# STAT 5010 Tutorial 2\*

Oct. 2022

### **Definitions**

1. A parametric family  $\{P_{\theta}: \theta \in \Theta\}$  dominated by a  $\sigma$ -finite measure  $\nu$  on  $(\Omega, \mathcal{F})$  is called an *exponential family* if and only if

 $\frac{dP_{\theta}}{d\nu}(\omega) = \exp\left\{ \left[ \eta(\theta) \right]^{\tau} T(\omega) - \xi(\theta) \right\} h(\omega), \quad \omega \in \Omega,$ 

where  $\exp\{x\} = e^x$ , T is a random p-vector with a fixed positive integer p,  $\eta$  is a function from  $\Theta$  to  $\mathcal{R}^p$ , h is a nonnegative Borel function on  $(\Omega, \mathcal{F})$ , and  $\xi(\theta) = \log\{\int_{\Omega} \exp\{[\eta(\theta)]^{\tau} T(\omega)\} h(\omega) d\nu(\omega)\}$ 

2. In an exponential family, consider the reparameterization  $\eta = \eta(\theta)$  and

$$f_{\eta}(\omega) = \exp \{ \eta^{\tau} T(\omega) - \zeta(\eta) \} h(\omega), \quad \omega \in \Omega,$$

where  $\zeta(\eta) = \log \left\{ \int_{\Omega} \exp \left\{ \eta^{\tau} T(\omega) \right\} h(\omega) d\nu(\omega) \right\}$ . This is the *canonical form* for the family, which is not unique. The new parameter  $\eta$  is called the *natural parameter*. The new parameter space  $\Xi = \{ \eta(\theta) : \theta \in \Theta \}$ , a subset of  $\mathcal{R}^p$ , is called the *natural parameter space*. An exponential family in canonical form is called a *natural exponential family*. If there is an open set contained in the natural parameter space of an exponential family, then the family is said to be of *full rank*.

- 3. A measurable function of X, T(X), is called a *statistic* if T(X) is a known value whenever X is known, i.e., the function T is a known function.
- 4. Let X be a sample from an unknown population  $P \in \mathcal{P}$ , where  $\mathcal{P}$  is a family of populations. A statistic T(X) is said to be *sufficient* for  $P \in \mathcal{P}$  (or for  $\theta \in \Theta$  when  $\mathcal{P} = \{P_{\theta} : \theta \in \Theta\}$  is a parametric family) if and only if the conditional distribution of X given T is known (does not depend on P or  $\theta$ ).
- 5. Let T be a sufficient statistic for  $P \in \mathcal{P}$ . T is called a *minimal sufficient* statistic if and only if, for any other statistic S sufficient for  $P \in \mathcal{P}$ , there is a measurable function  $\psi$  such that  $T = \psi(S)$  a.s.  $\mathcal{P}$
- 6. A statistic V(X) is said to be *ancillary* if its distribution does not depend on the population P and *first-order ancillary* if E[V(X)] is independent of P.
- 7. A statistic T(X) is said to be *complete* for  $P \in \mathcal{P}$  if and only if, for any Borel f, E[f(T)] = 0 for all  $P \in \mathcal{P}$  implies f(T) = 0 a.s.  $\mathcal{P}$ . T is said to be *boundedly complete* if and only if the previous statement holds for any bounded Borel f.

# Propositions and Theorems

1. If  $\eta_0$  is an interior point of the natural parameter space, then the m.g.f.  $\psi_{\eta_0}$  of  $P_{\eta_0} \circ T^{-1}$  is finite in a neighborhood of 0 and is given by

$$\psi_{\eta_0}(t) = \exp\left\{\zeta\left(\eta_0 + t\right) - \zeta\left(\eta_0\right)\right\}$$

2. (The factorization theorem) Suppose that X is a sample from  $P \in \mathcal{P}$  and  $\mathcal{P}$  is a family of probability measures on  $(\mathcal{R}^n, \mathcal{B}^n)$  dominated by a  $\sigma$ -finite measure  $\nu$ . Then T(X) is sufficient for  $P \in \mathcal{P}$  if and only if there are nonnegative Borel functions h (which does not depend on P) on  $(\mathcal{R}^n, \mathcal{B}^n)$  and  $g_P$  (which depends on P) on the range of T such that

$$\frac{dP}{du}(x) = g_P(T(x))h(x).$$

- 3. Suppose that  $\mathcal{P}$  contains p.d.f.'s  $f_P$  w.r.t. a  $\sigma$ -finite measure and that there exists a sufficient statistic T(X) such that, for any possible values x and y of X,  $f_P(x) = f_P(y)\phi(x,y)$  for all P implies T(x) = T(y), where  $\phi$  is a measurable function. Then T(X) is minimal sufficient for  $P \in \mathcal{P}$ .
- 4. If P is in an exponential family of full rank with p.d.f.'s given as in Definition. 2, then T(X) is complete and sufficient for  $\eta \in \Xi$ .

<sup>\*</sup>Dept. of Stat., CUHK. TA: YX

- 5. (Basu's theorem) Let V and T be two statistics of X from a population  $P \in \mathcal{P}$ . If V is ancillary and T is boundedly complete and sufficient for  $P \in \mathcal{P}$ , then V and T are independent w.r.t. any  $P \in \mathcal{P}$ .
- 6. A complete and sufficient statistic is also minimal sufficient. However, a minimal sufficient statistic is not necessarily complete.

## Question 1

1. Let X and Y be two random variables such that Y has the binomial distribution with size N and probability  $\pi$  and, given Y = y, X has the binomial distribution with size y and probability p. Suppose that  $p \in (0,1)$  and  $\pi \in (0,1)$  are unknown and N is known. Show that (X,Y) is minimal sufficient for  $(p,\pi)$ .

#### Solution:

Let  $A = \{(x, y) : x = 0, 1, \dots, y, y = 0, 1, \dots, N\}$ . The joint probability density of (X, Y) with respect to the counting measure is

$$\begin{pmatrix} N \\ y \end{pmatrix} \pi^y (1-\pi)^{N-y} \begin{pmatrix} y \\ x \end{pmatrix} p^x (1-p)^{y-x} I_A$$

$$= \exp\left\{ x \log \frac{p}{1-p} + y \log \frac{\pi(1-p)}{1-\pi} + N \log(1-\pi) \right\} \begin{pmatrix} N \\ y \end{pmatrix} \begin{pmatrix} y \\ x \end{pmatrix} I_A.$$

Hence, (X,Y) has a distribution from an exponential family of full rank (0 .This implies that <math>(X,Y) is minimal sufficient for  $(p,\pi)$ 

## Question 2

1. Let  $X_1, \ldots, X_n$  be i.i.d. random variables from  $P_{\theta}$ , the uniform distribution  $U(\theta, \theta + 1), \theta \in \mathcal{R}$ . Prove that  $T = (X_{(1)}, X_{(n)})$  is minimal sufficient.

**Solution:** Suppose that n > 1. The joint Lebesgue p.d.f. of  $(X_1, \ldots, X_n)$  is

$$f_{\theta}(x) = \prod_{i=1}^{n} I_{(\theta,\theta+1)}(x_i) = I_{(x_{(n)}-1,x_{(1)})}(\theta), \quad x = (x_1,\dots,x_n) \in \mathbb{R}^n,$$

where  $x_{(i)}$  denotes the *i* th smallest value of  $x_1, \ldots, x_n$ . By the factorization theorem,  $T = (X_{(1)}, X_{(n)})$  is sufficient for  $\theta$ . Note that

$$x_{(1)} = \sup \{\theta : f_{\theta}(x) > 0\}$$
 and  $x_{(n)} = 1 + \inf \{\theta : f_{\theta}(x) > 0\}$ .

If S(X) is a statistic sufficient for  $\theta$ , then by the factorization theorem, there are Borel functions h and  $g_{\theta}$  such that  $f_{\theta}(x) = g_{\theta}(S(x))h(x)$ . For x with h(x) > 0,  $x_{(1)} = \sup\{\theta : g_{\theta}(S(x)) > 0\}$  and  $x_{(n)} = 1 + \inf\{\theta : g_{\theta}(S(x)) > 0\}$ . Hence, there is a measurable function  $\psi$  such that  $T(x) = \psi(S(x))$  when h(x) > 0. Since h > 0 a.s.  $\mathcal{P}$ , we conclude that T is minimal sufficient.

## Proofs for Some propositions

### • Prop. 3.

*Proof.* From Bahadur (1957), there exists a minimal sufficient statistic S(X). The result follows if we can show that  $T(X) = \psi(S(X))$  a.s.  $\mathcal{P}$  for a measurable function  $\psi$ . By the factorization theorem, there are Borel functions  $g_P$  and h such that  $f_P(x) = g_P(S(x))h(x)$  for all P. Let  $A = \{x : h(x) = 0\}$ . Then P(A) = 0 for all P. For x and y such that  $S(x) = S(y), x \notin A$  and  $y \notin A$ ,

$$f_P(x) = g_P(S(x))h(x)$$
  
=  $g_P(S(y))h(x)h(y)/h(y)$   
=  $f_P(y)h(x)/h(y)$ 

for all P. Hence T(x) = T(y). This shows that there is a function  $\psi$  such that  $T(x) = \psi(S(x))$  except for  $x \in A$ . It remains to show that  $\psi$  is measurable. Since S is minimal sufficient, g(T(X)) = S(X) a.s.  $\mathcal{P}$  for a measurable function g. Hence g is one-to-one and  $\psi = g^{-1}$ . The measurability of  $\psi$  follows from Theorem 3.9 in Parthasarathy (1967).

#### • Prop. 4.

*Proof.* Obviously, T is sufficient. Suppose that there is a function f such that E[f(T)] = 0 for all  $\eta \in \Xi$ . Then,

$$\int f(t) \exp \left\{ \eta^{\tau} t - \zeta(\eta) \right\} d\lambda = 0 \quad \text{ for all } \eta \in \Xi,$$

where  $\lambda$  is a measure on  $(\mathcal{R}^p, \mathcal{B}^p)$ . Let  $\eta_0$  be an interior point of  $\Xi$ . Then

$$\int f_{+}(t)e^{\eta^{\tau}t}d\lambda = \int f_{-}(t)e^{\eta^{\tau}t}d\lambda \quad \text{ for all } \eta \in N(\eta_{0}),$$

where  $N(\eta_0) = \{ \eta \in \mathbb{R}^p : ||\eta - \eta_0|| < \epsilon \}$  for some  $\epsilon > 0$ . In particular,

$$\int f_{+}(t)e^{\eta_{0}^{\tau}t}d\lambda = \int f_{-}(t)e^{\eta_{0}^{\tau}t}d\lambda = c.$$

If c=0, then f=0 a.e.  $\lambda$ . If c>0, then  $c^{-1}f_+(t)e^{\eta_0^{\tau}t}$  and  $c^{-1}f_-(t)e^{\eta_0^{\tau}t}$  are p.d.f.'s w.r.t.  $\lambda$  and this implies that their m.g.f.'s are the same in a neighborhood of 0. Thus,  $c^{-1}f_+(t)e^{\eta_0^{\tau}t}=c^{-1}f_-(t)e^{\eta_0^{\tau}t}$ , i.e.,  $f=f_+-f_-=0$  a.e.  $\lambda$ . Hence T is complete.