## ECE368: Probabilistic Reasoning

Lab 2: Bayesian Linear Regression

Name: Ryan Song Student Number: 1003833658

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**)

we first need to realize that the posterior is Gaussian.

$$\Rightarrow \mathcal{N} = \left( \begin{array}{c} \chi^T \chi + \frac{\sigma^2}{B} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{array} \right)^{-1} \chi^T Z \\ Z_{alz} = \left( \begin{array}{c} \chi^T \chi + \frac{\sigma^2}{B} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{array} \right)^{-1} \sigma^2 \end{array}, \chi = \begin{bmatrix} 1 & \chi_1 \\ 1 & \chi_N \end{bmatrix}$$

$$\Rightarrow \mathcal{P}\left( \begin{array}{c} \Delta \left( \chi_1, Z_1, \dots, \chi_N, Z_N \right) \sim \mathcal{N} \right) \left( \begin{array}{c} \mathcal{N}, Z_{al2} \end{array} \right)$$

- 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,  $p(z|x, x_1, z_1, \ldots, x_N, z_N)$ . (1 **pt**)

The conditional probability 
$$P(Z|X, X_1, Z_1, ..., X_N, Z_N)$$
 is also a Gaussian distribution.

 $P(Z|X, X_1, Z_1, ..., X_N, Z_N) \sim N\left(\left[a_0 a_1\right]_{x}^{1}, \left[1 \times\right]_{x}^{1} \times \frac{\sigma^{2}}{B} I\right)_{x}^{1} \cdot \sigma^{2} + \sigma^{2}$ 

a is determined by training on  $X_1, Z_1, X_2, ..., X_N, Z_N$ .

- 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:
  - (a) The predictions are based on one training sample, i.e., based on  $p(z|x, x_1, z_1)$ .
  - (b) The predictions are based on 5 training samples, i.e., based on  $p(z|x, x_1, z_1, \dots, x_5, z_5)$ .
  - (c) 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**)