

**SOLUTION 9 15.12.2015**

Q1. We have to check the defining property  $(CE)$  (V.1, L18) for  $B = \emptyset$  and  $B = \Omega$ . For  $B = \emptyset$  both sides are zero; for  $B = \Omega$  both sides are  $EY$ . //

(ii) We have to check  $(CE)$  for *all* sets  $B \in \mathcal{A}$ . The only integrand that integrates like  $Y$  over *all* sets is  $Y$  itself, or a function agreeing with  $Y$  except on a set of measure zero.

(iii) Recall that  $Y$  is *always*  $\mathcal{A}$ -measurable (this is the definition of  $Y$  being a random variable). For  $\mathcal{B} \subset \mathcal{A}$ ,  $Y$  may not be  $\mathcal{B}$ -measurable, but if it is, the proof above applies with  $\mathcal{B}$  in place of  $\mathcal{A}$ .

(iv) (Tower property). If  $\mathcal{C} \subset \mathcal{B}$ ,  $E[E(Y|\mathcal{B})|\mathcal{C}] = E[Y|\mathcal{C}]$  a.s.

*Proof.*  $E_{\mathcal{C}}E_{\mathcal{B}}Y$  is  $\mathcal{C}$ -measurable, and for  $C \in \mathcal{C} \subset \mathcal{B}$ ,

$$\begin{aligned} \int_C E_{\mathcal{C}}[E_{\mathcal{B}}Y]dP &= \int_C E_{\mathcal{B}}YdP \quad (\text{definition of } E_{\mathcal{C}} \text{ as } C \in \mathcal{C}) \\ &= \int_C YdP \quad (\text{definition of } E_{\mathcal{B}} \text{ as } C \in \mathcal{B}). \end{aligned}$$

So  $E_{\mathcal{C}}[E_{\mathcal{B}}Y]$  satisfies the defining relation for  $E_{\mathcal{C}}Y$ . Being also  $\mathcal{C}$ -measurable, it *is*  $E_{\mathcal{C}}Y$  (a.s.). //

9 (iv') (Tower property). If  $\mathcal{C} \subset \mathcal{B}$ ,  $E[E(Y|\mathcal{C})|\mathcal{B}] = E[Y|\mathcal{C}]$  a.s.

*Proof.*  $E[Y|\mathcal{C}]$  is  $\mathcal{C}$ -measurable, so  $\mathcal{B}$ -measurable as  $\mathcal{C} \subset \mathcal{B}$ , so  $E[.|\mathcal{B}]$  has no effect, by (iii). //

*Corollary.*  $E[E(Y|\mathcal{C})|\mathcal{C}] = E[Y|\mathcal{C}]$  a.s.

So the operation  $E[.|\mathcal{C}]$  is linear and *idempotent* (doing it twice is the same as doing it once), so is a *projection*. So we can use what we know about projections, from Ch. IV, Linear Algebra, Functional Analysis etc.

Q2. (i) For  $t \neq 0$ ,  $X$  is Gaussian with zero mean (as  $B$  is), and continuous (again, as  $B$  is). The covariance of  $B$  is  $\min(s, t)$ . The covariance of  $X$  is

$$\begin{aligned} \text{cov}(X_s, X_t) &= \text{cov}(sB(1/s), tB(1/t)) = E[sB(1/s).tB(1/t)] = st.E[B(1/s)B(1/t)] \\ &= st.\text{cov}(B(1/s), B(1/t)) = st.\min(1/s, 1/t) = \min(t, s) = \min(s, t). \end{aligned}$$

This is the same covariance as Brownian motion. So, away from the origin,  $X$  *is* Brownian motion, as a Gaussian process is uniquely characterized by

its mean and covariance (from the properties of the multivariate normal distribution). So  $X$  is continuous. So we can define it at the origin by continuity. So  $X$  is Brownian motion everywhere –  $X$  is BM.

(ii) Since Brownian motion is 0 at the origin,  $X(0) = 0$ . Since Brownian motion is continuous at the origin,  $X(t) \rightarrow 0$  as  $t \rightarrow 0$ . This says that

$$tB(1/t) \rightarrow 0 \quad (t \rightarrow 0), \quad \text{i.e.} \quad B(t)/t \rightarrow 0 \quad (t \rightarrow \infty).$$

*Note.* For  $t$  integer, this is the Strong Law of Large Numbers applied to the distribution of  $B(1)$ , which is standard normal. The above neat proof by *time-inversion* follows from the proof of existence of Brownian motion (defined to be continuous), given in lectures by the PWZ wavelet expansion.

Q3. Brownian bridge  $X_t := B_t - tB_1$  ( $t \in [0, 1]$ ) is Gaussian (it is obtained from the Gaussian process  $B$  by linear operations – as in the multivariate normal distribution, IV.3). It has mean 0 (as  $B$  does), and covariance

$$\begin{aligned} E[X_s X_t] &= E[(B_s - sB_1)(B_t - tB_1)] = E[B_s B_t] - tE[B_s B_1] - sE[B_t B_1] + stE[B_1^2] \\ &= \min(s, t) - t \min(s, 1) - s \min(t, 1) + st = \min(s, t) - st - st + st = \min(s, t) - st. \end{aligned}$$

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