## Large Sample Theory

## Ferguson

## Exercises, Section 2, Partial Converses to Theorem 1.

1. Suppose $X$ has the gamma distribution $\mathcal{G}(\alpha, \beta)$ (density proportional to $e^{-x / \beta} x^{\alpha-1}$ on $x>0)$, and let $Y=(1 / \gamma) \log (X)$ where $\gamma \neq 0$.
(a) Find the density of $Y$. Make the change of parameter $\theta$ for $\beta$ through the equation $\log (\beta)=\gamma \theta-\log (\alpha)$, and note that $\theta$ is a location parameter of the resulting distribution.
(b) Let $\psi(\alpha)$ denote the digamma function, $\psi(\alpha)=(d / d \alpha) \log (\Gamma(\alpha))=\Gamma^{\prime}(\alpha) / \Gamma(\alpha)=$ $(1 / \Gamma(\alpha)) \int_{0}^{\infty} \log (x) e^{-x} x^{\alpha-1} d x$. What is the mean of $Y$ ? Let $\sigma^{2}$ denote the variance of $Y$ and show that $\sigma^{2} \gamma^{2}=\psi^{\prime}(\alpha)$. ( $\psi^{\prime}(\alpha)$ is sometimes called the trigamma function.)
(c) Denote the above distribution of $Y$ as $\mathcal{N}\left(\theta, \sigma^{2}, \gamma\right)$, defined for all $\theta$, all $\sigma^{2}>0$ and all $\gamma \neq 0$. Fill in the missing case, $\gamma=0$, by showing that as $\gamma \rightarrow 0, \mathcal{N}\left(\theta, \sigma^{2}, \gamma\right)$ converges in law to $\mathcal{N}\left(\theta, \sigma^{2}\right)$, the normal distribution with mean $\theta$ and variance $\sigma^{2}$. (You may use $\alpha \psi^{\prime}(\alpha) \rightarrow 1$ as $\alpha \rightarrow \infty$, and Stirling's formula.)
(d) Thus $\mathcal{N}\left(\theta, \sigma^{2}, \gamma\right)$ is a three parameter generalization of the normal distribution, with $\mathcal{N}\left(\theta, \sigma^{2}, 0\right)$ being the normal distributions. Show $\mathcal{N}\left(\theta, \sigma^{2}, \gamma\right)$ can also be defined at $\gamma= \pm \infty$, by showing that $\mathcal{N}\left(0, \sigma^{2}, \gamma\right)$ converges in law to the exponential distribution with mean $\sigma$ as $\gamma \rightarrow-\infty$. (You may use $\alpha \Gamma(\alpha) \rightarrow 1$ and $\alpha^{2} \psi^{\prime}(\alpha) \rightarrow 1$ as $\alpha \rightarrow 0$.)
2. Someone is walking on the integer lattice of the line starting at some unknown integer, $N$. Your task is to find him even though you are blindfolded. You may only ask questions of the form "Are you at integer $n$ ?" and he must answer truthfully. He may move at most two integers up or down between questions, so after the first question he can only move to one of $N-2, N-1, N, N+1, N+2$. Can you devise a sequence of questions so that you will eventually find him almost surely (i.e. with probability one) no matter where he starts and what he does?
3. Let $X_{1}, X_{2}, \ldots$ be a sequence of random variables such that $X_{1}$ is uniform on $[0,1]$, and where for $n=1,2, \ldots$, the conditional distribution of $X_{n+1}$ given $X_{1}, \ldots, X_{n}$, is uniform on $\left[0, c X_{n}\right]$ for some number $c$ such that $\sqrt{3}<c<2$.
(a) Find the expectation of $X_{n}^{r}$ for $r>0$.
(b) Show that $X_{n}$ converges to 0 in mean $(r=1)$, but not in quadratic mean, $(r=2)$.
(c) Does $X_{n}$ converge to 0 almost surely?
4. Let $X_{1}, X_{2}, \ldots$ be a sequence of i.i.d. random variables (with no assumptions of any finite moments). Does $X_{n} / n$ converge almost surely to 0 ? If so, show it; if not, give a counterexample.
5. Suppose that random variables $X_{1}, X_{2}, \ldots$ are uniformly bounded, i.e. suppose there is a constant $B$ such that $\left|X_{n}\right|<B$ a.s. for all $n$. Let $r>0$. Show that $X_{n} \xrightarrow{\mathrm{P}} X$ if and only if $X_{n} \xrightarrow{\mathrm{r}} X$.
6. Suppose that $X$ has a standard Cauchy distribution. Find a sequence of random variables $X_{n}$ for $n=1, \ldots$, such that $X_{n}$ converges in quadratic mean to $X$, but that $X_{n}$ does not converge to $X$ almost surely.
7. Let $X_{1}, X_{2}, \ldots$ be independent Bernoulli trials with $\mathrm{P}\left(X_{n}=1\right)=1 / n^{2}$ for $n=$ $1,2, \ldots$. Then, as noted in the text, $\mathrm{P}\left(X_{n}=1\right.$ i.o. $)=0$. Therefore, there is a last $n$ such that $X_{n}=1$. Let $N=\max \left\{n: X_{n}=1\right\}$ be the random time at which this occurs. Find the distribution of $N$, i.e. find $\mathrm{P}(N=n)$ for $n=1,2, \ldots$ What is $\mathrm{E}(N)$ ?
