

### 1 Problem 1

Let  $f_n(t) = \left(\frac{t}{T}\right)^n$  for  $t \in [0, T]$ . Then we have that  $f_n \in C[0, T]$  for  $n = 1, 2, \dots$ . Let  $K = \{f_n(t), n = 1, 2, \dots\}$ . In order to prove  $K$  is closed, it suffices to prove  $C[0, T] \setminus K$  is open. For any  $f \in C[0, T] \setminus K$ , assume  $\inf_n \|f - f_n\| = 0$ . Since  $\liminf_n \|f - f_n\| \leq \inf_n \|f - f_n\|$ , then we have  $\liminf_n \|f - f_n\| = 0$ . Then there exists a subsequence  $\{f_{n_i}, i = 1, 2, \dots\}$  such that  $\|f - f_{n_i}\| \rightarrow 0$  as  $i \rightarrow \infty$ . Thus, we have

$$f(t) = \begin{cases} 0, & \text{if } 0 \leq t < T \\ 1, & \text{if } t = T \end{cases}$$

However,  $f$  is not continuous and thus  $f \notin C[0, T] \setminus K$ . Therefore, we have  $\inf_n \|f - f_n\| > 0$ . There exists a  $\epsilon > 0$  such that  $B(f, \epsilon) \in C[0, T] \setminus K$ , namely,  $C[0, T] \setminus K$  is open. Also, we have that

$$\|f_n\| \leq 1, \quad n = 1, 2, \dots$$

Thus,  $K$  is also bounded. For two consecutive  $f_n(t)$  and  $f_{n+1}(t)$ , we have

$$\begin{aligned} \|f_n(t) - f_{n+1}(t)\| &= \sup_{0 \leq t \leq T} \left| \left( \frac{t}{T} \right)^n \left( 1 - \frac{t}{T} \right) \right| \\ &= \left( \frac{n}{n+1} \right)^n \frac{1}{1+n} \\ &> \left( \frac{1}{3} \right)^n \frac{1}{1+n} \quad \text{for } n = 1, 2, \dots \end{aligned}$$

Then

$$\cup_{n=1}^{\infty} \left\{ B_o \left( f_n(t), \left( \frac{1}{3} \right)^n \frac{1}{1+n} \right) \right\}$$

is an open cover of  $K$ , but in this case we can not find a finite subset

$$\cup_{i=1}^m \left\{ B_o \left( f_{n_i}(t), \left( \frac{1}{3} \right)^{n_i} \frac{1}{1+n_i} \right) \right\} \quad \text{where } m \text{ is finite.}$$

such that it is a finite subcover of  $K$ . Thus,  $K$  is not compact.

## 2 Problem 2

*Proof.* ( $\Rightarrow$ ) Suppose  $f : S_1 \rightarrow S_2$  is continuous. For any open set  $O \subset S_2$ , we have  $f^{-1}(O) \subset S_1$ . For a fixed  $x \in f^{-1}(O)$ , we have that  $f(x) \in O$ . Since  $O$  is open, there exists an  $\epsilon > 0$  such that  $B_o(f(x), \epsilon) \subset O$ . Since  $f$  is continuous, there exists a  $\delta > 0$  such that  $f(B_o(x, \delta)) \subset B_o(f(x), \epsilon) \subset O$ . Thus, we have  $B_o(x, \delta) \subset f^{-1}(O)$ . That is,  $f^{-1}(O)$  is open.

( $\Leftarrow$ ) Suppose  $f^{-1}(O)$  is open in  $S_1$  for every open set  $O \in S_2$ . For an  $x \in S_1$ , there is an  $\epsilon > 0$  such that  $B_o(f(x), \epsilon)$  is an open set in  $S_2$ . Thus, we have that  $f^{-1}(B_o(f(x), \epsilon))$  is an open set in  $S_1$ . For any  $x \in f^{-1}(B_o(f(x), \epsilon))$ , there exists an  $\delta > 0$  such that  $B_o(x, \delta) \subset f^{-1}(B_o(f(x), \epsilon))$  which yields  $f(B_o(x, \delta)) \subset B_o(f(x), \epsilon)$ . Thus,  $f$  is continuous.  $\square$

## 3 Problem 3

*Proof.* Suppose  $f_1, f_2, \dots$  is a cauchy sequence in  $C[0, T]$  with the uniform metric  $\|x - y\|$ . For an  $\epsilon > 0$ , there exists an  $N > 0$  such that for any  $n_1, n_2 > N$ , we have

$$\begin{aligned}\epsilon > \|f_{n_1} - f_{n_2}\| &= \sup_{t \in [0, T]} |f_{n_1}(t) - f_{n_2}(t)| \\ &\geq |f_{n_1}(t) - f_{n_2}(t)|, \text{ for any } t \in [0, T].\end{aligned}$$

Thus, for a fixed  $t \in [0, T]$ ,  $f_1(t), f_2(t), \dots$  is a Cauchy sequence in  $\mathbb{R}$ . Since  $\mathbb{R}$  is complete, we have  $f_n(t) \rightarrow f(t)$  as  $n \rightarrow \infty$ .

Next, we need to show that  $f(t)$  is continuous on  $[0, T]$ . For  $t_1 \in [0, T]$  and any  $\delta > 0$ , there exists an  $m$  large enough, and an  $\eta > 0$  such that for every  $t_2$  satisfying  $|t_1 - t_2| < \eta$ , we have that

$$\begin{aligned}|f(t_1) - f_m(t_1)| &< \frac{\delta}{3} \quad (\text{by convergence of } \{f_n(t)\} \text{ for a fixed } t.) \\ |f(t_2) - f_m(t_2)| &< \frac{\delta}{3} \quad (\text{by convergence of } \{f_n(t)\} \text{ for a fixed } t.) \\ |f_m(t_1) - f_m(t_2)| &< \frac{\delta}{3} \quad (\text{by the continuity of } f_m(t).)\end{aligned}$$

By the triangle inequality, we have

$$|f(t_1) - f(t_2)| \leq |f(t_1) - f_m(t_1)| + |f_m(t_1) - f_m(t_2)| + |f_m(t_2) - f(t_2)| < \delta$$

$f(t)$  is continuous on  $[0, T]$ , which completes the proof.  $\square$

#### 4 Problem 4

*Proof.* **Part a.**

$M(0) = E[e^0] = 1$ . If  $M(\theta) < \infty$  for some  $\theta > 0$ , then for any  $\theta' \in (0, \theta]$ , we have

$$\begin{aligned} M(\theta') &= \mathbb{E}(e^{\theta'X}) = \int_{-\infty}^{\infty} e^{\theta'x} dP(x) \\ &\leq \int_0^{\infty} e^{\theta'x} dP(x) + 1 \\ &\leq M(\theta) + 1 < \infty \end{aligned}$$

Likewise if  $M(\theta) < \infty$  for some  $\theta < 0$ , then for any  $\theta' \in [\theta, 0)$ , we have

$$\begin{aligned} M(\theta') &= \mathbb{E}(e^{\theta'X}) = \int_{-\infty}^{\infty} e^{\theta'x} dP(x) \\ &\leq \int_{-\infty}^0 e^{\theta'x} dP(x) + 1 \\ &\leq M(\theta) + 1 < \infty \end{aligned}$$

**Part b.**

Suppose  $X$  has Cauchy distribution, i.e. its density function is

$$f_X(x) = \frac{1}{\pi(1+x^2)}$$

However, for any  $\theta \neq 0$ ,  $\lim_{|x| \rightarrow +\infty} \frac{\exp(|\theta x|)}{x^2} = +\infty$ , and the function  $\frac{\exp(\theta x)}{1+x^2}$  is therefore not integrable.

**Part c.**

Let  $X$  be a random variable with the following probability density function.

$$f_X(x) = \begin{cases} A \exp(x - \sqrt{-x}), & \text{if } x \leq 0 \\ A \exp(-x - \sqrt{x}), & \text{if } x > 0 \end{cases}$$

where  $A \doteq 1.10045$  is a normalizing constant. For  $\theta \in [-1, 1]$ . It is readily verified that

$$M(\theta) = \mathbb{E}[\exp(\theta X)] = A \int_{-\infty}^0 \exp((1+\theta)x - \sqrt{-x}) dx + A \int_0^{\infty} \exp((\theta-1)x - \sqrt{x}) dx$$

is finite while for any  $\theta$  outside of  $[-1, 1]$ ,  $M(\theta)$  is not finite.

**Part d.**

Consider a Bernoulli random variable  $X \sim Be(1/2)$ , so  $\mathbb{E}(X) = 1/2$ .  $X$  satisfies all requirements with  $x_0 = 1$ . We compute that  $M(\theta) = \frac{1}{2}(1 + e^\theta)$  which is finite for all  $\theta$ . The rate function is

$$I(x) = \sup_{\theta} \{x\theta - \log(1 + \exp(\theta)) + \log 2\}$$

We differentiate the expression above and obtain

$$\frac{d}{d\theta} (\theta - \log(1 + \exp(\theta)) + \log 2) = x - \frac{\exp(\theta)}{1 + \exp(\theta)}$$

Solving for  $\theta$ , we obtain  $\exp(\theta) = \frac{x}{1-x}$ . For  $0 \leq x < 1$ , the equation admits solution  $\theta = \log(\frac{x}{1-x})$  and the rate function is  $I(x) = x \log(x) + (1-x) \log(1-x) + \log(2)$ . Let  $x = 1$ . Then  $\theta - \log(1 + \exp(\theta)) \leq \theta - \log(\exp(\theta)) \leq 0$ , so that  $\{\theta - \log(1 + \exp(\theta)) + \log 2\}$  is bounded by  $\log 2$  and admits a finite supremum, and  $I(1)$  is finite. For  $x > 1$  (and  $\theta > 0$ ),

$$\theta x - \log\left(\frac{1 + \exp(\theta)}{2}\right) \geq \theta x - \log(\exp(\theta)) \geq (x-1)\theta$$

Taking  $\theta \rightarrow +\infty$ , we obtain  $I(x) = +\infty$ .  $\square$

## 5 Problem 5

*Proof.* Consider two strictly positive sequences  $x_n > 0$  and  $y_n > 0$ . Since  $\limsup_n \frac{\log x_n}{n} \leq I$  and  $\limsup_n \frac{\log y_n}{n} \leq I$ , then for any  $\epsilon_1 > 0$ , there exists an  $N$  such that for any  $n > N$ , we have

$$\sup_{m \geq n} \frac{\log x_m}{m} \leq I + \epsilon_1, \quad \sup_{m \geq n} \frac{\log y_m}{m} \leq I + \epsilon_1$$

which yields that

$$\max\left\{\sup_{m \geq n} \frac{\log x_m}{m}, \sup_{m \geq n} \frac{\log y_m}{m}\right\} \leq I + \epsilon_1$$

Thus, for any  $m \geq n$ ,

$$\max\left\{\frac{\log x_m}{m}, \frac{\log y_m}{m}\right\} \leq I + \epsilon_1$$

Taking sup for both sides of the last inequality gives

$$\sup_{m \geq n} \left\{ \max\left\{\frac{\log x_m}{m}, \frac{\log y_m}{m}\right\} \right\} \leq I + \epsilon_1$$

For any  $\epsilon_2 > 0$ , we can choose  $n$  large enough such that

$$\frac{\log 2}{n} + \sup_{m \geq n} \left\{ \max \left\{ \frac{\log x_m}{m}, \frac{\log y_m}{m} \right\} \right\} \leq I + \epsilon_1 + \epsilon_2$$

which gives that

$$\sup_{m \geq n} \left\{ \max \left\{ \frac{\log(2x_m)}{m}, \frac{\log(2y_m)}{m} \right\} \right\} \leq \frac{\log 2}{n} + \sup_{m \geq n} \left\{ \max \left\{ \frac{\log x_m}{m}, \frac{\log y_m}{m} \right\} \right\} \leq I + \epsilon_1 + \epsilon_2$$

Since as  $n \rightarrow \infty$ , we can make  $\epsilon$  and  $\epsilon_2$  both approach to 0. Thus,

$$\lim_{n \rightarrow \infty} \sup_{m \geq n} \frac{\log(x_m + y_m)}{m} \leq I$$

that is,  $\limsup_{n \rightarrow \infty} \frac{\log(x_n + y_n)}{n} \leq I$ .  $\square$

## 6 Problem 6

*Proof.* For any  $x_1, x_2$  and  $\alpha \in [0, 1]$ , let  $x = \alpha x_1 + (1 - \alpha)x_2$ , and observe

$$\begin{aligned} I(x) &= I(\alpha x_1 + (1 - \alpha)x_2) = \sup_{\theta} (\theta(\alpha x_1 + (1 - \alpha)x_2) - \log M(\theta)) \\ &= \sup_{\theta} (\alpha(\theta x_1 - \log M(\theta)) + (1 - \alpha)(\theta x_2 - \log M(\theta))) \\ &\leq \alpha \sup_{\theta} (\theta x_1 - \log M(\theta)) + (1 - \alpha) \sup_{\theta} (\theta x_2 - \log M(\theta)) \\ &\leq \alpha I(x_1) + (1 - \alpha)I(x_2) \end{aligned} \tag{1}$$

If  $I(x)$  is not strictly convex, there exists  $x_1 \neq x_2$ , and  $\alpha \in (0, 1)$  such that

$$\begin{aligned} &\sup_{\theta} \{ \alpha(x_1\theta - \log M(\theta)) + (1 - \alpha)(x_2\theta - \log M(\theta)) \} \\ &= \alpha \sup_{\theta} \{ x_1\theta - \log M(\theta) \} + (1 - \alpha) \sup_{\theta} \{ x_2\theta - \log M(\theta) \} \end{aligned}$$

However, we know that for every  $x \in \mathbb{R}$  there exists  $\theta_0 \in \mathbb{R}$  such that  $I(x) = \theta_0 x - \log M(\theta_0)$ . Moreover,  $\theta_0$  satisfies

$$x = \frac{\dot{M}(\theta_0)}{M(\theta_0)}$$

Let  $\theta_0 \in \mathbb{R}$  such that  $\frac{\dot{M}(\theta_0)}{M(\theta_0)} = \alpha x_1 + (1 - \alpha)x_2$ . We have

$$\begin{aligned} I(x) &= \theta_0(\alpha x_1 + (1 - \alpha)x_2) - \log M(\theta_0) \\ &= \alpha(\theta_0 x_1 - \log M(\theta_0)) + (1 - \alpha)(\theta_0 x_2 - \log M(\theta_0)) \end{aligned}$$

Clearly, if either  $\theta_0 x_1 - \log M(\theta_0) < \sup_\theta (\theta x_1 - \log M(\theta))$  or  $\theta_0 x_2 - \log M(\theta_0) < \sup_\theta (\theta x_2 - \log M(\theta))$ , then the equality does not hold. Therefore  $\theta_0$  also achieves the maximum for both  $x_1$  and  $x_2$ . By first order conditions, we also obtain

$$\frac{\dot{M}(\theta_0)}{M(\theta_0)} = x_1, \quad \frac{\dot{M}(\theta_0)}{M(\theta_0)} = x_2$$

which implies that  $x_1 = x_2$  and thus gives a contradiction.

□

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