

EXERCISE 2 - Brownian Bridge (V/IV - Bonus):

Real process X , SDE :
(Real and \mathcal{C}^0 standard BM)
$$\begin{cases} dX_t = -\frac{1}{1-t} X_t dt + dB_t, & 0 \leq t \leq 1; \\ X_0 = 0 \end{cases}$$

$$1) \quad \begin{cases} dX_t = b(t, X_t) dt + \sigma(t, X_t) dB_t \\ X_0 = 0 \in \mathbb{R} \end{cases} \quad w/ : \quad \begin{cases} b(t, x) := -\frac{x}{1-t} \\ \sigma(t, x) := 1 \end{cases}$$

• We have that X solves SDE on $[0, 1)$ if \exists solves SDE on $[0, T]$, $\forall T \in [0, 1)$.

• SDE is of the form from ex I: $dX_t = \alpha(t)dt + \beta(t)X_t dt + \sigma(t)dB_t$

with: $\alpha(t) = 0$, $\beta(t) = -\frac{1}{1-t}$, $\sigma(t) = 1$. And $\sup_{[0, T]} |\beta(t)| = \frac{1}{1-T} < +\infty$.

→ For the same argument we used in ex I, we have that:

\exists pathwise unique strong solution to the SDE, $\forall 0 \leq T < 1$. \square

2) FIND THE SOLUTION X ?

Using the ex I we have:

$$\Lambda(t) = \int_0^t \beta(s) ds = - \int_0^t \frac{1}{1-s} ds = \log(1-t)$$

$$\Rightarrow X_t = e^{\Lambda(t)} \left[0 + \int_0^t e^{-\Lambda(s)} \sigma(s) dB_s \right]$$

$$X_t = (1-t) \int_0^t \frac{1}{1-s} dB_s \quad : \text{ THE UNIQUE SOLUTION OF SDE.}$$

... It seems familiar ... (Brownian Bridge) ...

3) $X_t = (1-t) \int_0^t \frac{1}{1-s} dB_s$ is a continuous gaussian process since $f(s) = \frac{1}{1-s}$ is deterministic and in $L^2([0, T])$ $\forall 0 \leq T \leq 1$.

Let's now precise its parameters.

$$\mathbb{E}[X_t] = 0 \quad (f \in L^2[0, T] \text{ so its S.I. is a mg.}).$$

$$\begin{aligned} \text{Cov}(X_t, X_s) &= \text{Cov}\left((1-t) \int_0^t \frac{1}{1-r} dB_r, (1-s) \int_0^s \frac{1}{1-r} dB_r\right) \\ &= (1-t)(1-s) \cdot \text{Cov}\left(\int_0^t \frac{1}{1-r} dB_r, \int_0^s \frac{1}{1-r} dB_r\right) \\ &= (1-t)(1-s) \cdot \left[\mathbb{E}\left[\int_0^t \frac{1}{1-r} dB_r \int_0^s \frac{1}{1-r} dB_r\right] - \underbrace{\mathbb{E}[\dots]\mathbb{E}[\dots]}_{=0} \right] \\ &= (1-t)(1-s) \mathbb{E}\left[\int_0^t \frac{1}{1-r} dB_r \int_0^s \frac{1}{1-r} dB_r\right] \\ &= (1-t)(1-s) \mathbb{E}\left[\int_0^T \frac{1_{[0,t]}(r)}{1-r} dB_r \int_0^T \frac{1_{[0,s]}(r)}{1-r} dB_r\right] \\ &= (1-t)(1-s) \mathbb{E}\left[\int_0^T \left(\frac{1}{1-r}\right)^2 1_{[0, t \wedge s]}(r) dr\right] \\ &= (1-t)(1-s) \int_0^{t \wedge s} \left(\frac{1}{1-r}\right)^2 dr = (1-t)(1-s) \left[\frac{1}{1-r}\right]_0^{t \wedge s} \\ &= (1-t)(1-s) \left[\frac{1}{1-t \wedge s} - \frac{1-t \wedge s}{1-t \wedge s}\right] = (1-t)(1-s) \frac{t \wedge s}{1-t \wedge s} \\ &= \cancel{(1-t \wedge s)}(1-t \vee s) \frac{t \wedge s}{\cancel{1-t \wedge s}} = t \wedge s (1-t \vee s) \end{aligned}$$

$$\boxed{\text{Cov}(X_t, X_s) = t \wedge s (1-t \vee s) \quad \forall s, t \in [0, T], \forall T \in (0, 1)} \quad \square$$

So :

$$\boxed{X_t \sim \mathcal{N}(0, t(1-t))} \quad \square$$

Let's introduce the "Brownian Bridge":

$$\tilde{X}_t := B_t - tB_1 \quad \forall t \in [0, 1].$$

4) \tilde{X}_t follows a gaussian law as a Σ of two gaussian-distributed RV, $\forall t \in [0, 1]$.

$$\cdot \mathbb{E}[\tilde{X}_t] = \mathbb{E}[B_t - tB_1] = \mathbb{E}[B_t] - t\mathbb{E}[B_1] = 0.$$

$$\begin{aligned} \cdot \text{Cov}(\tilde{X}_t, \tilde{X}_s) &= \mathbb{E}[\tilde{X}_t \tilde{X}_s] = \mathbb{E}[(B_t - tB_1)(B_s - sB_1)] \\ &= \mathbb{E}[B_t B_s] - s \mathbb{E}[B_t B_1] - t \mathbb{E}[B_1 B_s] + ts \mathbb{E}[B_1^2] \\ &= t \wedge s - s \cdot t - t \cdot s + ts = t \wedge s - st + st \\ &= t \wedge s (1 - ts) \end{aligned}$$

$$\text{So } \boxed{\tilde{X}_t \sim \mathcal{N}(0, t(1-t)) \quad \forall t \in [0, 1].}$$

In other words,

$$\boxed{X_t \text{ and } \tilde{X}_t, \quad t \in [0, 1), \text{ are equivalent.}} \quad \square$$

5) Indistinguishable would mean that: $\mathbb{P}(\forall t \in [0, 1), X_t = \tilde{X}_t) = 1$, i.e.,
 $\mathbb{P}(\{\omega \in \Omega \mid \forall t \in [0, 1), X_t(\omega) = \tilde{X}_t(\omega)\}) = 1.$

\rightarrow They are not indistinguishable since $(X_t)_t$ is \mathcal{F}_t -adapted (\mathcal{F}_t : filtration of the B.M.) but $(\tilde{X}_t)_t$ is not (clearly, \tilde{X}_t depends of B_1). \square

Eventually, we introduce:

$$\tilde{B}_t = B_t - \int_0^t \frac{B_u - B_u}{(1-u)} du, \quad 0 \leq t < 1$$

$$\tilde{\mathcal{F}}_t = \sigma(B_s, 0 \leq s \leq t) \vee \sigma(B_1) \vee \mathcal{N}.$$

6) $(\tilde{B}_t)_t$ is:

- $\tilde{B}_0 = 0$ a.s.;
- \tilde{B}_t is gaussian (standard argument: Riemann-Stieltjes sum);
- $\mathbb{E}[\tilde{B}_t] = 0 \quad \forall 0 \leq t < 1$;
- $\text{Cov}(\tilde{B}_t, \tilde{B}_s) = \dots = s \wedge t$.
 \uparrow
 very long!

As a conclusion:

$(\tilde{B}_t)_t$ is a continuous natural BM. \square

7) Suppose that \tilde{B}_t is $\tilde{\mathcal{F}}_t$ -measurable:

Then:

$$B_1 = \underbrace{\left(\underbrace{\tilde{B}_t - B_t}_{\substack{\mathcal{F}_t\text{-meas} \\ \text{by assumption}}} + \underbrace{\int_0^t \underbrace{B_u}_{\mathcal{F}_t\text{-meas}} \cdot \underbrace{\frac{1}{1-u}}_{\substack{\mathcal{F}_t\text{-meas} \\ \text{deterministic}}} du \right)}_{\mathcal{F}_t\text{-meas}}.$$

So B_1 would be \mathcal{F}_t -measurable $\forall 0 \leq t < 1$. \swarrow

Contradiction. So:

\tilde{B}_t is not \mathcal{F}_t -measurable. \square

8) • \tilde{B}_t is $\tilde{\mathcal{F}}_t$ -adapted by def of $\tilde{\mathcal{F}}_t$.

• $\tilde{\mathcal{F}}_t$ is standard since $N \subset \tilde{\mathcal{F}}_t$, it is right continuous (since the augmented Brownian filtration $\sigma(B_s | 0 \leq s \leq t) \vee N$ is right continuous).

• \tilde{B}_t is continuous by construction.

• $\tilde{B}_t - \tilde{B}_s \perp \tilde{\mathcal{F}}_s \quad \forall 0 \leq s \leq t \leq T < 1$. Indeed: $\forall 0 \leq s \leq t \leq T$

$$E[(\tilde{B}_t - \tilde{B}_s) B_r] = 0.$$

$$\text{Moreover: } E[(\tilde{B}_t - \tilde{B}_s) B_1] = 0.$$

So they are uncorrelated. And by the usual "Gaussian family" argument, we conclude that $\tilde{B}_t - \tilde{B}_s \perp \sigma(B_1) \vee \sigma(B_r | 0 \leq r \leq s)$ and so:

$$\tilde{B}_t - \tilde{B}_s \perp \tilde{\mathcal{F}}_s.$$

• The gaussianity of the increments follows from 6.

9) Show that \tilde{X}_t is the solution to:

$$(\tilde{B}\tilde{B}) \quad \begin{cases} d\tilde{X}_t = -\frac{1}{1-t} \tilde{X}_t dt + d\tilde{B}_t & , 0 \leq t < 1. \\ \tilde{X}_0 = 0 \end{cases}$$

⚠ Here we are studying the behaviour of an Itô Process under a change of filtration, with \mathbb{P} fixed. So this situation is \neq from the usual change of probability "Girsanov's style".

We showed that \tilde{B}_t was a \mathcal{G}^0 std. BM wrt $(\tilde{\mathcal{F}}_t)_t$.

What we can do is to try to write \tilde{X}_t as a fct. of \tilde{B}_t and show that it is Itô with a stochastic

differential given by $(\tilde{B}\tilde{B})$.

- $\tilde{X}_t = B_t - tB_1$ so \tilde{X}_t is $\tilde{\mathcal{F}}_t$ measurable.

$$\int_0^t \tilde{X}_t = \tilde{B}_t + \int_0^t \frac{B_1 - B_u}{1-u} du - tB_1.$$

→ It's an Itô process, w.r.t $\tilde{\mathcal{F}}_t$.

- Let's compute $d\tilde{X}_t$:

$$d\tilde{X}_t = d\tilde{B}_t + \frac{B_1 - B_t}{1-t} dt - B_1 dt$$

$$d\tilde{X}_t = d\tilde{B}_t - \frac{1}{1-t} \tilde{X}_t dt : \boxed{\tilde{X}_t \text{ solves } (\tilde{B}\tilde{B})}. \quad \square$$

⚠ X_t and \tilde{X}_t solve the same equation, with different B.M and filtration.