

ECE368: Probabilistic Reasoning

Lab 2: Bayesian Linear Regression

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You should hand in: 1) A scanned .pdf version of this sheet with your answers (file size should be under 2 MB); 2) four figures for Question 2 and three figures for Question 4 in the .pdf format; and 3) one Python file `regression.py` that contains your code. All these files should be uploaded to Quercus.

1. Express the posterior distribution $p(\mathbf{a}|x_1, z_1, \dots, x_N, z_N)$ using $\sigma^2, \beta, x_1, z_1, x_2, z_2, \dots, x_N, z_N$. (1 pt)

$$\begin{aligned}
 p(\tilde{\mathbf{a}}|x_1, z_1, \dots, x_N, z_N) &= \mathcal{N}(\mu_{\tilde{\mathbf{a}}}, \Sigma_{\tilde{\mathbf{a}}}) \\
 \Sigma_{\tilde{\mathbf{a}}} &= \Sigma_{\tilde{\mathbf{a}}\tilde{\mathbf{a}}|x_1, z_1, \dots, x_N, z_N} = (\Sigma_{\tilde{\mathbf{a}}\tilde{\mathbf{a}}}^{-1} + A^T \Sigma_{\mathbf{w}\mathbf{w}}^{-1} A)^{-1} \\
 \mu_{\tilde{\mathbf{a}}} &= \mu_{\tilde{\mathbf{a}}|x_1, z_1, \dots, x_N, z_N} = (\Sigma_{\tilde{\mathbf{a}}\tilde{\mathbf{a}}}^{-1} + A^T \Sigma_{\mathbf{w}\mathbf{w}}^{-1} A)^{-1} (A^T \Sigma_{\mathbf{w}\mathbf{w}}^{-1} (\tilde{\mathbf{z}} - \tilde{\mathbf{b}}) + \Sigma_{\tilde{\mathbf{a}}\tilde{\mathbf{a}}}^{-1} \mu_{\tilde{\mathbf{a}}}^0) \\
 \text{So, } p(\tilde{\mathbf{a}}|x_1, z_1, \dots, x_N, z_N) &= \Sigma_{\tilde{\mathbf{a}}\tilde{\mathbf{a}}|x_1, z_1, \dots, x_N, z_N} \cdot (A^T \Sigma_{\mathbf{w}\mathbf{w}}^{-1} \cdot \tilde{\mathbf{z}}) \\
 &= \mathcal{N} \left(([\begin{smallmatrix} \beta & 0 \\ 0 & \beta \end{smallmatrix}])^{-1} + [\tilde{\mathbf{r}} \ \tilde{\mathbf{x}}]^T \cdot \frac{1}{\sigma^2} \cdot [\tilde{\mathbf{r}} \ \tilde{\mathbf{x}}])^{-1} \cdot ([\tilde{\mathbf{r}} \ \tilde{\mathbf{x}}]^T \cdot \frac{1}{\sigma^2} \cdot [\tilde{\mathbf{z}}_1 \ \tilde{\mathbf{z}}_2 \dots \tilde{\mathbf{z}}_N]^T), ([\begin{smallmatrix} \beta & 0 \\ 0 & \beta \end{smallmatrix}])^{-1} + [\tilde{\mathbf{r}} \ \tilde{\mathbf{x}}]^T \cdot \frac{1}{\sigma^2} \cdot [\tilde{\mathbf{r}} \ \tilde{\mathbf{x}}])^{-1} \right)
 \end{aligned}$$

2. Let $\sigma^2 = 0.1$ and $\beta = 1$. Draw four contour plots corresponding to the distributions $p(\mathbf{a})$, $p(\mathbf{a}|x_1, z_1)$, $p(\mathbf{a}|x_1, z_1, \dots, x_5, z_5)$, and $p(\mathbf{a}|x_1, z_1, \dots, x_{100}, z_{100})$. In all contour plots, the x-axis represents a_0 , and the y-axis represents a_1 . Please save the figures with names **prior.pdf**, **posterior1.pdf**, **posterior5.pdf**, **posterior100.pdf**, respectively. (1.5 pt)
3. Suppose that there is a new input x , for which we want to predict the corresponding target value z . Write down the distribution of the prediction z , i.e. $p(z|x, x_1, z_1, \dots, x_N, z_N)$. (1 pt)

$$\begin{aligned}
 p(\tilde{z}|x, x_1, z_1, \dots, x_N, z_N) &= \mathcal{N}(\tilde{z}|\mu_{\tilde{z}}, \tilde{\Sigma}_{\tilde{z}}) \\
 \mu_{\tilde{z}} &= A' \cdot \mu_{\tilde{\mathbf{a}}} \\
 \tilde{\Sigma}_{\tilde{z}} &= \Sigma_{\mathbf{w}\mathbf{w}} + A' \cdot \Sigma_{\tilde{\mathbf{a}}} \cdot A'^T \\
 \text{Recall that } \tilde{\Sigma}_{\tilde{\mathbf{a}}} &= (\tilde{\Sigma}_{\tilde{\mathbf{a}}\tilde{\mathbf{a}}}^{-1} + A^T \tilde{\Sigma}_{\mathbf{w}\mathbf{w}}^{-1} A)^{-1}, \tilde{\Sigma}_{\mathbf{w}\mathbf{w}} = \sigma^2 \\
 \text{and, } \mu_{\tilde{\mathbf{a}}} &= \tilde{\Sigma}_{\tilde{\mathbf{a}}} \cdot (A^T \tilde{\Sigma}_{\mathbf{w}\mathbf{w}}^{-1} (\tilde{\mathbf{z}} - \tilde{\mathbf{b}}) + \tilde{\Sigma}_{\tilde{\mathbf{a}}\tilde{\mathbf{a}}}^{-1} \mu_{\tilde{\mathbf{a}}}^0) \\
 \text{So, } p(\tilde{z}|x, x_1, z_1, \dots, x_N, z_N) &= \mathcal{N}([1 \ x] \cdot \mu_{\tilde{\mathbf{a}}}, \sigma^2 + [1 \ x] \cdot \tilde{\Sigma}_{\tilde{\mathbf{a}}} \cdot [1 \ x]^T) \\
 \text{Note that } \tilde{z}|x, x_1, z_1, \dots, x_N, z_N &= \mu_{\tilde{z}} \\
 \text{and } \mu_{\tilde{z}} &= A' \cdot \mu_{\tilde{\mathbf{a}}|x_1, z_1, \dots, x_N, z_N} = A' \cdot \mu_{\tilde{\mathbf{a}}}
 \end{aligned}$$

4. Let $\sigma^2 = 0.1$ and $\beta = 1$. Given a set of new inputs $\{-4, -3.8, \dots, 3.8, 4\}$, plot three figures, whose x-axis is the input and y-axis is the prediction, corresponding to three cases:
- The predictions are based on one training sample, i.e., based on $p(z|x, x_1, z_1)$.
 - The predictions are based on 5 training samples, i.e., based on $p(z|x, x_1, z_1, \dots, x_5, z_5)$.
 - The predictions are based on 100 training samples, i.e., based on $p(z|x, x_1, z_1, \dots, x_{100}, z_{100})$.

The range of each figure is set as $[-4, 4] \times [-4, 4]$. Each figure should contain the following three components: 1) the new inputs and the corresponding predicted targets; 2) a vertical interval at each predicted target, indicating the range within one standard deviation; 3) the training sample(s) that are used for the prediction. Use `plt.errorbar` for 1) and 2); use `plt.scatter` for 3). Please save the figures with names **predict1.pdf**, **predict5.pdf**, **predict100.pdf**, respectively. (1.5 pt)