

reference priors

Dr. Merlise Clyde

objective prior distributions

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let prior sample size n_0 go to zero, s_0^2 go to zero and
 $v_0 = n_0 - 1$ go to -1

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$$s_n^2 = \frac{1}{v_n} \left[s_0^2 v_0 + s^2 (n - 1) + \frac{n_0 n}{n_n} (\bar{Y} - m_0)^2 \right] \rightarrow s^2$$

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- ▶ Non-Generative or “improper prior distribution”
- ▶ “Formal” posterior distribution, proper distribution if $n \geq 2$

$$\text{NormalGamma}(\bar{Y}, n, s^2, n - 1)$$

$$\mu \mid \sigma^2, \text{data} \sim \text{N}(\bar{Y}, \sigma^2/n)$$

$$1/\sigma^2 \mid \text{data} \sim \text{Gamma}((n - 1)/2, s^2(n - 1)/2)$$

reference analysis

- ▶ under the reference prior $p(\mu, \sigma^2) \propto 1/\sigma^2$

$$\frac{\mu - \bar{Y}}{\sqrt{s^2/n}} \mid \text{data} \sim t(n-1, 0, 1)$$

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- ▶ intervals estimates $(\bar{Y} - t_{1-\alpha/2}s/\sqrt{n}, \bar{Y} + t_{1-\alpha/2}s/\sqrt{n})$

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- ▶ Bayesian allows probability statements after seeing the data

$$P(\bar{Y} - t_{1-\alpha/2}s/\sqrt{n} < \mu < \bar{Y} + t_{1-\alpha/2}s/\sqrt{n}) = 1 - \alpha$$

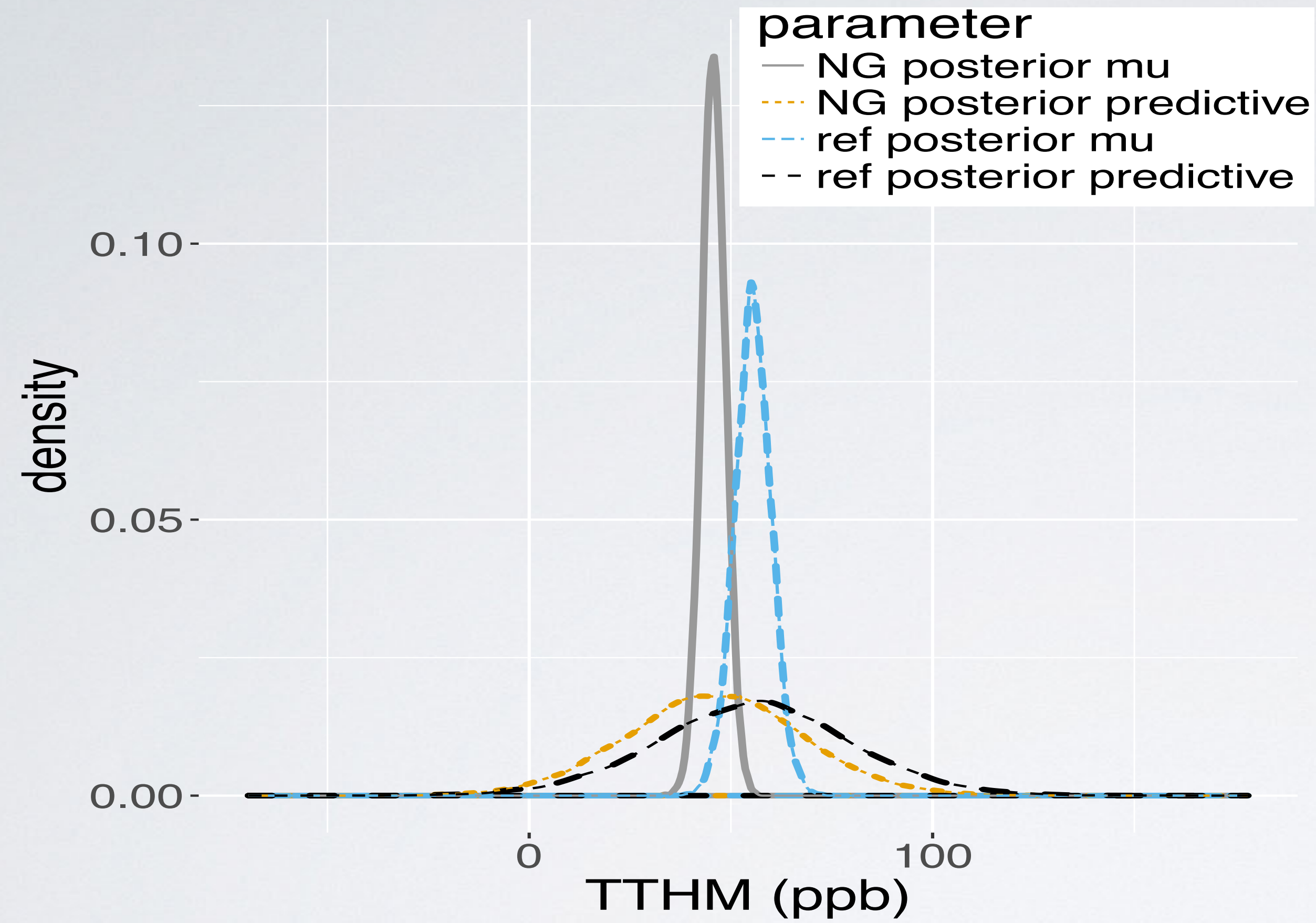
tap water example: reference analysis

R Code

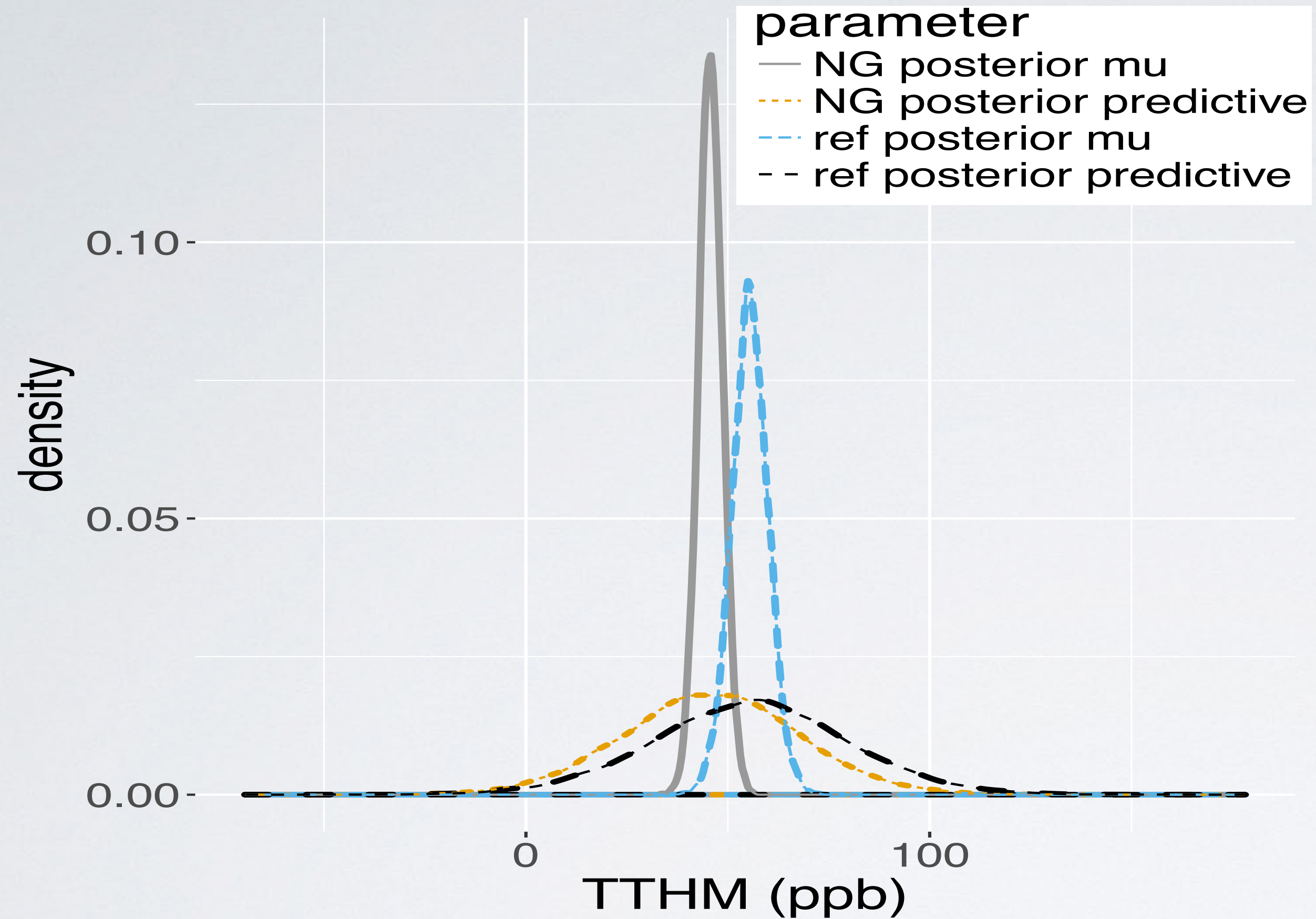
```
phi = rgamma(10000, (n-1)/2, s2*(n-1)/2)
sigma = 1/sqrt(phi)
post_mu = rnorm(10000, mean=ybar, sd=sigma/(sqrt(n)))
pred_y = rnorm(10000, post_mu, sigma)
quantile(pred_y, c(.025, .975))

##          2.5%          97.5%
## 6.692877 104.225954
```

comparison of posterior densities



comparison of posterior densities



```
sum(pred_y > 80)/length(pred_y) #  $P(Y > 80 \mid data)$ 
```

```
## [1] 0.1534
```

summary

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- ▶ sensitivity analysis

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- ▶ check assumptions of model and prior

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next: mixtures of conjugate priors,
robustness and MCMC