
**PQIC Mathematical Notes, Analytical Series, Number 79:
Univariate Continuous F Distribution Random Variate
Generation Using Distribution Ratio Methodology**

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Abstract

The purpose of this memorandum is to document the analytical methods by which random variates from a univariate continuous F probability distribution may be generated regardless of the relative sizes of its parameters, and to document alternative calculation methods for estimating F distribution values when one or the other (or both) positive parameters increase without bound.

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1. Introduction

The univariate continuous F probability distribution is used within PQICSTAT™ as the discriminator of choice for reducing designed experiment linear models to the fundamental core of significant factors based on a given dataset. The iterative steps of re-allocating the sum of squares to reflect the annexation of non-significant factors to the error term requires repeated calculation of an F distributed statistic with increasing second degree of freedom. To control the extent to which factors are accepted or rejected in this process, it is important to understand the power of these decisions, which must be evaluated through simulated results, i.e., through the generation of large numbers of random variates, to study the behavior of the F distributions under a variety of alternative hypotheses. In this respect, random variates from the F distribution must not only be available for simulated results in this and other contexts, but also asymptotic estimation is needed when the degrees of freedom become very large in extensive datasets.

The purpose of this memorandum is to document the analytical methods by which random variates from a univariate continuous F probability distribution may be generated regardless of the relative sizes of its parameters, and to document alternative calculation methods for estimating the F distribution values when one or the other (or both) positive parameters increase without bound.

Definition 1 A univariate continuous random variable X is said to have an F Distribution With $r > 0$ and $s > 0$ Degrees Of Freedom, symbolized as $F(r, s)$, if r and s are integers, and its density function $f_X(x; r, s)$ is given by

$$f_X(x; r, s) = \frac{\Gamma\left(\frac{r+s}{2}\right)}{\Gamma\left(\frac{r}{2}\right)\Gamma\left(\frac{s}{2}\right)} \left(\frac{r}{s}\right)^{\frac{r}{2}} x^{\frac{r}{2}-1} \left(1 + \frac{r}{s}x\right)^{-\frac{r+s}{2}}, \quad x \geq 0$$

Claim 2 The function $f_X(x; r, s)$ is a density function of a probability distribution.

Proof. For r even, we have

$$\begin{aligned} \int_0^\infty f_X(x; r \text{ even}, s) dx &= \frac{\Gamma\left(\frac{r+s}{2}\right)}{\Gamma\left(\frac{r}{2}\right)\Gamma\left(\frac{s}{2}\right)} \left(\frac{r}{s}\right)^{\frac{r}{2}} \int_0^\infty x^{\frac{r}{2}-1} \left(1 + \frac{r}{s}x\right)^{-\frac{r+s}{2}} dx \\ &= \frac{\Gamma\left(\frac{r+s}{2}\right)}{\Gamma\left(\frac{r}{2}\right)\Gamma\left(\frac{s}{2}\right)} \left(\frac{r}{s}\right)^{\frac{r}{2}} \left(\begin{array}{c} \frac{s}{r} x^{\frac{r}{2}-1} \frac{(1+\frac{r}{s}x)^{1-\frac{r+s}{2}}}{1-\frac{r+s}{2}} \Big|_{x=0}^{x \rightarrow \infty} \\ - \left(\frac{s}{r}\right) \frac{\left(\frac{r}{2}-1\right)}{1-\frac{r+s}{2}} \int_0^\infty x^{\frac{r}{2}-2} \left(1 + \frac{r}{s}x\right)^{1-\frac{r+s}{2}} dx \end{array} \right), \\ &\quad \left(\begin{array}{c} u = x^{\frac{r}{2}-1}, dv = \left(1 + \frac{r}{s}x\right)^{-\frac{r+s}{2}} dx \\ du = \left(\frac{r}{2}-1\right) x^{\frac{r}{2}-2} dx, v = \left(\frac{s}{r}\right) \frac{(1+\frac{r}{s}x)^{1-\frac{r+s}{2}}}{1-\frac{r+s}{2}} \end{array} \right) \\ &= \frac{\Gamma\left(\frac{r+s}{2}\right)}{\Gamma\left(\frac{r}{2}\right)\Gamma\left(\frac{s}{2}\right)} \left(\frac{r}{s}\right)^{\frac{r}{2}} \left(\begin{array}{c} \left(2 \frac{s}{r(2-(r+s))} \lim_{x \rightarrow \infty} \frac{\left(\frac{x}{1+\frac{r}{s}x}\right)^{\frac{r}{2}-1}}{\left(1+\frac{r}{s}x\right)^{\frac{s}{2}}} = 0 \text{ since } s > 0 \right) \\ + \left(\frac{s}{r}\right) \frac{\frac{r}{2}-1}{\frac{r+s}{2}-1} \int_0^\infty x^{\frac{r}{2}-2} \left(1 + \frac{r}{s}x\right)^{1-\frac{r+s}{2}} dx \end{array} \right) \\ &= \frac{\Gamma\left(\frac{r+s}{2}\right)}{\Gamma\left(\frac{r}{2}\right)\Gamma\left(\frac{s}{2}\right)} \left(\frac{r}{s}\right)^{\frac{r}{2}-1} \frac{\frac{r}{2}-1}{\frac{r+s}{2}-1} \int_0^\infty x^{\frac{r}{2}-2} \left(1 + \frac{r}{s}x\right)^{1-\frac{r+s}{2}} dx \\ &= \dots \\ &= \frac{\Gamma\left(\frac{r+s}{2}\right)}{\Gamma\left(\frac{r}{2}\right)\Gamma\left(\frac{s}{2}\right)} \left(\frac{r}{s}\right)^{\frac{r}{2}-\left(\frac{r}{2}-1\right)} \left(\begin{array}{c} \left(\frac{\frac{r}{2}-1}{\frac{r+s}{2}-1}\right) \dots \\ \times \left(\frac{\frac{r}{2}-\left(\frac{r}{2}-1\right)}{\frac{r+s}{2}-\left(\frac{r}{2}-1\right)}\right) \end{array} \right) \int_0^\infty x^{\frac{r}{2}-\frac{r}{2}} \left(1 + \frac{r}{s}x\right)^{\left(\frac{r}{2}-1\right)-\frac{r+s}{2}} dx \\ &= \frac{s}{2} \left(\frac{\left(\frac{r}{2}-1\right)!}{\Gamma\left(\frac{r}{2}\right)}\right) \left(\frac{\Gamma\left(\frac{r+s}{2}\right)}{\frac{s}{2}\Gamma\left(\frac{s}{2}\right) \prod_{k=1}^{\frac{r}{2}-1} \left(\frac{r+s}{2}-k\right)}\right) \int_0^\infty \frac{r}{s} \left(1 + \frac{r}{s}x\right)^{-\frac{s}{2}-1} dx \\ &= -\frac{s}{2} \left(1 + \frac{r}{s}x\right)^{-\frac{s}{2}} \Big|_{x=0}^{x \rightarrow \infty} \end{aligned}$$

= 1, since $s > 0$

When r is odd, then $r - 1$ is even, so we have¹

$$\begin{aligned}
 \int_0^\infty f_X(x; r \text{ odd}, s) dx &= \frac{\Gamma\left(\frac{r+s}{2}\right)}{\Gamma\left(\frac{r}{2}\right)\Gamma\left(\frac{s}{2}\right)} \left(\frac{r}{s}\right)^{\frac{r}{2}} \int_0^\infty x^{\frac{r}{2}-1} \left(1 + \frac{r}{s}x\right)^{-\frac{r+s}{2}} dx \\
 &= \text{same steps as for } r \text{ even} \\
 &= \left(\frac{\Gamma\left(\frac{r+s}{2}\right)}{\Gamma\left(\frac{r}{2}\right)\Gamma\left(\frac{s}{2}\right)} \left(\frac{r}{s}\right)^{\frac{r}{2}-(\frac{r-1}{2}-1)} \left(\frac{\frac{r}{2}-1}{\frac{r+s}{2}-1}\right) \dots \right. \\
 &\quad \left. \times \left(\frac{\frac{r}{2}-(\frac{r-1}{2}-1)}{\frac{r+s}{2}-(\frac{r-1}{2}-1)}\right) \int_0^\infty x^{\frac{r}{2}-1-(\frac{r-1}{2}-1)} \left(1 + \frac{r}{s}x\right)^{(\frac{r-1}{2}-1)-\frac{r+s}{2}} dx \right) \\
 &= \left(\frac{\Gamma\left(\frac{r+s}{2}\right)}{\Gamma\left(\frac{r}{2}\right)\Gamma\left(\frac{s}{2}\right)} \left(\frac{r}{s}\right)^{\frac{3}{2}} \left(\frac{\frac{r}{2}-1}{\frac{r+s}{2}-1}\right) \dots \right) \\
 &\quad \left. \times \left(\frac{\frac{3}{2}}{\frac{r+s}{2}-(\frac{r-1}{2}-1)}\right) \int_0^\infty x^{\frac{1}{2}} \left(1 + \frac{r}{s}x\right)^{-\frac{s+3}{2}} dx \right) \\
 &= \left(\frac{2}{\sqrt{\pi}} \left(\frac{\Gamma\left(\frac{r+s}{2}\right) = \Gamma\left(\frac{s}{2} + \frac{3}{2}\right) \prod_{j=1}^{\frac{r-1}{2}-1} \left(\frac{r+s}{2}-j\right)}{\Gamma\left(\frac{s}{2}\right) \prod_{k=1}^{\frac{r-1}{2}-1} \left(\frac{r+s}{2}-k\right)} \right) \left(\frac{\frac{1}{2}\Gamma\left(\frac{1}{2}\right) \prod_{k=1}^{\frac{r-1}{2}-1} \left(\frac{r}{2}-k\right)}{\Gamma\left(\frac{r}{2}\right)} \right) \right) \\
 &\quad \times \int_0^\infty \frac{r}{s} \left(\frac{r}{s}x\right)^{\frac{1}{2}} \left(1 + \frac{r}{s}x\right)^{-\frac{s+3}{2}} dx \\
 &= \frac{4}{\sqrt{\pi}} \frac{\Gamma\left(\frac{s+3}{2}\right)}{\Gamma\left(\frac{s}{2}\right)} \int_0^{\frac{\pi}{2}} \tan^2 u \sec^2 u \left(1 + \tan^2 u\right)^{-\frac{s+3}{2}} du, \\
 &\quad \left(\begin{aligned} \frac{r}{s}x &= \tan^2 u, \\ dx &= 2\frac{s}{r} \tan u \sec^2 u du \end{aligned} \right) \tag{1} \\
 &= \frac{4}{\sqrt{\pi}} \frac{\Gamma\left(\frac{s+3}{2}\right)}{\Gamma\left(\frac{s}{2}\right)} \int_0^{\frac{\pi}{2}} (\sec^2 u - 1) \sec^{-s-1} u du \\
 &= \frac{4}{\sqrt{\pi}} \frac{\Gamma\left(\frac{s+3}{2}\right)}{\Gamma\left(\frac{s}{2}\right)} \left(\int_0^{\frac{\pi}{2}} \cos^{s-1} u du - \int_0^{\frac{\pi}{2}} \cos^{s+1} u du \right) \\
 &= \frac{4}{\sqrt{\pi}} \frac{\Gamma\left(\frac{s+3}{2}\right)}{\Gamma\left(\frac{s}{2}\right)} \left(\int_0^{\frac{\pi}{2}} \cos^{s-1} u du - \int_0^{\frac{\pi}{2}} \cos^{(s+2)-1} u du \right) \\
 &= \frac{4}{\sqrt{\pi}} \frac{\Gamma\left(\frac{s+3}{2}\right)}{\Gamma\left(\frac{s}{2}\right)} \left(\frac{1}{2}\sqrt{\pi} \frac{\Gamma\left(\frac{s}{2}\right)}{\Gamma\left(\frac{s+1}{2}\right)} - \frac{1}{2}\sqrt{\pi} \frac{\Gamma\left(\frac{s+2}{2}\right)}{\Gamma\left(\frac{s+3}{2}\right)} \right) \tag{2} \\
 &= 2 \left(\frac{\Gamma\left(\frac{s+3}{2}\right)}{\Gamma\left(\frac{s+1}{2}\right)} - \frac{\Gamma\left(\frac{s+2}{2}\right)}{\Gamma\left(\frac{s}{2}\right)} \right) \\
 &= 2 \left(\frac{s+1}{2} - \frac{s}{2} \right) \\
 &= 1
 \end{aligned}$$

■

¹ The step labeled (2) in this proof was proven in Claim 2 in *PQIC Mathematical Notes, Analytical Series, Number 78: Univariate Continuous Student-T Distribution Random Variate Generation Using Mixtures Methodology*, December 2020, where, for integer n , we have

$$\int_{-\frac{\pi}{2}}^{\frac{\pi}{2}} \cos^{n-1} \psi d\psi = 2 \int_0^{\frac{\pi}{2}} \cos^{n-1} \psi d\psi$$

since the cosine is an even function.

2. Main Theorem

Theorem 3 If $X_i \sim \Gamma\left(\frac{\alpha_i}{2}, \frac{\beta_i}{2}\right)$, $i = 1, 2$, $\alpha_i, \beta_i > 0$, are independent, then $\frac{\alpha_2\beta_1 X_1}{\alpha_1\beta_2 X_2} \sim F(\alpha_1, \alpha_2)$.

Proof. As was proven in *PQIC Mathematical Notes, Analytical Series, Number 76: Univariate Continuous Gamma Distribution Random Variate Generation Using Integral Transform Methodology*, October 2020, we have

$$X_i \sim \Gamma\left(\frac{\alpha_i}{2}, \frac{\beta_i}{2}\right) \implies \frac{\beta_i}{\alpha_i} X_i \sim \Gamma\left(\frac{\alpha_i}{2}, \frac{\alpha_i}{2}\right), i = 1, 2$$

so that if $Y_i = \frac{\beta_i}{\alpha_i} X_i$, $i = 1, 2$, then by the independence of the $X_i > 0$, the $Y_i > 0$ are also independent, which means

$$\begin{aligned} P\left(\frac{\alpha_2\beta_1 X_1}{\alpha_1\beta_2 X_2} \leq t > 0\right) &= \int_0^t P\left(\frac{Y_1}{Y_2} = x > 0\right) dx \\ &= \int_0^t \int_0^\infty P(Y_1 = xy_2 | Y_2 = y_2) P(Y_2 = y_2) \frac{d(xy_2)}{dx} dy_2 dx \\ &= \int_0^t \int_0^\infty y_2 P(Y_1 = xy_2) P(Y_2 = y_2) dy_2 dx \\ &= \int_0^t \int_0^\infty y_2 \left(\frac{\left(\frac{\alpha_1}{2}\right)^{\frac{\alpha_1}{2}}}{\Gamma\left(\frac{\alpha_1}{2}\right)} (xy_2)^{\frac{\alpha_1}{2}-1} e^{-\frac{\alpha_1}{2}(xy_2)}\right) \left(\frac{\left(\frac{\alpha_2}{2}\right)^{\frac{\alpha_2}{2}}}{\Gamma\left(\frac{\alpha_2}{2}\right)} y_2^{\frac{\alpha_2}{2}-1} e^{-\frac{\alpha_2}{2}y_2}\right) dy_2 dx \\ &= \frac{\left(\frac{\alpha_1}{2}\right)^{\frac{\alpha_1}{2}} \left(\frac{\alpha_2}{2}\right)^{\frac{\alpha_2}{2}}}{\Gamma\left(\frac{\alpha_1}{2}\right) \Gamma\left(\frac{\alpha_2}{2}\right)} \int_0^t x^{\frac{\alpha_1}{2}-1} \int_0^\infty y_2^{\frac{\alpha_1+\alpha_2}{2}-1} e^{-\left(\frac{\alpha_1}{2}x + \frac{\alpha_2}{2}\right)y_2} dy_2 dx \\ &= \left(\frac{\left(\frac{\alpha_1}{2}\right)^{\frac{\alpha_1}{2}} \left(\frac{\alpha_2}{2}\right)^{\frac{\alpha_2}{2}}}{\Gamma\left(\frac{\alpha_1}{2}\right) \Gamma\left(\frac{\alpha_2}{2}\right)} \int_0^t \frac{x^{\frac{\alpha_1}{2}-1}}{\left(\frac{\alpha_1}{2}x + \frac{\alpha_2}{2}\right)^{\frac{\alpha_1+\alpha_2}{2}}}\right) \left(\int_0^\infty v^{\frac{\alpha_1+\alpha_2}{2}-1} e^{-v} dv\right), \left(v = \left(\frac{\alpha_1}{2}x + \frac{\alpha_2}{2}\right)y_2\right) \\ &\quad \left(dy_2 = \frac{dv}{\frac{\alpha_1}{2}x + \frac{\alpha_2}{2}}\right) \\ &= \frac{\left(\frac{\alpha_1}{2}\right)^{\frac{\alpha_1}{2}} \left(\frac{\alpha_2}{2}\right)^{\frac{\alpha_2}{2}}}{\Gamma\left(\frac{\alpha_1}{2}\right) \Gamma\left(\frac{\alpha_2}{2}\right)} \int_0^t x^{\frac{\alpha_1}{2}-1} \left(\frac{\alpha_1}{2}x + \frac{\alpha_2}{2}\right)^{-\frac{\alpha_1+\alpha_2}{2}} \Gamma\left(\frac{\alpha_1+\alpha_2}{2}\right) dx \\ &= \frac{\Gamma\left(\frac{\alpha_1+\alpha_2}{2}\right)}{\Gamma\left(\frac{\alpha_1}{2}\right) \Gamma\left(\frac{\alpha_2}{2}\right)} \left(\left(\frac{\alpha_1}{2}\right)^{\frac{\alpha_1}{2}} \left(\frac{\alpha_2}{2}\right)^{\frac{\alpha_2}{2}-\frac{\alpha_1+\alpha_2}{2}}\right) \int_0^t x^{\frac{\alpha_1}{2}-1} \left(1 + \frac{\alpha_1}{\alpha_2}x\right)^{-\frac{\alpha_1+\alpha_2}{2}} dx \\ &= \frac{\Gamma\left(\frac{\alpha_1+\alpha_2}{2}\right)}{\Gamma\left(\frac{\alpha_1}{2}\right) \Gamma\left(\frac{\alpha_2}{2}\right)} \left(\frac{\alpha_1}{\alpha_2}\right)^{\frac{\alpha_1}{2}} \int_0^t x^{\frac{\alpha_1}{2}-1} \left(1 + \frac{\alpha_1}{\alpha_2}x\right)^{-\frac{\alpha_1+\alpha_2}{2}} dx \end{aligned}$$

so that

$$P\left(\frac{\alpha_2\beta_1 X_1}{\alpha_1\beta_2 X_2} = t > 0\right) = \frac{\Gamma\left(\frac{\alpha_1+\alpha_2}{2}\right)}{\Gamma\left(\frac{\alpha_1}{2}\right) \Gamma\left(\frac{\alpha_2}{2}\right)} \left(\frac{\alpha_1}{\alpha_2}\right)^{\frac{\alpha_1}{2}} t^{\frac{\alpha_1}{2}-1} \left(1 + \frac{\alpha_1}{\alpha_2}t\right)^{-\frac{\alpha_1+\alpha_2}{2}}$$

■

Remark 4 Note that this result does not depend on the specific values of β_i , $i = 1, 2$, so that convenient values (such as $\beta_1 = 2 = \beta_2$) may be used at the discretion of the implementing analyst. In that particular convenient case, we would have

$$X_i \sim \Gamma\left(\frac{\alpha_i}{2} > 0, 1\right) \implies \frac{\alpha_2 X_1}{\alpha_1 X_2} \sim F(\alpha_1, \alpha_2) \text{ and } \frac{\alpha_1 X_2}{\alpha_2 X_1} \sim F(\alpha_2, \alpha_1)$$

Algorithm 5 The following steps generates N random variates (with precision tol) from the univariate continuous F distribution with r and s degrees of freedom.

1. Set $A = \emptyset$ and $C = 1$.
2. Generate N random variates \mathbf{R} from the $\Gamma\left(\frac{r}{2}, 1\right)$ distribution with precision tol .²
3. Generate N random variates \mathbf{S} from the $\Gamma\left(\frac{s}{2}, 1\right)$ distribution with precision tol .

² Note this is a special case of the general analytical methods used to generate random variates from a $\Gamma(\alpha, \beta)$ distribution as documented in *PQIC Mathematical Notes, Analytical Series, Number 76: Univariate Continuous Gamma Distribution*

4. Annex the value of $\frac{sR[C]}{rS[C]}$ to the Acceptance Set A .
5. Add 1 to C .
6. If $C = N + 1$ then skip to Step 7; otherwise skip to Step 4.
7. Return A as the set of N -many random variates with precision tol from the $F(r, s)$ distribution.

Note that this algorithm does not use any asymptotic estimation for “large” parameter values (see next section). It shall be PQICSTAT™ policy to leave the use of such alternative generation methods, e.g., decide when a parameter value is “large,” to the discretion of the implementing analyst.

3. Simplifying Expression When r Or s Becomes Large

It may be a practical consideration to use asymptotic inference when r or s becomes large, as the calculation of the components of the density function may be impractical. For example, for an $F(4, 10000)$ distribution, even though $\frac{\Gamma(5002)}{\Gamma(2)\Gamma(5000)} \left(\frac{4}{10000}\right)^2$ admits to significant simplification, there is still a need to calculate $\left(1 + \frac{4}{10000}x\right)^{-5002}$, which, even for large x , such as $x = 100$, only has significant digits starting in position 86. And this situation also applies when r is large relative to s , such as for an $F(10000, 4)$ distribution, for then $\left(1 + \frac{10000}{4}x\right)^{-5002}$ must still be evaluated.

Therefore, the following claims provide an analytical means for evaluating such F distribution random variates by asymptotic methods. See also the appendix where the relationship between the Student-T and F distributions is explicitly demonstrated.

4. When s Becomes Large

Claim 6 If $X \sim F(r, s)$, then $\lim_{s \rightarrow \infty} rX \sim \chi_r^2$.

Proof. We have

$$\begin{aligned} \lim_{s \rightarrow \infty} \left(1 + \frac{1}{s}u\right)^{-\frac{r+s}{2}} &= \left(\lim_{s \rightarrow \infty} \left(1 + \frac{1}{s}u\right)^{-s}\right)^{\frac{1}{2}} \lim_{s \rightarrow \infty} \left(1 + \frac{1}{s}u\right)^{-\frac{r}{2}} \\ &= (e^{-u})^{\frac{1}{2}} (1)^{-\frac{r}{2}} \\ &= e^{-\frac{u}{2}} \end{aligned}$$

and using Stirling’s Approximation, we have

$$\begin{aligned} \lim_{s \rightarrow \infty} \frac{\Gamma\left(\frac{r+s}{2}\right)}{\Gamma\left(\frac{s}{2}\right)} \left(\frac{1}{s}u\right)^{\frac{r}{2}} &= u^{\frac{r}{2}} \lim_{s \rightarrow \infty} \frac{\Gamma\left(\frac{r+s}{2}\right)}{s^{\frac{r}{2}} \Gamma\left(\frac{s}{2}\right)} \\ &= u^{\frac{r}{2}} \lim_{s \rightarrow \infty} \frac{\sqrt{2\pi} \left(\frac{r+s}{2} - 1\right) \left(\frac{\frac{r+s}{2} - 1}{e}\right)^{\frac{r+s}{2} - 1}}{s^{\frac{r}{2}} \sqrt{2\pi} \left(\frac{s}{2} - 1\right) \left(\frac{\frac{s}{2} - 1}{e}\right)^{\frac{s}{2} - 1}} \\ &= u^{\frac{r}{2}} \lim_{s \rightarrow \infty} e^{-\frac{r}{2}} \frac{\sqrt{\frac{r+s-2}{s-2}} \left(\frac{r+s}{2} - 1\right)^{\frac{r+s}{2} - 1}}{s^{\frac{r}{2}} \left(\frac{s}{2} - 1\right)^{\frac{s}{2} - 1}} \\ &= u^{\frac{r}{2}} e^{-\frac{r}{2}} \lim_{s \rightarrow \infty} \sqrt{1 + \frac{r}{s-2}} \left(\frac{\frac{r+s}{2} - 1}{s}\right)^{\frac{r}{2}} \left(\frac{\frac{r+s}{2} - 1}{\frac{s}{2} - 1}\right)^{\frac{s}{2} - 1} \end{aligned}$$

Random Variate Generation Using Integral Transform Methodology, October 2020. Since $\beta = 1$ in this case, we have

$$f_{\Gamma\left(\frac{r}{2}, 1\right)}(x) = \frac{1}{\Gamma\left(\frac{r}{2}\right)} x^{\frac{r}{2} - 1} e^{-x}, \quad x > 0$$

This simplifies to

$$\frac{x^{\frac{r}{2} - 1} e^{-x}}{\left(\frac{r}{2} - 1\right)!} \quad (r \text{ even}) \quad \text{and} \quad \frac{x^{\frac{r}{2} - 1} e^{-x}}{\sqrt{\pi} \left(\frac{r}{2} - 1\right) \left(\frac{r}{2} - 2\right) \cdots \frac{1}{2}} \quad (r \text{ odd})$$

Nevertheless, it shall be PQICSTAT™ policy to use the analytical methods in *op. cit.* to generate the gamma distribution random variates even in this special case.

$$\begin{aligned}
&= u^{\frac{r}{2}} e^{-\frac{r}{2}} \lim_{s \rightarrow \infty} \left(\sqrt{1 + \frac{r}{s-2}} \right) \lim_{s \rightarrow \infty} \left(\frac{r+s-2}{2s} \right)^{\frac{r}{2}} \lim_{s \rightarrow \infty} \left(\frac{r+s-2}{s-2} \right)^{\frac{s}{2}-1} \\
&= u^{\frac{r}{2}} \frac{1}{2^{\frac{r}{2}}} e^{-\frac{r}{2}} (1) \left(\lim_{s \rightarrow \infty} \left(1 + \frac{r-2}{s} \right)^{\frac{r}{2}} \right) \lim_{s \rightarrow \infty} \left(1 + \frac{r}{(2s+2)-2} \right)^{\frac{2s+2}{2}-1} \\
&= \left(\frac{u}{2} \right)^{\frac{r}{2}} e^{-\frac{r}{2}} (1) (1) \lim_{s \rightarrow \infty} \left(1 + \frac{r}{s} \right)^s \\
&= \left(\frac{u}{2} \right)^{\frac{r}{2}} e^{-\frac{r}{2}} e^{\frac{r}{2}} \\
&= \left(\frac{u}{2} \right)^{\frac{r}{2}}
\end{aligned}$$

$$\begin{aligned}
\lim_{s \rightarrow \infty} P(rF(r, s) \leq x) &= \lim_{s \rightarrow \infty} P\left(F(r, s) \leq \frac{x}{r}\right) \\
&= \lim_{s \rightarrow \infty} \int_0^{\frac{x}{r}} \frac{\Gamma\left(\frac{r+s}{2}\right)}{\Gamma\left(\frac{r}{2}\right)\Gamma\left(\frac{s}{2}\right)} \left(\frac{r}{s}\right)^{\frac{r}{2}} t^{\frac{r}{2}-1} \left(1 + \frac{r}{s}t\right)^{-\frac{r+s}{2}} dt \\
&= \lim_{s \rightarrow \infty} \int_0^x \frac{1}{r} \frac{\Gamma\left(\frac{r+s}{2}\right)}{\Gamma\left(\frac{r}{2}\right)\Gamma\left(\frac{s}{2}\right)} \left(\frac{r}{s}\right)^{\frac{r}{2}} \left(\frac{u}{r}\right)^{\frac{r}{2}-1} \left(1 + \frac{1}{s}u\right)^{-\frac{r+s}{2}} du, \quad \left(\begin{array}{l} u = rt, \\ dt = \frac{1}{r} du \end{array} \right) \\
&= \int_0^x \frac{1}{u\Gamma\left(\frac{r}{2}\right)} \left(\lim_{s \rightarrow \infty} \frac{\Gamma\left(\frac{r+s}{2}\right)}{\Gamma\left(\frac{s}{2}\right)} \left(\frac{1}{s}\right)^{\frac{r}{2}} \right) \left(\lim_{s \rightarrow \infty} \left(1 + \frac{1}{s}u\right)^{-\frac{r+s}{2}} \right) du \\
&= \int_0^x \frac{1}{u\Gamma\left(\frac{r}{2}\right)} \left(\left(\frac{u}{2}\right)^{\frac{r}{2}} \right) \left(e^{-\frac{u}{2}} \right) du \\
&= \frac{1}{2^{\frac{r}{2}}\Gamma\left(\frac{r}{2}\right)} \int_0^x u^{\frac{r}{2}-1} e^{-\frac{u}{2}} du
\end{aligned}$$

which means

$$\lim_{s \rightarrow \infty} P(rF(r, s) = x) = \frac{1}{2^{\frac{r}{2}}\Gamma\left(\frac{r}{2}\right)} x^{\frac{r}{2}-1} e^{-\frac{x}{2}}$$

and this is the density function for a χ_r^2 distribution. ■

This proof also shows that the rate of convergence of this approximation depends on the two limits given by

$$\lim_{s \rightarrow \infty} \frac{\Gamma\left(\frac{r+s}{2}\right)}{\Gamma\left(\frac{s}{2}\right)} \left(\frac{1}{s}\right)^{\frac{r}{2}} = \left(\frac{u}{2}\right)^{\frac{r}{2}} \quad \text{and} \quad \lim_{s \rightarrow \infty} \left(1 + \frac{1}{s}u\right)^{-\frac{r+s}{2}} = e^{-\frac{u}{2}}$$

The first limit depends on r and u , while the second limit only depends on u – the particular value of r is not relevant to the limit, but it is to the rate of convergence. The quicker these terms converge, the closer the chi-square approximation comes to the actual F distribution value. However, the complicated interplay between these factors means such approximations must be considered “close” to the target distribution without claiming, or relying on, the false premise that the approximating value are actually coming from an F distribution.

5. When r Becomes Large

Corollary 7 If $X \sim F(r, s)$, then $\lim_{r \rightarrow \infty} s \frac{1}{X} \sim \chi_s^2$.

Proof. First note that

$$X \sim F(r, s) \implies \frac{1}{X} \sim F(s, r)$$

[Proof: If $X \sim F(r, s)$, then

$$P(X \leq x) = \frac{\Gamma\left(\frac{r+s}{2}\right)}{\Gamma\left(\frac{r}{2}\right)\Gamma\left(\frac{s}{2}\right)} \left(\frac{r}{s}\right)^{\frac{r}{2}} \int_0^x t^{\frac{r}{2}-1} \left(1 + \frac{r}{s}t\right)^{-\frac{r+s}{2}} dt$$

so that

$$\begin{aligned}
P\left(\frac{1}{X} \leq x\right) &= P\left(X \geq \frac{1}{x}\right) \\
&= 1 - P\left(X \leq \frac{1}{x}\right) \\
&= 1 - \frac{\Gamma\left(\frac{r+s}{2}\right)}{\Gamma\left(\frac{r}{2}\right)\Gamma\left(\frac{s}{2}\right)} \left(\frac{r}{s}\right)^{\frac{r}{2}} \int_0^{\frac{1}{x}} t^{\frac{r}{2}-1} \left(1 + \frac{r}{s}t\right)^{-\frac{r+s}{2}} dt \\
&= 1 - \frac{\Gamma\left(\frac{r+s}{2}\right)}{\Gamma\left(\frac{r}{2}\right)\Gamma\left(\frac{s}{2}\right)} \left(\frac{r}{s}\right)^{\frac{r}{2}} \int_x^\infty u^{-1-\frac{r}{2}} \left(1 + \frac{r}{su}\right)^{-\frac{r+s}{2}} du, \quad \left(\begin{array}{l} u = \frac{1}{t} \\ dt = -\frac{du}{u^2} \end{array}\right) \\
&= 1 - \frac{\Gamma\left(\frac{r+s}{2}\right)}{\Gamma\left(\frac{r}{2}\right)\Gamma\left(\frac{s}{2}\right)} \left(\frac{r}{s}\right)^{\frac{r}{2}} \int_x^\infty u^{-1-\frac{r}{2}} \left(\frac{r}{su}\right)^{-\frac{r+s}{2}} \left(\frac{s}{r}u + 1\right)^{-\frac{r+s}{2}} du \\
&= 1 - \frac{\Gamma\left(\frac{r+s}{2}\right)}{\Gamma\left(\frac{r}{2}\right)\Gamma\left(\frac{s}{2}\right)} \left(\frac{s}{r}\right)^{\frac{s}{2}} \int_x^\infty u^{\frac{s}{2}-1} \left(1 + \frac{s}{r}u\right)^{-\frac{r+s}{2}} du \\
&= \frac{\Gamma\left(\frac{r+s}{2}\right)}{\Gamma\left(\frac{r}{2}\right)\Gamma\left(\frac{s}{2}\right)} \left(\frac{s}{r}\right)^{\frac{s}{2}} \int_0^x u^{\frac{s}{2}-1} \left(1 + \frac{s}{r}u\right)^{-\frac{r+s}{2}} du
\end{aligned}$$

so that $\frac{1}{X} \sim F(s, r)$.]

Then by Claim 6, we have $\lim_{r \rightarrow \infty} s \frac{1}{X} \sim \chi_s^2$. ■

These claims may be used interchangeably by the implementing analyst as analytical circumstances require.

6. Special Case When r And s Are Even Positive Integers

It is possible to reduce the size of the values involved in generating random variates from an F distribution when r and s are even integers. In this case, both $r = 2k$ and $s = 2l$ are positive even integers, and the density function becomes

$$\begin{aligned}
f_X(x; 2k, 2l) &= \frac{\Gamma\left(\frac{2k+2l}{2}\right)}{\Gamma\left(\frac{2k}{2}\right)\Gamma\left(\frac{2l}{2}\right)} \left(\frac{2k}{2l}\right)^{\frac{2k}{2}} x^{\frac{2k}{2}-1} \left(1 + \frac{2k}{2l}x\right)^{-\frac{2k+2l}{2}} \\
&= l \frac{(k+l-1)!}{(k-1)!l!} \left(\frac{k}{l}\right)^k x^{k-1} \left(1 + \frac{k}{l}x\right)^{-(k+l)} \\
&= l \binom{k+l-1}{k-1} \left(\frac{k}{l}\right)^k x^{k-1} \left(1 + \frac{k}{l}x\right)^{-(k+l)} \tag{3}
\end{aligned}$$

Furthermore, if $k = l$ in (3), then we have

$$f_X(x; 2k, 2k) = k \binom{2k-1}{k-1} x^{k-1} (1+x)^{-2k}$$

which means

$$\int_0^\infty x^{k-1} (1+x)^{-2k} dx = \left(k \binom{2k-1}{k-1}\right)^{-1}$$

and if $l = k - 1$ in (3), then we have

$$f_X(x; 2(l+1), 2l) = l \binom{2l}{l} (1+l^{-1})^{l+1} x^l (1+(1+l^{-1})x)^{-(2l+1)}$$

We also have a special convenience if $r = 2^w$ and $s = 2^q$ for positive integers w and q , since then (3) may be applied iteratively until $f_X(x; 2^w, 2^q)$ is expressed only in terms of w and q .

7. MAPLE Implementation

The following MAPLE® module implements the analytical methods documented in this memorandum.

```

1 GenFRV:=proc(r,s,N,dg,M)
2   local ii,F::Vector,R::Vector,S::Vector,val;
3   description "Generates N Random Variates From F Distribution";
4   options 'Copyright 2020 PQI Consulting All Rights Reserved';
5   Digits:=dg;
6   R:=Vector(N); S:=Vector(N); F:=Vector(N);
7   R:=GenGRV(r/2,1,N,dg,M); S:=GenGRV(s/2,1,N,dg,M);
8   for ii from 1 to N do: F[ii]:=(s*R[ii])/(r*S[ii]); end do;
9   return evalm(F);
10 end proc;
```

8. MMIX Implementation

The following MMIX code is the implementation of the MAPLE code of the previous section for Rational Arithmetic And Conversions (RAC) values. The reference to calculating univariate (continuous) standard normal random variates is found in *PQIC Mathematical Notes, Analytical Series, Number 77: Univariate Continuous Standard Normal Distribution Random Variate Generation Using Paired Methodology*, November 2020, and to calculating univariate (continuous) gamma distribution random variates is found in *PQIC Mathematical Notes, Analytical Series, Number 76: Univariate Continuous Gamma Distribution Random Variate Generation Using Integral Transform Methodology*, October 2020, *et seq.*. All other subroutine references are to the PQICSTAT™ Fundamental Instruction Set Operation Codes (FISOC) module library.

```

1 ; PQICSTAT(tm) MMIX implementation for Generating Random Variates
2 ;   From a Univariate Continuous F Distribution with Two
3 ;   Parameter Degrees Of Freedom
4 ;
5 ; Application Specific Implementation Description
6
7 ; Copyright 2024 PQI Consulting LLC All Rights Reserved
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10 ; distributed, transmitted, displayed, published, broadcast, included within
11 ; any other software code, product, or functional unit, whether compiled,
12 ; interpreted, or otherwise made executable, nor utilized in any way whatsoever
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16 ; LLC code documentation making reference to this document may not be removed
17 ; from copies of the content.
18
19 ; Version 1.0 Build 20240115A, et seq.
20
21         PREFIX :
22 t         IS      $255
23 Buf       IS      Data_Segment
24 NVAL      GREG    10           Number of generated random variates
25 XVAL      GREG    520
```

MAPLE is a registered trademark of Maplesoft (a division of Waterloo Maple Inc.), 615 Kumpf Drive, Waterloo, Ontario, Canada, N2V 1K8. The MAPLE version used to produce the results found in this memorandum is 2017.1, June, 19, 2017, Maple Build ID 1238644.

26	RVAL	GREG	1040
27	cNVAL	BYTE	10
28	cXVAL	WYDE	520
29	cRVAL	WYDE	1040
30			
31		LOC	Buf
32		GREG	@
33	BufZZ	OCTA	@+(256-@)&255
34			
35		LOC	BufZZ
36	cTWO	GREG	@
37		BYTE	#1
38		LOC	@+1
39		WYDE	1
40		LOC	@+4
41		OCTA	#0,#0,#0,#0,#0,#0,#0,#0
42		OCTA	#0,#0,#0,#0,#0,#0,#0,#0
43		OCTA	#0,#0,#0,#0,#0,#0,#0,#0
44		OCTA	#0,#0,#0,#0,#0,#0,#0,#0
45		OCTA	#0,#0,#0,#0,#0,#0,#0,#0
46		OCTA	#0,#0,#0,#0,#0,#0,#0,#0
47		OCTA	#0,#0,#0,#0,#0,#0,#0,#0
48		OCTA	#0,#0,#0,#0,#0,#0,#0,#2
49			
50		LOC	BufZZ+8*cXVAL
51	KONE	GREG	@
52		BYTE	#1
53		LOC	@+1
54		WYDE	1
55		LOC	@+4
56		OCTA	#0,#0,#0,#0,#0,#0,#0,#0
57		OCTA	#0,#0,#0,#0,#0,#0,#0,#0
58		OCTA	#0,#0,#0,#0,#0,#0,#0,#0
59		OCTA	#0,#0,#0,#0,#0,#0,#0,#0
60		OCTA	#0,#0,#0,#0,#0,#0,#0,#0
61		OCTA	#0,#0,#0,#0,#0,#0,#0,#0
62		OCTA	#0,#0,#0,#0,#0,#0,#0,#0
63		OCTA	#0,#0,#0,#0,#0,#0,#0,#1
64			
65		LOC	BufZZ+8*2*cXVAL
66	RDF	GREG	@ First F Degrees Of Freedom
67		BYTE	#1
68		LOC	@+1
69		WYDE	1
70		LOC	@+4
71		OCTA	#0,#0,#0,#0,#0,#0,#0,#0
72		OCTA	#0,#0,#0,#0,#0,#0,#0,#0
73		OCTA	#0,#0,#0,#0,#0,#0,#0,#0
74		OCTA	#0,#0,#0,#0,#0,#0,#0,#0
75		OCTA	#0,#0,#0,#0,#0,#0,#0,#0
76		OCTA	#0,#0,#0,#0,#0,#0,#0,#0
77		OCTA	#0,#0,#0,#0,#0,#0,#0,#0
78		OCTA	#0,#0,#0,#0,#0,#0,#0,#0
79			

```

80          LOC      BufZZ+8*3*cXVAL
81 SDF      GREG      @          Second F Degrees Of Freedom
82          BYTE     #1
83          LOC      @+1
84          WYDE     1
85          LOC      @+4
86          OCTA     #0,#0,#0,#0,#0,#0,#0,#0
87          OCTA     #0,#0,#0,#0,#0,#0,#0,#0
88          OCTA     #0,#0,#0,#0,#0,#0,#0,#0
89          OCTA     #0,#0,#0,#0,#0,#0,#0,#0
90          OCTA     #0,#0,#0,#0,#0,#0,#0,#0
91          OCTA     #0,#0,#0,#0,#0,#0,#0,#0
92          OCTA     #0,#0,#0,#0,#0,#0,#0,#0
93          OCTA     #0,#0,#0,#0,#0,#0,#0,#0
94
95          LOC      BufZZ+8*4*cXVAL
96 RMEM     GREG      @
97
98          LOC      BufZZ+8*4*cXVAL+cNVAL*cRVAL
99 SMEM     GREG      @
100
101         LOC      BufZZ+8*4*cXVAL+2*cNVAL*cRVAL
102 FMEM     GREG      @
103
104         LOC      BufZZ+8*4*cXVAL+3*cNVAL*cRVAL
105 GFRVA    GREG      @
106
107         LOC      BufZZ+8*4*cXVAL+(3*cNVAL+1)*cRVAL
108 GFRVB    GREG      @
109
110         LOC      BufZZ+8*4*cXVAL+(3*cNVAL+2)*cRVAL
111 GFRVC    GREG      @
112
113         LOC      BufZZ+8*4*cXVAL+(3*cNVAL+3)*cRVAL
114 GFRVD    GREG      @
115
116         LOC      #2000
117 ;       PREFIX   :pqicstat:
118 rRES     IS       $0
119
120 Main     SWYM     0
121         SET      $1,:RDF      First F degree of freedom
122         SET      $2,:SDF      Second F degree of freedom
123         SET      $3,:FMEM     Destination of NVAL-many F(r,s) random variates
124         PUSHJ    $0,:pqicstat:RAC:pqicGFRV:Start
125 Quit     TRAP     0,Halt,0
126
127         PREFIX   :pqicstat:RAC:pqicGFRV:
128 rRA      IS       $0
129 rSB      IS       $1
130 rDEST    IS       $2
131 rNUM     IS       $3
132 rDEN     IS       $4
133 rSRR     IS       $5

```

134	rSSS	IS	\$6
135	rMEMT	IS	\$7
136	rMEMB	IS	\$8
137	rCNT	IS	\$9
138	rLMT	IS	\$10
139	rCMP	IS	\$11
140	rJ	IS	\$12
141	rTMPA	IS	\$13
142	rTMPB	IS	\$14
143	rTMPC	IS	\$15
144	rTMPD	IS	\$16
145	rTMPE	IS	\$17
146	rTMPF	IS	\$18
147	rTMPG	IS	\$19
148	rTMPAA	IS	\$20
149	rTMPBB	IS	\$21
150	rTMPCC	IS	\$22
151	rTMPDD	IS	\$23
152	rTMPEE	IS	\$24
153	rTMPFF	IS	\$25
154	rTMPGG	IS	\$26
155			
156	Start	GET	rJ, :rJ
157		SET	rNUM, :RMEM
158		SET	rDEN, :SMEM
159		SET	rSRR, :GFRVA
160		SET	rSSS, :GFRVB
161		SET	rMEMT, :GFRVC
162		SET	rMEMB, :GFRVD
163		MULU	rLMT, :RVAL, :NVAL
164		SET	rTMPB, rRA
165		SET	rTMPC, :KONE
166		SET	rTMPD, :cTWO
167		SET	rTMPE, :KONE
168		SET	rTMPF, rMEMT
169		SET	rTMPG, rMEMB
170		PUSHJ	rTMPA, :pqicstat:RAC:pqicRDIV:Start
171		SET	rTMPB, rMEMT
172		SET	rTMPC, rSRR
173		PUSHJ	rTMPA, :pqicstat:RAC:pqicXCPX:Start
174		SET	rTMPB, rMEMB
175		ADDU	rTMPC, rSRR, :XVAL
176		PUSHJ	rTMPA, :pqicstat:RAC:pqicXCPX:Start
177		SET	rTMPB, rSB
178		SET	rTMPC, :KONE
179		SET	rTMPD, :cTWO
180		SET	rTMPE, :KONE
181		SET	rTMPF, rMEMT
182		SET	rTMPG, rMEMB
183		PUSHJ	rTMPA, :pqicstat:RAC:pqicRDIV:Start
184		SET	rTMPB, rMEMT
185		SET	rTMPC, rSSS
186		PUSHJ	rTMPA, :pqicstat:RAC:pqicXCPX:Start
187		SET	rTMPB, rMEMB

```

188      ADDU      rTMPC,rSSS,:XVAL
189      PUSHJ     rTMPA,:pqicstat:RAC:pqicXCPX:Start
190      SET       rTMPB,rSRR
191      ADDU      rTMPC,rSRR,:XVAL
192      SET       rTMPD,:KONE
193      SET       rTMPE,:KONE
194      SET       rTMPF,rNUM      Numerator Gamma Random Variates
195      ADDU      rTMPG,rNUM,:XVAL
196      PUSHJ     rTMPA,:pqicstat:RAC:RACGenGRV:Start
197      SET       rTMPB,rSSS
198      ADDU      rTMPC,rSSS,:XVAL
199      SET       rTMPD,:KONE
200      SET       rTMPE,:KONE
201      SET       rTMPF,rDEN      Denominator Gamma Random Variates
202      ADDU      rTMPG,rDEN,:XVAL
203      PUSHJ     rTMPA,:pqicstat:RAC:RACGenGRV:Start
204      SET       rCNT,0
205 Cycle  ADDU      rTMPB,rNUM,rCNT
206      ADDU      rTMPD,rDEN,rCNT
207      ADDU      rTMPF,rDEST,rCNT
208      ADDU      rCNT,rCNT,:XVAL
209      ADDU      rTMPC,rNUM,rCNT
210      SET       rTMPBB,rTMPB
211      SET       rTMPCC,rTMPC
212      SET       rTMPDD,rSB
213      SET       rTMPEE,:KONE
214      SET       rTMPFF,rMEMT
215      SET       rTMPGG,rMEMB
216      PUSHJ     rTMPAA,:pqicstat:RAC:pqicRMUL:Start
217      SET       rTMPBB,rMEMT
218      SET       rTMPCC,rSRR
219      PUSHJ     rTMPAA,:pqicstat:RAC:pqicXCPX:Start
220      SET       rTMPB,rSRR
221      SET       rTMPBB,rMEMB
222      SET       rTMPCC,rSSS
223      PUSHJ     rTMPAA,:pqicstat:RAC:pqicXCPX:Start
224      SET       rTMPC,rSSS
225      ADDU      rTMPE,rDEN,rCNT
226      SET       rTMPBB,rTMPD
227      SET       rTMPCC,rTMPE
228      SET       rTMPDD,rRA
229      SET       rTMPEE,:KONE
230      SET       rTMPFF,rMEMT
231      SET       rTMPGG,rMEMB
232      PUSHJ     rTMPAA,:pqicstat:RAC:pqicRMUL:Start
233      SET       rTMPB,rSRR
234      SET       rTMPC,rSSS
235      SET       rTMPD,rMEMT
236      SET       rTMPE,rMEMB
237      SET       rTMPF,rDEST
238      ADDU      rTMPG,rDEST,rCNT      F Random Variate Calculation
239      PUSHJ     rTMPA,:pqicstat:RAC:pqicRDIV:Start
240      ADDU      rCNT,rCNT,:XVAL
241      CMPU      rCMP,rCNT,rLMT

```

242	BN	rCMP,Cycle
243	Quit	PUT :rJ,rJ
244	POP	0,0

9. Appendix: The Relationship Between The Student-T And F Distributions

When using the F distribution as the discriminator of choice in reducing linear models in designed experiments, it is commonly the case to use an $F(1, n)$ distribution, where the factor in question has only one degree of freedom with which to make a significance decision.

There is a close relationship between the (unscaled) Student-T distribution $t_n(0, 1, 1)$ density function and the $F(1, n)$ distribution density function. In particular, we have

$$P(t_n(0, 1, 1) = x) = \frac{\Gamma(\frac{n+1}{2})}{\sqrt{n\pi}\Gamma(\frac{n}{2})} \left(\frac{1}{1 + \frac{1}{n}x^2}\right)^{\frac{n+1}{2}}$$

and

$$\begin{aligned} P(F(1, n) = x) &= \frac{\Gamma(\frac{n+1}{2})}{\Gamma(\frac{1}{2})\Gamma(\frac{n}{2})} \left(\frac{1}{n}\right)^{\frac{1}{2}} x^{\frac{1}{2}-1} \left(1 + \frac{1}{n}x\right)^{-\frac{n+1}{2}} \\ &= \frac{\Gamma(\frac{n+1}{2})}{\sqrt{n\pi}\Gamma(\frac{n}{2})} \left(\frac{1}{\sqrt{x}}\right) \left(\frac{1}{1 + \frac{1}{n}x}\right)^{\frac{n+1}{2}} \end{aligned} \tag{4}$$

so that

$$\begin{aligned} \frac{P(F(1, n) = x)}{P(t_n(0, 1, 1) = x)} &= \frac{\frac{\Gamma(\frac{n+1}{2})}{\sqrt{n\pi}\Gamma(\frac{n}{2})} \left(\frac{1}{\sqrt{x}}\right) \left(\frac{1}{1 + \frac{1}{n}x}\right)^{\frac{n+1}{2}}}{\frac{\Gamma(\frac{n+1}{2})}{\sqrt{n\pi}\Gamma(\frac{n}{2})} \left(\frac{1}{1 + \frac{1}{n}x^2}\right)^{\frac{n+1}{2}}} \\ &= \frac{1}{\sqrt{x}} \left(\frac{x^2 + n}{x + n}\right)^{\frac{n+1}{2}} \end{aligned}$$

Since

$$\frac{x^2 + n}{x + n} = 1 + x \frac{x - 1}{x + n}$$

we have

$$\begin{aligned} \lim_{n \rightarrow \infty} \left(\frac{x^2 + n}{x + n}\right)^{\frac{n+1}{2}} &= \sqrt{\lim_{n \rightarrow \infty} \frac{x^2 + n}{x + n}} \sqrt{\lim_{n \rightarrow \infty} \left(\frac{x^2 + n}{x + n}\right)^n} \\ &= \sqrt{\lim_{n \rightarrow \infty} \left(1 + x \frac{x - 1}{x + n}\right)^n} \\ &= \sqrt{\left(\lim_{n \rightarrow \infty} \left(1 + x \frac{x - 1}{x + n}\right)^{\frac{x+n}{x(x-1)}}\right)^{x(x-1)}} \\ &= \sqrt{\left(\lim_{n \rightarrow \infty} \left(1 + x \frac{x - 1}{x + n}\right)^{\frac{x}{x(x-1)}}\right)^{x(x-1)}} \\ &= \sqrt{e^{x(x-1)}} \\ &= e^{\frac{1}{2}x(x-1)} \end{aligned}$$

This means random variates from $t_n(0, 1, 1)$ may be generated at the same time as for $F(1, n)$ through

$$P(F(1, n) = x) = \frac{1}{\sqrt{x}} \left(\frac{x^2 + n}{x + n}\right)^{\frac{n+1}{2}} P(t_n(0, 1, 1) = x)$$

Hence, if x_0 is a random variate from $t_n(0, 1, 1)$, then³

³ It shall be PQICSTAT™ policy to calculate the value $x_0^* = f_{F(1,n)}^{-1}(y_0)$ through the analytical method IBSFTI documented in *PQIC Mathematical Notes, Analytical Series, Number 26: The Implicit Bracketing System For The Inverse*, October 2008.

$$x_0^* = f_{F(1,n)}^{-1} \left(\frac{1}{\sqrt{x_0}} \left(\frac{x_0^2 + n}{x_0 + n} \right)^{\frac{n+1}{2}} P(t_n(0,1,1) = x_0) \right)$$

is a random variate from $F(1, n)$, and if x_1 is a random variate from $F(1, n)$, then⁴

$$x_1^* \in f_{t_n(0,1,1)}^{-1} \left(\sqrt{x_1} \left(\frac{x_1 + n}{x_1^2 + n} \right)^{\frac{n+1}{2}} P(F(1, n) = x_1) \right) \quad (5)$$

is a random variate from $t_n(0, 1, 1)$. Note that a single x_1 most likely will produce a choice of two x_1^* values (one corresponding to the “upswing” part of the density function, and the other to the “downswing” part), one of which may be unbiasedly chosen at random; furthermore, since the density function of a $t_n(0, 1, 1)$ distribution is an even function, i.e., we have

$$P(t_n(0, 1, 1) = -x) = P(t_n(0, 1, 1) = x)$$

then if $x_1^{*(1)}$ and $x_1^{*(2)}$ are the two values that come from (5), then

$$x_1^{*(1)} = -x_1^{*(2)}$$

However, since the density function of an $F(1, n)$ distribution is always decreasing, x_0^* is always unique. [Proof: From (4) we have

$$\begin{aligned} \frac{d}{dx} P(F(1, n) = x > 0) &\propto \left(\begin{array}{l} x^{-\frac{1}{2}} \frac{d}{dx} \left(1 + \frac{1}{n}x \right)^{-\frac{n+1}{2}} \\ + \left(1 + \frac{1}{n}x \right)^{-\frac{n+1}{2}} \frac{d}{dx} x^{-\frac{1}{2}} \end{array} \right) \\ &= \left(\begin{array}{l} -\frac{n+1}{2n} x^{-\frac{1}{2}} \left(1 + \frac{1}{n}x \right)^{-\frac{n+3}{2}} \\ -\frac{1}{2} x^{-\frac{3}{2}} \left(1 + \frac{1}{n}x \right)^{-\frac{n+1}{2}} \end{array} \right) \\ &= -\frac{1}{2n} ((n+2)x + n) x^{-\frac{3}{2}} \left(1 + \frac{1}{n}x \right)^{-\frac{n+3}{2}} \\ &< 0 \end{aligned}$$

since $n, x > 0$.]

We also have

$$\lim_{n \rightarrow \infty} \frac{P(F(1, n) = x)}{P(t_n(0, 1, 1) = x)} = \frac{e^{\frac{1}{2}x(x-1)}}{\sqrt{x}}$$

and since $\lim_{n \rightarrow \infty} P(F(1, n) = x) \sim \chi_1^2 = \frac{e^{-\frac{1}{2}x}}{\sqrt{2\pi x}}$ by Claim 6, we have

$$\lim_{n \rightarrow \infty} P(t_n(0, 1, 1) = x) = \sqrt{x} e^{-\frac{1}{2}x(x-1)} \lim_{n \rightarrow \infty} P(F(1, n) = x)$$

While the lower bound for this method is clearly 0, the upper bound must be found iteratively by increasing an argument x_c until $f_{F(1,n)}(x_c) < y_0$. This cycle ends when the first instance of such a value x_{c^*} is found. The bracketing interval to find x_0^* is then $[0, x_{c^*}]$. The incremental order of magnitude in the iterative step is at the discretion of the implementing analyst.

⁴ It shall be PQICSTAT™ policy to find the positive value of $0 < x_1^* = f_{t_n(0,1,1)}^{-1}(y_0)$ through the analytical method **IBSFTI** documented in *op. cit.*, with a lower bound of 0 and upper bound found by the same iterative method as would be used for the $F(1, n)$ density function case. The lower bound is the peak value of this density function, which marks the value at which the inverse function is a single point. [Proof: We have

$$\begin{aligned} 0 &= \frac{d}{dx} f_{t_n(0,1,1)}(x_p) \propto \frac{d}{dx} \left(\frac{1}{1 + \frac{1}{n}x_p^2} \right)^{\frac{n+1}{2}} \\ &= \frac{n+1}{2} \left(\frac{2x_p}{n} \right) \left(\frac{1}{1 + \frac{1}{n}x_p^2} \right)^{\frac{n-1}{2}} \\ &= \left(1 + \frac{1}{n} \right) x_p \left(\frac{1}{1 + \frac{1}{n}x_p^2} \right)^{\frac{n-1}{2}} \\ &\implies x_p = 0 \end{aligned}$$

since $1 + \frac{1}{n}x^2 \geq 1 > 0$ for all $x \in \mathbb{R}$.

$$\begin{aligned}
&= \left(\sqrt{x} e^{-\frac{1}{2}x(x-1)} \right) \left(\frac{e^{-\frac{1}{2}x}}{\sqrt{2\pi x}} \right) \\
&= \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x^2} \\
&\sim N(0, 1)
\end{aligned}$$

so that asymptotically the $t_n(0, 1, 1)$ distribution agrees with the standard normal distribution (and may be used to approximate $t_n(0, 1, 1)$ for large n , although the convergence is especially slow for large x).

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