# Problem Set 10

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## February 26, 2020

### Problem 1.

1. We have PDF:  $p(x) = p^x(1-p)^{1-x} = e^{x\log(p)+(1-x)\log(1-p)}$ . Hence we have:

$$\begin{cases} \eta_1 = log(p) \\ \eta_2 = log(1-p) \\ T_1(x) = x \\ T_2(x) = 1 - x \\ h(x) = 1 \\ B(p) = 0 \end{cases}$$

2. We have PDF  $p(x)=e^{\mu x-\frac{\mu^2}{2}}\frac{e^{\frac{-x^2}{2}}}{\sqrt{2\pi}}.$ Hence

$$\begin{cases} \eta = \mu \\ T(x) = x \\ h(x) = \frac{e^{-\frac{x^2}{2}}}{\sqrt{2\pi}} \\ B(p) = \frac{\mu^2}{2} \end{cases}$$

3. We have PDF  $p(x)=e^{\frac{\mu}{\sigma^2}-\frac{1}{2\sigma^2}x^2-\frac{\mu^2}{2\sigma^2}}\frac{1}{\sigma\sqrt{2\pi}}.$  Hence

$$\begin{cases} \eta_1 = \frac{\mu}{\sigma^2} \\ \eta_2 = -\frac{1}{2\sigma^2} \\ T_1(x) = x \\ T_2(x) = x^2 \\ h(x) = 1 \\ B(p) = \frac{\mu^2}{2\sigma^2} + \log(\sigma\sqrt{2\pi}) \end{cases}$$

4. We have  $p(x) = \lambda e^{-\lambda x}$ . Hence

$$\begin{cases} \eta = -\lambda \\ T(x) = x \\ h(x) = \lambda \\ B(p) = 0 \end{cases}$$

5. We have  $p(x) = \frac{1}{v} = e^{-log(v)}$ . Hence We have  $p(x) = \lambda e^{-\lambda x}$ . Hence

$$\begin{cases} \eta = -log(\upsilon) \\ T(x) = 1 \\ h(x) = 1 \\ B(p) = 0 \end{cases}$$

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6. We have 
$$p(x)=rac{eta^{lpha}}{\mathbb{I}(lpha)}x^{lpha-1}e^{-eta x}=e^{-eta x+(lpha-1)log(x)}rac{eta^{lpha}}{\mathbb{I}(lpha)}.$$
 Hence

$$\begin{cases} \eta_1 = -\beta \\ \eta_2 = \alpha - 1 \\ T_1(x) = x \\ T_2(x) = log(x) \\ h(x) = 1 \\ B(p) = -log(\frac{\beta^{\alpha}}{\mathbb{I}(\alpha)}) \end{cases}$$

7. We have 
$$p(x) = \frac{\lambda^x e^{-\lambda}}{x!} = e^{-\lambda + \log(\lambda)x} \frac{1}{x!}$$
. Hence

$$\begin{cases} \eta_1 = log(\lambda) \\ T(x) = x \\ h(x) = \frac{1}{x!} \\ B(p) = -\lambda \end{cases}$$

#### Problem 2.

1.

a)

b) We have

$$\mathbb{E}\left[\frac{\partial log(f_{\theta}(x))}{\partial \theta}\right] = \mathbb{E}\left[\frac{\frac{\partial}{\partial \theta}f_{\theta}(x)}{f_{\theta}(x)}\right] = \int \frac{\frac{\partial}{\partial \theta}f_{\theta}(x)}{f_{\theta}(x)}f_{\theta}(x) = \frac{\partial}{\partial \theta}\int f_{\theta}(x)dx = 0$$

$$\mathbb{E}\left[\frac{\partial^2}{\partial \theta^2}log(f_{\theta}(x))\right] = \int \frac{\left[\frac{\partial^2}{\partial \theta^2}f_{\theta}(x)\right]f_{\theta}(x) - \left[\frac{\partial}{\partial \theta}f_{\theta}(x)\right]^2}{(f_{\theta}(x))^2}f_{\theta}(x) = \int \frac{\partial^2}{\partial \theta^2}f_{\theta}(x) + \int \left(\frac{\left[\frac{\partial}{\partial \theta}f_{\theta}(x)\right]^2}{f_{\theta}(x)}dx\right) = \mathbb{E}\left[\left(\frac{\frac{\partial}{\partial \theta}f_{\theta}(x)}{f_{\theta}(x)}\right)^2\right]$$

We have  $l(\theta) = -\theta x + b(\theta) + log(a(x))$ . Hence  $l'(\theta) = -x + b'(\theta)$  and  $l''(\theta) = b''(\theta)$ . From previous quantities,  $\mathbb{E}[l'(\theta)] = -\mathbb{E}[x] + b'(\theta) = 0$ 

From previous quantities,  $\mathbb{E}\left[\frac{\partial^2}{\partial \theta^2}log(f_{\theta}(x))\right] = \mathbb{E}[l''(\theta)] = \mathbb{E}[(l'(\theta))^2]$ . Because  $\mathbb{E}[l'(\theta)] = 0$ , hence  $\mathbb{E}[(l'(\theta))^2] = 0$  $var(l'(\theta)) = -\mathbb{E}[l''(\theta)]$ 

c) Fisher Information.

2.

From previous questions, we have  $l'(\theta) = -x + b'(\theta)$  and  $l''(\theta) = b''(\theta)$ 

From previous result  $\mathbb{E}[l'(\theta)] - \mathbb{E}[x] + b'(\theta) = 0 \Rightarrow \mathbb{E}[x] = b'(\theta)$ 

From previous result  $\mathbb{E}[(l'(\theta))^2] = \mathbb{E}[(-x + b'(\theta))^2] = \mathbb{E}[(-x + E[x])^2] = var(x) = -b''(\theta)$ 

a) Sample space  $X = (0, +\infty)$ 

b)  $a(x) = x^{\alpha-1}$  and  $b(\theta) = log(\frac{\theta^{\alpha}}{\mathbb{I}(\alpha)})$  c) From previous result,  $\mathbb{E}[x] = b'(\theta) = \frac{\alpha}{\theta}$  and  $Var(x) = -b''(\theta) = \frac{\alpha}{\theta^2}$ 

#### Problem 3.

- 1.  $Z_1$  is either 0 or 1, hence  $Z_1$  is Bernoulli distribution conditional on  $X_i$  with parameter  $p = 1 F(-X^T\beta)$ .
- 2. We have  $\mathbb{E}[Z] = p = F(-X^T\beta)$ . Hence  $g(1 F(-X^T\beta)) = X^T\beta \Rightarrow g(x) = (1 F(-x))^{-1}$ .
- 3. If  $\epsilon$  is standard Gaussian, then  $g(x) = (1 F(-x))^{-1} = (1 \phi(-x))^{-1} = \phi(x)^{-1}$ .

a) 
$$F(t) = \int \frac{e^{-t}}{(1+e^{-1})^2} = -\frac{e^{-t}}{1+e^{-t}}$$
  
b)  $g(x) = 1 - F^{-1}(x) = 1 - \log(\frac{x}{1-x})$ 

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