

MATH68181: EXTREME VALUES FIRST SEMESTER ANSWERS TO IN CLASS TEST

ANSWER TO QUESTION 1 If there are norming constants $a_n > 0$, b_n and a nondegenerate G such that the cdf of a normalized version of M_n converges to G , i.e.

$$\Pr \left(\frac{M_n - b_n}{a_n} \leq x \right) = F^n(a_n x + b_n) \rightarrow G(x) \quad (1)$$

as $n \rightarrow \infty$ then G must be of the same type as (cdfs G and G^* are of the same type if $G^*(x) = G(ax + b)$ for some $a > 0$, b and all x) as one of the following three classes:

$$\begin{aligned} I & : \Lambda(x) = \exp \{-\exp(-x)\}, \quad x \in \mathbb{R}; \\ II & : \Phi_\alpha(x) = \begin{cases} 0 & \text{if } x < 0, \\ \exp \{-x^{-\alpha}\} & \text{if } x \geq 0 \end{cases} \\ & \text{for some } \alpha > 0; \\ III & : \Psi_\alpha(x) = \begin{cases} \exp \{-(-x)^\alpha\} & \text{if } x < 0, \\ 1 & \text{if } x \geq 0 \end{cases} \\ & \text{for some } \alpha > 0. \end{aligned}$$

The necessary and sufficient conditions for the three extreme value distributions are:

$$\begin{aligned} I & : \exists \gamma(t) > 0 \text{ s.t. } \lim_{t \uparrow w(F)} \frac{1 - F(t + x\gamma(t))}{1 - F(t)} = \exp(-x), \quad x \in \mathbb{R}, \\ II & : w(F) = \infty \text{ and } \lim_{t \uparrow \infty} \frac{1 - F(tx)}{1 - F(t)} = x^{-\alpha}, \quad x > 0, \\ III & : w(F) < \infty \text{ and } \lim_{t \downarrow 0} \frac{1 - F(w(F) - tx)}{1 - F(w(F) - t)} = x^\alpha, \quad x > 0. \end{aligned}$$

First, suppose that G belongs to the max domain of attraction of the Gumbel extreme value distribution. Then, there must exist a strictly positive function, say $h(t)$, such that

$$\lim_{t \rightarrow w(G)} \frac{1 - G(t + xh(t))}{1 - G(t)} = \exp(-x)$$

for every $x \in (-\infty, \infty)$. But, using L'Hopital's rule, we note that

$$\begin{aligned} & \lim_{t \rightarrow w(F)} \frac{1 - F(t + xh(t))}{1 - F(t)} \\ &= \lim_{t \rightarrow w(F)} \frac{(1 + xh'(t))f(t + xh(t))}{f(t)} \\ &= \lim_{t \rightarrow w(G)} \frac{(1 + xh'(t))\{-\log[1 - G(t + xh(t))]\}^2 g(t + xh(t))}{-\log[1 - G(t)]^2 g(t)} \end{aligned}$$

$$\begin{aligned}
&= \lim_{t \rightarrow w(G)} \frac{(1 + xh'(t))g(t + xh(t))}{g(t)} \left\{ \frac{\log[1 - G(t + xh(t))]}{\log[1 - G(t)]} \right\}^2 \\
&= \lim_{t \rightarrow w(G)} \frac{(1 + xh'(t))g(t + xh(t))}{g(t)} \left\{ \frac{1 - G(t)}{1 - G(t + xh(t))} \frac{(1 + xh'(t))g(t + xh(t))}{g(t)} \right\}^2 \\
&= \lim_{t \rightarrow w(G)} \frac{(1 + xh'(t))g(t + xh(t))}{g(t)} \left\{ \frac{g(t)}{(1 + xh'(t))g(t + xh(t))} \frac{(1 + xh'(t))g(t + xh(t))}{g(t)} \right\}^2 \\
&= \lim_{t \rightarrow w(G)} \frac{(1 + xh'(t))g(t + xh(t))}{g(t)} \\
&= \lim_{t \rightarrow w(G)} \frac{1 - G(t + xh(t))}{1 - G(t)} \\
&= \exp(-x)
\end{aligned}$$

for every $x \in (-\infty, \infty)$. So, it follows that F also belongs to the max domain of attraction of the Gumbel extreme value distribution with

$$\lim_{n \rightarrow \infty} \Pr \{a_n(M_n - b_n) \leq x\} = \exp\{-\exp(-x)\}$$

for some suitable norming constants $a_n > 0$ and b_n .

Second, suppose that G belongs to the max domain of attraction of the Fréchet extreme value distribution. Then, there must exist a $\beta > 0$, such that

$$\lim_{t \rightarrow \infty} \frac{1 - G(tx)}{1 - G(t)} = x^{-\beta}$$

for every $x > 0$. But, using L'Hopital's rule, we note that

$$\begin{aligned}
\lim_{t \rightarrow \infty} \frac{1 - F(tx)}{1 - F(t)} &= \lim_{t \rightarrow \infty} \frac{xf(tx)}{f(t)} \\
&= \lim_{t \rightarrow \infty} \frac{x\{-\log[1 - G(tx)]\}^2 g(tx)}{\{-\log[1 - G(t)]\}^2 g(t)} \\
&= \lim_{t \rightarrow \infty} \frac{xg(tx)}{g(t)} \left\{ \frac{\log[1 - G(tx)]}{\log[1 - G(t)]} \right\}^2 \\
&= \lim_{t \rightarrow \infty} \frac{xg(tx)}{g(t)} \left\{ \frac{1 - G(t)}{1 - G(tx)} \frac{xg(tx)}{g(t)} \right\}^2 \\
&= \lim_{t \rightarrow \infty} \frac{xg(tx)}{g(t)} \left\{ \frac{g(t)}{xg(tx)} \frac{xg(tx)}{g(t)} \right\}^2 \\
&= \lim_{t \rightarrow \infty} \frac{xg(tx)}{g(t)} \\
&= \lim_{t \rightarrow \infty} \frac{1 - G(tx)}{1 - G(t)} \\
&= x^\beta
\end{aligned}$$

for every $x > 0$. So, it follows that F also belongs to the max domain of attraction of the Fréchet extreme value distribution with

$$\lim_{n \rightarrow \infty} \Pr \{a_n(M_n - b_n) \leq x\} = \exp(-x^{-\beta})$$

for some suitable norming constants $a_n > 0$ and b_n .

Third, suppose that G belongs to the max domain of attraction of the Weibull extreme value distribution. Then, there must exist a $\alpha > 0$, such that

$$\lim_{t \rightarrow 0} \frac{1 - G(w(G) - tx)}{1 - G(w(G) - t)} = x^\alpha$$

for every $x > 0$. But, using L'Hopital's rule, we note that

$$\begin{aligned} \lim_{t \rightarrow 0} \frac{1 - F(w(F) - tx)}{1 - F(w(F) - t)} &= \lim_{t \rightarrow 0} \frac{xf(w(F) - tx)}{f(w(F) - t)} \\ &= \lim_{t \rightarrow 0} \frac{x \{-\log[1 - G(w(F) - tx)]\}^{a-1} g(w(F) - tx)}{\{-\log[1 - G(w(F) - tx)]\}^2 g(w(F) - t)} \\ &= \lim_{t \rightarrow 0} \frac{xg(w(F) - tx)}{g(w(F) - t)} \left\{ \frac{\log[1 - G(w(F) - tx)]}{\log[1 - G(w(F) - t)]} \right\}^2 \\ &= \lim_{t \rightarrow 0} \frac{xg(w(F) - tx)}{g(w(F) - t)} \left\{ \frac{1 - G(w(F) - t)}{1 - G(w(F) - tx)} \frac{xg(w(F) - tx)}{g(w(F) - t)} \right\}^2 \\ &= \lim_{t \rightarrow 0} \frac{xg(w(F) - tx)}{g(w(F) - t)} \left\{ \frac{g(w(F) - t)}{xg(w(F) - tx)} \frac{xg(w(F) - tx)}{g(w(F) - t)} \right\}^2 \\ &= \lim_{t \rightarrow 0} \frac{xg(w(F) - tx)}{g(w(F) - t)} \\ &= \lim_{t \rightarrow 0} \frac{1 - G(w(F) - tx)}{1 - G(w(F) - t)} \\ &= x^\alpha. \end{aligned}$$

So, it follows that F also belongs to the max domain of attraction of the Weibull extreme value distribution with

$$\lim_{n \rightarrow \infty} \Pr \{a_n(M_n - b_n) \leq x\} = \exp\{-(-x)^\alpha\}$$

for some suitable norming constants $a_n > 0$ and b_n .

ANSWER TO QUESTION 2 i) Note that $w(F) = \infty$ and take $\gamma(t) = 1$. Then

$$\lim_{t \uparrow \infty} \frac{1 - F(t + x)}{1 - F(t)} = \lim_{t \uparrow \infty} \frac{1 - [1 - \exp(-t - x)]^2}{1 - [1 - \exp(-t)]^2} = \lim_{t \uparrow \infty} \frac{2 \exp(-t - x)}{2 \exp(-t)} = \exp(-x).$$

. So, the exponentiated exponential cdf $F(x) = [1 - \exp(-x)]^2$ belongs to the Gumbel domain of attraction.

ii) Note that $w(F) = \infty$. Then

$$\lim_{t \rightarrow \infty} \frac{1 - F(t + x\gamma(t))}{1 - F(t)} = \lim_{t \rightarrow \infty} \frac{1 - \frac{\{1 - \exp(-2(t + x\gamma(t)))\}^2}{0.5 + 0.5\{1 - \exp(-2(t + x\gamma(t)))\}^2}}{1 - \frac{\{1 - \exp(-2t)\}^2}{0.5 + 0.5\{1 - \exp(-2t)\}^2}}$$

$$\begin{aligned}
&= \lim_{t \rightarrow \infty} \frac{1 - \frac{\{1 - \exp(-2(t + x\gamma(t)))\}^2}{0.5 + 0.5}}{1 - \frac{\{1 - \exp(-2t)\}^2}{0.5 + 0.5}} \\
&= \lim_{t \rightarrow \infty} \frac{1 - \{1 - \exp(-2(t + x\gamma(t)))\}^2}{1 - \{1 - \exp(-2t)\}^2} \\
&= \lim_{t \rightarrow \infty} \frac{1 - \{1 - 2\exp(-2(t + x\gamma(t)))\}}{1 - \{1 - 2\exp(-2t)\}} \\
&= \lim_{t \rightarrow \infty} \frac{2\exp(-2(t + x\gamma(t)))}{2\exp(-2t)} \\
&= \lim_{t \rightarrow \infty} \exp(-2x\gamma(t)) \\
&= \exp(-x)
\end{aligned}$$

if $\gamma(t) = 0.5$. So, the exponentiated exponential geometric cdf $F(x) = \frac{\{1-\exp(-2x)\}^2}{1-0.5+0.5\{1-\exp(-2x)\}^2}$ belongs to the Gumbel domain of attraction.

iii) Note that $w(F) = \infty$. Then

$$\begin{aligned}
\lim_{t \rightarrow \infty} \frac{1 - F(t + x\gamma(t))}{1 - F(t)} &= \lim_{t \rightarrow \infty} \frac{1 - \left\{ 1 - \frac{(0.5)^2 \exp(-4(t + x\gamma(t)))}{[1 - 0.5 \exp(-2(t + x\gamma(t)))]^2} \right\}}{1 - \left\{ 1 - \frac{(0.5)^2 \exp(-4t)}{[1 - 0.5 \exp(-2t)]^2} \right\}} \\
&= \lim_{t \rightarrow \infty} \frac{\frac{(0.5)^2 \exp(-4(t + x\gamma(t)))}{[1 - 0.5 \exp(-2(t + x\gamma(t)))]^2}}{\frac{(0.5)^2 \exp(-4t)}{[1 - 0.5 \exp(-2t)]^2}} \\
&= \lim_{t \rightarrow \infty} \frac{\frac{\exp(-4(t + x\gamma(t)))}{[1 - 0.5 \exp(-2(t + x\gamma(t)))]^2}}{\frac{\exp(-4t)}{[1 - 0.5 \exp(-2t)]^2}} \\
&= \lim_{t \rightarrow \infty} \frac{\frac{\exp(-4x\gamma(t))}{[1 - 0.5 \exp(-2(t + x\gamma(t)))]^2}}{\frac{1}{[1 - 0.5 \exp(-2t)]^2}} \\
&= \lim_{t \rightarrow \infty} \frac{\frac{\exp(-4x\gamma(t))}{[1 - 0]^2}}{\frac{1}{[1 - 0]^2}} \\
&= \lim_{t \rightarrow \infty} \exp(-4x\gamma(t)) \\
&= \exp(-x)
\end{aligned}$$

if $\gamma(t) = 1/4$. So, the exponential-negative binomial distribution cdf $F(x) = 1 - \frac{(1-0.5)^2 \exp(-4x)}{[1-0.5 \exp(-2x)]^2}$ belongs to the Gumbel domain of attraction.

iv) For the degenerate distribution,

$$p(k) = \begin{cases} 1, & \text{if } k = 2, \\ 0, & \text{if } k \neq 2. \end{cases}$$

So,

$$F(k) = \begin{cases} 1, & \text{if } k \geq 2, \\ 0, & \text{if } k < 2, \end{cases}$$

and

$$\frac{\Pr(X = k)}{1 - F(k-1)} = \begin{cases} 1/1, & \text{if } k = 2, \\ 0/1, & \text{if } 2 < k < 3, \\ 0/0, & \text{if } k \geq 3, \\ 0/1, & \text{if } k < 2. \end{cases}$$

Hence, there can be no sequences $a_n > 0$ and b_n such that $(M_n - b_n)/a_n$ has a non-degenerate limiting distribution.

v) For the Poisson distribution,

$$\begin{aligned} \frac{\Pr(X = k)}{1 - F(k-1)} &= \frac{2^k/k!}{\sum_{j=k}^{\infty} 2^j/j!} \\ &= \frac{1}{1 + \sum_{j=k+1}^{\infty} k!2^{j-k}/j!}. \end{aligned}$$

The term in the denominator of the last term can be rewritten as

$$\sum_{j=1}^{\infty} \frac{2^j}{(k+1)(k+2)\cdots(k+j)} \leq \sum_{j=1}^{\infty} \left(\frac{2}{k}\right)^j = \frac{2/k}{1 - 2/k}$$

(when $k > 2$) and the bound tends to 0 as $k \rightarrow \infty$ and so it follows that $p_k/(1 - F(k-1)) \rightarrow 1$. Hence, there can be no sequences $a_n > 0$ and b_n such that $(M_n - b_n)/a_n$ has a non-degenerate limiting distribution.

ANSWER TO QUESTION 3 i) The cdf of $M = \max(X, Y, Z)$ is

$$\begin{aligned} F_M(m) &= \Pr(X < m, Y < m, Z < m) \\ &= 1 - \Pr(X > m) - \Pr(Y > m) - \Pr(Z > m) + \Pr(X > m, Y > m) \\ &\quad + \Pr(X > m, Z > m) + \Pr(Y > m, Z > m) - \Pr(X > m, Y > m, Z > m) \\ &= 1 - \left[1 + \frac{m}{2}\right]^{-2} - \left[1 + \frac{m}{2}\right]^{-2} - \left[1 + \frac{m}{2}\right]^{-2} + [1+m]^{-2} \\ &\quad + [1+m]^{-2} + [1+m]^{-2} - \left[1 + \frac{3m}{2}\right]^{-2}. \end{aligned}$$

ii) Differentiating $F_M(m)$ with respect to m gives the pdf of M as

$$\begin{aligned} f_M(m) &= \left[1 + \frac{m}{2}\right]^{-3} + \left[1 + \frac{m}{2}\right]^{-3} + \left[1 + \frac{m}{2}\right]^{-3} - 2[1+m]^{-3} \\ &\quad - 2[1+m]^{-3} - 2[1+m]^{-3} \\ &\quad + 3\left[1 + \frac{3m}{2}\right]^{-3}. \end{aligned}$$

iii) The n th moment of M can be calculated as

$$\begin{aligned}
E(M^n) &= \int_0^\infty m^n \left[1 + \frac{m}{2}\right]^{-3} dm + \int_0^\infty m^n \left[1 + \frac{m}{2}\right]^{-3} dm \\
&\quad + \int_0^\infty m^n \left[1 + \frac{m}{2}\right]^{-3} dm - 2 \int_0^\infty m^n [1+m]^{-3} dm \\
&\quad - 2 \int_0^\infty m^n [1+m]^{-3} dm \\
&\quad - 2 \int_0^\infty m^n [1+m]^{-3} dm \\
&\quad + 3 \int_0^\infty m^n \left[1 + \frac{3m}{2}\right]^{-3} dm \\
&= 2^{n+1} \int_0^1 x^{1-n} (1-x)^n dx + 2^{n+1} \int_0^1 x^{1-n} (1-x)^n dx \\
&\quad + 22^{n+1} \int_0^1 x^{1-n} (1-x)^n dx - \int_0^1 x^{1-n} (1-x)^n dx \\
&\quad - 2 \int_0^1 x^{1-n} (1-x)^n dx \\
&\quad - 2 \int_0^1 x^{1-n} (1-x)^n dx \\
&\quad + 3 \left(\frac{3}{2}\right)^{-n-1} \int_0^1 x^{1-n} (1-x)^n dx \\
&= 22^n B(2-n, n+1) + 22^n B(2-n, n+1) \\
&\quad + 22^n B(2-n, n+1) - 2B(2-n, n+1) \\
&\quad - 2B(2-n, n+1) \\
&\quad - 2B(2-n, n+1) \\
&\quad + 2 \left(\frac{3}{2}\right)^{-n} B(2-n, n+1).
\end{aligned}$$

iv) The cdf of $L = \min(X, Y, Z)$ is

$$\begin{aligned}
F_L(l) &= 1 - \Pr(L > l) \\
&= 1 - \Pr(X > l, Y > l, Z > l) \\
&= 1 - \left[1 + \frac{3l}{2}\right]^{-2}.
\end{aligned}$$

v) Differentiating $F_L(l)$ with respect to l gives the pdf of L as

$$f_L(l) = 2 \left(\frac{3}{2}\right) \left[1 + \frac{3l}{2}\right]^{-3}.$$

vi) The n th moment of L can be calculated as

$$\begin{aligned}
E(L^n) &= 2 \left(\frac{3}{2}\right) \int_0^\infty l^n \left[1 + \frac{3l}{2}\right]^{-3} dl \\
&= 2 \left(\frac{3}{2}\right)^{-n} B(2-n, n+1).
\end{aligned}$$