

4.63

Since $X = e^Z$ and $g(z) = e^z$ is convex, by Jensen's Inequality $EX = Eg(Z) \geq g(EZ) = e^0 = 1$. In fact, there is equality in Jensen's Inequality if and only if there is an interval I with $P(Z \in I) = 1$ and $g(z)$ is linear on I . But e^z is linear on an interval only if the interval is a singlepoint. So $EX > 1$, unless $P(Z = EZ = 0) = 1$.

5.3

Note that $Y_i \sim \text{Bernoulli}$ with $p_i = P(X_i \geq \mu) = 1 - F(\mu)$ for each i . Since the Y_i s are iid Bernoulli, $\sum_{i=1}^n Y_i \sim \text{binomial}(n, 1 - F(\mu))$.

5.11

Let $g(s) = s^2$. Since $g(\cdot)$ is a convex function, we know from Jensen's inequality that $Eg(S) \geq g(ES)$, which implies $\sigma^2 = ES^2 \geq (ES)^2$. Taking square roots, $\sigma \geq ES$. It is clear that the inequality will be strict unless there is an interval I such that g is linear on I and $P(X \in I) = 1$. Since s^2 is linear only on single points, we have $ET^2 > (ET)^2$ for any random variable T , unless $P(T = ET) = 1$.

5.12

$$\begin{aligned}
 E\bar{Y}_1 &= E|\bar{X}| \\
 &= \int_{-\infty}^{\infty} |x| \frac{1}{\sqrt{2\pi/n}} e^{-\frac{nx^2}{2}} dx \quad \text{since } |\bar{X}| \sim N(0, 1/n) \\
 &= 2 \int_0^{\infty} x \frac{1}{\sqrt{2\pi/n}} e^{-\frac{nx^2}{2}} dx \\
 &= \sqrt{\frac{2}{n\pi}} \\
 E|X_i| &= \int_{-\infty}^{\infty} |x_i| \frac{1}{\sqrt{2\pi}} e^{-\frac{x_i^2}{2}} dx_i \\
 &= \frac{1}{\sqrt{2\pi}} \cdot 2 \int_0^{\infty} x_i e^{-\frac{x_i^2}{2}} dx_i \\
 &= \sqrt{\frac{2}{\pi}} \\
 E\bar{Y}_2 &= \frac{1}{n} \sum_1^n E|X_i| = \sqrt{\frac{2}{\pi}}
 \end{aligned}$$

Thus $E\bar{Y}_1 < E\bar{Y}_2$.

$$\begin{aligned}
E(c\sqrt{S^2}) &= c\sqrt{\frac{\sigma^2}{n-1}}E\left(\sqrt{\frac{S^2(n-1)}{\sigma^2}}\right) \\
&= c\sqrt{\frac{\sigma^2}{n-1}}\int_0^\infty \sqrt{q}\frac{1}{\Gamma(\frac{n-1}{2})2^{(n-1)/2}}q^{\frac{n-1}{2}-1}e^{-q/2}dq \\
&\quad \sqrt{S^2(n-1)/\sigma^2} \text{ is the square root of a } \chi_{n-1}^2 \text{ random variable} \\
&= c\sqrt{\frac{\sigma^2}{n-1}}\frac{\Gamma(\frac{n}{2})2^{n/2}}{\Gamma(\frac{n-1}{2})2^{(n-1)/2}}\int_0^\infty \frac{1}{\Gamma(\frac{n}{2})2^{n/2}}q^{\frac{n}{2}-1}e^{-q/2}dq \\
&= c\sqrt{\frac{\sigma^2}{n-1}}\frac{\Gamma(\frac{n}{2})2^{n/2}}{\Gamma(\frac{n-1}{2})2^{(n-1)/2}} \quad \text{use pdf of } \chi_n^2 \\
&= c\sigma\sqrt{\frac{2}{n-1}}\frac{\Gamma(\frac{n}{2})}{\Gamma(\frac{n-1}{2})} \equiv \sigma
\end{aligned}$$

Thus, $c = \sqrt{\frac{n-1}{2}}\frac{\Gamma(\frac{n-1}{2})}{\Gamma(\frac{n}{2})}$.

5.16

a. $\sum_{i=1}^3 \left(\frac{X_i-i}{i}\right)^2 \sim \chi_3^2$.

b. $\frac{X_1-1}{\sqrt{\sum_{i=2}^3 \left(\frac{X_i-i}{i}\right)^2/2}} \sim t_2$.

c. $\left(\frac{X_1-1}{\sqrt{\sum_{i=2}^3 \left(\frac{X_i-i}{i}\right)^2/2}}\right)^2 \sim F_{1,2}$.

Additional Prob

$$\begin{aligned}
M_{X_\alpha}(t) &= \left(\frac{1}{1-t}\right)^\alpha, \\
M_{Z_\alpha}(t) &= Ee^{tZ_\alpha} \\
&= Ee^{t\frac{X_\alpha - EX_\alpha}{\sqrt{Var(X_\alpha)}}} \\
&= Ee^{t\frac{X_\alpha - \alpha}{\sqrt{\alpha}}} \\
&= e^{-t\sqrt{\alpha}}M_{X_\alpha}(t/\sqrt{\alpha}) \\
&= e^{-t\sqrt{\alpha}}\left(\frac{1}{1-t/\sqrt{\alpha}}\right)^\alpha
\end{aligned}$$

$$\begin{aligned} \ln M_{X_\alpha}(t) &= -t\sqrt{\alpha} - \alpha \ln(1 - t/\sqrt{\alpha}) \\ &= -t\sqrt{\alpha} - \alpha \left(-\frac{t}{\sqrt{\alpha}} - \frac{t^2}{2\alpha} + R\left(\frac{t}{\sqrt{\alpha}}\right) \right) \\ &= \frac{t^2}{2} + \alpha R\left(\frac{t}{\sqrt{\alpha}}\right) \\ &\rightarrow \frac{t^2}{2} \quad \text{as } \alpha \rightarrow \infty. \end{aligned}$$