
Exercise 1

$\mu_j \ll \mu_k$] whether distribution of μ_j is absolutely continuous with respect to μ_k

This means μ_j does not have more 0 probability events than μ_k

$X_1: \text{Normal}(2, 7)$] Continuous distribution in \mathbb{R}

$X_2: \text{Binomial}(7, 0.3)$] Discrete distribution with possible values 0 to 7

$X_3: \text{Poisson}(2)$] Discrete distribution over non-negative integers

$X_4: e^{X_1}$] X_4 is log normal since X_1 is normal

: continuous distribution & in range $(0, \infty)$

$X_5: \text{Uniform}(0, 1)$: Continuous distribution on $[0, 1]$

$X_6: X_1 \times X_3 + X_2$: Mixed distribution

Note:- A continuous & discrete distribution are mutually singular i.e. they are not absolutely continuous

- 2 continuous distributions are absolutely continuous only if their ranges/supports overlap

$\mu_4 \ll \mu_1 \rightarrow$ If μ_1 assigns 0 prob to a set A , then μ_4 also assigns 0 prob to the set A

\rightarrow Absolutely continuous

because $A \subseteq (0, \infty)$

Let $A = [1, 2] \rightarrow$ range

$$\mu_1(A) > 0$$

$$\mu_4(A) > 0$$

Let $A = 1$ [single point]

$$\mu_1(A) = 0$$

$$\mu_4(A) = 0$$

$\mu_1 \ll \mu_4 \rightarrow$ Not absolutely continuous
because $A \subseteq (-\infty, 0)$

$$\mu_1(A) > 0, \mu_4(A) = 0$$

[since out of range]

Similarly, $\mu_5 \ll \mu_1$ is absolutely continuous
as support of μ_5 is within the support
of μ_1

, $\mu_5 \ll \mu_4$ is absolutely continuous with the
same reason above

, $\mu_2 \ll \mu_3 \rightarrow$ though discrete, the poisson process
support contains the binomial process
support. \therefore absolutely
continuous

$\mu_5 \ll \mu_6 \rightarrow$ Absolutely continuous

$$X_6 = X_1 X_3 + X_2$$

[] []
 continuous discrete
 component component

$X_1 X_3 \rightarrow$ Normal distribution scaled by X_3
For each $X_3 = k \geq 1$, $X_1 k$ is Normal with mean $2k$
& variance $7k^2$
but still spans all real numbers

The continuous component of μ_6 dominates μ_5 on $[0,1]$

Similarity,

$\mu_4 \ll \mu_6 \rightarrow$ Absolutely continuous

Because the continuous component of μ_6 dominates

μ_4 on $[0, \infty]$

$\mu_2 \ll \mu_6 \rightarrow$ Absolutely continuous

The discrete component of μ_6 has the same support
as μ_2 at $[0, \dots, 7]$

Therefore, the only absolutely continuous pairs are \rightarrow

- $\mu_4 \ll \mu_1$
- $\mu_5 \ll \mu_4$
- $\mu_4 \ll \mu_6$
- $\mu_5 \ll \mu_1$
- $\mu_2 \ll \mu_3$
- $\mu_2 \ll \mu_6$

Other pairs are not absolutely continuous

Exercise 2

$W_n \rightarrow$ Winnings at time n

$$M_n = 1 - W_n$$

$$M_0 = 1, T = \min(n : M_n = 0)$$

Let F_n be the info in M_0, \dots, M_n

i) why M_n is a non-negative martingale

For any n , if we have lost all games upto n

$$W_n = 1 - 2^n$$

$$\therefore M_n = 1 - (1 - 2^n) = 2^n$$

If we have won the n^{th} game

$$W_n = 1$$

$$M_n = 1 - 1 = 0$$

$\therefore M_n$ will always be positive

$$\begin{aligned} E[M_{n+1} / F_n] &= E[(1 - W_{n+1}) / F_n] \\ &= [1 - E(W_{n+1} / F_n)] \\ &= [1 - (\frac{1}{2} \times 1 + \frac{1}{2} \times (1 - 2^{n+1}))] \\ &= [1 - (1 - 2^n)] = 2^n = M_n \end{aligned}$$

$\therefore M_n$ is a non negative martingale.

$$2) Q_n(v) = E[M_n \mathbf{1}_v]$$

To show $\Rightarrow Q_m(v) = Q_n(v)$ where $m < n$

$$\begin{aligned} Q_n(v) &= E[M_n \mathbf{1}_v] = E[E[M_n / F_m] \mathbf{1}_v] \\ &= E[M_m \mathbf{1}_v] = Q_m(v) \end{aligned} \quad \boxed{\text{Using Tower Property}}$$

$$3) Q\{M_{n+1} = 2^{n+1} / M_n = 2^n\}$$

$$P(M_{n+1} = 2^{n+1} / M_n = 2^n) = \frac{1}{2}$$

$$\text{let } A = (M_{n+1} = 2^{n+1})$$

$$B = (M_n = 2^n)$$

$$Q(A/B) = \frac{Q(A \cap B)}{Q(B)} = \frac{Q(M_{n+1} = 2^{n+1}, M_n = 2^n)}{Q(M_n = 2^n)}$$

$$= \frac{E^P[M_{n+1} \mathbf{1}_{(M_{n+1} = 2^{n+1}, M_n = 2^n)}]}{E^P[M_n \mathbf{1}_{(M_n = 2^n)}]}$$

$$= \frac{2^{n+1} P(M_{n+1} = 2^{n+1}, M_n = 2^n)}{2^n P(M_n = 2^n)}$$

$$= \frac{2^{n+1}}{2^n} P(M_{n+1} = 2^{n+1} / M_n = 2^n)$$

$$= 2 \times \frac{1}{2} = 1$$

4) Q probability that $T < \infty$

From part 3,

M_n always transitions to 2^{n+1} (never to 0)

Thus, $T = \infty$ with probability 1 under \mathbb{Q}

$$\therefore \mathbb{Q}(T < \infty) = 0$$

$$\begin{aligned} 5) E[M_{n+1}/F_n] &= E[2M_n/F_n] = 2E[M_n/F_n] \\ &= 2M_n \\ &= 2^{n+1} \end{aligned}$$

$$E[M_{n+1}/F_n] \neq M_n$$

$\therefore M_n$ is not a martingale under \mathbb{Q}

Exercise 3

B_t is a standard BM on (Ω, P)

Is there prob measure \mathbb{Q} such that X_t ($0 \leq t \leq 1$) is standard BM

$$B_0 = 0, X_0 = 0$$

$$1) dX_t = 2dt + dB_t$$

drift = 2 & variance parameter = 1
(μ)

Novikov's condition \rightarrow

$$E^P \left[\exp \left(\frac{1}{2} \int_0^1 \mu^2 dt \right) \right] = \exp(2) < \infty$$

$\therefore \mathbb{Q}$ exists

We can apply Girsanov's theorem as volatility matches standard brownian motion ($\sigma = 1$), we can remove the drift

The Radon-Nikodym derivative $\frac{d\mathbb{Q}}{dP}$ is predefined as:

$$\frac{d\mathbb{Q}}{dP} = \exp \left(-\mu B_t - \frac{1}{2} \mu^2 T \right)$$

where $\mu \rightarrow \text{drift}$ & $T \rightarrow \text{Time horizon (variance parameter)}$

$$\frac{dQ}{dP} = \exp\left(-2B_t - \frac{1}{2} \times 2^2 \times 1\right) = \exp(-2B_t - 2)$$

The term $\exp(-2B_t - 2)$ reweights probabilities under the new measure Q . It tilts the original measure P to cancel the drift $2dt$ in X_t .

$$dX_t = 2dt + dB_t$$

By defining $\frac{dQ}{dP} = \exp(-2B_t - 2)$, we define BM B_t^Q

$$dB_t^Q = dB_t + \mu dt = dB_t + 2dt \quad [\text{Girsanov Transformation}]$$

Substituting $dX_t = dB_t^Q - \mu dt$ into original SDE

$$dX_t = dB_t^Q$$

2) $dX_t = 2dt + 6dB_t$

drift (μ) = 2, variance parameter = 6

Girsanov's Theorem only adjusts the drift, not the volatility.

$\therefore Q$ does not exist

3) $dX_t = 2B_t dt + dB_t$

drift (μ) = $2B_t$, variance parameter = 1

Novikov's Condition \rightarrow

$$E^P \left[\exp \left(\frac{1}{2} \int_0^1 (2B_t)^2 dt \right) \right] = E^P \left[\exp \left(2 \int_0^1 B_t^2 dt \right) \right]$$

The integral has a heavy tailed behaviour because of the coefficient 2. \therefore it fails Novikov's condition
 $\therefore Q$ does not exist. [Note: If coefficient $< \frac{1}{2T}$, the expectation would be finite]

Exercise 4

B_t is standard BM with $B_0 = 0$

$$X_t = e^{-mB_t^2} \quad \text{where } m > 0$$

1) $M_t = X_t \exp\left(\int_0^t g(B_s) ds\right)$ then M_t is a local martingale

$$\text{Let } Y_t = \exp\left(\int_0^t g(B_s) ds\right)$$

$$\therefore M_t = X_t Y_t$$

$$dX_t = e^{-mB_t^2} \times -2mB_t dB_t + \frac{1}{2} \left(e^{-mB_t^2} \times -2m + (-2mB_t)^2 e^{-mB_t^2} \right) dt$$

$$= -2mX_t B_t dB_t + (-mX_t + 2m^2 B_t^2 X_t) dt$$

$$dY_t = \exp\left(\int_0^t g(B_s) ds\right) g(B_t) dt$$

$$= g(B_t) Y_t dt$$

[since Y_t is a time integral & not stochastic integral]

$$dM_t = Y_t dX_t + X_t dY_t + \underbrace{dX_t dY_t}_0$$

$$= Y_t \left[-2mX_t B_t dB_t - mX_t dt + 2m^2 B_t^2 X_t dt \right]$$

$$+ X_t \left[g(B_t) Y_t dt \right]$$

$$= M_t \left[-2mB_t dB_t + (2m^2 B_t^2 - m) dt \right] + M_t \left[g(B_t) dt \right]$$

$$dM_t = M_t [-2mB_t dB_t + (2m^2 B_t^2 - m + g(B_t)) dt]$$

For M_t to be a local martingale

$$2m^2 B_t^2 - m + g(B_t) = 0$$

$$\therefore g(B_t) = m - 2m^2 B_t^2$$

2) SDE for M_t

$$dM_t = -2mB_t M_t dB_t$$

3) $Q(V) = E[M_t \mathbf{1}_A]$

Find SDE satisfied by B_t wrt Q -Brownian motion

Girsanov's Theorem states that if M_t is a martingale under P , then under the new measure Q , the process:

$$\tilde{B}_t = B_t + \int_0^t \theta_s ds \quad \text{is a standard BM}$$

Since SDE of $M_t = dM_t = -2mB_t M_t dB_t$

$$\theta_t = 2mB_t \quad \left[\begin{array}{l} \text{Since } dM_t = -\theta_t M_t dB_t \text{ where} \\ \theta_t \text{ is the drift adjustment} \end{array} \right]$$

$$\tilde{B}_t = B_t + \int_0^t 2m B_s ds$$

$$d\tilde{B}_t = dB_t + 2m B_t dt$$

$$\therefore dB_t = d\tilde{B}_t - 2m B_t dt$$

4) Why M_t is a martingale & not local martingale

To prove M_t is a martingale, we can use Novikov's condition

Norikov's condition states that if

$$E \left[\exp \left(\frac{1}{2} \int_0^t \theta_s^2 ds \right) \right] < \infty$$

Then,

$$M_t = \exp \left(\int_0^t \theta_s dB_s - \frac{1}{2} \int_0^t \theta_s^2 ds \right)$$

is a true martingale

$$M_t = X_t Y_t$$

$$\begin{aligned} &= e^{-m B_t^2 + \int_0^t m - 2m^2 B_s^2 ds} \\ &= e^{-m B_t^2 + mt - 2m^2 \int_0^t B_s^2 ds} \end{aligned}$$

$$\text{Let } f(B_t) = -m B_t^2$$

$$d(-m B_t^2) = -2m B_t dB_t - m dt$$

$$-m B_t^2 = -2m \int_0^t B_s dB_s - mt$$

$$-2m \int_0^t B_s dB_s = -m B_t^2 + mt$$

$$M_t = e^{-2m \int_0^t B_s dB_s - 2m^2 \int_0^t B_s^2 ds}$$

$$= e^{\int_0^t -2m B_s dB_s - \frac{1}{2} \int_0^t (2m B_s)^2 ds}$$

This matches Norikov's martingale expression

If $E \left[\exp \left(\frac{1}{2} \int_0^t (2m B_s)^2 ds \right) \right] < \infty$, then M_t is martingale

$$= E \left[\exp \left(2m^2 \int_0^t B_s^2 ds \right) \right]$$

For small m , the coefficient $2m^2$ can be small enough to ensure light tails, making expectation finite.

Exercise 5

B_t is standard BM with $B_0 = 1$

$$T = \min\{t : B_t = 0\}$$

$$r > 0$$

$$X_t = B_t^r$$

$$1) M_t = X_t \exp \left\{ \int_0^t g(B_s) ds \right\}$$

$$\text{Let } Y_t = \exp \left\{ \int_0^t g(B_s) ds \right\}$$

$$dY_t = g(B_t) Y_t dt \quad [\text{from last question}]$$

$$dX_t = r B_t^{r-1} dB_t + \frac{1}{2} r(r-1) B_t^{r-2} dt$$

$$dM_t = Y_t dX_t + X_t dY_t + \underbrace{dX_t dY_t}_0$$

$$= Y_t \left[r B_t^{r-1} dB_t + \frac{1}{2} r(r-1) B_t^{r-2} dt \right]$$

$$+ B_t^r \left[g(B_t) Y_t dt \right]$$

$$= Y_t \left[r B_t^{r-1} dB_t + \left(\frac{1}{2} r(r-1) B_t^{r-2} + B_t^r g(B_t) \right) dt \right]$$

For M_t to be local martingale

$$\frac{1}{2} r(r-1) B_t^{r-2} + B_t^r g(B_t) = 0$$

$$g(B_t) = -\frac{1}{2} r(r-1) B_t^{r-2} = -\frac{1}{2} r(r-1) B_t^{-2}$$

2) SDE of M_t

$$dM_t = Y_t \left[r B_t^{r-1} dB_t + \left(\frac{1}{2} r(r-1) B_t^{r-2} + B_t^r \left(-\frac{1}{2} r(r-1) B_t^{-2} \right) \right) dt \right]$$

$$= Y_t [r B_t^{r-1} dB_t] = M_t r B_t^{-1} dB_t$$

3) SDE satisfied by B_t wrt \mathbb{Q} brownian motion

Girsanov's theorem states that if M_t is a martingale under P , then under the new measure \mathbb{Q} , the process:

$$\tilde{B}_t = B_t + \int_0^t \theta_s ds \quad \text{is a standard BM}$$

$$dM_t = M_t r B_t^{-1} dB_t$$

$$dM_t = -\theta M_t dB_t \quad] \quad \therefore \theta = -r B_t^{-1}$$

$$\tilde{B}_t = B_t + \int_0^t -r B_s^{-1} ds$$

$$d\tilde{B}_t = dB_t - r B_t^{-1} dt$$

$$dB_t = \tilde{B}_t + r B_t^{-1} dt$$

4) $\mathbb{Q}(T < \infty) = 0$

$$dB_t = d\tilde{B}_t + r B_t^{-1} dt$$

This follows a Bessel process $dX_t = \frac{\delta-1}{2} X_t^{\delta-1} dt + dW_t$

$$r = \frac{\delta-1}{2} \quad \therefore \delta = 2r+1$$

If $r \geq 2$, the process never hits 0
If $r < 2$, the process can hit 0

$$d \geq 2, r \geq \frac{2-1}{2} \geq \frac{1}{2}$$

\therefore if $r \geq \frac{1}{2}$, the process never hits 0

Hence $Q(T < \infty) = 0$

