

MATH 556 - ASSIGNMENT 4 SOLUTIONS

1. The cdf is

$$F_X(x) = 1 - \exp\{-\lambda(x - \eta)\} \quad x > \eta$$

and zero otherwise. Therefore $Y_{n,1}$ has cdf

$$F_{Y_{n,1}}(y) = 1 - \exp\{-n\lambda(y - \eta)\} \quad y > \eta$$

and as $n \rightarrow \infty$ we have

$$F_{Y_{n,1}}(y) \rightarrow \begin{cases} 0 & y \leq \eta \\ 1 & y > \eta \end{cases}$$

Therefore the limiting distribution is degenerate at η ; the limit function is not right continuous, but clearly for any $\epsilon > 0, P[Y_{n,1} < \epsilon] \rightarrow 1$.

2 Marks

2. We have

$$F_X(x) = 1 - \frac{1}{(1+x)} \quad x > 0$$

and zero otherwise. Then $Y_{n,n}$ has cdf

$$F_{Y_{n,n}}(y) = \left(1 - \frac{1}{1+y}\right)^n \quad y > 0$$

Let $V_n = Y_{n,n}/n$. Then for $v > 0$

$$F_{V_n}(v) = \left(1 - \frac{1}{1+nv}\right)^n$$

and in the limit we have that

$$F_{V_n}(v) \rightarrow \begin{cases} 0 & v \leq 0 \\ \exp\{-1/v\} & v > 0 \end{cases}$$

4 Marks

3. By Markov's inequality, for $r = 2$ and $\epsilon > 0$,

$$P_X [X^2 \geq \epsilon] \leq \frac{\mathbb{E}_X [X^2]}{\epsilon}.$$

Consider the sequence of random variables $\{X_n - X, n \geq 1\}$. Then this inequality becomes

$$P_{X_n, X} [(X_n - X)^2 \geq \epsilon] \leq \frac{\mathbb{E}_{X_n, X} [(X_n - X)^2]}{\epsilon}.$$

But the right hand side goes to zero as $n \rightarrow \infty$. Therefore, for any $\epsilon > 0$,

$$P_{X_n, X} [(X_n - X)^2 \geq \epsilon] \rightarrow 0$$

which confirms that $X_n \xrightarrow{p} X$.

4 Marks

4. (a) We have from the Weak Law of Large Numbers and Central Limit Theorem that

$$\bar{X}_n \xrightarrow{p} \phi \quad \sqrt{n}(\bar{X}_n - \phi) \xrightarrow{d} \mathcal{N}(0, \phi(1 - \phi))$$

so using the Delta Method for transform $g(t) = t^2$, so that $\dot{g}(t) = 2t$, we have

$$S_n = \{\bar{X}_n\}^2 \xrightarrow{p} \phi^2$$

and

$$\sqrt{n}(S_n - \phi^2) \xrightarrow{d} \mathcal{N}(0, 4\phi^3(1 - \phi))$$

so that for large n

$$S_n \dot{\sim} \mathcal{N}(\phi^2, 4\phi^3(1 - \phi)/n).$$

5 Marks

(b) We have that

$$T_n = \log M_n = \frac{1}{n} \sum_{i=1}^n \log U_i = \frac{1}{n} \sum_{i=1}^n Y_i$$

say. Note that, by parts

$$\mathbb{E}_{Y_i}[Y_i] = \int_0^1 \log u \, du = [-u(1 - \log u)]_0^1 = -1$$

and also by parts

$$\mathbb{E}_{Y_i}[Y_i^2] = \int_0^1 (\log u)^2 \, du = [u(\log u)^2]_0^1 - 2 \int_0^1 \log u \, du = 2$$

so that $\text{Var}_{Y_i}[Y_i] = 1$. Therefore by the CLT

$$\sqrt{n}(T_n + 1) \xrightarrow{d} \mathcal{N}(0, 1).$$

Using the Delta Method for transform $g(t) = e^t$, so that $\dot{g}(t) = e^t$, we have

$$\sqrt{n}(M_n - e^{-1}) \xrightarrow{d} \mathcal{N}(0, e^{-2})$$

so that for large n

$$M_n \dot{\sim} \mathcal{N}(e^{-1}, e^{-2}/n).$$

5 Marks

Here is a Monte Carlo study: the code simulates the distribution of the standardized version of M_n , that is

$$Z_n = \sqrt{n} \frac{M_n - e^{-1}}{e^{-1}}$$

for $n = 5, 10, 20, 50, 100$.

```

1 set.seed(2323)
2 nvec<-c(5,10,20,50,100)
3 nreps<-5000
4 Xsamp<-matrix(0,nrow=nreps,ncol=length(nvec))
5 for(i in 1:length(nvec)){
6     Xsamp[,i]<-replicate(nreps,exp(mean(log(runif(nvec[i])))))
7     Xsamp[,i]<-sqrt(nvec[i])*(Xsamp[,i]-exp(-1))/exp(-1)
8 }
9 Xsamp<-cbind(Xsamp,rnorm(nreps))
10 boxplot(Xsamp,names=c(nvec,Inf))

```

The boxplots for increasing n seem to match the asymptotic limit for $n \geq 20$.

5000 samples from the distribution of Z_n

