

Solutions to Exercise Set 2.

$$\begin{aligned}
2.7. \quad \mathbb{P}(N = n) &= \mathbb{P}(X_n = 1, X_{n+1} = 0, X_{n+2} = 0, \dots) \\
&= \mathbb{P}(X_n = 1)\mathbb{P}(X_{n+1} = 0)\mathbb{P}(X_{n+2} = 0) \cdots \\
&= \frac{1}{n^2} \cdot \frac{n(n+2)}{(n+1)^2} \cdot \frac{(n+1)(n+3)}{(n+2)^2} \cdots = \frac{1}{n(n+1)}
\end{aligned}$$

So $\mathbb{E}(N) = \sum_1^\infty n/(n(n+1)) = \sum_1^\infty 1/(n+1) = \infty$.

3.5. (a) The characteristic function is

$$\begin{aligned}
\frac{1}{\Gamma(\alpha)\beta^\alpha} \int_0^\infty x^{\alpha-1} e^{itx-(x/\beta)} dx &= \frac{1}{\Gamma(\alpha)\beta^\alpha} \int_0^\infty x^{\alpha-1} e^{-((1-it\beta)/\beta)x} dx \\
&= \frac{1}{\Gamma(\alpha)\beta^\alpha} \Gamma(\alpha) \left(\frac{\beta}{1-it\beta} \right)^\alpha = \frac{1}{(1-it\beta)^\alpha}
\end{aligned}$$

(b) The characteristic function of $X_n - n$ is

$$\varphi_{X_n-n}(t) = e^{-itn} (1 - (it/n))^{-n^2}.$$

It is sufficient to show that $\log \varphi_{X_n-n}(t) \rightarrow -t^2/2$.

$$\begin{aligned}
\log \varphi_{X_n-n}(t) &= -itn - n^2 \log(1 - (it/n)) \\
&= -itn + n^2 [(it/n) + (1/2)(it/n)^2 + O(n^{-3})] \\
&= -t^2/2 + O(n^{-1}) \rightarrow -t^2/2
\end{aligned}$$

4.1. (a) The least squares estimate of λ is that value of λ that minimizes $\sum_1^n (X_i - \lambda z_i)^2$. Taking a derivative of this sum with respect to λ , setting to zero and solving gives $\hat{\lambda}_{LS} = \sum_1^n X_i z_i / \sum_1^n z_i^2$. The weighted least squares estimate of λ is that value of λ that minimizes $\sum_1^n (X_i - \lambda z_i)^2 / z_i$. Similarly, we find $\hat{\lambda}_W = \sum_1^n X_i / \sum_1^n z_i$.

(b) Since $\mathbb{E}\hat{\lambda}_{LS} = \lambda$, it is sufficient to find for what values of the z_i we have $\text{Var}(\hat{\lambda}_{LS}) \rightarrow 0$ as $n \rightarrow \infty$ (see Exercise 1.5). But

$$\text{Var} \hat{\lambda}_{LS} = \text{Var} \frac{\sum_1^n X_i z_i}{\sum_1^n z_i^2} = \frac{\text{Var} \sum_1^n X_i z_i}{(\sum_1^n z_i^2)^2} = \frac{\sum_1^n z_i^2 \text{Var} X_i}{(\sum_1^n z_i^2)^2} = \frac{\lambda \sum_1^n z_i^3}{(\sum_1^n z_i^2)^2}$$

Therefore, λ_{LS} is consistent in quadratic mean if and only if $\sum_1^n z_i^3 / (\sum_1^n z_i^2)^2 \rightarrow 0$ as $n \rightarrow \infty$.

(c) $\hat{\lambda}_W$ is also unbiased, and $\text{Var}(\hat{\lambda}_W) = \lambda \sum_1^n z_i / (\sum_1^n z_i)^2 = \lambda / \sum_1^n z_i$. So $\hat{\lambda}_W$ is consistent in quadratic mean if and only if $\sum_1^n z_i \rightarrow \infty$ as $n \rightarrow \infty$.

(d) If $\hat{\lambda}_W$ is not consistent in quadratic mean, i.e. if $\sum_1^\infty z_i < \infty$, then both $\sum_1^\infty z_i^2 < \infty$ and $\sum_1^\infty z_i^3 < \infty$, so $\hat{\lambda}_{LS}$ is not consistent either. However if $z_i = 1/i$, then $\sum_1^\infty z_i = \infty$ so that $\hat{\lambda}_W$ is consistent in quadratic mean, but $\sum_1^\infty z_i^2 < \infty$ and $\sum_1^\infty z_i^3 < \infty$ so that $\hat{\lambda}_{LS}$ is not consistent.