

# LTCC: MEASURE-THEORETIC PROBABILITY

## EXAMINATION SOLUTIONS, 2011

Q1. (i) For a random variable  $X \sim U[0, 1]$ , take its dyadic expansion  $X = \sum_1^\infty \epsilon_n/2^n$ . Thus  $\epsilon_1 = 0$  iff  $X \in [0, 1/2)$ ,  $1$  iff  $X \in [1/2, 1]$  (or  $[1/2, 1]$ : we can omit  $1$ , as it carries  $0$  probability). If  $\epsilon_1, \dots, \epsilon_{n-1}$  are already defined, on the dyadic intervals  $[k/2^{n-1}, (k+1)/2^{n-1})$ , and independent fair coin-tosses (Bernoulli  $B(\frac{1}{2})$ ), split each interval into two halves:  $\epsilon_n = 0$  on the left half,  $1$  on the right half. Then  $\epsilon_n$  is again  $B(\frac{1}{2})$ , and is independent of  $\epsilon_1, \dots, \epsilon_{n-1}$ :

$$\begin{aligned} P(\epsilon_n = 0, \epsilon_1 = e_1, \dots, \epsilon_{n-1} = e_{n-1}) &= \frac{1}{2} P(\epsilon_1 = e_1, \dots, \epsilon_{n-1} = e_{n-1}) \\ &= P(\epsilon_n = 0) P(\epsilon_1 = e_1, \dots, \epsilon_{n-1} = e_{n-1}), \end{aligned}$$

and similarly for  $\epsilon_n = 1$ . By induction,  $\epsilon_n$  ( $n = 1, 2, \dots$ ) are independent  $B(\frac{1}{2})$ . Conversely, given  $\epsilon_n$  independent coin tosses, form  $X := \sum_1^\infty \epsilon_n/2^n$ . Then  $X_n := \sum_1^n \epsilon_k/2^k \rightarrow X$  a.s. The distribution function  $F_n$  of  $X_n$  has jumps  $1/2^n$  at  $k/2^n$ ,  $k = 0, 1, \dots, 2^n - 1$ . This ‘saw-tooth jump function’ converges to  $x$  on  $[0, 1]$ , the distribution function of  $U[0, 1]$  ( $\sup |F_n(x) - x| = 2^{-n} \rightarrow 0$  as  $n \rightarrow \infty$ ). So  $X \sim U[0, 1]$ . So if  $X = \sum_1^\infty \epsilon_n/2^n$ ,  $X \sim U[0, 1]$  iff  $\epsilon_n$  are independent coin tosses – the Lebesgue probability space models *both* (a) length on the unit interval *and* (b) infinitely many independent coin tosses.

(ii) From the given  $U[0, 1]$ , we get by dyadic expansion as above a sequence of independent coin-tosses  $\epsilon_n$ . Rearrange these into a two-suffix array  $\epsilon_{jk}$  (as with Cantor’s proof of 1873 that the rationals are countable). The  $\epsilon_{jk}$  are all independent, so the  $X_j := \sum \epsilon_{jk}/2^k$  are independent, and  $U[0, 1]$  by above. So from *one*  $U(0, 1)$ , we get in this way *infinitely many copies*.

(iii) The standard normal distribution function  $\Phi$  is continuous and strictly increasing, so its inverse function  $\Phi^{-1}(t) := \inf\{\Phi(x) \geq t\}$  ( $0 < t < 1$ ) is continuous and strictly increasing. Then if  $U \sim U[0, 1]$ ,  $X := \Phi^{-1}(U) \sim \Phi$ : for,  $\{X \leq x\} = \{\Phi^{-1}(U) \leq x\} = \{U \leq \Phi(x)\}$ , which has probability  $\Phi(x)$  as  $U$  is uniform. Hence by (ii) above we can then generate infinitely many independent standard normals. We can hence simulate a Brownian motion  $B = (B_t)$  from  $B_t = \sum_0^\infty \lambda_n Z_n \Delta_n(t)$ , with  $Z_n$  independent standard normals,  $\Delta_n(t)$  the Schauder functions and  $\lambda_n$  suitable normalising constants.

(iv) Similarly, given one  $U[0, 1]$ , we can split it as in (ii) into infinitely many independent  $U[0, 1]$ s. From each, we can generate a Brownian motion by (iii), giving infinitely many independent Brownian motions.

Q2. (i) The *Itô isometry* states that for  $f \in H^2 := H^2(0, T) := \{f : E[\int_0^T f^2(\omega, t)dt] < \infty\}$ ,

$$E[(\int_0^t f^2(\omega, u)dB_u)^2] = E[\int_0^t f^2(\omega, u)du].$$

(ii) *Conditional Itô isometry*. For  $0 \leq s \leq t \leq T$ ,

$$E[(\int_s^t f^2(\omega, u)dB_u)^2|\mathcal{F}_s] = E[\int_s^t f^2(\omega, u)du|\mathcal{F}_s].$$

*Proof.* It suffices to show that for all  $A \in \mathcal{F}_s$ ,

$$E[I(A)(\int_s^t f^2(\omega, u)dB_u)^2] = E[I(A) \int_s^t f^2(\omega, u)du].$$

This follows from the unconditional Itô isometry, applied to the integrand  $g(\omega, u) := fI_A(\omega)I_{(s,t]}(u)$ .

(iii) For  $s \leq t$ ,  $M_t := (\int_s^t f(\omega, u)dB_u)^2 - \int_0^t f^2(\omega, u)du$ ,

$$\begin{aligned} E[M_t|\mathcal{F}_s] &= E[\{(\int_0^s + \int_s^t)f_u dB_u\}^2|\mathcal{F}_s] - \int_0^s f_u^2 du - E[\int_s^t f_u^2 du|\mathcal{F}_s] \\ &= E[(\int_0^s f_u dB_u)^2] + 2(\int_0^s f_u dB_u)E[\int_s^t f_u dB_u|\mathcal{F}_s] + E[(\int_s^t f_u dB_u)^2|\mathcal{F}_s] - \int_0^s f_u^2 du - E[\int_s^t f_u^2 du|\mathcal{F}_s]. \end{aligned}$$

The first and fourth terms give  $M_s$ . The third and fifth terms cancel, by the conditional Itô isometry (ii). The second factor in the second term involves an Itô integral, which (for an integral  $f \in H^2$ ) is a martingale, so has constant expectation, which is 0 on taking  $t = s$ , so the second term is 0. Combining, the RHS is  $M_s$ , which proves that  $M$  is a martingale.

(iv) Taking  $f \equiv 1$  gives  $M_t := B_t^2 - t$  is a martingale.

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