## 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)

$$\hat{\mu}_{\alpha} = \mu_{\alpha} \left[ \begin{array}{c} \chi_{1} \\ \chi_{1} \\ \chi_{2} \\ \chi_{3} \\ \chi_{4} \\ \chi_{5} \\ \chi_{7} \\ \chi_{1} \\ \chi_{2} \\ \chi_{1} \\ \chi_{1} \\ \chi_{2} \\ \chi_{1} \\ \chi_{1} \\ \chi_{2} \\ \chi_{1} \\ \chi_{2} \\ \chi_{1} \\ \chi_{1} \\ \chi_{2} \\ \chi_{1} \\ \chi_{2} \\ \chi_{1} \\ \chi_{2} \\ \chi_{1} \\ \chi_{1} \\ \chi_{1} \\ \chi_{1} \\ \chi_{2} \\ \chi_{1} \\ \chi_{1} \\ \chi_{1} \\ \chi_{2} \\ \chi_{1} \\ \chi_{1} \\ \chi_{1} \\ \chi_{2} \\ \chi_{1} \\ \chi_{2} \\ \chi_{1} \\ \chi_{1} \\ \chi_{2} \\ \chi_{1} \\ \chi_{1} \\ \chi_{2} \\ \chi_{2} \\ \chi_{2} \\ \chi_{1} \\ \chi_{2} \\ \chi_$$

- 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)

$$Z_n = [a, a_{\alpha}] \begin{bmatrix} x \end{bmatrix} + w$$
 but  $w \sim N(0, \sigma^2)$ , so  $w$  does not affect mean of  $Z$ .  

$$\therefore \mu_z = E[z] = E[a, x] + E[a_{\alpha}] + E[w] = \mu_{\alpha} \times \sigma [a_{\alpha} + a_{\alpha}] \times \sigma [a_{$$

- 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)