

**Stat 260/CS 294 Homework Assignment 3** (due March 29)

- Count data are often modeled with a Poisson distribution. The mean and variance of  $Y \sim \text{Pois}(\lambda)$  are both  $\lambda$ , but in real life count data often exhibit overdispersion, where the variance is greater than the mean. Let  $\lambda$  be a random variable, distributed according to  $\text{Ga}(a, b)$ . What amount of extra variance in  $Y$  does this mixture induce?
- Consider the following model:

$$\begin{aligned} y_i | \theta_i &\stackrel{iid}{\sim} \text{Pois}(\theta_i) \\ \theta_i | \alpha, \beta &\stackrel{iid}{\sim} \text{Ga}(\alpha, \beta) \\ \alpha &\sim \text{Exp}(a) \\ \beta &\sim \text{Ga}(b, c), \end{aligned}$$

where  $a, b$  and  $c$  are treated as fixed constants. Conditioning on  $y = (y_1, \dots, y_n)$ , find closed-form expressions for the conditional distributions of individual components of the parameter vector  $(\beta, \theta_1, \dots, \theta_n)$ , conditioning on  $y, \alpha$  and all other components of the parameter vector. Show also that the conditional distribution of  $\alpha$  is log concave.

- In the setting of linear regression, consider the null hypothesis  $H_0 : R\beta = 0$ , where  $R$  is a  $(q, p)$  matrix of rank  $q$ . Show that the restricted model on  $y$  given  $X$  can be represented as

$$y | \beta_0, \sigma_0^2, X_0 \sim N(X_0\beta_0, \sigma_0^2 I).$$

where  $X_0$  is an  $(n, p - q)$  matrix and  $\beta_0$  is a  $(p - q)$  dimensional vector. (Hint: Give the form of  $X_0$  and  $\beta_0$  in terms of  $X$  and  $\beta$ .) Under the hypothesis-specific prior,  $\beta_0 | H_0, \sigma_0^2 \sim N(\tilde{\beta}_0, \sigma_0^2(M_0)^{-1})$  and  $\sigma_0^2 \sim \text{IG}(a_0, b_0)$ , construct the Bayes factor associated with the test of  $H_0$  (where the prior under the full model is  $\beta | \sigma^2 \sim N(\tilde{\beta}, \sigma^2(M)^{-1})$  and  $\sigma^2 \sim \text{IG}(a, b)$ ).

- Consider the probit regression model,

$$P(y | X, \beta) = \prod_{i=1}^n \Phi(x_i^T \beta)^{y_i} (1 - \Phi(x_i^T \beta))^{1-y_i},$$

where  $\Phi$  is the Gaussian cumulative distribution function. Place a hierarchical prior on  $\beta$ :

$$\beta | \sigma^2, X \sim N(0, \sigma^2(X^T X)^{-1}) \quad \text{and} \quad \pi(\sigma^2 | X) \propto \sigma^{-3/2}.$$

- Find the posterior distribution  $\pi(\beta | y, X)$ , obtained by integrating out  $\sigma^2$ .
- Find conditions on  $\sum_i y_i$  and  $\sum_i (1 - y_i)$  for the posterior to be proper.