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 CS 542        Problem Set 3  
 Discussed with classmates  
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# 1. Written Problems

## a. Bishop 5.3

$$p(T|X, w, \Sigma) = \prod_{n=1}^N N(t_n|y(x_n, w), \Sigma)$$

Let  $y_n = y(x_n, w)$  and  $K = \text{dimensionality of } y \text{ and } t$ , then

$$\ln p(T|X, w, \Sigma) = -\frac{N}{2}(\ln|\Sigma| + K\ln(2\pi)) - \frac{1}{2}\sum_{n=1}^N (t_n - y_n)^T \Sigma^{-1} (t_n - y_n)$$

If  $\Sigma$  is fixed and known, then we would have

$$E(w) = \frac{1}{2} \sum_{n=1}^N (t_n - y_n)^T \Sigma^{-1} (t_n - y_n)$$

If otherwise, then we have

$$\begin{aligned} \ln p(T|X, w, \Sigma) &= -\frac{N}{2} \ln|\Sigma| - \frac{1}{2} \sum_{n=1}^N (t_n - y_n)^T \Sigma^{-1} (t_n - y_n) \\ &= -\frac{N}{2} \ln|\Sigma| - \frac{1}{2} \text{Tr}[\Sigma^{-1} \sum_{n=1}^N (t_n - y_n)(t_n - y_n)^T] \end{aligned}$$

By setting derivative w.r.t.  $\Sigma^{-1} = 0$ ,

$$\Sigma = \frac{1}{N} \sum_{n=1}^N (t_n - y_n)(t_n - y_n)^T$$

## b. Bishop 5.4

Let  $k \in \{0,1\}$  be the true class label.

$$\begin{aligned} p(t = 1|x) &= p(t = 1|k = 0)p(k = 0|x) + p(t = 1|k = 1)p(k = 1|x) \\ &= (1 - \epsilon)y(x, w) + \epsilon(1 - y(x, w)) \end{aligned}$$

$$p(t|x) = p(t = 1|x)^t (1 - p(t = 1|x))^{1-t}$$

$$\begin{aligned} E(w) &= - \sum_{n=1}^N t_n \ln[(1 - \epsilon)y(x_n, w) + \epsilon(1 - y(x_n, w))] \\ &\quad + (1 - t_n) \ln[1 - (1 - \epsilon)y(x_n, w) - \epsilon(1 - y(x_n, w))] \end{aligned}$$

## c. Bishop 5.26

$$\Omega_n = \frac{1}{2} \sum_k (\mathcal{G}y_k)^2 \Big|_{x_n}$$

Since  $J_{ki} \equiv \frac{\partial y_k}{\partial x_i}$ , we have

$$\Omega_n = \frac{1}{2} \sum_k \left( \sum_i \tau_{ni} \frac{\partial y_{nk}}{\partial x_{ni}} \right)^2, \text{ which is equivalent to (5.128)}$$

Now, we have  $z_j = h(a_j)$ ,  $a_j = \sum_i w_{ji} z_i$ ,  $\alpha_j = h'(\alpha_j) \beta_j$ ,  $\beta_j = \sum_i w_{ji} \alpha_i$ ,  $\alpha_j \equiv \mathcal{G}z_j$ ,  $\beta_j \equiv \mathcal{G}a_j$

$$\beta_{nj} = \sum_i w_{ji} \alpha_{ni} = \sum_i w_{ji} \mathcal{G}x_{ni} = \sum_i w_{ji} \sum_{i'} \tau_{ni'} \frac{\partial y_{nk}}{\partial x_{ni'}} = \sum_i w_{ji} \tau_{ni}$$

As  $\tau_n$  is propagated forward by  $\alpha_j = h'(\alpha_j)\beta_j, \beta_j = \sum_i w_{ji}\alpha_i$ , we have  $\beta_{nl}$  for the output layer, and  $\Omega_n$  can be computed as

$$\Omega_n = \frac{1}{2} \sum_k (\mathcal{G}y_k)^2 = \frac{1}{2} \sum_k \alpha_{nk}^2$$

For the derivatives,

$$\text{Let } \delta_{kr} \equiv \frac{\partial y_k}{\partial a_r}, \phi_{kr} \equiv \mathcal{G}\delta_{kr}$$

$$\text{Then } \frac{\partial \Omega_n}{\partial w_{rs}} = \sum_k (\mathcal{G}y_{nk}) \mathcal{G}(\delta_{nkr} z_{ns}) = \sum_k \alpha_{nk} (\phi_{kr} z_{ns} + \delta_{kr} \alpha_s)$$

With 5.74, we can get

$$\delta_{nkr} = h'(a_{nr}) \sum_l w_{lr} \delta_{nkl}$$

$$\phi_{nkr} = \mathcal{G}\delta_{nkr} = \mathcal{G}\left(h'(a_{nr}) \sum_l w_{lr} \delta_{nkl}\right) = h''(a_{nr}) \beta_{nr} \sum_l w_{lr} \delta_{nkl} + h'(a_{nr}) \sum_l w_{lr} \phi_{nkl}$$

## 2. Programming

Available in 'starter' folder.