

# MAS651 Theory of Stochastic Processes

## Homework #10

20150597 Jeonghwan Lee

Department of Mathematical Sciences, KAIST

June 11, 2021

Throughout this homework, let  $\mathbb{Z}_+$  denote the set of all non-negative integers,  $\mathbb{R}_+$  be the set of all non-negative real numbers, and  $[a : b] := \{a, a + 1, \dots, b - 1, b\}$  for  $a, b \in \mathbb{Z}$  with  $a \leq b$ . We also write  $[n] := [1 : n]$  for  $n \in \mathbb{N}$ . Moreover,  $\uplus$  denotes the *disjoint union*, and given a set  $A$  and  $k \in \mathbb{Z}_+$ ,  $\binom{A}{k} := \{B \subseteq A : |B| = k\}$ .

**Problem 1** (*Exercise 8.1.1* in [1]).

Given any  $-\infty < a < 0 \leq b < +\infty$ , let  $T_{a,b} := \inf \{t \in \mathbb{R}_+ : B(t) \in \mathbb{R} \setminus (a, b)\}$ , where  $\{B(t) : t \in \mathbb{R}_+\}$  is a standard one-dimensional Brownian motion defined on a probability space  $(\Omega, \mathcal{F}, \mathbb{P})$ . From *Exercise 7.5.4* in [1], we know that

$$\mathbb{E}[T_{a,b}^2] \leq 4 \cdot \mathbb{E}[B(T_{a,b})^4]. \quad (1)$$

We know that *Exercise 7.5.4* in [1] was one of the problems in *Homework #9* of this course. Let  $(U, V)$  be an  $((-\infty, 0) \times [0, +\infty), \mathcal{B}((-\infty, 0) \times [0, +\infty)))$ -valued random variable on  $(\Omega, \mathcal{F}, \mathbb{P})$ , which is independent of  $\{B(t) : t \in \mathbb{R}_+\}$ , and has the following distribution:

$$\mathbb{P}\{(U, V) \in A\} = \frac{1}{c} \iint_A (v - u) dF(u) dF(v), \quad \forall A \in \mathcal{B}((-\infty, 0) \times [0, +\infty)), \quad (2)$$

where  $F(\cdot) : \mathbb{R} \rightarrow [0, 1]$  is the probability distribution function of the random variable  $X$  defined on  $(\Omega, \mathcal{F}, \mathbb{P})$  such that  $\mathbb{E}[X] = 0$  and  $\mathbb{E}[X^2] < +\infty$ , and  $c \in [0, +\infty)$  is the normalization constant given by

$$c := \int_{[0, +\infty)} v dF(v) \stackrel{(a)}{=} - \int_{(-\infty, 0)} u dF(u),$$

where the step (a) follows from the condition  $\mathbb{E}[X] = 0$ . At this point, we may assume that  $\mathbb{E}[X^2] > 0$  as otherwise trivial. For this case, we know that  $c > 0$  and one can check that the function

$$A \in \mathcal{B}((-\infty, 0) \times [0, +\infty)) \mapsto \frac{1}{c} \iint_A (v - u) dF(u) dF(v) \in \mathbb{R}_+$$

is a probability measure defined on  $((-\infty, 0) \times [0, +\infty), \mathcal{B}((-\infty, 0) \times [0, +\infty)))$ . We note that this proce-

sure guarantees the existence of a random variable  $(U, V)$  satisfying (2). To this end,

$$\begin{aligned}
\iint_{(-\infty, 0) \times [0, +\infty)} (v - u) dF(u) dF(v) &= \int_{[0, +\infty)} \left[ v \cdot \mathbb{P}\{X < 0\} - \underbrace{\int_{(-\infty, 0)} u dF(u)}_{=c} \right] dF(v) \\
&= \mathbb{P}\{X < 0\} \underbrace{\int_{[0, +\infty)} v dF(v)}_{=c} + c \cdot \mathbb{P}\{X \geq 0\} \\
&= c \cdot \mathbb{P}\{X < 0\} + c \cdot \mathbb{P}\{X \geq 0\} \\
&= c,
\end{aligned}$$

and this establishes our claim. We have shown in the proof of *Theorem 8.1.1* in [1] that for every bounded measurable function  $\varphi(\cdot) : (\mathbb{R}, \mathcal{B}(\mathbb{R})) \rightarrow (\mathbb{R}, \mathcal{B}(\mathbb{R}))$ ,

$$\int_{\mathbb{R}} \varphi(x) dF(x) = \mathbb{E}[\varphi(X)] = \mathbb{E}_{(U, V)} \left[ \int_{\{U, V\}} \varphi(x) \mu_{U, V}(dx) \right], \quad (3)$$

where the probability measure  $\mu_{u, v}$  on  $(\{u, v\}, 2^{\{u, v\}})$ ,  $-\infty < u < 0 \leq v < +\infty$ , is defined by

$$\mu_{u, v}(\{u\}) = \frac{v}{v - u} \quad \text{and} \quad \mu_{u, v}(\{v\}) = \frac{-u}{v - u}.$$

Note that for any  $-\infty < u < 0 \leq v < +\infty$ , we see that

$$\begin{aligned}
\mathbb{P}\{B(T_{u, v}) = u\} &= \mathbb{P}\{T_u < T_v\} \stackrel{(b)}{=} \frac{v}{v - u} = \mu_{u, v}(\{u\}); \\
\mathbb{P}\{B(T_{u, v}) = v\} &= \mathbb{P}\{T_u > T_v\} \stackrel{(c)}{=} \frac{-u}{v - u} = \mu_{u, v}(\{v\}),
\end{aligned}$$

where the step (b) and the step (c) holds by *Theorem 7.5.3* in [1] (we note that the case  $-\infty < u < 0 < v < +\infty$  makes use of *Theorem 7.5.3* in [1], and the case  $-\infty < u < 0 = v < +\infty$  is trivial). Therefore,  $\mu_{u, v}(\cdot)$  is the probability distribution of  $B(T_{u, v})$  under  $\mathbb{P}$  for every  $-\infty < u < 0 \leq v < +\infty$ . Thus for every bounded measurable function  $\varphi(\cdot) : (\mathbb{R}, \mathcal{B}(\mathbb{R})) \rightarrow (\mathbb{R}, \mathcal{B}(\mathbb{R}))$ ,

$$\begin{aligned}
\mathbb{E}[\varphi(X)] &= \mathbb{E} \left[ \mathbb{E} \left[ \int_{\{U, V\}} \varphi(x) \mu_{U, V}(dx) \middle| (U, V) \right] \right] \\
&= \mathbb{E}[\mathbb{E}[\varphi\{B(T_{U, V})\} | (U, V)]] \\
&= \mathbb{E}[\varphi\{B(T_{U, V})\}],
\end{aligned}$$

thereby  $X \stackrel{d}{=} B(T_{U, V})$  under  $\mathbb{P}$ . Hence,

$$\begin{aligned}
\mathbb{E}[T_{U, V}^2] &= \mathbb{E}[\mathbb{E}[T_{U, V}^2 | (U, V)]] \\
&\stackrel{(d)}{\leq} \mathbb{E}\left[4 \cdot \mathbb{E}[B(T_{U, V})^4 | (U, V)]\right] \\
&= 4 \cdot \mathbb{E}[B(T_{U, V})^4] \\
&\stackrel{(e)}{\leq} 4 \cdot \mathbb{E}[X^4],
\end{aligned}$$

where the step (d) follows from the inequality (1), and the step (e) is due to the fact  $X \stackrel{d}{=} B(T_{U,V})$  under  $\mathbb{P}$ . We finally note that  $T_{U,V}$  is a stopping time with respect to the natural filtration  $\{\mathcal{F}(t) : t \in \mathbb{R}_+\}$  generated by  $\{B(t) : t \in \mathbb{R}_+\}$ , conditionally given  $(U, V) = (u, v) \in (-\infty, 0) \times [0, +\infty)$ .

**Problem 2** (*Exercise 8.1.2.* in [1]).

Let  $X$  be a random variable defined on  $(\Omega, \mathcal{F})$  with  $\mathbb{E}[X] = 0$  and  $\mathbb{E}[X^2] = \sigma^2 < +\infty$ , and  $\{X_n\}_{n=1}^\infty \stackrel{\text{i.i.d.}}{\sim} X$  under  $\mathbb{P}$ . Let  $S_0 := 0$  and  $S_n := \sum_{k=1}^n X_k$  for  $n \in \mathbb{N}$ . From this one-dimensional simple random walk  $\{S_n\}_{n=0}^\infty$  defined on  $(\Omega, \mathcal{F}, \mathbb{P})$ , we consider the following continuous-time stochastic process  $\{S(t) : t \in \mathbb{R}_+\}$  on  $(\Omega, \mathcal{F})$  by

$$S(t) := \{1 - (t - [t])\} S_{[t]} + (t - [t]) S_{[t]+1}, \quad \forall t \in \mathbb{R}_+.$$

Given any  $n \in \mathbb{N}$ , let  $\{\hat{S}_n(t) : t \in \mathbb{R}_+\}$  be a continuous-time stochastic process  $\{S(t) : t \in \mathbb{R}_+\}$  defined on  $(\Omega, \mathcal{F})$  by

$$\hat{S}_n(t) := \frac{S(nt)}{\sqrt{n}}, \quad \forall t \in \mathbb{R}_+.$$

The Donsker's invariance principle (*Theorem 8.1.4* in [1]) implies

$$\pi_1 \left\{ \hat{S}_n(\cdot) \right\} \xrightarrow{d} \sigma \cdot \pi_1 \{B(\cdot)\} \quad \text{as } n \rightarrow \infty,$$

where  $\pi_L : (C([0, +\infty), \mathbb{R}), \mathcal{C}([0, +\infty), \mathbb{R})) \rightarrow (C([0, L], \mathbb{R}), \mathcal{C}([0, L], \mathbb{R}))$  is the *cutting-off map* defined by  $\pi_L(f(\cdot)) := f|_{[0, L]}$  for each  $L \in \mathbb{R}_+$ , and  $B(\cdot) : (\Omega, \mathcal{F}) \rightarrow (C([0, +\infty), \mathbb{R}), \mathcal{C}([0, +\infty), \mathbb{R}))$  is a map defined by

$$B(\cdot)(\omega)(t) := \begin{cases} B_t(\omega) & \text{if } \omega \in \mathcal{B}; \\ 0 & \text{otherwise,} \end{cases}$$

where  $\{B(t) : t \in \mathbb{R}_+\}$  is a standard one-dimensional Brownian motion defined on  $(\Omega, \mathcal{F}, \mathbb{P})$ , and

$$\mathcal{B} := \{\omega \in \Omega : \text{the sample path of } \{B(t) : t \in \mathbb{R}_+\} \text{ is continuous everywhere.}\} \in \mathcal{F}.$$

Now, it's time to solve the given problem. One can see that

$$\begin{aligned} R_n &:= 1 + \max \{S_k : k \in [0 : n]\} - \min \{S_k : k \in [0 : n]\} \\ &\stackrel{\text{(a)}}{=} 1 + \max \{S(t) : t \in [0, n]\} - \min \{S(t) : t \in [0, n]\} \\ &= 1 + \max \{S(nt) : t \in [0, 1]\} - \min \{S(nt) : t \in [0, 1]\}, \end{aligned}$$

where the step (a) follows from the fact that  $\{S(t) : t \in [0, n]\}$  attains its maximum and minimum at some integer values of  $t \in [0, n]$ , since the continuous-time stochastic process  $\{S(t) : t \in \mathbb{R}_+\}$  is defined by interpolating the given random walk  $\{S_n\}_{n=0}^\infty$  linearly. Thus, we have

$$\begin{aligned} \frac{R_n}{\sigma\sqrt{n}} &= \frac{1}{\sigma\sqrt{n}} + \max \left\{ \frac{S(nt)}{\sigma\sqrt{n}} : t \in [0, 1] \right\} - \min \left\{ \frac{S(nt)}{\sigma\sqrt{n}} : t \in [0, 1] \right\} \\ &= \frac{1}{\sigma\sqrt{n}} + \max \left\{ \sigma^{-1} \hat{S}_n(t) : t \in [0, 1] \right\} - \min \left\{ \sigma^{-1} \hat{S}_n(t) : t \in [0, 1] \right\} \end{aligned} \quad (4)$$

for every  $n \in \mathbb{N}$ . At this point, we define a functional  $\varphi : C([0, 1], \mathbb{R}) \rightarrow \mathbb{R}$  by

$$\varphi(f(\cdot)) := \max \{f(t) : t \in [0, 1]\} - \min \{f(t) : t \in [0, 1]\}, \quad \forall f(\cdot) \in C([0, 1], \mathbb{R}).$$

We claim that the map  $\varphi : C([0, 1], \mathbb{R}) \rightarrow \mathbb{R}$  is continuous under the uniform convergence topology, *i.e.*, the metric topology on  $C([0, 1], \mathbb{R})$  induced by the sup-norm. We recall that the sup-norm  $\|\cdot\|_{\text{sup}} : C(X, \mathbb{R}) \rightarrow \mathbb{R}_+$  defined on the space of continuous functions from a compact topological space  $X$  to  $\mathbb{R}$  is defined by

$$\|f\|_{\text{sup}} := \sup \{|f(x)| : x \in X\} = \max \{|f(x)| : x \in X\}.$$

In order to prove this claim, given any  $f(\cdot), g(\cdot) \in C([0, 1], \mathbb{R})$ , we may observe that

$$f(t) = g(t) + \{f(t) - g(t)\} \leq g(t) + \|g - f\|_{\text{sup}} \leq \max \{g(t) : t \in [0, 1]\} + \|g - f\|_{\text{sup}}, \quad \forall t \in [0, 1],$$

thereby we arrive at

$$\max \{f(t) : t \in [0, 1]\} \leq \max \{g(t) : t \in [0, 1]\} + \|g - f\|_{\text{sup}}, \quad \forall t \in [0, 1]. \quad (5)$$

On the other hand, one can see that

$$f(t) = g(t) + \{f(t) - g(t)\} \geq g(t) - \|g - f\|_{\text{sup}} \geq \min \{g(t) : t \in [0, 1]\} - \|g - f\|_{\text{sup}}, \quad \forall t \in [0, 1],$$

thereby we reach

$$\min \{f(t) : t \in [0, 1]\} \geq \min \{g(t) : t \in [0, 1]\} - \|g - f\|_{\text{sup}}, \quad \forall t \in [0, 1]. \quad (6)$$

Taking two pieces (5) and (6) collectively yields

$$\begin{aligned} \varphi(f(\cdot)) &= \max \{f(t) : t \in [0, 1]\} - \min \{f(t) : t \in [0, 1]\} \\ &\leq \max \{g(t) : t \in [0, 1]\} - \min \{g(t) : t \in [0, 1]\} + 2\|g - f\|_{\text{sup}} \\ &= \varphi(g(\cdot)) + 2\|g - f\|_{\text{sup}}. \end{aligned} \quad (7)$$

By interchanging the roles of  $f(\cdot)$  and  $g(\cdot)$  in the inequality (7), we obtain

$$\varphi(g(\cdot)) \leq \varphi(f(\cdot)) + 2\|g - f\|_{\text{sup}}. \quad (8)$$

So combining two bounds (7) and (8) produces

$$|\varphi(g(\cdot)) - \varphi(f(\cdot))| \leq 2\|g - f\|_{\text{sup}}, \quad \forall f(\cdot), g(\cdot) \in C([0, 1], \mathbb{R}). \quad (9)$$

From the inequality (9), we see that the function  $\varphi : C([0, 1], \mathbb{R}) \rightarrow \mathbb{R}$  is Lipschitz continuous with Lipschitz constant 2. So in particular, the function  $\varphi : C([0, 1], \mathbb{R}) \rightarrow \mathbb{R}$  is continuous under the uniform convergence topology on  $C([0, 1], \mathbb{R})$ , *i.e.*, the metric topology on  $C([0, 1], \mathbb{R})$  induced by the sup-norm. Due to *Theorem 8.1.5* in [1], a simple consequence of the Donsker's invariance principle (*Theorem 8.1.4* in [1]) together with the continuous mapping theorem (*Theorem 3.2.10* in [1]), we see from  $\mathbb{P}\{\pi_1\{B(\cdot)\} \in \mathcal{D}(\varphi)\} = 0$ , where

$$\mathcal{D}(\varphi) := \{f(\cdot) \in C([0, 1], \mathbb{R}) : \varphi : C([0, 1], \mathbb{R}) \rightarrow \mathbb{R} \text{ is not continuous at } f(\cdot)\}$$

denotes the discontinuity set of  $\varphi(\cdot)$ , that

$$\varphi \left\{ \pi_1 \left( \frac{\hat{S}_n(\cdot)}{\sigma} \right) \right\} \xrightarrow{d} \varphi \{ \pi_1 (B(\cdot)) \} \quad \text{as } n \rightarrow \infty. \quad (10)$$

Here, we note that  $\mathcal{D}(\varphi) = \emptyset$ . Hence,

$$\begin{aligned}
\frac{R_n}{\sqrt{n}} &= \sigma \left( \frac{R_n}{\sigma\sqrt{n}} - \frac{1}{\sigma\sqrt{n}} \right) + \frac{1}{\sqrt{n}} \\
&\stackrel{\text{(b)}}{=} \sigma \cdot \varphi \left\{ \pi_1 \left( \frac{\hat{S}_n(\cdot)}{\sigma} \right) \right\} + \frac{1}{\sqrt{n}} \\
&\stackrel{\text{(c)}}{\xrightarrow{d}} \sigma \cdot \{ \pi_1(B(\cdot)) \} \\
&= \sigma (\max \{B(t) : t \in [0, 1]\} - \min \{B(t) : t \in [0, 1]\}),
\end{aligned}$$

where the step (b) follows from (4), and the step (c) makes use of the result (10) together with the converging together lemma (*Exercise 3.2.13* in [1]). This establishes our desired result!

## References

- [1] Rick Durrett. *Probability: theory and examples*, volume 49. Cambridge university press, 2019.