

## Bayesian Exam Study Guide

$$P(\theta|x) \propto P(x|\theta)P(\theta)$$

- Summarizing posterior

$$\hat{\theta} = \mathbb{E}[\theta|x] \quad \hat{\theta}_{\text{map}} = \underset{\Theta}{\operatorname{argmax}} P(\theta|x)$$

- Interval summaries

$$(F_{\alpha/2}^{-1}, F_{1-\alpha/2}^{-1}) \quad \text{"credibility interval"}$$

$$\{\theta : P(\theta|x) > k_\alpha\} \quad k_\alpha : P(\{\theta : P(\theta|x) > k_\alpha\}) = 1 - \alpha$$

"Highest posterior density" credibility interval

- $X|\theta \sim \text{Binom}(n, \theta) \quad \theta \sim \text{Beta}(\alpha, \beta)$
- $\theta|x \sim \text{Beta}(x + \alpha, n - x + \beta)$

- Noninformative Priors

Jeffrey's Prior: invariant to reparametrization

$$\begin{aligned} P(\theta) &\propto I(\theta)^{-1/2} \quad I(\theta) = \mathbb{E}_{X|\theta} \left[ \left( \frac{\partial \ell}{\partial \theta} \right)^2 \right] \\ &= -\mathbb{E}_{X|\theta} \left[ \frac{\partial^2 \ell}{\partial \theta^2} \right] \end{aligned}$$

Ex:  $X_i \sim \text{Binom}(n, \theta)$  If  $\theta \sim \text{Beta}(\alpha, \beta)$

Examples from Exp Families

(a)  $X_i | \mu \sim N(\mu, \sigma^2)$   $\sigma^2$  known

$$\mu \sim N(\mu_0, \tau_0^{-2})$$

$$\mu | X \sim N\left(\frac{\mu_0 / \tau_0^{-2} + \bar{x} / \sigma^2 n}{1 / \tau_0^{-2} + 1 / \sigma^2 n}, \frac{1}{\frac{1}{\tau_0^{-2}} + \frac{1}{\sigma^2 n}}\right)$$

$$\hat{\mu} | X \sim N(\mu_0, \tau_0^{-2} + \sigma^2)$$

(b)  $X_i | \sigma^2 \sim N(\mu, \sigma^2)$   $\mu$  known

$$\sigma^2 \sim \text{Inv-}X^2(v_0, \tau_0^{-2}) \quad \left[ (\sigma^2)^{-(v_0/2-1)} \exp\left(-\frac{v_0 \tau_0^{-2}}{2\sigma^2}\right)\right]$$

$$\sigma^2 | X \sim \text{Inv-}X^2\left[n+v_0, \frac{S(x) + v_0 \tau_0^{-2}}{n+v_0}\right]$$

(c)  $Y_i | \theta \sim P_0(X_i | \theta)$   $X_i$  exposures

$$\theta \sim \text{Gamma}(\alpha, \beta) \quad \left[ e^{\alpha-1} e^{-\beta \theta} \right]$$

$$\theta | Y \sim \text{Gamma}(\alpha + n\bar{y}, \beta + n\bar{x})$$

$$\hat{\theta} | Y \sim NB\left(\alpha + n\bar{y}, \frac{\bar{x}}{\bar{x} + \beta + n\bar{x}}\right)$$

Multiparameter priors

$$x_1, \dots, x_n \stackrel{\text{iid}}{\sim} N(\mu, \sigma^2)$$

$$\mu | \sigma^2 \sim N(\mu_0, \sigma^2/k_0)$$

$$\sigma^2 \sim \text{Inv-}\chi^2(r_0, \sigma_0^2)$$

$$\mu, \sigma^2 | y \sim N\text{-Inv-}\chi^2(\mu_0, \sigma_0^2/k_0; r_0, \sigma_0^2)$$

$$\mu_n = \frac{k_0 \mu_0 + n \bar{y}}{k_0 + n}$$

$$k_n = k_0 + n$$

$$r_n = r_0 + n$$

$$\sigma_n \sigma_n^2 = r_0 \sigma_0^2 + (n-1) s^2 + \frac{k_0 n}{k_0 + n} (\bar{y} - \mu_0)^2$$

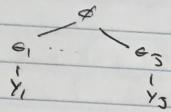
Laplace Approx

$$\theta | X \approx N(\theta^*, I_{\text{obs}}^{-1}(\theta^*))$$

$$\theta^* = \underset{\Theta}{\text{argmax}} P(\theta | X)$$

$$I_{\text{obs}}(\theta^*) = \left. \frac{\partial^2 \ell(\theta)}{\partial \theta^2} \right|_{\theta=\theta^*}$$

### Hierarchical Models



$$P(\epsilon, \alpha | y) \propto P(y|\epsilon) P(\epsilon|\alpha) P(\alpha)$$

Useful things to remember

- Change of variables:

Let  $X \sim f_X$      $Y = h(X)$  with  $h(\cdot)$  monotone.

$$\text{Then } f_Y(y) = f_X(h^{-1}(y)) \cdot |(h^{-1})'(y)|$$

### Conjugate Distributions

Binomial	Beta( $\alpha, \beta$ )	Beta( $\alpha+x, \beta+n-x$ )
Poisson	Gamma( $\alpha, \beta$ )	Gamma( $n\bar{x}+\alpha, n+\beta$ )
Geometric	Beta( $\alpha, \beta$ )	Beta( $n+\alpha, n(\bar{x}-1)+\beta$ )
Exponential	Gamma( $\alpha, \beta$ )	Gamma( $n+\alpha, n\bar{x}+\beta$ )
Neg. Binomial	Gamma( $\alpha, \beta$ )	Gamma( $\alpha+n\bar{x}, n+\beta$ )