2.Log-normal distribution and the Gini coefficient

We consider a sample of 8 observations representing monthly incomes (in thousands of Swedish Krona). Assume the data follow a log-normal distribution:

$$y_1, ..., y_n | \mu, \sigma^2 \sim \log N(\mu, \sigma^2)$$

with known $\mu = 3.65$ and unknown variance σ^2 . For the variance, we use a non-informative prior:

$$p(\sigma^2) \propto 1/\sigma^2$$

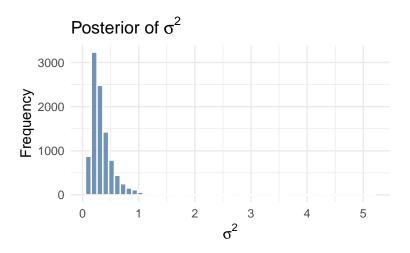
Under this model, the posterior distribution for σ^2 follows a scaled inverse-chi-squared distribution:

Scale-inv-
$$\chi^2(n, \tau^2)$$
, where $\tau = \frac{\sum_{i=1}^n (\log y_i - \mu)^2}{n}$

To draw samples from this posterior, we use the fact that if:

$$\sigma \sim \text{Scale-inv-}\chi^2(n,\tau^2), \text{ then } \sigma^2 = \frac{n\tau^2}{\chi_n^2}$$

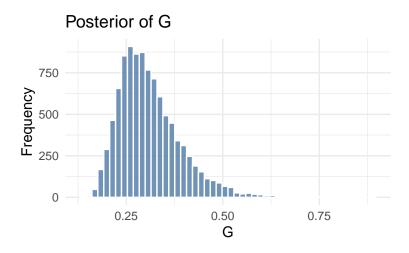
We generate 10000 posterior draws for σ^2 and visualize the resulting distribution below.



Then, the Gini coefficient is calculated using the formula:

$$G = 2\Phi(\sigma/\sqrt{2}) - 1$$

where Φ denotes the cumulative distribution function of the standard normal distribution, and $\sigma = \sqrt{\sigma^2}$. Using the posterior draws of σ^2 , we obtain the posterior distribution of the Gini coefficient:



Using the posterior distribution of the Gini coefficient G, we compute both a 95% equal-tail credible interval and a 95% Highest Posterior Density Interval (HPDI) for G.

Table 1: Confidence intervals for G

	Lower.bound	Upper.bound
Equal tail	0.1930103	0.5080196
HPDI	0.1757738	0.4726963

The equal-tail interval removes 2.5% of the posterior mass from each tail, while the HPDI represents the narrowest interval containing 95% of the posterior mass. In this case, the HPDI is slightly narrower and starts at a lower bound compared to the equal-tail interval. This suggests that the posterior distribution of G is slightly right-skewed, and the HPDI better captures the region of highest probability. Therefore, if interpretability and precision are important, the HPDI may be preferred.