

# ONLINE SUPPLEMENT

## Spectral backtests unbounded and folded\*

Michael B. Gordy

Federal Reserve Board, Washington DC

Alexander J. McNeil

School for Business and Society, University of York

15 July 2024

### 1 Standardizing the Fernández-Steel t distribution

Recall the density of the Fernández-Steel t (FS-t) distribution:

$$f_{FS}(x; \gamma, \zeta) = \frac{2\gamma}{\sigma(1 + \gamma^2)} \begin{cases} f_t\left(\frac{\gamma(x-\mu)}{\sigma}; \zeta\right) & x \leq \mu \\ f_t\left(\frac{x-\mu}{\gamma\sigma}; \zeta\right) & x > \mu \end{cases} \quad (\text{OS.1})$$

for location parameter  $\mu$  and scale parameter  $\sigma$  and where  $f_t$  is the density of the t distribution. In this appendix, we show how these parameters are specified to obtain a standardized distribution of mean zero and unit variable.

Let

$$m_1(\zeta) = \frac{2\Gamma(\frac{\zeta+1}{2})\zeta}{\sqrt{\pi\zeta}\Gamma(\frac{\zeta}{2})(\zeta-1)}$$
$$m_2(\zeta) = \frac{\zeta}{\zeta-2}.$$

These are the first and (non-central) second moments of the absolute value of a Student t distributed random variable with  $\zeta$  degrees of freedom. In the case of a normal distribution

---

\*The opinions expressed here are our own, and do not reflect the views of the Board of Governors or its staff. Address correspondence to Alexander J. McNeil, School, University of York, Freboys Lane, York YO10 5GD, UK, +44 (0) 1904 325307, [alexander.mcneil@york.ac.uk](mailto:alexander.mcneil@york.ac.uk).

they simplify to  $m_1(\infty) = \sqrt{2/\pi}$  and  $m_2(\infty) = 1$ . The required scaling and location parameters in (OS.1) to give a distribution with mean 0 and variance 1 are

$$\begin{aligned}\sigma(\zeta, \gamma) &= ((m_2(\zeta) - m_1(\zeta)^2)(\gamma^2 + \gamma^{-2}) + 2m_1(\zeta)^2 - m_2(\zeta))^{-1/2} \\ \mu(\zeta, \gamma) &= -\sigma(\zeta, \gamma)m_1(\zeta)(\gamma - \gamma^{-1}).\end{aligned}$$

## 2 Special cases of the hypergeometric ${}_3F_2(1)$ function

We seek solutions to the second moments and cross-moments of the transformed PIT values when the standardized kernel densities take the form

$$\tilde{g}_\nu(u) = u^{a-1}(1-u)^{b-1}$$

for parameters ( $a > 0, b > -1/2$ ). As discussed in Appendix C of the main text, this involves the integral

$$M(a_1, b_1, a_2, b_2) = \int_0^1 (1-u)\tilde{g}_1(u)\tilde{G}_2(u)du \quad (\text{OS.2})$$

where  $\tilde{G}_i$  denotes the transform function for kernel  $i$  and has parameters  $(a_i, b_i)$ . The general solution to  $M$  can be written in two ways:

$$M(a_1, b_1, a_2, b_2) = \frac{B(a_1 + a_2, 1 + b_1)}{a_2} {}_3F_2(a_2, a_1 + a_2, 1 - b_2; 1 + a_2, 1 + a_1 + a_2 + b_1; 1) \quad (\text{OS.3})$$

$$= \frac{B(a_1 + a_2, 1 + b_1 + b_2)}{a_2} {}_3F_2(1, a_1 + a_2, a_2 + b_2; 1 + a_2, 1 + a_1 + a_2 + b_1 + b_2; 1) \quad (\text{OS.4})$$

In this appendix we document special cases for which the  ${}_3F_2(1)$  terms have known closed-form solution.

### 2.1 Case $b_1 = 0$

When  $b_1 = 0$  and  $b_2 \neq 0$ , we apply Wolfram Research (2023, 07.27.03.0119.01) to (C.3) to obtain

$$M(a_1, 0, a_2, b_2) = \frac{1}{a_1} (B(a_2, b_2) - B(a_1 + a_2, b_2)) \quad (\text{OS.5})$$

When  $b_1 = b_2 = 0$ , we instead apply Wolfram Research (2023, 07.27.03.0035.01) to (OS.4):

$$M(a_1, 0, a_2, 0) = \frac{1}{a_1} (\psi(a_1 + a_2) - \psi(a_2)) \quad (\text{OS.6})$$

where  $\psi$  is the digamma function. One can derive this by noting that Wolfram Research (2023, 06.14.26.0001.01 and 06.18.06.0004.01) together imply the limit

$$\lim_{b \rightarrow 0} \left( \frac{1}{b} - B(a, b) \right) = \psi(a) + \gamma$$

where  $\gamma$  is the Euler-Mascheroni constant.

## 2.2 Case $b_1 = n \in \mathbb{N}$

Binomial expansion of the term  $(1 - u)^{b_1}$  in the integral form (C.5) of  $M$  leads to

$$M(a_1, n, a_2, b_2) = \sum_{k=0}^n (-1)^k \binom{n}{k} M(a_1 + k, 0, a_2, b_2) \quad (\text{OS.7})$$

This approach would be practical for small values of  $n$ .

## 2.3 Case $a_2 = n \in \mathbb{N}$

For the moment, assume  $b_2 \neq 0$ . We have from Wolfram Research (2023, 06.19.03.0003.01):

$$B(u; n, b) = B(n, b) \left( 1 - (1 - u)^b \sum_{k=0}^{n-1} \frac{(b)_k u^k}{k!} \right) \quad (\text{OS.8})$$

where  $(b)_k$  is Pochhammer's notation defined recursively by  $(b)_0 = 1$ ,  $(b)_{m+1} = (b+m) \cdot (b)_m$ . This leads to

$$M(a_1, b_1, n, b_2) = \frac{(n-1)!}{(b_2)_n} \left( B(a_1, 1 + b_1) - \sum_{k=0}^{n-1} \frac{(b_2)_k}{k!} B(a_1 + k, 1 + b_1 + b_2) \right). \quad (\text{OS.9})$$

This can be computed more effectively in recursive form:

$$M(a_1, b_1, 1, b_2) = \frac{1}{b_2} (B(a_1, 1 + b_1) - B(a_1, 1 + b_1 + b_2)) \quad (\text{OS.10})$$

$$M(a_1, b_1, m + 1, b_2) = \frac{1}{b_2 + m} (m M(a_1, b_1, m, b_2) - B(a_1 + m, 1 + b_1 + b_2)) \quad (\text{OS.11})$$

To take the limit of (OS.10) as  $b_2 \rightarrow 0$ , we have by Wolfram Research (2023, 06.18.06.0013.01) that

$$B(a, b + z) = B(a, b) (1 + z(\psi(b + 1) - \psi(a + b + 1)) + \mathcal{O}(z^2))$$

which implies

$$M(a_1, b_1, 1, 0) = B(a_1, 1 + b_1) (\psi(1 + a_1 + b_1) - \psi(1 + b_1)) \quad (\text{OS.12})$$

Observe that (OS.11) is well-behaved at  $b_2 = 0$ .

## 2.4 Case $b_2 = n \in \mathbb{N}$

We have from Wolfram Research (2023, 06.19.03.0001.01):

$$B(u; a, 1 + m) = B(a, 1 + m) u^a \sum_{k=0}^m \frac{(a)_k (1 - u)^k}{k!} \quad (\text{OS.13})$$

Substituting into the integral for  $M$ :

$$M(a_1, b_1, a_2, n) = \frac{(n-1)!}{(a_2)_n} \sum_{k=0}^{n-1} \frac{(a_2)_k}{k!} B(a_1 + a_2, b_1 + k + 1). \quad (\text{OS.14})$$

This could also be expressed compactly in recursive form. An especially simple subcase is

$$M(a_1, b_1, a_2, 1) = \frac{1}{a_2} B(a_1 + a_2, b_1 + 1). \quad (\text{OS.15})$$

### 3 Examples of TLSF score tests

**Test based on normal distribution:** We have  $R = \Phi$ ,  $\rho = \phi$ ,  $\lambda_\rho(x) = x$ , and  $\lambda'_\rho(x) = 1$ . Consequently, integrals of type  $\mathcal{B}_{\rho,k}$  have solution  $\mathcal{B}_{\rho,0}(\alpha) = \alpha$ ,  $\mathcal{B}_{\rho,1}(\alpha) = \mathcal{A}_{\rho,0}(\alpha)$ , and  $\mathcal{B}_{\rho,2}(\alpha) = \mathcal{A}_{\rho,1}(\alpha)$ .

**Test based on logistic distribution:** When  $R(x)$  is the logistic function  $S(x)$ , we have  $\rho(x) = S(x)S(-x)$ ,  $\lambda_\rho(x) = S(x) - S(-x)$  and  $\lambda'_\rho(x) = 2\rho(x)$ . The inverse cdf is  $R^{-1}(p) = \ln(p/(1-p))$ , which leads to the convenient expressions  $\rho(R^{-1}(\alpha)) = \alpha(1-\alpha)$ ,  $C_1(R^{-1}(\alpha)) = 1-\alpha$ ,  $\bar{C}_1(R^{-1}(\alpha)) = -\alpha$ ,  $C_2(R^{-1}(\alpha)) = \bar{C}_2(R^{-1}(\alpha)) = -\alpha(1-\alpha)$ .

Integrals of type  $\mathcal{B}_{\rho,k}$  have solutions

$$\mathcal{B}_{\rho,0}(\alpha) = \frac{1}{3}\alpha^2(3-2\alpha) \quad (\text{OS.16})$$

$$\mathcal{B}_{\rho,1}(\alpha) = \begin{cases} \frac{1}{3}(\alpha(1-\alpha) + \ln(1-\alpha)) + \mathcal{B}_{\rho,0}(\alpha)R^{-1}(\alpha) & \alpha < 1 \\ 0 & \alpha = 1 \end{cases} \quad (\text{OS.17})$$

$$\mathcal{B}_{\rho,2}(\alpha) = \begin{cases} \frac{-2}{3}(\alpha + \text{Li}_2(\frac{-\alpha}{1-\alpha})) + 2\mathcal{B}_{\rho,1}(\alpha)R^{-1}(\alpha) - \mathcal{B}_{\rho,0}(\alpha)R^{-1}(\alpha)^2 & \alpha < 1 \\ \frac{2}{3}(\frac{\pi^2}{6} - 1) & \alpha = 1. \end{cases} \quad (\text{OS.18})$$

The dilogarithm function  $\text{Li}_2$  is available in the GSL package.

For the lower bound on permissible  $\alpha_1$ , (14) simplifies to the condition  $xR(x) = 1$ , which has solution  $x_1 = W_0(1/e) + 1$ , where  $W_0$  here denotes the Lambert  $W$ -function. This leads to the bound  $\alpha_1 \geq 1/(1 + W_0(1/e)) \approx 0.782$ .

**Tests based on Gumbel distribution:** For the standard Gumbel, we have  $R(x) = \exp(-\exp(-x))$  and  $\rho(x) = \exp(-x)R(x)$ , so  $\lambda_\rho(x) = 1 - \exp(-x) = 1 - C_1(x)$ . The inverse cdf is  $R^{-1}(p) = -\ln(-\ln(p))$ , which leads to  $\rho(R^{-1}(\alpha)) = -\alpha \ln(\alpha)$  and  $\lambda_\rho(R^{-1}(\alpha)) = 1 + \ln(\alpha)$ . For  $\mathcal{B}_{\rho,0}(\alpha)$  and  $\mathcal{B}_{\rho,1}(\alpha)$  we have

$$\mathcal{B}_{\rho,0}(\alpha) = \begin{cases} 0 & \alpha = 0 \\ \alpha(1 - \ln(\alpha)) & \alpha > 0 \end{cases}$$

$$\mathcal{B}_{\rho,1}(\alpha) = \begin{cases} \text{li}(\alpha) - \alpha + \mathcal{B}_{\rho,0}(\alpha)R^{-1}(\alpha) & \alpha < 1 \\ \gamma - 1 & \alpha = 1 \end{cases}$$

where  $\text{li}$  is the logarithmic integral function and  $\gamma$  is the Euler-Mascheroni constant.  $\mathcal{B}_{\rho,2}(\alpha)$  is best solved numerically.

For the complementary Gumbel, we apply Remark (3.6) to obtain  $R_c$ ,  $\lambda_c$ , etc., and equations (D.16)–(D.17) to provide solutions to  $\mathcal{A}_{\rho,k}^c$  and  $\mathcal{B}_{\rho,k}^c$ .

**Tests based on logistic-beta distribution:** The logistic-beta distribution has density

$$\rho(x; a, b) = \frac{S(x)^a S(-x)^b}{B(a, b)}$$

for parameters  $a > 0, b > 0$ .  $S(x) = 1/(1 + \exp(-x))$  is the logistic function and  $B(a, b)$  is the beta function.

The logistic function satisfies  $S(-x) = 1 - S(x)$  and  $S'(x) = S(x)S(-x)$  from which it follows that

$$\begin{aligned}\lambda_\rho(x) &= bS(x) - aS(-x) \\ \lambda'_\rho(x) &= (a + b)S(x)S(-x) > 0\end{aligned}$$

The cumulants of the distribution are

$$\kappa_m(a, b) = \psi^{(m-1)}(a) + (-1)^m \psi^{(m-1)}(b)$$

where  $\psi^{(j)}$  is the  $j^{\text{th}}$  derivative of the digamma function. An implication is that the skew of the distribution has the same sign as  $a - b$ , since  $\psi^{(2)}(x)$  is strictly increasing for  $x > 0$ .

The cdf is a composition of the beta cdf  $\mathcal{I}(z; a, b)$  and the logistic function, i.e.,  $R(x) = \mathcal{I}(S(x); a, b)$ . The inverse cdf is therefore

$$R^{-1}(p) = \text{logit}(\mathcal{I}^{-1}(p; a, b))$$

where  $\text{logit}(u) = \ln(u/(1 - u))$ . The well-known symmetry for the beta distribution implies a symmetry of the same form for the inverse cdf:

$$1 - \mathcal{I}^{-1}(p; a, b) = \mathcal{I}^{-1}(1 - p; b, a)$$

Consequently, we can write

$$R^{-1}(p) = \ln(\mathcal{I}^{-1}(p; a, b)) - \ln(\mathcal{I}^{-1}(1 - p; b, a)). \quad (\text{OS.19})$$

Observe that

$$\begin{aligned}\lambda'_\rho(x; a, b)\rho(x; a, b) &= \frac{(a + b)}{B(a, b)} S(x)^{a+1} S(-x)^{b+1} \\ &= \frac{(a + b)B(a + 1, b + 1)}{B(a, b)} \rho(x; a + 1, b + 1) = \frac{ab}{1 + a + b} \rho(x; a + 1, b + 1)\end{aligned}$$

which implies that

$$\begin{aligned}\mathcal{B}_{\rho,0}(1) &= \frac{ab}{1 + a + b} \\ \mathcal{B}_{\rho,1}(1) &= \mathcal{B}_{\rho,0}(1)\kappa_1(a + 1, b + 1) \\ \mathcal{B}_{\rho,2}(1) &= \mathcal{B}_{\rho,0}(1) (\kappa_2(a + 1, b + 1) + \kappa_1(a + 1, b + 1)^2)\end{aligned}$$

For any  $\alpha$ , we have

$$\mathcal{B}_{\rho,0}(\alpha) = \mathcal{B}_{\rho,0}(1)R(R^{-1}(\alpha; a, b); a + 1, b + 1).$$

For  $k = 1, 2$ , we can obtain solutions to  $\mathcal{B}_{\rho,k}(\alpha)$  in terms of  ${}_3F_2$  and  ${}_4F_3$  hypergeometric functions. In practice, it is easiest to pre-calculate interpolation tables for these functions for select values of  $(a, b)$ . The numeric integrals appear to be straightforward and knowing the end-value of  $\mathcal{B}_{\rho,k}(1)$  ensures a high degree of accuracy.

One special case of possible special interest is the skew sech distribution. The density is usually written

$$\rho^{ss}(x; \theta) = \frac{\cos(\theta)}{2} \exp(\theta x) \operatorname{sech}\left(\frac{\pi}{2}x\right)$$

where the skew parameter  $|\theta| < \pi/2$  and  $\theta = 0$  corresponds to the standard sech distribution. We can easily show that if  $X$  is distributed skew sech with parameter  $\theta$  then  $\pi X$  is distributed logistic-beta with parameters  $(a, 1 - a)$  for  $a = 1/2 + \theta/\pi \in (0, 1)$ .

## References

**Wolfram Research**, “Mathematical Functions Site,” <https://functions.wolfram.com/>, as of 2023-07-23 2023.