Statistical Modelling and Inference: Week 3 Prior Modelling

Prior Modelling ** 1. Extend your program to also learn q from the data. Work with the Normal.Gamma prior with specification**

$$a = -\frac{1}{2}, b = 0, \mu = 0, D = \delta I$$

• Produce a figure that includes posterior draws of the linear predictor for this model

2.1 Priors that penalize the L norm of the mean of a Gaussian and MAP estimation

1. Suppose for simplicity that a > 0 (a < 0 can be handled in the same way), assume also that $\lambda > 0$ and consider the function

$$f(\mu) = (\mu - a)^2 + \lambda |\mu|$$

Show that μ is minimised at $(a - \lambda/2)^+$ where x^+ denotes the positive part of x.

We take the derivative and equate to zero to find the λ that maximizes the L norm

$$\frac{\partial f(\mu)}{\partial \mu} = 2(\mu - a)^2 + \lambda |\mu| = 0$$

$$\lambda = -2(\mu - a)$$

$$\mu = a - \frac{\lambda}{2}$$

And we can check that it is indeed a minimum as the second derivative is > 0:

$$\frac{\partial}{\partial \mu} \left(\frac{\partial f(\mu)}{\partial \mu} \right) = 2 > 0$$

2. Find \mathbf{w}_{MAP} under this model in closed form

 $\mathbf{w}_{MAP} = arg \ max \ log \ p(w|t) = arg \ max \ log \ p(t|w) + log \ p(w)$

$$= arg \ max \ log \ \prod_{i=1}^{N} p(t_n|w) + log \ e^{-\frac{\delta}{2} \sum_{i=1}^{M+1} |w_i|}$$

$$= arg \ max \ \sum_{i=1}^{N} \ log \ N(w_n, q^{-1}I) + log \ e^{-\frac{\delta}{2} \sum_{i=1}^{M+1} |w_i|}$$

$$= arg \ max \ \sum_{i=1}^{N} \ c \ log \ e^{-\frac{1}{2}q(t_n-w)^T(t-w)} + log \ e^{-\frac{\delta}{2}\sum_{i=1}^{M+1}|w_i|}$$

$$= arg \ max \ \sum_{i=1}^{N} \ -\frac{c}{2}q(t_n - w)^T(t - w) - \frac{\delta}{2} \sum_{i=1}^{M+1} |w_i|$$

$$\propto min \sum_{i=1}^{N} q(t_n - w)^T (t - w) + \delta \sum_{i=1}^{M+1} |w_i|$$

$$= min \ q \sum_{i=1}^{N} \sum_{i=1}^{M+1} (t_n - w)^T (t - w) + \delta \sum_{i=1}^{M+1} |w_i|$$

Taking the derivative and equating to 0,

$$\frac{\partial}{\partial w_i} = 0$$

$$-2q \sum_{i=1}^{N} (t_{ni} - w_i) + \delta = 0$$

$$-Nw_i + \sum_{i=1}^{N} t_{ni} = \frac{\delta}{2} q^{-1}$$

$$w_i = \frac{1}{N} \left(\sum_{i=1}^{N} t_{ni} - \frac{\delta}{2} q^{-1} \right)$$