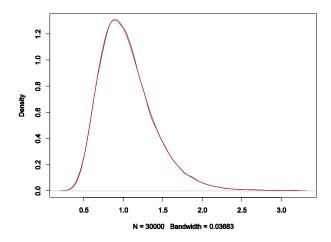
I observed the similarity between the lognormal distribution and the distribution of the sample median that comes from a lognormal population. The chart below shows the density (in black) of the sample median (with sample size 15) coming from LogN(0,1) and the fitted lognormal density (in red) using method of moments, with $\mu=-0.0022$, $\sigma=0.3206$.



Since this is not a journal article, I don't write it in an academic manner. Still, for the sake of readiness, I use the following notations for certain special functions.

The CDF of normal distribution	$\Phi(w) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{w} \exp(-t^2) dt$
Hypergeometric function	${}_{2}F_{1}(a,b,c;w) = \sum_{n=0}^{\infty} \frac{(a)_{n}(b)_{n}}{(c)_{n}} \frac{w^{n}}{n!}; (a)_{n} = \begin{cases} 1 & n=0\\ a(a+1)\cdots(a+n-1) & n>0 \end{cases}$
Error function	$erf(w) = \frac{2}{\sqrt{\pi}} \int_0^w \exp(-t^2) dt$
Beta function	$B(a,b) = \int_0^1 t^{a-1} (1-t)^{b-1} dt$

Also, given the CDF of the parent population G(x), the sample median distribution is given by,

$$\int_{-\infty}^{x} \frac{1}{B(k+1,k+1)} G^{k}(y) [1 - G(y)]^{k} dG(y)$$

where sample size n=2k+1. In this paper, $G(x)=\Phi\left(\frac{\ln(x)-\mu}{\sigma}\right)$. Also, having this substituted then solving this integral gives the exact sample median distribution,

$$\frac{\Phi^{k+1}\left(\frac{\ln(x)-\mu}{\sigma}\right)}{(k+1)B(k+1,k+1)} \, {}_2F_1\left(k+1,-k,k+2;\Phi\left(\frac{\ln(x)-\mu}{\sigma}\right)\right)$$

Formally I found the lognormal approximation to the distribution of the sample median that comes from a lognormal parent,

$$\lim_{k \to \infty} \frac{\Phi^{k+1}\left(\frac{\ln(x) - \mu}{\sigma}\right)}{(k+1)B(k+1, k+1)} {}_{2}F_{1}\left(k+1, -k, k+2; \Phi\left(\frac{\ln(x) - \mu}{\sigma}\right)\right) = \Phi\left(\frac{\ln(x) - \mu}{\sigma B(k+1, k+1)4^{k}\sqrt{2}}\right)$$

That is, the sample median distribution can be approximated by a lognormal distribution with the two parameters as μ and $\sigma B(k+1,k+1)4^k$.

Alternatively, in terms of PDF,

$$\frac{\exp\left\{-\frac{1}{2}\left(\frac{\ln(x)-\mu}{\sigma}\right)^{2}\right\}}{x\sigma\sqrt{2\pi}B(k+1,k+1)}\Phi\left(\frac{\ln(x)-\mu}{\sigma}\right)^{k}\left[1-\Phi\left(\frac{\ln(x)-\mu}{\sigma}\right)\right]^{k} \approx \frac{\exp\left\{-\frac{1}{2}\left(\frac{\ln(x)-\mu}{\sigma B(k+1,k+1)4^{k}}\right)^{2}\right\}}{x\sigma B(k+1,k+1)4^{k}\sqrt{2\pi}}$$

The proof requires the following approximation.

Proposition (Approximating the error function):

$$\operatorname{erf}(w) \approx \operatorname{sgn}(w) \sqrt{1 - \exp\left\{-w^2 \frac{4}{\pi}\right\}}$$

with the maximum absolute error of 6.29×10^{-3} at $w \approx 1.169$ and $sgn(\cdot)$ is the sign function.

This approximation is based on the uniform approximation in Winitzki (2013) with the use of Padé approximant which is given by,

$$\operatorname{erf}(w) \approx \operatorname{sgn}(w) \sqrt{1 - \exp\left\{-w^2 \frac{P(w)}{Q(w)}\right\}}$$

where P(w) and Q(w) are polynomials.

Taking P(w) and Q(w) with the degree of two, Nealy, et al (2010) then approximates that,

$$\operatorname{erf}(w) \approx \operatorname{sgn}(w) \sqrt{1 - \exp\left\{-w^2 \frac{\frac{4}{\pi} + aw^2}{1 + aw^2}\right\}}$$

where $a=\frac{8(\pi-3)}{3\pi(4-\pi)}$. This approximation used here further reduces the degree of the polynomials to one.

Following the above error function approximation, the main proof is given below,

$$\frac{\exp\left\{-\frac{1}{2}\left(\frac{\ln(x)-\mu}{\sigma}\right)^{2}\right\}}{x\sigma\sqrt{2\pi}B(k+1,k+1)}\Phi\left(\frac{\ln(x)-\mu}{\sigma}\right)^{k}\left[1-\Phi\left(\frac{\ln(x)-\mu}{\sigma}\right)\right]^{k}$$

$$=\frac{\exp\left\{-\frac{1}{2}\left(\frac{\ln(x)-\mu}{\sigma}\right)^{2}\right\}}{x\sigma\sqrt{2\pi}B(k+1,k+1)}\left[1-erf^{2}\left(\frac{\ln(x)-\mu}{\sigma\sqrt{2}}\right)\right]^{k}$$

Then,

$$\frac{\exp\left\{-\frac{1}{2}\left(\frac{\ln(x)-\mu}{\sigma}\right)^{2}\right\}}{x\sigma\sqrt{2\pi}B(k+1,k+1)}\left[1-erf^{2}\left(\frac{\ln(x)-\mu}{\sigma\sqrt{2}}\right)\right]^{k}\approx\frac{\exp\left\{-\frac{\left(\ln(x)-\mu\right)^{2}}{2\sigma^{2}B^{2}(k+1,k+1)4^{2k}}\right\}}{xB(k+1,k+1)4^{k}\sigma\sqrt{2\pi}}$$

$$\Rightarrow\left[1-erf^{2}\left(\frac{\ln(x)-\mu}{\sigma\sqrt{2}}\right)\right]^{k}\approx\frac{1}{xB(k+1,k+1)4^{k}\sigma\sqrt{2\pi}}\exp\left\{-\frac{\left(\ln(x)-\mu\right)^{2}}{2\sigma^{2}B^{2}(k+1,k+1)4^{2k}}\right\}\frac{B(k+1,k+1)4^{k}}{f(x;\mu,\sigma)}$$

$$\Rightarrow\left[1-erf^{2}\left(\frac{\ln(x)-\mu}{\sigma\sqrt{2}}\right)\right]^{k}\approx\frac{\exp\left\{-\frac{\left(\ln(x)-\mu\right)^{2}}{2\sigma^{2}B^{2}(k+1,k+1)4^{2k}}\right\}}{\exp\left\{-\frac{\left(\ln(x)-\mu\right)^{2}}{2\sigma^{2}}\right\}}$$

$$\Rightarrow\left[1-erf^{2}\left(\frac{\ln(x)-\mu}{\sigma\sqrt{2}}\right)\right]^{k}\approx\left[\exp\left\{-\frac{\left(\ln(x)-\mu\right)^{2}}{2\sigma^{2}}\right\}\right]^{\left(\frac{1}{H}-1\right)}$$

$$\Rightarrow\left[1-erf^{2}\left(\frac{\ln(x)-\mu}{\sigma\sqrt{2}}\right)\right]^{k}\approx\left[\exp\left\{-\frac{\left(\ln(x)-\mu\right)^{2}}{2\sigma^{2}}\right\}\right]^{\left(\frac{1}{H}-1\right)}$$

where $H = 4^{2k} B^2 (k+1, k+1)$.

For x cannot be negative, apply the error function approximation on the LHS and rearrange to get,

$$\exp\left\{-\frac{(\ln(x)-\mu)^{2}}{2\sigma^{2}}\frac{4k}{\pi}\right\} \approx \left[\exp\left\{-\frac{(\ln(x)-\mu)^{2}}{2\sigma^{2}}\right\}\right]^{\left(\frac{1}{H}-1\right)}$$

$$\Rightarrow \left[\exp\left\{-\frac{(\ln(x)-\mu)^{2}}{2\sigma^{2}}\right\}\right]^{\frac{4k}{\pi}} \approx \left[\exp\left\{-\frac{(\ln(x)-\mu)^{2}}{2\sigma^{2}}\right\}\right]^{\left(\frac{1}{H}-1\right)}$$

As noted earlier, the maximum absolute error of the error function approximation is 6.29×10^{-3} . The resulting error is not as large as it seems; $\left| \operatorname{erf}(w) \right|$ is strictly smaller than 1 hence the error resulted from $\left[1 - \operatorname{erf}(w) \right]^k$ is much smaller as k increases.

We now have to show that $\frac{4k}{\pi} \sim \frac{1}{H} - 1$. Given that $H = 2^{4k} B^2 (k+1,k+1)$, we first simplify the beta function.

$$B(k+1,k+1) = B(k,k) \frac{k^2}{4k^2+2k}$$
 . Then by Stirling's formula,

$$B(k+1,k+1) = B(k,k) \frac{k^2}{4k^2 + 2k} \approx \sqrt{2\pi} \frac{k^{2k-1}}{(2k)^{2k-0.5}} \frac{k^2}{4k^2 + 2k} = \frac{\sqrt{\pi k}}{2^{2k}(2k+1)}$$

Therefore,
$$H=2^{4k}B^2(k+1,k+1)\approx \frac{\pi k}{(2k+1)^2}$$
 and it is straight forward that $\lim_{k\to\infty}\frac{\frac{1}{H}-1}{4k/\pi}=1$

A final note regarding the approximation error and the rate of convergence:

- The maximum absolute error depends only on the sample size.
- It has faster rate of convergence than the asymptotically normal sample median distribution introduced in Cramer (1946, p.369).
- (Hence, I still haven't shown how the lognormal approximation proposed here eventually converges to the one in Cramer (1946, p.369), recalling the normal approximation to lognormal distribution)

References:

Cramer H. (1946) Mathematical Methods of Statistics, Princeton: Princeton University Press.

Nealy JE, Chang CK, Norman RB, Blattnig SR, Badavi FF, Adamczyk AM. (2010) A deterministic transport code for space environment electrons, NASA Technical Documents, NASA/TP-2010-216168, Hampton: NASA Langley Research Center.

Winitzki S. (2013) Uniform approximations for transcendental functions in Lecture Notes in Computer Science, 2667:780-789.