## Homework 4

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**1.** Let  $\theta_2 > \theta_1$ ,

$$\frac{g_T(t|\theta_2)}{g_T(t|\theta_1)} = \frac{h(t)c(\theta_2)e^{w(\theta_2)t}}{h(t)c(\theta_1)e^{w(\theta_1)t}} = \frac{c(\theta_2)}{c(\theta_1)}e^{(w(\theta_2)-w(\theta_1))t}$$

Since  $g_T$  is an Exponential family,  $c(\theta_2) \geq 0, c(\theta_1) \geq 0$ . Since w is a nondecreasing function,  $w(\theta_2) - w(\theta_1) > 0$ , thus  $\frac{g_T(t|\theta_2)}{g_T(t|\theta_1)}$  is increasing function of t. Therefore, this family has an MLR.

**2.**  $T(\boldsymbol{X}) = \bar{X} \sim \mathcal{N}(\mu, \sigma_0^2/n)$ . Thus, for any  $\mu_2 > \mu_1$ ,

$$\frac{g_T(t|\mu_2)}{g_T(t|\mu_1)} = \frac{\frac{\sqrt{n}}{\sqrt{2\pi\sigma_0}} e^{-\frac{(t-\mu_2)^2}{2\sigma_0^2/n}}}{\frac{\sqrt{n}}{\sqrt{2\pi\sigma_0}} e^{-\frac{(t-\mu_1)^2}{2\sigma_0^2/n}}} = e^{-\frac{n}{2\sigma_0^2}[(t-\mu_2)^2 - (t-\mu_1)^2]} = e^{-\frac{n}{2\sigma_0^2}[(\mu_2^2 - \mu_1^2) + 2(\mu_2 - \mu_1)t]}$$

Since  $\mu_2 > \mu_1$ ,  $\frac{g_T(t|\mu_2)}{g_T(t|\mu_1)}$  is an increasing function of t. Thus the family of the distribution for  $\bar{X}$  has an MLR.

3.

By Neyman-Pearson Lemma, supposing that the rejection region is  $R:\{f_{\mathbf{X}}(\mathbf{x}|\theta=2)>kf_{\mathbf{X}}(\mathbf{x}|\theta=1)\}$  for some  $k\geq 0$ . Then the UMP level  $\alpha$  test is

$$\alpha = P_{\theta=1}(\boldsymbol{x} \in R)$$

$$= P_{\theta=1}(\frac{f_{\boldsymbol{X}}(\boldsymbol{x}|\theta=2)}{f_{\boldsymbol{X}}(\boldsymbol{x}|\theta=1)} > k)$$

$$= P_{\theta=1}(\frac{k}{2^n} < \prod_{i=1}^n x_i < 1)$$

$$= P_{\theta=1}(0 < -\sum_{i=1}^n \log x_i < n \log 2 - \log k)$$

When  $\theta = 1$ ,  $f_X(x|\theta = 1) = I(0 < x < 1)$ , thus  $x \sim \text{Unif}(0,1)$ . Let  $y = -\log x$ ,  $f_Y(y|\theta = 1) = f_X(e^{-y}|\theta = 1)e^{-y} = e^{-y}$ , thus  $y \sim \text{Expo}(1)$ ,  $\sum_{i=1}^n y_i = -\sum_{i=1}^n \log x_i \sim \text{Gamma}(n,1)$ .

Let  $k_{1-\alpha}$  be the cutoff point that  $1-\alpha=P_{\theta=1}(-\sum_{i=1}^n\log x_i>k_{1-\alpha})$ , then the test that rejects  $H_0$  if  $P_{\theta=1}(-\sum_{i=1}^n\log x_i< k_{1-\alpha})$  is an UMP level  $\alpha$  test.

4.

**a.** By Neyman-Pearson Lemma, supposing that the rejection region is  $R:\{f_X(x|\theta=2)>kf_X(x|\theta=1)\}$  for some  $k\geq 0$ . Then the UMP level  $\alpha$  test is

$$\alpha = P_{\theta=1}(\boldsymbol{x} \in R) = P_{\theta=1}(\frac{f_{\boldsymbol{X}}(\boldsymbol{x}|\theta=2)}{f_{\boldsymbol{X}}(\boldsymbol{x}|\theta=1)} > k) = P_{\theta=1}(\frac{\frac{1}{2^n}e^{-\sum_{i=1}^n x_i/2}}{e^{-\sum_{i=1}^n x_i}} > k)$$
$$= P_{\theta=1}(e^{\sum_{i=1}^n x_i/2} > 2^n k) = P_{\theta=1}(\sum_{i=1}^n x_i > 2(n \log 2 + \log k))$$

Since each  $x_i \sim \text{Expo}(1)$ ,  $\sum_{i=1}^n x_i \sim \text{Gamma}(n,1)$ . Supposing that  $k_{\alpha}$  is the cutoff point such that  $\alpha = P_{\theta=1}(\sum_{i=1}^n x_i > k_{\alpha})$ . Then the test that rejects  $H_0$  if  $\sum_{i=1}^n X_i > k_{\alpha}$  is an UMP level  $\alpha$  test.

**b.** By definition,

$$f_{\mathbf{X}}(\mathbf{x}|\theta) = \frac{1}{\theta^n} e^{-\frac{1}{\theta} \sum_{i=1}^n x_i} \prod_{i=1}^n I(x_i > 0)$$

Let  $T = \sum_{i=1}^{n} X_i$ ,  $g(t|\theta) = \frac{1}{\theta^n} e^{-\frac{1}{\theta} \sum_{i=1}^{n} x_i}$ ,  $h(\boldsymbol{x}) = \prod_{i=1}^{n} I(x_i > 0)$ . By Factorization Theorem,  $f_{\boldsymbol{X}}(\boldsymbol{x}|\theta) = g(t|\theta)h(\boldsymbol{x})$ . Thus T is a sufficient statistic.

**c.** Since  $X_i \sim \text{Expo}(\frac{1}{\theta})$ ,  $T(X) = \sum_{i=1}^n X_i \sim \text{Gamma}(n, \frac{1}{\theta})$ . For any  $\theta_2 > \theta_1$ ,

$$\frac{g_T(t|\theta_2)}{g_T(t|\theta_2)} = \frac{\frac{1}{\Gamma(n)\theta_2^n} x^{n-1} e^{-\frac{1}{\theta_2} x}}{\frac{1}{\Gamma(n)\theta_2^n} x^{n-1} e^{-\frac{1}{\theta_1} x}} = (\frac{\theta_1}{\theta_2})^n e^{(\frac{1}{\theta_1} - \frac{1}{\theta_2})x}$$

Since  $\theta_2 > \theta_1 \implies \frac{1}{\theta_1} - \frac{1}{\theta_2} > 0$ ,  $\frac{g_T(t|\theta_2)}{g_T(t|\theta_2)}$  is an increasing function of t. Thus T has a MLR. Let  $k_\alpha$  be the cutoff point that  $\alpha = P_{\theta_0}(T > t_\alpha)$ . By Karlin-Rubin Theorem,

$$\alpha = P_{\theta_0}(T > t_0) \implies t_0 = k_{\alpha}$$

Thus the test that rejects  $H_0$  if  $T > k_{\alpha}$  is an UMP level  $\alpha$  test.

5.

Let  $T(X) = \sum_{i=1}^{n} X_i$ . Since each  $X_i \sim \text{Pois}(\theta)$ ,  $T \sim \text{Pois}(n\theta)$ .

Since

$$f_{\mathbf{X}}(\mathbf{x}|\theta) = \frac{\theta^{\sum_{i=1}^{n} x_i} e^{-n\theta}}{\prod_{i=1}^{n} x_i!} = \frac{\theta^t e^{-n\theta}}{\prod_{i=1}^{n} x_i!}$$

Let  $g(t|\theta) = \theta^t e^{-n\theta}$ ,  $h(\mathbf{x}) = \frac{1}{\prod_{i=1}^n x_i!}$ . By Factorization Theorem,  $f_{\mathbf{X}}(\mathbf{x}|\theta) = g(t|\theta)h(\mathbf{x})$ . Thus T is a sufficient statistic.

For any  $\theta_2 > \theta_1$ ,

$$\frac{g_T(t|\theta_2)}{g_T(t|\theta_1)} = \frac{\frac{(n\theta_2)^t e^{-n\theta_2}}{t!}}{\frac{(n\theta_1)^t e^{-n\theta_1}}{t!}} = (\frac{\theta_2}{\theta_1})^t e^{n(\theta_1 - \theta_2)}$$

Since  $\theta_2 > \theta_1$ ,  $\frac{g_T(t|\theta_2)}{g_T(t|\theta_1)}$  is a decreasing function of t. Thus T has a MLR.

By Karlin-Rubin Theorem,

$$\alpha = P_{\theta_0}(T > t_0) = \sum_{t=t_0}^{\infty} \frac{(n\theta_0)^t e^{-n\theta}}{t!}$$

Since Poisson is a discrete distribution, let  $t' = \min\{t_0 | \sum_{t=t_0}^{\infty} \frac{(n\theta_0)^t e^{-n\theta}}{t!} \le \alpha\}$ , then the test that rejects  $H_0$  if T > t' is a UMP level  $\alpha$  test.