample each process
# and you want to run the process position times.
while( length(samples) < positions )
{
  #This is the number computed in the problem.
  #The n needed to produce the minimum
  num.samples <- 12
```

```

# The uniform distribution from 0 to 1
U          <-  runif(num.samples,0,1)

# The pdf
Y <-  exp(-U)

# The inverse cdf function
X          <-  -log(1-U)

# take the minimum
sortedx <-  sort(X)
minx = min(sortedx)
samples <-  append(samples,minx)
}

# Get the ratio of samples <0.2 to total samples.
# This is your probability.
less <-  sum(samples <0.2)
ratio <-  less/positions

# plot
hist(samples, breaks=30, freq=F, xlab='X', main='Generating_Exponential')

```

```
curve(dexp(X, rate=2) , 0, 3, lwd=2, xlab = "", ylab = "", add = T)
```