

## MTH 515a: Inference-II Assignment No. 1: Invariance

1. Let  $X_1, \dots, X_n$  ( $n \geq 2$ ) be a random sample from an  $N(\theta, 1)$  distribution, where  $\theta \in \Theta = \mathbb{R}$ . Consider estimation of  $\theta$  under the squared error loss function. For any  $\underline{x} \in \chi$ , consider the randomized decision rule  $\delta_0((-\infty, t]|\underline{x}) = \Phi(t - x_1)$ .
  - (i) Find a randomized estimator (say  $\delta_1$ ) that is a function of a minimal-sufficient statistic and has the same risk function as that of  $\delta_0$ .
  - (ii) Find a non-randomized estimator that dominates the estimators  $\delta_0$  and  $\delta_2(\underline{X}) = X_1$ .
2. Let  $X \sim \text{Bin}(n, \theta)$ , where  $\theta$  ( $\in \Theta = (0, 1)$ ) is unknown and  $n$  is a fixed positive integer. For estimating  $\theta$ , under the squared error loss function, consider the randomized estimator

$$\delta_0(a|x) = \begin{cases} \frac{1}{2}, & \text{if } a = \frac{1}{2} \\ \frac{1}{2}, & \text{if } a = \frac{x}{n} \end{cases}, x = 0, 1, \dots, n.$$

Find a non-randomized estimator that has uniformly smaller risk than  $\delta_0$ .

3. Let  $X_1, \dots, X_{2n+1}$  be i.i.d.  $N(\theta, 1)$  random variables, where  $\theta$  ( $\in \Theta = \mathbb{R}$ ) is unknown. Consider estimation of  $\theta$  under a loss function  $L(\theta, a)$ ,  $\theta \in \Theta$ ,  $a \in \mathcal{A} = \mathbb{R}$ .
  - (a) Find a randomized estimator based on the complete-sufficient statistic  $\bar{X}$  which has the same risk function as the sample median  $\tilde{X}$ .
  - (b) Suppose that, for every fixed  $\theta \in \Theta$ , the loss function  $L(\theta, a)$  is strictly convex in  $a \in \mathcal{A}$ . Find an estimator that dominates the sample median  $\tilde{X}$ .
4. For an invariant estimation problem  $(\mathcal{P}, \mathcal{A}, L)$ , show that  $\bar{\mathcal{G}} = \{\bar{g} : g \in \mathcal{G}\}$  is a group of transformations of  $\mathcal{P}$  into itself.
5. For an invariant estimation problem  $(\mathcal{P}, \mathcal{A}, L)$ , show that  $\tilde{\mathcal{G}} = \{\tilde{g} : g \in \mathcal{G}\}$  is a group of transformations of  $\mathcal{A}$  into itself.
6. Let  $\mathcal{P} = \{F_\theta : \theta \in \Theta\}$ , where  $\Theta = \mathbb{R}$  and  $F_\theta$  is the distribution function of  $X \sim N(\theta, 1)$ ,  $\theta \in \Theta$ . Let  $\mathcal{A} = \Theta$  and consider estimating  $\theta$  under the loss function

$$L(\theta, a) = W(|a - \theta|), \quad a \in \mathcal{A}, \theta \in \Theta,$$

where  $W$  is some non-negative function defined on  $\mathbb{R}_+ = [0, \infty)$ . Let  $\chi = \mathbb{R}$ .

- (a) Show that the estimation/decision problem is invariant under the additive group  $\mathcal{G} = \{g_c : c \in \mathbb{R}\}$ , where  $g_c(x) = x + c$ ,  $x \in \chi$ ,  $c \in \mathbb{R}$ ;

- (b) Show that a decision rule  $\delta$  is invariant under  $\mathcal{G}$  if, and only if,  $\delta(A|x) = \delta(A + c|x + c), \forall A \in \mathcal{F}_{\mathcal{A}}, x \in \chi, c \in \mathbb{R}$ , where  $A + c = \{a + c : a \in A\}$  (or  $Y_x + c \stackrel{d}{=} Y_{x+c}, \forall x \in \chi, c \in \mathbb{R}$ , where, for each  $x \in \chi$ ,  $Y_x$  is a random variable corresponding to probability measure  $\delta(\cdot|x)$ );
- (c) Show that a non-randomized decision rule  $\delta$  is invariant if, and only if,  $\delta(x) = x + c, x \in \chi$ , for some  $c \in \mathbb{R}$ ;
- (d) For any invariant decision rule  $\delta$ , show that the risk function  $R_\delta(\theta)$  is constant (does not depend on  $\theta \in \Theta$ ).

**(Note:** The loss function  $L(\theta, a) = (a - \theta)^2, a \in \mathcal{A}, \theta \in \Theta$ , corresponding to the choice  $W(x) = x^2, x \in \mathbb{R}_+$ , is called the squared error loss function.)

7. Let  $\mathcal{P} = \{F_\theta : \theta \in \Theta\}$ , where  $\Theta = (0, 1)$  and  $F_\theta$  is the distribution function of  $X \sim \text{Bin}(n, \theta), \theta \in \Theta$ ; here  $n$  is a known positive integer. Let  $\mathcal{A} = \Theta, \chi = \{0, 1, \dots, n\}$  and consider the problem of estimating  $\theta$  under loss function

$$L(\theta, a) = W(|a - \theta|), \quad a \in \mathcal{A}, \theta \in \Theta,$$

where  $W : \mathbb{R}_+ \rightarrow \mathbb{R}_+$  is some function.

- (a) Find a suitable group of transformations  $\mathcal{G}$  under which the problem of estimating  $\theta$  is invariant;
- (b) Show that a decision rule  $\delta$  is invariant under  $\mathcal{G}$  if, and only if,  $\delta(A|x) = \delta(1 - A|n - x), \forall A \in \mathcal{F}_{\mathcal{A}}, x \in \chi$ , where  $1 - A = \{1 - a : a \in A\}$  (or  $1 - Y_x \stackrel{d}{=} Y_{n-x}, \forall x \in \chi$ , where, for each  $x \in \chi$ ,  $Y_x$  is a random variable corresponding to probability measure  $\delta(\cdot|x)$ );
- (c) Show that a non-randomized decision rule  $\delta$  is invariant if, and only if,  $1 - \delta(x) = \delta(n - x), x \in \chi$ ;
- (d) For any invariant decision rule  $\delta$ , show that  $R_\delta(\theta) = R_\delta(1 - \theta), \forall \theta \in \Theta$ ;
- (e) Does  $\overline{\mathcal{G}}$  acts transitively on  $\mathcal{P}$ ?
8. Let  $\mathcal{P} = \{F_\theta : \theta \in \Theta\}$ , where  $\Theta = \mathbb{R}_{++} = (0, \infty)$  and  $F_\theta(x) = F(\frac{x}{\theta}), x \in \chi = \mathbb{R}_+ = [0, \infty), \theta \in \mathbb{R}_{++}$ , for some distribution function  $F$ . Consider the problem of estimating  $\theta$  under the loss function

$$L(\theta, a) = W(|\frac{a}{\theta} - 1|), \quad a \in \mathcal{A} = \Theta, \theta \in \Theta,$$

where  $W : \mathbb{R}_+ \rightarrow \mathbb{R}_+$  is some function.

- (a) Show that the estimation/decision problem is invariant under the scale (multiplicative) group  $\mathcal{G} = \{g_c : c \in \mathbb{R}\}$ , where  $g_c(x) = cx, x \in \chi, c \in \mathbb{R}_{++}$ ;

- (b) Show that a decision rule  $\delta$  is invariant under  $\mathcal{G}$  if, and only if,  $\delta(A|x) = \delta(cA|cx), \forall A \in \mathcal{F}_{\mathcal{A}}, x \in \chi, c \in \mathbb{R}_{++}$ , where  $cA = \{ca : a \in A\}$  (or  $cY_x \stackrel{d}{=} Y_{cx}, \forall x \in \chi, c \in \mathbb{R}_+$ , where, for each  $x \in \chi$ ,  $Y_x$  is a random variable corresponding to probability measure  $\delta(\cdot|x)$ );
- (c) Show that a non-randomized decision rule  $\delta$  is invariant if, and only if,  $\delta(x) = cx, x \in \chi$ , for some  $c \in \mathbb{R}$ ;
- (d) For any invariant decision rule  $\delta$ , show that the risk function  $R_{\delta}(\theta)$  is constant (does not depend on  $\theta \in \Theta$ ).

**(Note:** The loss function  $L(\theta, a) = (\frac{a}{\theta} - 1)^2, a \in \mathcal{A}, \theta \in \Theta$ , corresponding to the choice  $W(x) = x^2, x \in \mathbb{R}_+$ , is called the scaled squared error loss function.)

9. Let  $\mathcal{P} = \{F_{\theta} : \theta \in \Theta\}$ , where  $\Theta = \mathbb{R}$  and  $F_{\theta}(x) = F(x - \theta), x \in \chi = \mathbb{R}, \theta \in \Theta$ , for some distribution function  $F$ . Consider the problem of estimating  $\theta$  under the loss function

$$L(\theta, a) = W(|a - \theta|), a \in \mathcal{A} = \Theta, \theta \in \Theta,$$

where  $W : \mathbb{R}_+ \rightarrow \mathbb{R}_+$  is some function.

- (a) Show that the estimation/decision problem is invariant under the additive group  $\mathcal{G} = \{g_c : c \in \mathbb{R}\}$ , where  $g_c(x) = x + c, x \in \chi, c \in \mathbb{R}$ ;
- (b) Show that a decision rule  $\delta$  is invariant under  $\mathcal{G}$  if, and only if,  $\delta(A|x) = \delta(A + c|x + c), \forall A \in \mathcal{F}_{\mathcal{A}}, x \in \chi, c \in \mathbb{R}$ , where  $A + c = \{a + c : a \in A\}$  (or  $Y_x + c \stackrel{d}{=} Y_{x+c}, \forall x \in \chi, c \in \mathbb{R}$ , where, for each  $x \in \chi$ ,  $Y_x$  is a random variable corresponding to probability measure  $\delta(\cdot|x)$ );
- (c) Show that a non-randomized decision rule  $\delta$  is invariant if, and only if,  $\delta(x) = x + c, x \in \chi$ , for some  $c \in \mathbb{R}$ ;
- (d) For any invariant decision rule  $\delta$ , show that the risk function  $R_{\delta}(\theta)$  is constant (does not depend on  $\theta \in \Theta$ ).