

Stochastic Processes: Homework 2

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February 22, 2024

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Exercise 1

Let $f(x) = \frac{1}{2} \sin(x) \mathbb{1}_{[0, \pi]}(x)$.

1. For $n = 4$ **calculate** the first moment of the order statistic $X_{i:n}$ for every $i \in 1, \dots, 4$.
2. **Sketch a telling drawing** with the original density, the densities $f_{i:4}$ and $\mathbf{E}[X_{i:4}]$

Solution Part 1

In the first place, for $x \in [0, \pi]$

$$\begin{aligned} F(x) &= \frac{1}{2} \int_{-\infty}^x \sin(t) \mathbb{1}(t) dt = \frac{1}{2} \int_0^x \sin(t) dt \\ &= \cos(x) \Big|_0^x = \frac{1}{2} (1 - \cos(x)) \cdot \mathbb{1}_{[0, \pi]}(x) \end{aligned}$$

Thus,

$$\overline{F}(x) = \frac{1}{2} (1 + \cos(x)) \cdot \mathbb{1}_{[0, \pi]}(x),$$

and,

$$\begin{aligned} f_{i:n}(x) &= i \cdot \binom{n}{i} f(x) \cdot F^{i-1}(x) \cdot \overline{F}^{n-i}(x) \\ &= i \cdot \binom{n}{i} \frac{1}{2^n} \cdot \sin(x) \cdot (1 - \cos(x))^{i-1} \cdot (1 + \cos(x))^{n-i}. \end{aligned}$$

For simplicity, define $C_{i:n} = i \binom{n}{i} 2^{-n}$. Now let $n = 4$. In order to calculate the exact expected value formula, we must use the following angle identities,

$$\sin(x) \cos(x) = \frac{1}{2} \sin(2x), \quad \cos^2(x) = \frac{1}{2} (1 + \cos(2x)),$$

and the result of these integrals

$$\int x \sin(kx) dx = \frac{\sin(kx) - kx \cos(kx)}{k^2}.$$

From now on I'm going to simplify the trigonometric expressions using the TR8 algorithm provided by the [package sympy](#). Also, use the package to integrate and evaluate the final expression. As always, the code is included with this document.

• $i = 1$:

$$C_{1:4}^{-1} \cdot f_{1:4} = \frac{7 \sin(x)}{4} + \frac{7 \sin(2x)}{4} + \frac{3 \sin(3x)}{4} + \frac{\sin(4x)}{8},$$

$$\begin{aligned} C_{1:4}^{-1} \cdot \mathbf{E}[X_{1:4}] &= C_{1:4}^{-1} \cdot \int_0^\pi x f_{1:4} dx. \\ &= -\frac{7x \cos(x)}{4} - \frac{7x \cos(2x)}{8} - \frac{x \cos(3x)}{4} - \frac{x \cos(4x)}{32} \Big|_0^\pi \\ &\quad + \frac{7 \sin(x)}{4} + \frac{7 \sin(2x)}{16} + \frac{\sin(3x)}{12} + \frac{\sin(4x)}{128} \Big|_0^\pi = \frac{35\pi}{32}. \end{aligned}$$

• $i = 2$:

$$C_{2:4}^{-1} \cdot f_{2:4} = \frac{3 \sin(x)}{4} + \frac{\sin(2x)}{4} - \frac{\sin(3x)}{4} - \frac{\sin(4x)}{8},$$

$$\begin{aligned} C_{2:4}^{-1} \cdot \mathbf{E}[X_{2:4}] &= C_{2:4}^{-1} \cdot \int_0^\pi x f_{2:4} dx. \\ &= -\frac{3x \cos(x)}{4} - \frac{x \cos(2x)}{8} + \frac{x \cos(3x)}{12} + \frac{x \cos(4x)}{32} \Big|_0^\pi \\ &\quad + \frac{3 \sin(x)}{4} + \frac{\sin(2x)}{16} - \frac{\sin(3x)}{36} - \frac{\sin(4x)}{128} \Big|_0^\pi = \frac{55\pi}{96}. \end{aligned}$$

• $i = 3$:

$$C_{3:4}^{-1} \cdot f_{3:4} = \frac{3 \sin(x)}{4} - \frac{\sin(2x)}{4} - \frac{\sin(3x)}{4} + \frac{\sin(4x)}{8},$$

$$\begin{aligned} C_{3:4}^{-1} \cdot \mathbf{E}[X_{3:4}] &= C_{3:4}^{-1} \cdot \int_0^\pi x f_{3:4} dx. \\ &= -\frac{3x \cos(x)}{4} + \frac{x \cos(2x)}{8} + \frac{x \cos(3x)}{12} - \frac{x \cos(4x)}{32} \Big|_0^\pi \\ &\quad + \frac{3 \sin(x)}{4} - \frac{\sin(2x)}{16} - \frac{\sin(3x)}{36} + \frac{\sin(4x)}{128} \Big|_0^\pi = \frac{73\pi}{96}. \end{aligned}$$

• $i = 4$:

$$C_{4:4}^{-1} \cdot f_{4:4} = \frac{7 \sin(x)}{4} - \frac{7 \sin(2x)}{4} + \frac{3 \sin(3x)}{4} - \frac{\sin(4x)}{8},$$

$$\begin{aligned} C_{4:4}^{-1} \cdot \mathbf{E}[X_{4:4}] &= C_{4:4}^{-1} \cdot \int_0^\pi x f_{4:4} dx. \\ &= \left. \frac{7x \cos(x)}{4} + \frac{7x \cos(2x)}{8} - \frac{x \cos(3x)}{4} + \frac{x \cos(4x)}{32} \right|_0^\pi \\ &\quad - \left. \frac{7 \sin(x)}{4} - \frac{7 \sin(2x)}{16} + \frac{\sin(3x)}{12} - \frac{\sin(4x)}{128} \right|_0^\pi = \frac{93\pi}{32}. \end{aligned}$$

Therefore,

$$\begin{aligned} \mathbf{E}[X_{1:4}] &= \frac{35\pi}{128}, & \mathbf{E}[X_{2:4}] &= \frac{55\pi}{128}, \\ \mathbf{E}[X_{3:4}] &= \frac{73\pi}{128}, & \mathbf{E}[X_{4:4}] &= \frac{93\pi}{128}. \end{aligned}$$

Solution Part 2

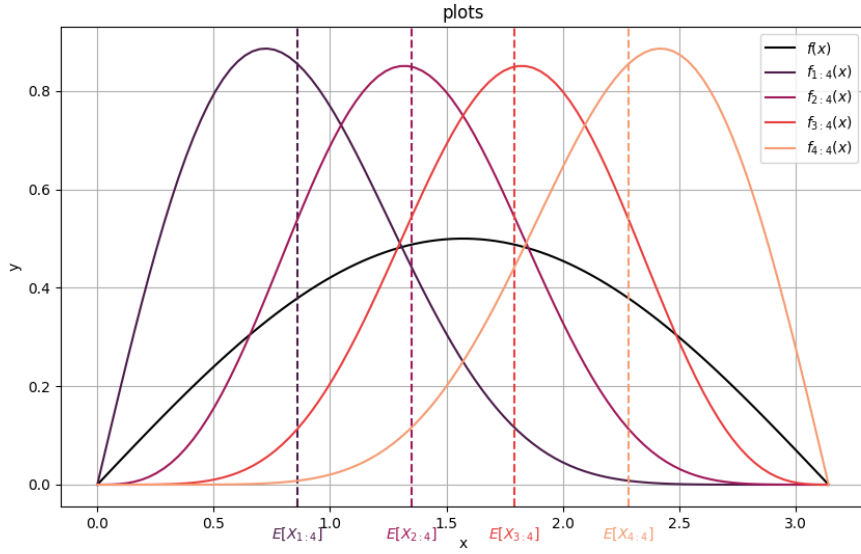


Figure 1: This is a floating figure with an image.

Exercise 2

1. **Formulate** the criterion that characterizes the existence of an extreme distribution (in terms of \overline{F}).
2. **Determine and justify** by this criterion if the following distribution has an extreme distribution. The cumulative distribution function is given by

$$F(n) := 1 - \frac{C}{(n+1)^{\ln(n+1)}}, \quad n \in \mathbb{N}.$$

3. In the case that it has a limit distribution, **argue** about which should be the limit distribution.

Solution Part 1

The criterion states that for every distribution F that satisfies the following condition:

$$\lim_{x \rightarrow x_F^-} \frac{\overline{F}(x)}{\overline{F}(x-)} = 1,$$

then, there exist sequences d_n and c_n such that, for the extreme $M_n = X_{n:n}$,

$$\frac{M_n - d_n}{c_n} \xrightarrow{d} H,$$

for a non-degenerate distribution H . Furthermore, for the discrete case, the condition would be that

$$\lim_{n \rightarrow \infty} \frac{\overline{F}(n)}{\overline{F}(n-1)} = 1,$$

which is equivalent to

$$\lim_{n \rightarrow \infty} \frac{f(n)}{\overline{F}(n)} = 0.$$

Solution Part 2

$$\overline{F}(n) = \frac{C}{(n+1)^{\ln(n+1)}}.$$

$$\begin{aligned}
\ln \left(\lim_{n \rightarrow \infty} \frac{\overline{F}(n)}{\overline{F}(n-1)} \right) &= \lim_{n \rightarrow \infty} (\ln \overline{F}(n) - \ln \overline{F}(n-1)) \\
&= \lim_{x \rightarrow \infty} \ln^2(x+1) - \ln^2(x) \\
&= \lim_{x \rightarrow \infty} (\ln(x+1) + \ln(x)) \cdot (\ln(x+1) - \ln(x)) \\
&= \lim_{x \rightarrow \infty} \frac{\ln \frac{x+1}{x}}{\ln^{-1}(x^2+x)} \\
(\text{L'Hôpital}) &= \lim_{x \rightarrow \infty} \frac{-(x^2+x)^{-1}}{-(2x+1)(x^2+x)^{-1} \ln^{-2}(x^2+x)} \\
&= \lim_{x \rightarrow \infty} \frac{\ln^2(x^2+x)}{2x+1} \\
(\text{L'Hôpital}) &= \lim_{x \rightarrow \infty} \frac{(2x+1)(x^2+x)^{-1}}{2} \\
&= \lim_{x \rightarrow \infty} \frac{2x+1}{2(x^2+x)} = 0.
\end{aligned}$$

This implies that

$$\lim_{n \rightarrow \infty} \frac{\overline{F}(n)}{\overline{F}(n-1)} = \exp \left[\ln \left(\lim_{n \rightarrow \infty} \frac{\overline{F}(n)}{\overline{F}(n-1)} \right) \right] = e^0 = 1.$$

Solution Part 3

It's clear that F is not in the domain of attraction of the Weibull since its extreme isn't bounded. In order to see it cannot be either Fréchet, let's take a look to the continuous case,

$$\overline{G}(x) = \frac{C}{x^{\ln(x)}} \cdot \mathbb{1}_{[1, \infty)}(x) = e^{\ln^2(x)} \mathbb{1}_{[1, \infty)}(x)$$

$$\begin{aligned}
\ln \left(\lim_{x \rightarrow \infty} \frac{\overline{G}(\lambda \cdot x)}{\overline{G}(x)} \right) &= \lim_{x \rightarrow \infty} \ln^2(\lambda x) - \ln^2(x) \\
&= \lim_{x \rightarrow \infty} (\ln(\lambda x) + \ln(x)) \cdot (\ln(\lambda x) - \ln(x)) \\
&= \lim_{x \rightarrow \infty} \ln(\lambda x^2) \ln(\lambda) \\
&= \infty.
\end{aligned}$$

Thus, it's clear that $G \in DAM(\Lambda)$ since it's not regular. Finally, note that since $F(n) = G(\lceil x \rceil)$ for $x \in (n-1, n]$, it follows that F and G have equivalent tails, and thus, F inherits the same domain of attraction that G has.

Exercise 3

Let $(X_i)_{i \in \mathbb{N}}$ be a sequence of i.i.d. random variables with $X_1 \sim \mu$,

$$M_n := \max \{X_1, \dots, X_n\}, \quad \text{and} \quad N_n := \min \{X_1, \dots, X_n\}.$$

1. For $\mu = \text{Gamma}(\alpha, \beta)$, $\alpha, \beta > 0$, **determine and justify** the extreme distribution of (M_n) and (N_n) .
2. For $\mu = \text{Beta}(\alpha, \beta)$, $\alpha, \beta > 0$, that is $f(x) = C_{\alpha, \beta} x^{\alpha-1} (1-x)^{\beta-1}$, $\alpha, \beta > 0$, **determine and justify** the extreme distribution of (M_n) and (N_n) .
3. For μ such that for $\alpha > 0$,

$$F(x) = \begin{cases} 0, & \text{for } x < 1, \\ \frac{\ln(x)}{x^\alpha}, & \text{for } x \geq 1. \end{cases}$$

determine and justify the extreme distribution of (M_n) and (N_n)

Solution Part 1

The gamma distribution with parameters $\alpha, \beta > 0$ has the following density function:

$$f(x) = \frac{\beta^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\beta x}, \quad x \in [0, \infty).$$

It's clear that the Gamma distribution has extreme distribution convergence since it's absolutely continuous. The extreme distribution cannot be Weibull because the population maximum $x_F = \infty$. Now, we are going to prove that the tails are not regular, and thus, the extremes do not converge to Fréchet (using L'Hôpital rule):

$$\begin{aligned} \lim_{x \rightarrow \infty} \frac{\overline{F}(\lambda \cdot x)}{\overline{F}(x)} &= \lim_{x \rightarrow \infty} \frac{\lambda \cdot f(\lambda x)}{f(x)} \\ &= \lim_{x \rightarrow \infty} \frac{\lambda^\alpha x^{\alpha-1} e^{-\beta \lambda x}}{x^{\alpha-1} e^{-\beta x}} \\ &= \lim_{x \rightarrow \infty} \lambda^\alpha \exp(-\beta x(\lambda - 1)) \\ &= 0, \quad \forall \lambda > 1. \end{aligned}$$

Therefore, the only option left for M_n is to be in the Gumbell's domain of attraction, that is $F \in DAM(\Lambda)$.

Now, for N_n , we are going to use the von Mises condition to prove that F is in the domain of attraction of the Weibull distribution. In the first place, note that the population minimum x_F equals 0 and we are approaching it from the right. Thus, after reflecting the whole distribution over the y axis, we obtain

$$\begin{aligned} f'(x) &= (\alpha - 1)x^{\alpha-2}e^{-\beta x^2} - x^{\alpha-1} \cdot 2\beta x e^{-\beta x^2} \\ &= x^{\alpha-2}e^{-\beta x^2}((\alpha - 1) - 2\beta x^2) \end{aligned}$$

$$\begin{aligned} \lim_{x \rightarrow 0^-} \frac{xf(x)}{F(x)} &= \lim_{x \rightarrow 0^-} \frac{xf'(x) + f(x)}{f(x)} \\ &= \lim_{x \rightarrow 0^-} \frac{x^{\alpha-1}e^{-\beta x^2}((\alpha - 1) - 2\beta x^2)}{x^{\alpha-1}e^{-\beta x^2}} + \frac{f(x)}{f(x)} \\ &= \lim_{x \rightarrow 0^-} (\alpha - 1) - 2\beta x^2 + 1 = \alpha. \end{aligned}$$

Thus, $\bar{F}(-\cdot) \in DAM(\Psi_\alpha)$.

Solution Part 2

Both M_n and N_n will be in the domain of attraction of Weibull. For M_n the von Mises condition is satisfied on the following limit:

$$\begin{aligned} f'(x) &= (\alpha - 1)x^{\alpha-2}(1 - x)^{\beta-1} - (\beta - 1)x^{\alpha-1}(1 - x)^{\beta-2} \\ &= x^{\alpha-2}(1 - x)^{\beta-2}((\alpha - 1)(1 - x) - (\beta - 1)x) \end{aligned}$$

$$\begin{aligned}
\lim_{x \rightarrow 1^+} \frac{(1-x)f(x)}{\bar{F}(x)} &= \lim_{x \rightarrow 1^+} \frac{(1-x)f'(x) - f(x)}{-f(x)} \\
&= \lim_{x \rightarrow 1^+} \frac{(1-x) \cdot x^{\alpha-2}(1-x)^{\beta-2}((\alpha-1)(1-x) - (\beta-1)x)}{-x^{\alpha-1}(1-x)^{\beta-1}} + 1 \\
&= \lim_{x \rightarrow 1^+} -(\alpha-1) \underbrace{\frac{1-x}{x}}_{\rightarrow 0} + (\beta-1) + 1 \\
&= \beta - 1 + 1 = \beta.
\end{aligned}$$

Now, for N_n , after reflecting the whole distribution, we obtain:

$$\begin{aligned}
\lim_{x \rightarrow 0^-} \frac{xf(x)}{F(x)} &= \lim_{x \rightarrow 0^-} \frac{xf'(x) + f(x)}{f(x)} \\
&= \lim_{x \rightarrow 0^-} \frac{x \cdot x^{\alpha-2}(1-x)^{\beta-2}((\alpha-1)(1-x) - (\beta-1)x)}{x^{\alpha-1}(1-x)^{\beta-1}} + 1 \\
&= \lim_{x \rightarrow 0^-} (\alpha-1) - (\beta-1) \underbrace{\frac{x}{1-x}}_{\rightarrow 0} + 1 \\
&= \alpha - 1 + 1 = \alpha.
\end{aligned}$$

For M_n is Weibull with parameter β and for N_n is Weibull with parameter α .

$$F \in DAM(\Psi_\beta), \quad \bar{F}(-\cdot) \in DAM(\Psi_\alpha).$$

Solution Part 3

The provided distribution function cannot be a distribution function since is not monotonically increasing, as in fact, it goes to 0 when $x \rightarrow \infty$. I'm going to assume that it's the density function, but even then, we would encounter some problems. After some tinkering, I believe the correct density and distribution functions should be

$$f(x) = \alpha^2 \frac{\ln(x)}{x^{\alpha+1}} \mathbb{1}_{[1, \infty)}(x),$$

$$F(x) = 1 - \frac{\alpha \ln(x) + 1}{x^\alpha} \mathbb{1}_{[1, \infty)}(x).$$

For M_n we are using the von Mises condition to prove it is in the domain of attraction of Fréchet:

$$\begin{aligned}
\lim_{x \rightarrow \infty} \frac{xf(x)}{\overline{F}(x)} &= \lim_{x \rightarrow \infty} \alpha^2 \frac{\ln(x)}{x^\alpha} \cdot \frac{x^\alpha}{\alpha \ln(x) + 1} \\
&= \lim_{x \rightarrow \infty} \frac{\alpha^2 \ln(x)}{\alpha \ln(x) + 1} \\
(\text{L'Hôpital}) &= \lim_{x \rightarrow \infty} \frac{\alpha^2 x^{-1}}{\alpha x^{-1}} \\
&= \alpha > 0.
\end{aligned}$$

Thus, $F \in DAM(\Phi_\alpha)$. Now, for N_n we are going to use the von Mises condition for the Weibull domain of attraction. After re-centering and reflecting over the y axis we obtain,

$$\begin{aligned}
\lim_{x \rightarrow 1-} \frac{(x-1)f(x)}{F(x)} &= \lim_{x \rightarrow 1-} \frac{\alpha^2(x-1)\ln(x)}{x^{\alpha+1}} \cdot \frac{x^\alpha}{x^\alpha - \alpha \ln(x) + 1} \\
&= \lim_{x \rightarrow 1-} \frac{1}{x} \cdot \frac{\alpha^2 \ln(x)x - \ln(x)}{x^\alpha - \alpha \ln(x) - 1} \\
&= \frac{1}{1} \cdot \frac{0}{1-1}.
\end{aligned}$$

After applying L'Hôpital rule,

$$f'(x) = \frac{\alpha^2(1 - (a+1)\ln(x))}{x^{\alpha+2}}$$

$$\begin{aligned}
\lim_{x \rightarrow 1-} \frac{(x-1)f(x)}{F(x)} &= \lim_{x \rightarrow 1-} \frac{(x-1)f'(x) + f(x)}{f(x)} \\
&= \lim_{x \rightarrow 1-} \frac{(x-1)\alpha^2(1 - (\alpha+1)\ln(x))}{x^{\alpha+2}} \cdot \frac{x^{\alpha+1}}{\alpha^2 \ln(x)} + 1 \\
&= \lim_{x \rightarrow 1-} \frac{x-1}{x} \cdot \frac{1 - (\alpha+1)\ln(x)}{\ln(x)} + 1 \\
&= \lim_{x \rightarrow 1-} \frac{(x-1)}{x \ln(x)} - (\alpha+1) \frac{(x-1)\ln(x)}{x \ln(x)} + 1 \\
(\text{L'Hôpital}) &= \lim_{x \rightarrow 1-} \frac{1}{\ln(x) + 1} - (\alpha+1) \frac{1 + \ln(x) - x^{-1}}{\ln(x) + 1} + 1 \\
&= \frac{1}{0+1} - (\alpha+1) \frac{1+0-1}{0+1} + 1 \\
&= 2.
\end{aligned}$$

In the end we obtained that N_n is in the domain of attraction of the Weibull, that is $\overline{F}(-\cdot) \in DAM(\Psi_2)$