

PREAMBLE

1. Helen Xiao, Iñaki Arango, Alex Cai
2. 9 hours (THEORY: 5, COMPUTATIONAL: 4)
3. ChatGPT
4. I have not shared my code with anyone and have not used anyone's code.

PROBLEM 1

$$B_t \sim N(0, t) \Rightarrow \sqrt{2\lambda} B_t \sim \sqrt{2\lambda} N(0, t)$$

PDF: $u(\lambda, t, x) = \frac{1}{2\sqrt{\lambda t}} e^{-\frac{x^2}{4\lambda t}}$

NEED TO SHOW: $\frac{d}{dt} u(\lambda, t, x) = \lambda \frac{\partial^2}{\partial x^2} u(\lambda, t, x)$

$$\begin{aligned} \textcircled{1} \quad \frac{d}{dt} u(\lambda, t, x) &= \frac{1}{2\sqrt{\lambda t}} \left[-\frac{1}{2} t^{-\frac{3}{2}} e^{-\frac{x^2}{4\lambda t}} + t^{-\frac{1}{2}} e^{-\frac{x^2}{4\lambda t}} \left(-\frac{x^2}{4\lambda} (-1)t^{-2} \right) \right] \\ &= \frac{1}{2\sqrt{\lambda t}} e^{-\frac{x^2}{4\lambda t}} \left[-\frac{1}{2t} + \frac{x^2}{4\lambda t^2} \right] \end{aligned}$$

$$\begin{aligned} \textcircled{2} \quad \frac{d}{dx} u(\lambda, t, x) &= \frac{1}{2\sqrt{\lambda t}} e^{-\frac{x^2}{4\lambda t}} \left(-\frac{x}{2\lambda t} \right) \\ \frac{d^2}{dx^2} u(\lambda, t, x) &= \frac{1}{2\sqrt{\lambda t}} \left[-\frac{1}{2\lambda t} e^{-\frac{x^2}{4\lambda t}} + e^{-\frac{x^2}{4\lambda t}} \left(-\frac{x}{2\lambda t} \right) \left(-\frac{x}{2\lambda t} \right) \right] \\ &= \frac{1}{2\sqrt{\lambda t}} e^{-\frac{x^2}{4\lambda t}} \left[-\frac{1}{2\lambda t} + \frac{x^2}{4\lambda^2 t^2} \right] \end{aligned}$$

$$\lambda \frac{\partial^2}{\partial x^2} u(\lambda, t, x) = \frac{1}{2\sqrt{\lambda t}} e^{-\frac{x^2}{4\lambda t}} \left[-\frac{1}{2t} + \frac{x^2}{4\lambda t^2} \right]$$

Since $\frac{d}{dt} u(\lambda, t, x) = \lambda \frac{\partial^2}{\partial x^2} u(\lambda, t, x)$, the PDF of $\sqrt{2\lambda} B_t$ satisfies the diffusion equation.

PROBLEM 2

THEOREM: $\langle B \rangle_T = T$

$$\text{PROOF: } \langle B \rangle_T = \sum_{i=1}^n (B_{t_i} - B_{t_{i-1}})^2$$

$$\begin{aligned} E[\langle B \rangle_T] &= \sum_{i=1}^n E(B_{t_i} - B_{t_{i-1}})^2 \\ &= \sum_{i=1}^n t_i - t_{i-1} \\ &= T \end{aligned}$$

$* B_{t_i} - B_{t_{i-1}} \sim N(0, t_i - t_{i-1})$
 $\frac{\sigma}{\sqrt{t_i - t_{i-1}}} \sim N(0, 1)$

NEED TO SHOW: $\text{Var}(\langle B \rangle_T) \rightarrow 0$

$$\begin{aligned} \text{Var}(\langle B \rangle_T) &= E[(\langle B \rangle_T - E[\langle B \rangle_T])^2] \\ &= E\left[\sum_{i=1}^n [(B_{t_i} - B_{t_{i-1}})^2 - (t_i - t_{i-1})]^2\right] \\ &= \sum_{i=1}^n E[(B_{t_i} - B_{t_{i-1}})^2 - (t_i - t_{i-1})]^2 + \underbrace{2 \sum_{i,j} E[(B_{t_i} - B_{t_{i-1}})^2 - (t_i - t_{i-1})][(B_{t_j} - B_{t_{j-1}})^2 - (t_j - t_{j-1})]}_{\text{blue bracket}} \end{aligned}$$



$$\begin{aligned} &= \sum_{i,j} E[(B_{t_i} - B_{t_{i-1}})^2 (B_{t_j} - B_{t_{j-1}})^2] - E[(B_{t_i} - B_{t_{i-1}})^2 - (t_i - t_{i-1})] - E[(B_{t_j} - B_{t_{j-1}})^2 - (t_j - t_{j-1})] + E[(t_i - t_{i-1})(t_j - t_{j-1})] \\ &= (t_i - t_{i-1})(t_j - t_{j-1}) - (t_i - t_{i-1})(t_i - t_{i-1}) - (t_j - t_{j-1})(t_i - t_{i-1}) + (t_i - t_{i-1})(t_j - t_{j-1}) \\ &= 0 \end{aligned}$$

$$\begin{aligned} &= \sum_{i=1}^n [E[(B_{t_i} - B_{t_{i-1}})^4] - 2E[(B_{t_i} - B_{t_{i-1}})^2](t_i - t_{i-1}) + (t_i - t_{i-1})^2] \\ &= \sum_{i=1}^n [3(t_i - t_{i-1})^2 - 2(t_i - t_{i-1})^2 + (t_i - t_{i-1})^2] \\ &= 2 \sum_{i=1}^n (t_i - t_{i-1})^2 \\ &= 2 \sum_{i=1}^n \frac{1}{n^2} \rightarrow 0 \text{ as } n \rightarrow \infty \quad \checkmark \end{aligned}$$

PROBLEM 3

$$(1) \frac{dp}{dt} + p \left(1 - \frac{p}{K}\right) = \frac{p(K-p)}{K}$$

$$\int \frac{1}{p(K-p)} dp = \int \frac{1}{K} dt$$

$$\frac{1}{K} \int \left(\frac{1}{p} + \frac{1}{K-p} \right) dp = \frac{t}{K} + C$$

$$\ln p - \ln(K-p) = t + C$$

$$\ln\left(\frac{p}{K-p}\right) = t + C$$

$$\frac{p}{K-p} = Ae^t \quad * \quad t=0 \Rightarrow A = \frac{p_0}{K-p_0}$$

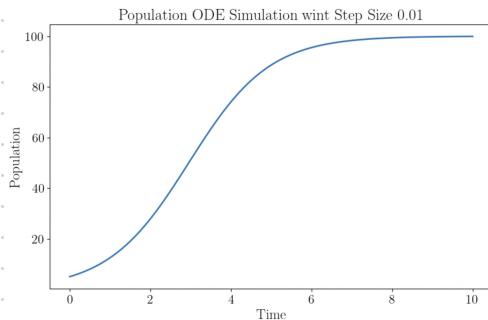
$$p = K Ae^t - p A e^t$$

$$p(t) = \frac{K A e^t}{1 + A e^t} \Rightarrow p(t) = \frac{p_0 K e^t}{(K-p_0) + p_0 e^t}$$

SIMULATION:

$$p(t) = \frac{5(100)e^t}{(100-5)+5e^t} = \frac{500e^t}{95+5e^t}$$

$$\text{SOLUTION FOR } t=1: p(t) = \frac{500e}{95+5e} \approx 12.516$$



```

def dpdt(p, K):
    return p * (1 - p/K)

def euler(p0, K, t_max, step_size):
    t = [0]
    p = [p0]
    for i in range(int(t_max/step_size)):
        t.append(t[i] + step_size)
        p_new = p[i] + step_size * dpdt(p[i], K)
        p.append(p_new)
    return t, p

p0 = 5
K = 100
t_max = 10
step_size = 0.01

t, p = euler(p0, K, t_max, step_size)
t[100], p[100]
✓ 0.0s
(1.0000000000000007, 12.470668118998324)

step_size = 1
t, p = euler(p0, K, t_max, step_size)
t[1], p[1]
✓ 0.0s
(1, 9.75)

```

We can see that larger step sizes are less accurate in approximating solutions.

$$(2) \frac{dx}{dt} = \frac{1}{1-t}$$

$$\int dx = \int \frac{1}{1-t} dt$$

$$x = -\ln|1-t| + C$$

$$x = -\ln|1-t| + C$$

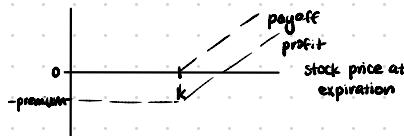
$$\begin{cases} x_0 = 0 \\ 0 = -\ln|1-0| + C \\ C = 0 \end{cases}$$

↳ this blows up in finite time since x becomes undefined when $t=1$

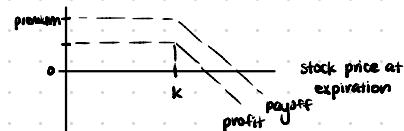
PROBLEM 4

(1)

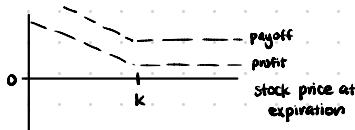
LONG CALL



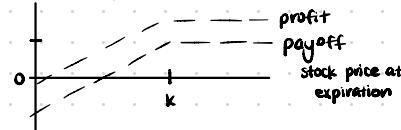
SHORT CALL



LONG PUT



SHORT PUT



$$(2) \text{ PUT-CALL PARITY: } C + k e^{-r(t-t)} = P + S$$

call price + PV of strike price ↓ put price spot price

- ① find price of an asset + prices of call option + put option w/ same strike price + expiration
- ② verify put-call parity identity

$$2.5 + 195 e^{0.05(0.5)} = 4.5 + 150 \rightarrow 161.42 \neq 154.5$$

(3) SPOT PRICE: \$169.4

STRIKE K		
	call option	put option
deep in the money	140	190
out of the money	180	150

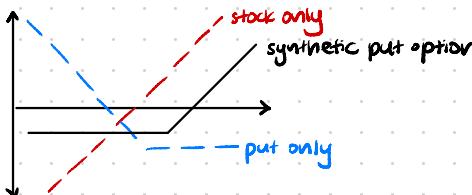
(4) buy a call option at k_1 (in the money): $k_1 < S$

2 call options (at the money): sell at S

buy a call option at k_2 (out of the money): $k_2 > S$

payoff function: $(S_T - k_1) + 2(S - S_T) + 0$

(5) buy stock + at-the-money put option to protect against depreciation in stock price



PROBLEM 5

$$d(f(t, B_t)) = d_t f(t, B_t) dt + d_x f(t, B_t) dB_t + \frac{1}{2} d_{xx}^2 f(t, B_t) dt$$

(1) $X_t = B_t^4 \rightarrow f(x) = x^4$

$$d(B_t^4) = 4B_t^3 dB_t + 6B_t^2 dt$$

(2) $X_t = tB_t^3 \rightarrow f(t, x) = t x^3$

$$d(tB_t^3) = B_t^3 dt + 3t B_t^2 dB_t + 3t B_t dt$$

(3) $X_t = t \cos(B_t) \rightarrow f(t, x) = t \cos(x)$

$$d(t \cos B_t) = \cos(B_t) dt - t \sin B_t dB_t - \frac{t}{2} \cos B_t dt$$

PROBLEM 6

ITÔ'S FORMULA 2.0:

$$d(f(t, B_t)) = d_t f(t, B_t) dt + \underbrace{d_x f(t, B_t) dB_t}_{\text{martingale}} + \frac{1}{2} d_{xx}^2 f(t, B_t) dt$$

* if $d_t f = -\frac{1}{2} d_{xx}^2 f$, $d(f(t, B_t)) = \text{martingale}$

(1) $M_t = e^{(B_t - \frac{t}{2})}$

$$\textcircled{1} \quad d_t M_t = e^{(B_t - \frac{t}{2})} (-\frac{1}{2}) = -\frac{1}{2} e^{(B_t - \frac{t}{2})}$$

$$\textcircled{2} \quad d_{B_t} M_t = d_{B_t B_t} M_t = e^{(B_t - \frac{t}{2})}$$

$$-\frac{1}{2} d_{B_t B_t} M_t = -\frac{1}{2} e^{(B_t - \frac{t}{2})}$$

Since $d_t M_t = -\frac{1}{2} d_{B_t B_t} M_t$, M_t is a martingale.

(2) $E[M_{t+1} | F_t] = E[e^{(B_{t+1} - \frac{t+1}{2})} | F_t]$

$$= E[e^{B_{t+1}} | F_t] e^{-(\frac{t+1}{2})}$$

$$= E[e^{B_{t+1} - B_t + B_t} | F_t] e^{-(\frac{t+1}{2})}$$

$$= E[e^{B_{t+1} - B_t} | F_t] E[e^{B_t} | F_t] e^{-(\frac{t+1}{2})}$$

$$= e^{\frac{1}{2}} e^{B_t} e^{-(\frac{t+1}{2})}$$

$$= e^{B_t - \frac{t}{2}}$$

$$= M_t \checkmark$$

PROBLEM 7

$$m_t = \int_0^t f(\omega_u) dB_u$$

THEOREM:

$$\text{Var}(m_t) = E[m_t^2] = \int_0^t E[f^2(\omega_u)] du$$

$$(1) X_t = \int_0^t u dB_u$$

$$\text{Var}(X_t) = \int_0^t E[u^2] du = \int_0^t u^2 du = \frac{u^3}{3} \Big|_0^t = \boxed{\frac{t^3}{3}}$$

$$(2) X_t = \int_0^t \cos(u) dB_u$$

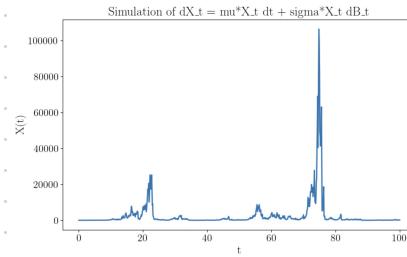
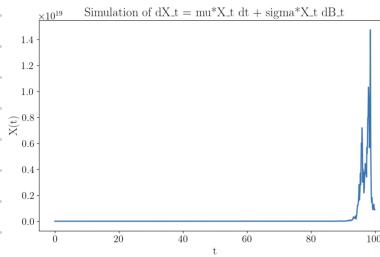
$$\text{Var}(X_t) = \int_0^t E[\cos^2(u)] du = \int_0^t \cos^2(u) du = \frac{u}{2} + \frac{1}{4} \sin(2u) \Big|_0^t = \boxed{\frac{t}{2} + \frac{1}{4} \sin(2t)}$$

$$(3) X_t = \int_0^t B_s^2 dB_s$$

$$\text{Var}(X_t) = \int_0^t E[B_s^4] ds = \int_0^t 3s^2 ds = s^3 \Big|_0^t = \boxed{t^3}$$

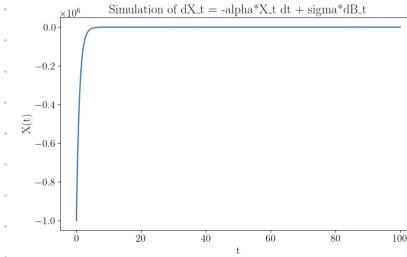
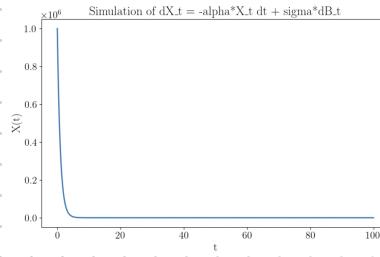
PROBLEM 8

(1)

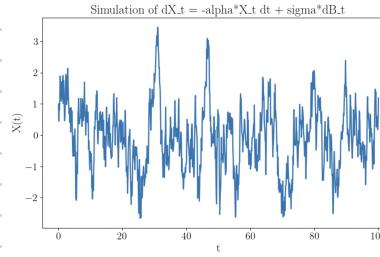


as $t \rightarrow \infty$, $X_t \rightarrow \infty$

(2)



X_t reverts to its long term mean of 0 for all X_0 .



* histogram in jupyter notebook pdf

PROBLEM 9

$$(1) f_x(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$$

$$v(x) = -\frac{x^2}{2}$$

$$SDE: dx_t = -x_t dt + \sqrt{2} dB_t$$

↳ 8.2 !!

$$(2) f_x(x) \propto e^{-\frac{1}{2}(x-3)^2}$$

$$\boxed{\mu=3 \quad \sigma=1}$$

$$V(x) = -\frac{1}{2}(x-3)^2$$

$$V'(x) = -x+3$$

$$SDE: dx_t = -x_t + 3 + \sqrt{2} dB_t$$

* simulation in jupyter notebook pdf

$$(3) f_x(x) \propto \frac{1}{2}e^{-\frac{1}{2}(x-3)^2} + \frac{1}{2}e^{-\frac{1}{2}(x+3)^2}$$

* simulation in jupyter notebook pdf