Mathematical Exercises 1: problem 1

Let $X = (X_1, ..., X_n)$, where n = 30 and s = 3 $P(\theta) = 1 \implies \theta \sim \text{Befa}(1,1)$

 $P(x|\theta) = \theta^{s} \cdot (1-\theta)^{f} = \theta^{3} \cdot (1-\theta)^{27}$ $P(\theta|x) \propto \theta^{3} \cdot (1-\theta)^{27}, \text{ i.e. } \theta|x \sim \text{Beta}\left(x+s=4, \beta+f=28\right)$

 $P(s|\theta) = {n \choose s} \cdot \theta^{s} \cdot (1-\theta)^{\frac{1}{2}} \propto \theta^{3} \cdot (1-\theta)^{27}$

• Hence, same inference as in a

 $P(n|\theta) = {n-1 \choose 5-1} \theta^{5} \cdot (1-\theta)^{5} \propto \theta^{3} \cdot (1-\theta)^{27}$ Hence, same inference as in a

$$\begin{array}{c} \underbrace{\begin{array}{c} M_{\text{od}} h_{\text{emodelical}} \quad E_{\text{xercises}} \quad 1: \quad p_{\text{viblem}} \quad 2 \\ \\ \text{Let} \quad x = \left(x_{1}, \dots, x_{n}\right) \\ \\ \uparrow \left(x_{11}, \dots, x_{n}\right) \quad \varphi \left(x_{11}, \dots, x_{n}\right) \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta^{2} + \frac{x_{1}}{4\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{n}{2\sigma^{2}} \cdot \theta^{2} + \frac{x_{1}}{4\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{n}{2\sigma^{2}} \cdot \theta^{2} + \frac{x_{1}}{4\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma_{n}^{2}} \cdot \left(\theta - \mu_{n}\right)^{2}\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma_{n}^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma_{n}^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma_{n}^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma_{n}^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma_{n}^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma_{n}^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta^{2} + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-\frac{1}{2\sigma^{2}} \cdot \theta + \frac{\mu_{n}}{\sigma^{2}} \cdot \theta\right] \\ \\ \Rightarrow \exp \left[-$$

Mathematical Exercises 1: problem 3

Det $x = (x_1, ..., x_{10})$ $\theta \mid x \sim N(\mu_n, \tau_n^2)$ We have that n = 10, $T^2 = 1$, X = 1.873, $\mu_0 = 0$, $T_0^2 = 5$ Use the formulas for μ_n and τ_n^2 on stide 4, lecture 2. $W = \frac{10|1}{10/1 + \frac{1}{5}} \quad \mu_n = W \cdot 1.873 + (1-w) \cdot 0 = 1.84$ $\frac{1}{T_n^2} = \frac{10}{1} + \frac{1}{5} \implies T_n^2 = 0.098$ by Let $y = (y_1, \dots, y_{10})$ $p(\theta|x,y) \propto p(x,y|\theta) \cdot p(\theta) = p(y|x,\theta) \cdot p(x|\theta) \cdot p(\theta)$ $\propto \gamma(e|x)$ Hence, use $p(\theta|x)$ as the prior before obtaining the second sample y. Use the formular for Mn and Tn again. $\mu_{n} = \frac{10 \cdot 2 \cdot 3^{2}}{\frac{10}{2} + \frac{1}{0.098}} \cdot \frac{0.582}{y} + \frac{\frac{10}{0.098}}{\frac{10}{2} + \frac{1}{0.098}} \cdot 1.84 = 1.43$ $\frac{1}{T_n^2} = \frac{10}{2} + \frac{1}{0.098} \implies \gamma_n^2 = 0.066$ $\frac{\tau_n^2}{\varphi(\theta|x_1y_1z)} \propto \varphi(x_1y_1z|\theta) \cdot \varphi(\theta) = \underbrace{\varphi(z|x_1y_1\theta)}_{\varphi(z|\theta)} \cdot \underbrace{\varphi(x_1y_1\theta)}_{\varphi(z|\theta)} \cdot \underbrace{\varphi(x_1y_1\theta)}_{\varphi(\theta|x_1y_1)}$ Hence, p(O|xiV) is the prior before obtaining the third sample Z. The libelihood p(Z/0) can be written as $P(Z_1 = Z_1, ..., Z_8 = Z_8, Z_9 > 3, Z_{10} > 3 | \theta) = P(Z_1 = Z_{11} ..., Z_8 = Z_9 | \theta)$ $P(Z_q > 3, Z_w > 3|\theta) = P(Z_1, ..., Z_q|\theta) \cdot \prod_{j=1}^{q} P(Z_j > 3|\theta) \rightarrow 1 - \overline{\mathbb{D}}[3]\theta$

Mathematical Exercises 1: problem 4

$$p(\theta) \propto \theta^{\alpha-1} e^{-\beta \theta}$$

$$p(x|\theta) = \underset{i=1}{\overset{h}{\prod}} p(x_i|\theta) = \underset{i=1}{\overset{h}{\prod}} \theta e^{-\theta x_i} = \theta^n \cdot e^{-\frac{y_i}{\sum_{i=1}^n x_i} \cdot \theta}$$

$$\varphi(\Theta|X) \propto \varphi(X|\Theta) \cdot \varphi(\Theta) \propto \Theta^{n} \cdot e^{-\frac{n}{2}X_{t}} \cdot \Theta \cdot \Theta^{\kappa-1} e^{-\Re\Theta} = \Theta^{\kappa+n-1} e^{-\frac{n}{2}X_{t}}$$

Hence,
$$\Theta | X \sim Gamma \left(x + n, \beta + \sum_{i=1}^{n} x_i \right)$$
, i.e.

• the gamma prior for o is the conjugate prior.