# Signle-Paprameter Models (Supp.)

Asymptotic property of the posterior

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# Conjugate Prior Distribution I

■ Let  $\mathcal{F}$  be a class of sampling distribution  $p(y|\theta)$ ,  $\mathcal{P}$  be a class of prior distribution  $p(\theta)$ .

Asymptotic property of the posterior

■ The class  $\mathcal P$  is conjugate for  $\mathcal F$  if

$$p(\theta|y) \in \mathcal{P}$$

for all  $p(\cdot|\theta) \in \mathcal{F}$  and  $p(\cdot) \in \mathcal{P}$ .

# Conjugate Prior Distribution II

- Exponential family and sufficient statistics
  - Exponential family

$$p(y_i|\theta) = f(y_i)g(\theta) \exp\left(\phi(\theta)^T u(y_i)\right).$$

Asymptotic property of the posterior

For i.i.d.  $y_1, \ldots, y_n$ , the density is

$$\left(\prod_{i=1}^{n} f(y_i)\right) g(\theta)^n \exp\left(\phi(\theta)^T \sum_{i=1}^{n} u(y_i)\right).$$

# Conjugate Prior Distribution III

3 If 
$$p(\theta) \propto g(\theta)^{\eta} \exp(\phi(\theta)^T \nu)$$
, then

$$p(\theta|y) \propto g(\theta)^{n+\eta} \exp(\phi(\theta)^T (v + t(y))),$$

Asymptotic property of the posterior

where 
$$t(y) = \sum_{i=1}^{n} u(y_i)$$
.

# Negative Binomial Distribution I

If  $p(y|\theta) = \exp(-\theta) \theta^y/y!$  and  $\theta \sim Gamma(\alpha, \beta)$ , then

$$p(y) = \frac{p(y|\theta)p(\theta)}{p(\theta|y)}$$

$$= \frac{\theta^{y} \exp(-\theta) \theta^{\alpha-1} \exp(-\beta\theta)\beta^{\alpha}\Gamma(\alpha+y)}{y!\theta^{\alpha+y-1} \exp(-(1+\beta)\theta)(1+\beta)^{\alpha+y}\Gamma(\alpha)}$$

$$= \frac{\Gamma(\alpha+y)\beta^{\alpha}}{\Gamma(\alpha)y!(1+\beta)^{\alpha+y}} = {\alpha+y-1 \choose y} \left(\frac{\beta}{\beta+1}\right)^{\alpha} \left(\frac{1}{\beta+1}\right)^{y}.$$

Asymptotic property of the posterior

### Negative Binomial Distribution II

Thus, we have

Neg-bin
$$(y|\alpha,\beta) = \int Poisson(y|\theta)Gamma(\theta|\alpha,\beta)d\theta$$
.

Asymptotic property of the posterior

■ Usually, this is related to over-dispersion model, it means that the variance is larger than mean. In Poisson model, mean and variance are equal.

# Asymptotic property of the posterior I

In limit probability theory can be put in a Bayesian context to show:

$$\left(\frac{\theta - E\left[\theta|y\right]}{\sqrt{\operatorname{Var}\left(\theta|y\right)}}\middle|y\right) \to N(0,1),$$

where  $\theta \in \mathbb{R}$ .

$$\left(\operatorname{Cov}(\theta|y)^{-1/2}(\theta - E[\theta|y])\middle|y\right) \to N(0, I_p),$$

where  $\theta \in \mathbb{R}^p$ .



Asymptotic property of the posterior

### Example of Data Analysis I

- Cancer of kidney/ureter in U. S. at 1980-1989.
  - Data are gathered in each county.
  - Population is varying through the counties.

# Example of Data Analysis II

#### Highest kidney cancer death rates



Figure 2.7 The counties of the United States with the highest 10% agestandardized death rates for cancer of kidney/ureter for U.S. white males, 1980–1989. Why are most of the shaded counties in the middle of the country? See Section 2.8 for discussion.

# Example of Data Analysis III

#### Lowest kidney cancer death rates



Figure 2.8 The counties of the United States with the lowest 10% agestandardized death rates for cancer of kidney/ureter for U.S. white males, 1980–1989. Surprisingly, the pattern is somewhat similar to the map of the highest rates, shown in Figure 2.7.

# Example of Data Analysis IV

■ This data implies that variations are different along with the populations of the counties.

Asymptotic property of the posterior

■ Counties in the Great Plains in the middle of the country appear in the lowest 10% and highest 10% simultaneously.

### Example of Data Analysis V

Conjugate Prior Distribution

■ Thus we consider the model of

$$y_j \sim Pois(10n_j\theta_j),$$

where  $n_i$  is the population of the county, and prior of

$$\theta_{i} \sim Gamma(20, 430000).$$

### Example of Data Analysis VI

Conjugate Prior Distribution

■ Then  $\theta_i|y_i \sim Gamma(20 + y_i, 430000 + 10n_i)$ , it implies

$$E(\theta_j|y_j) = \frac{20 + y_j}{43000 + 10n_j},$$

$$Var(\theta_j|y_j) = \frac{20 + y_j}{(43000 + 10n_i)^2}.$$

Small  $n_i$  results the relatively large weight of the prior in the posterior mean.

### Example of Data Analysis VII

■ (Prior) predictive distribution gives that

$$y_j \sim \text{Neg-bin}(20, 430000/(10n_j)),$$

and variance of  $y_i/(10n_i)$  is larger when  $n_i$  is small, note that

$$E[y_j] = 10n_j \frac{\alpha}{\beta},$$

$$Var(y_j) = 10n_j \frac{\alpha}{\beta} + (10n_j)^2 \frac{\alpha}{\beta^2}.$$

### Example of Data Analysis VIII

- How to choose  $\alpha = 20$  and  $\beta = 430000$ ? Data driven.
- In some case, we consider the parameter should be estimated, empirical Bayes approach (not full Bayesian).

Asymptotic property of the posterior

### Example of Data Analysis IX

We have

Conjugate Prior Distribution

$$y_j \sim \text{Neg-bin}\left(\alpha, \beta/(10n_j)\right)$$
,

and

$$E[y_j] = 10n_j \frac{\alpha}{\beta},$$

$$Var(y_j) = 10n_j \frac{\alpha}{\beta} + (10n_j)^2 \frac{\alpha}{\beta^2}.$$

Using such as the moment method based on data.

