## BART update for tree changes

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Suppose there are n observations in a terminal node and suppose that the (partial) residuals in this terminal node are denoted  $R_1, \ldots, R_n$ . The prior distribution for these residuals is:

$$R_1,\ldots,R_n|\mu,\tau\sim N(\mu,\tau^{-1})$$

Furthermore the prior on  $\mu$  is:

$$\mu \sim N(0, \tau_{\mu}^{-1})$$

Using  $\pi$  to denote a probability distribution, we want to find:

$$\pi(R_1, \dots, R_n | \sigma) = \int \pi(R_1, \dots, R_n | \mu, \tau) \pi(\mu) \partial \mu$$

$$\propto \prod_{i=1}^n \tau^{1/2} e^{-\frac{\tau}{2} (R_i - \mu)^2} \tau_{\mu}^{1/2} e^{-\frac{\tau_{\mu}}{2} \mu^2} \partial \mu$$

$$= \int \tau^{n/2} e^{-\frac{\tau}{2} \sum (R_i - \mu)^2} \tau_{\mu}^{1/2} e^{-\frac{\tau_{\mu}}{2} \mu^2} \partial \mu$$

$$= \int \tau^{n/2} \tau_{\mu}^{1/2} e^{-\frac{1}{2} \left[\tau \left\{\sum R_i^2 + n\mu^2 - 2\mu n\bar{R}\right\} + \tau_{\mu}\mu^2\right]} \partial \mu$$

$$= \tau^{n/2} \tau_{\mu}^{1/2} e^{-\frac{1}{2} \left[\tau \sum R_i^2\right]} \int e^{-\frac{1}{2}Q} \partial \mu$$

where

$$\begin{split} Q = & \tau n \mu^2 - 2\tau n \mu \bar{R} + \tau_\mu \mu^2 \\ = & (\tau_\mu + n\tau) \mu^2 - 2\tau n \mu \bar{R} \\ = & (\tau_\mu + n\tau) \left[ \mu^2 - \frac{2\tau n \mu \bar{R}}{\tau_\mu + n\tau} \right] \\ = & (\tau_\mu + n\tau) \left[ \left( \mu - \frac{2\tau n \bar{R}}{\tau_\mu + n\tau} \right)^2 - \left( \frac{\tau n \bar{R}}{\tau_\mu + n\tau} \right)^2 \right] \\ = & (\tau_\mu + n\tau) \left( \mu - \frac{2\tau n \bar{R}}{\tau_\mu + n\tau} \right)^2 - \frac{(n\tau \bar{R})^2}{\tau_\mu + n\tau} \end{split}$$

so therefore:

$$\int e^{-\frac{1}{2}Q} \partial \mu = \int \exp\left[-\frac{\tau_{\mu} + n\tau}{2} \left(\mu - \frac{2\tau n\bar{R}}{\tau_{\mu} + n\tau}\right)^{2} + \frac{(n\tau\bar{R})^{2}}{2(\tau_{\mu} + n\tau)}\right] \partial \mu$$

$$\propto \exp\left[\frac{1}{2} \frac{(\tau n\bar{R})^{2}}{\tau_{\mu} + n\tau}\right] (\tau_{\mu} + n\tau)^{-1/2}$$

And finally:

$$\pi(R_1, \dots, R_n | \tau) \propto (\tau_\mu + n\tau)^{-1/2} \tau^{n/2} \tau_\mu^{1/2} \exp\left[\frac{1}{2} \frac{(\tau n\bar{R})^2}{\tau_\mu + n\tau}\right] \exp\left[-\frac{\tau}{2} \sum R_i^2\right]$$
$$= \tau^{n/2} \left(\frac{\tau_\mu}{\tau_\mu + n\tau}\right)^{1/2} \exp\left[-\frac{\tau}{2} \left\{\sum R_i^2 - \frac{\tau(n\bar{R})^2}{\tau_\mu + n\tau}\right\}\right]$$

## Including multiple terminal nodes

When we put back in terminal nodes we write  $R_{ji}$  where j is the terminal node and i is still the observation, so in terminal node j we have partial residuals  $R_{j1}, \ldots, R_{jn_j}$ . When we have  $j = 1, \ldots, b$  terminal nodes the full conditional distribution is then:

$$\prod_{j=1}^{b} \pi(R_{j1}, \dots, R_{jn_j} | \tau) \propto \prod_{j=1}^{b} \left\{ \tau^{n_j/2} \left( \frac{\tau_{\mu}}{\tau_{\mu} + n_j \tau} \right)^{1/2} \exp \left[ -\frac{\tau}{2} \left\{ \sum_{j=1}^{n_j} R_{ji}^2 - \frac{\tau(n_j \bar{R}_j)^2}{\tau_{\mu} + n_j \tau} \right\} \right] \right\}$$

which on the log scale gives:

$$\sum_{j=1}^{b} \left\{ \frac{n_j}{2} \log(\tau) + \frac{1}{2} \log\left(\frac{\tau_{\mu}}{\tau_{\mu} + n_j \tau}\right) - \frac{\tau}{2} \left[ \sum_{i=1}^{n_j} R_{ji}^2 - \frac{\tau(n_j \bar{R}_j)^2}{\tau_{\mu} + n_j \tau} \right] \right\}$$

## Updating $\mu$

The full conditional for  $\mu_j$  (the terminal node parameters for node j) is similar to the above but without the integration:

$$\pi(\mu_j|\dots) \propto \prod_{i=1}^{n_j} \tau^{1/2} e^{-\frac{\tau}{2}(R_{ji} - \mu_j)^2} \tau_{\mu}^{1/2} e^{-\frac{\tau_{\mu}}{2}\mu_j^2}$$

$$\propto e^{-\frac{\tau}{2} \sum_{i=1}^{n_j} (R_{ji} - \mu_j)^2} e^{-\frac{\tau_{\mu}}{2}\mu_j^2}$$

$$\propto e^{-\frac{\tau}{2} \left[ n_j \mu_j^2 - 2\mu_j \sum_{i=1}^{n_j} R_{ji} \right] - \frac{\tau_{\mu}}{2} \mu_j^2}$$

$$\propto e^{-\frac{Q}{2}}$$

Write  $S_R = \sum_{i=1}^{n_j} R_{ji}$  and so:

$$\begin{split} Q = & n_j \tau \mu_j^2 - 2\mu_j \tau S_R + \tau_\mu \mu_j^2 \\ = & (n_j \tau + \tau_\mu) \mu_j^2 - 2\tau \mu_j S_R \\ = & (n_j \tau + \tau_\mu) \left[ \mu_j^2 - \frac{2\tau \mu_j S_R}{n_j \tau + \tau_\mu} \right] \\ \propto & (n_j \tau + \tau_\mu) \left[ \mu_j - \frac{\tau \mu_j S_R}{n_j \tau + \tau_\mu} \right]^2 \end{split}$$

so therefore:

$$\mu_j | \dots \sim N \left( \frac{\tau S_R}{n_j \tau + \tau_\mu}, (n_j \tau + \tau_\mu)^2 \right)$$

## Update for $\tau$

I am using the scale/rate parameterisation of the gamma with prior  $\tau \sim Ga(\nu/2, \nu\lambda/2)$ . Letting  $\mu_i$  be the prediction of the *i*th observation we get:

$$\pi(\tau|\ldots) \propto \prod_{i=1}^{n} \tau^{1/2} e^{-\frac{\tau}{2}(y_i - \mu_i)^2} \tau^{\nu/2} e^{-\tau\nu\lambda/2}$$

Letting  $S = \sum_{i=1}^{n} (y_i - \mu_i)^2$  we get:

$$\pi(\tau|\ldots) \propto \tau^{n/2} e^{-\frac{\tau}{2}S} \tau^{\nu/2} e^{-\tau\nu\lambda/2}$$
$$= \tau^{(n+\nu)/2} e^{-\frac{\tau}{2}(S+\nu\lambda)}$$

so

$$\tau | \ldots \sim Ga\left(\frac{n+\nu}{2}, \frac{S+\nu\lambda}{2}\right)$$