## Homework 1

Let  $(W_t)_{t\geq 0}$  be a standard Brownian motion throughout.

1. Write down a formula for  $u(x,t) := \mathbb{E}(f(W_T)|W_t = x)$  for a general function f.

**Solution.** From the definition of BM, the conditional distribution of  $W_T$  given  $W_t = x$  is N(x, T - t), so

$$\mathbb{E}(f(W_T)|W_t = x) = \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi(T-t)}} e^{-\frac{(y-x)^2}{2(T-t)}} f(y) dy = (p_{T-t} * f)(x)$$

where  $p_t(.)$  is the density of  $W_t$  and \* denotes **convolution** (the convolution of two functions f and g is  $(f*g)(x) = \int_{-\infty}^{\infty} f(y)g(x-y)dy$ . Note by setting x-y=u so y=x-u and dy=-du, we see that  $(f*g)(x)=-\int_{-\infty}^{\infty} f(x-u)g(u)du=\int_{-\infty}^{\infty} f(x-u)g(u)du=(g*f)(x)$ , i.e. f\*g=g\*f.

2. Simulating random variables with a given distribution. Let X be a random variable with a continuous strictly increasing distribution function  $F_X(x)$ . What is the distribution of  $F_X^{-1}(U)$ , where U is a standard Uniform random variable on [0,1]? Explain the significance of this result.

Solution.

$$\mathbb{P}(F_X^{-1}(U) \le x) = \mathbb{P}(U \le F_X(x)) = F_X(x) \tag{1}$$

since  $F_X(F_X^{-1}(x)) = x$ , i.e.  $F_X^{-1}(U) \sim X$ , where we have used that  $\mathbb{P}(U \leq x) = x$ . So the conclusion here is that  $F_X^{-1}(U)$  has the same distribution as X. This is how we typically generate a random variable with a given distribution in practice on a computer.

**3**. Let

$$B_t = (1-t)W_{\underline{t}}$$

for  $0 \le t < 1$ . Compute  $\mathbb{E}(B_s B_t)$  for 0 < s < t < 1. What do you notice at t = 1?

**Solution**.  $\mathbb{E}(B_sB_t)=(1-s)(1-t)\frac{s}{1-s}=s(1-t)$ . We note that  $\mathbb{E}(B_1^2)=0$ , hence  $B_1=0$  a.s. B is known as the **Brownian bridge** (see Figure 1 below), which is Brownian motion conditioned to be at zero at time 1.

**4.** Let  $S_t = S_0 e^{(\mu - \frac{1}{2}\sigma^2)t + \sigma W_t}$  (this is the famous **Black-Scholes model**). Compute  $\mathbb{E}(S_t^p)$  (hint: re-write  $S_t^p$  as  $S_0^p e^{pX_t}$  where  $X_t = (\mu - \frac{1}{2}\sigma^2)t + \sigma W_t$ ).

Solution.

$$\mathbb{E}(S_t^p) = S_0^p \, \mathbb{E}(e^{pX_t}) = S_0^p \mathbb{E}(e^{p((\mu - \frac{1}{2}\sigma^2)t + \sigma W_t)}) \,.$$

But  $(\mu - \frac{1}{2}\sigma^2)t + \sigma W_t \sim N(\mu_1, \sigma_1^2)$ , where  $\mu_1 = (\mu - \frac{1}{2}\sigma^2)t$  and  $\sigma_1^2 = \sigma^2 t$ . Thus

$$\mathbb{E}(S_t^p) = S_0^p e^{\mu_1 p + \frac{1}{2}\sigma_1^2 p^2} = S_0^p e^{(\mu - \frac{1}{2}\sigma^2)pt + \frac{1}{2}\sigma^2 p^2 t}$$

where we have used that the mgf of a general Normal  $N(\mu_1, \sigma^2)$  random variable is  $e^{\mu_1 p + \frac{1}{2}\sigma_1^2 p^2}$  from Applied Probability Revision chapter. Note that  $\mathbb{E}(S_t^p) < \infty$  for all  $p \in \mathbb{R}$ , i.e. all moments of  $S_t$  are finite.

5. Let  $X_t = \sum_{i=1}^n (W_t^{(i)})^2$ , where  $W^{(i)}$  are n independent standard Brownian motions. Using that  $\mathbb{E}(e^{-\lambda Z^2}) = \frac{1}{(1+2\lambda)^{\frac{1}{2}}}$  for  $\lambda > 0$  where  $Z \sim N(0,1)$ , compute  $\mathbb{E}(e^{-\lambda X_t})$  for  $\lambda > 0$ . X is known as a **Bessel squared process** of dimension n

**Solution**. Using that  $B_t^{(i)} \sim \sqrt{t}Z$  and the independence of the  $\delta$ -Brownian motions, we see that

$$\mathbb{E}(e^{-\lambda X_t}) = \prod_{i=1}^{\delta} \mathbb{E}(e^{-\lambda (B_t^{(i)})^2}) = \prod_{i=1}^{\delta} \mathbb{E}(e^{-\lambda (\sqrt{t}Z)^2}) = (\mathbb{E}(e^{-\lambda tZ^2}))^{\delta} = \frac{1}{(1+2\lambda t)^{\frac{1}{2}\delta}}.$$

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6. Portfolio optimization, relevant for two of the summer projects. Consider a financial market defined on a probability space  $(\Omega, \mathcal{F}, \mathbb{P})$  with d assets with random payoffs  $(\pi^1, ..., \pi^d)$  at time T (which are linearly independent) with market prices  $p_i$  at t = 0 (these assets can include European call/put options). Let  $Y_i = \pi_i - p_i$ , and assume a financial agent can only trade at time zero.

Derive the first order optimality condition for an agent to maximize their **expected utility**  $\mathbb{E}(U(b \cdot Y))$  over  $b \in \mathbb{R}^d$ , where  $U : \mathbb{R} \to \mathbb{R}$  is a strictly increasing function with U''(x) < 0 and  $b_i$  is the position in the *i*'th asset and we assume that  $\mathbb{E}(U(b \cdot Y)) < \infty$  for all  $b \in \mathbb{R}^d$ .

**Solution**. As in first year calculus, we set compute derivatives wrt each  $b_i$  and then set the answer to zero:

$$\frac{\partial}{\partial b_i} \mathbb{E}(U(b \cdot Y)) = \mathbb{E}(Y_i U'(b \cdot Y)) = 0$$

for i = 1..d, i.e. we have d equations for the d unknowns  $b_1^*, ..., b_d^*$  for the optimal portfolio allocation  $b^*$ . Note we can re-write this as

$$\mathbb{E}^{\mathbb{Q}}(Y_i) = 0 \tag{2}$$

where we define a **new probability measure** as  $\mathbb{Q}(A) = \mathbb{E}^{\mathbb{P}}(\frac{U'(b^* \cdot Y)}{\mathbb{E}^{\mathbb{P}}(U'(b^* \cdot Y))}1_A)$  for events  $A \in \mathcal{F}$ , and (as a sanity check) we note that  $\mathbb{Q}(\Omega) = 1$ .

Under the moment condition stated in the question, it turns out that a unique solution  $b^*$  exists if the **noarbitrage** condition is satisfied: if  $\mathbb{P}(b \cdot Y > 0) > 0$  then  $\mathbb{P}(b \cdot Y < 0) > 0$ . In this case, from (2), we see that under  $\mathbb{Q}$ , all contracts are priced according to the market, i.e.  $\mathbb{E}^{\mathbb{Q}}(\pi^i) = p_i$ .  $\mathbb{Q}$  is known as a **risk-neutral measure**. We can solve this maximization problem numerically using e.g. MOSEK convex optimization package in Python (used for summer project).

A common choice is the **exponential utility function**  $U(x) = -e^{-\lambda x}$ , in which case (for  $\lambda = 1$ ) we are computing  $\max_b(-\mathbb{E}(e^{-b\cdot Y})) = \max_b(-\mathbb{E}(e^{b\cdot Y})) = -\min_b\mathbb{E}(e^{b\cdot Y})$ , i.e. minus the **minimum of the mgf** of Y.

For the 1d case d = 1, if  $\pi_1 = f(S)$  with market price p and S has density p(S), then we can re-write expected utility as

$$-\int_0^\infty e^{-\lambda b(f(S)-p)}p(S)dS.$$

We can evaluate this integral explicitly in certain cases, e.g. if  $S \sim \text{Exp}(1)$  and f(S) = S, we see that

$$\mathbb{E}(U(b(S-p))) = \mathbb{E}(-e^{-\lambda b(S-p)}) = \int_0^\infty -e^{-\lambda b(S-p)}e^{-S}dS = -\frac{e^{pb\lambda}}{1+b\lambda}$$

if  $\lambda b + 1 > 0$ , and  $-\infty$  otherwise. Differentiating this expression wrt b and setting the answer to zero, we find that

$$b^* = \frac{1-p}{p\lambda} \tag{3}$$

and the risk-neutral density  $\mathbb{Q}$  corresponding to  $b^*$  is the density of a  $\operatorname{Exp}(\frac{1}{p})$  random variable, under which  $\mathbb{E}^{\mathbb{Q}}(S) = p$  as claimed.

Note the "fair price" of the stock is  $\int_0^\infty Se^{-S}dS = \mathbb{E}(S) = 1$ , so (3) says that we **buy** stock when the stock is **underpriced**, and **sell** when the stock is **overpriced**, and  $|b^*|$  is smaller when  $\lambda$  is larger i.e. when the trader is more risk-averse.

We can also modify the problem to account for interest rates, bid-ask spreads, finite liquidity or a limit order book structure.

7. Compute the **conditional distribution** of  $W_t$  given the known value of  $W_s$ , for s, t > 0 (you do not have to assume that s < t). You may use that for two correlated Normal random variables  $X \sim N(\mu_X, \sigma_X^2)$  and  $Y \sim N(\mu_Y, \sigma_Y^2)$  with  $Corr(X, Y) = \rho$ ,

$$Y|X \sim N(\mu_Y + \rho \frac{\sigma_Y}{\sigma_X}(X - \mu_X), (1 - \rho^2)\sigma_Y^2)$$

and recall that the correlation of two random variables X and Y is defined as  $\operatorname{Corr}(X,Y) = \frac{\mathbb{E}((X-\mu_X)(Y-\mu_Y)}{\sigma_X\sigma_Y}$ .

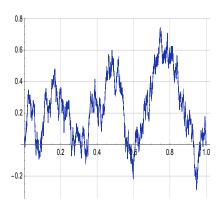


Figure 1: Simulation of a Brownian bridge on [0, 1].

**Solution**. For our case here,  $X = W_s$ ,  $Y = W_t$ ,  $\mu_X = 0$ ,  $\mu_Y = 0$ ,  $\sigma_X = \sqrt{s}$ ,  $\sigma_Y = \sqrt{t}$  and  $\rho = \frac{\min(s,t)}{\sqrt{st}}$ , and recall that we have shown in the lecture notes that  $\mathbb{E}(W_sW_t) = \min(s,t)$ . Thus

$$W_t|W_s \sim N(\rho \frac{\sqrt{t}}{\sqrt{s}} W_s, (1-\rho^2)t)$$
.

**8.** Compute  $\mathbb{E}(W_t^3|W_s=x)$  for  $0 \le s \le t$ .

**Solution**.  $W_t - W_s \sim N(0, t - s)$ , so

$$\mathbb{E}((x+W_t-W_s)^3|W_s=x) = \mathbb{E}(x^3+3x^2(W_t-W_s)+3x(W_t-W_s)^2+(W_t-W_s)^3)|W_s=x)$$

$$= x^3+3x(t-s).$$

We can generalize this computation to compute  $\mathbb{E}(W_t^n|W_s=x)$  for any  $n\in\mathbb{N}$ , since (from the **binomial theorem**) we know that  $(x+W_t-W_s)^n=\sum_{i=0}^n x^{n-i}(W_t-W_s)^i\binom{n}{i}$ , and we also know that all odd moments of  $W_t-W_s$  are zero