

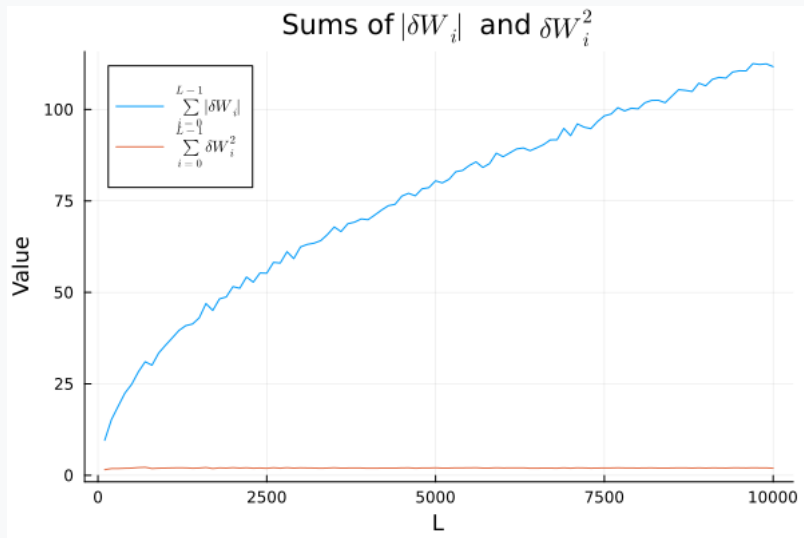
# **MATH 512 - Project 4**

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# Question 1 Overview

- Let  $W_t$  be a standard Wiener process, with drift parameter zero and variance parameter  $\sigma^2 = 1$ .
- We divide the interval  $[0, 2]$  into  $L$  subintervals  $[t_i, t_{i+1}]$ , where  $t_i = i\delta t$  and  $\delta t = 2/L$ .
- Let  $W_i = W(t_i)$  and  $\delta W_i = W_{i+1} - W_i$ .
- We verify numerically that:
  - $\sum_{i=0}^{L-1} |\delta W_i|$  is unbounded as  $\delta t$  goes to zero.
  - $\sum_{i=0}^{L-1} \delta W_i^2$  converges to 2 in probability as  $\delta t$  goes to zero.

# Question 1 Response



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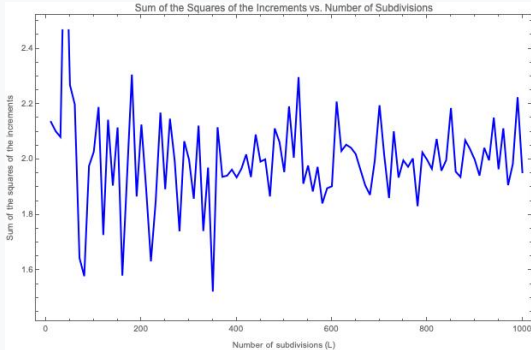


Figure:  $\delta W_i^2$

## Question 2a

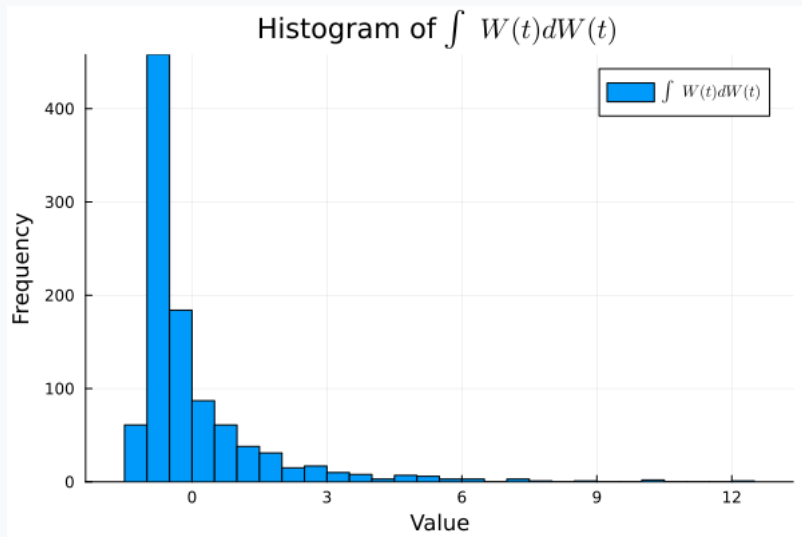


Figure: 2a

## Question 2b

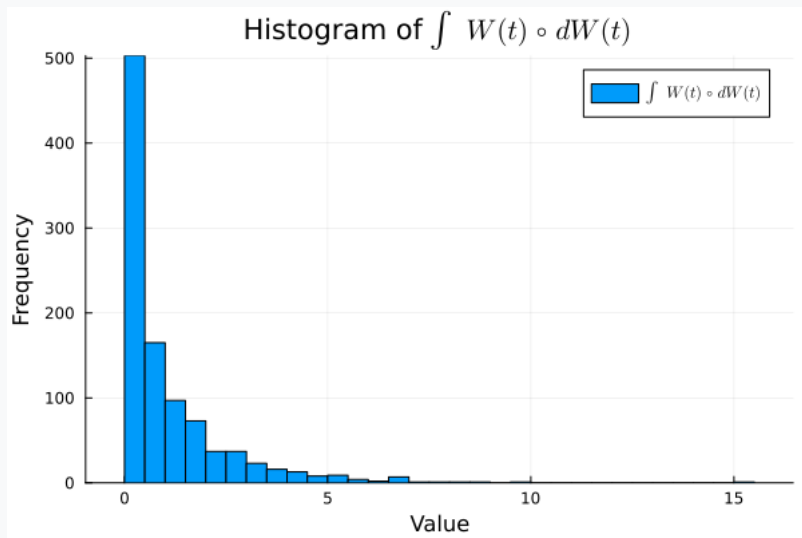


Figure: 2b

## Question 2c

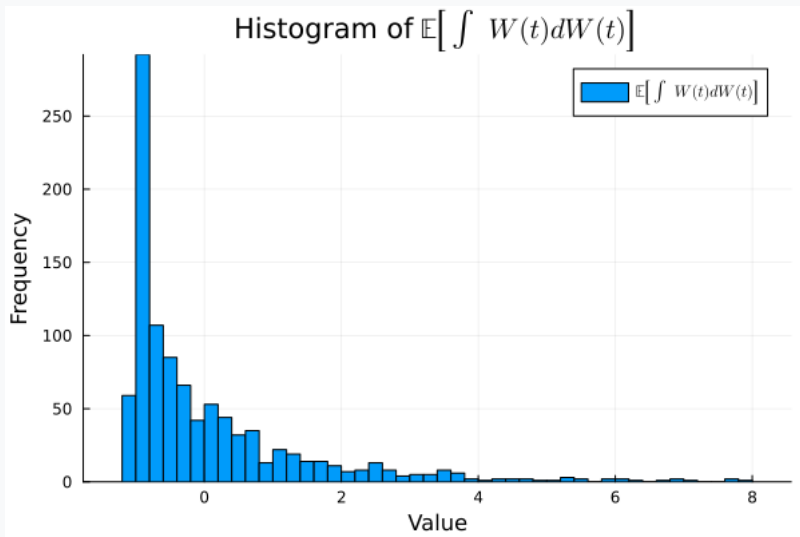


Figure: 2c

## Question 2d

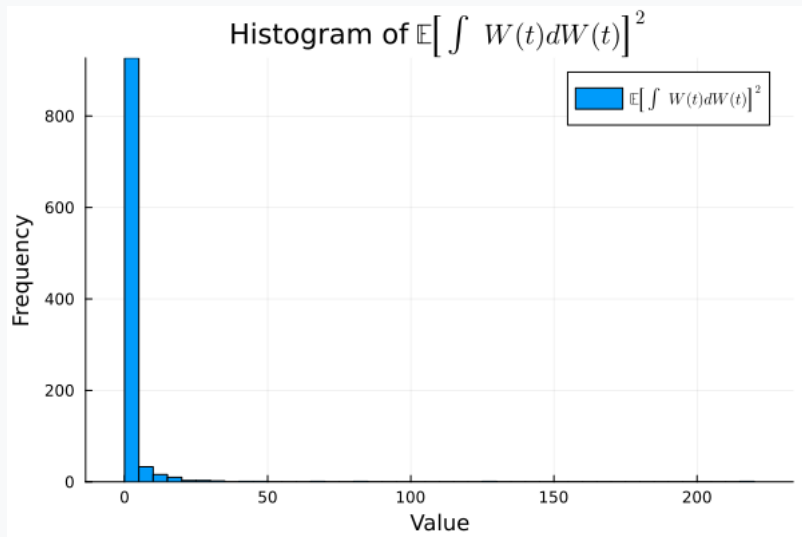


Figure: 2d



## Question 2e

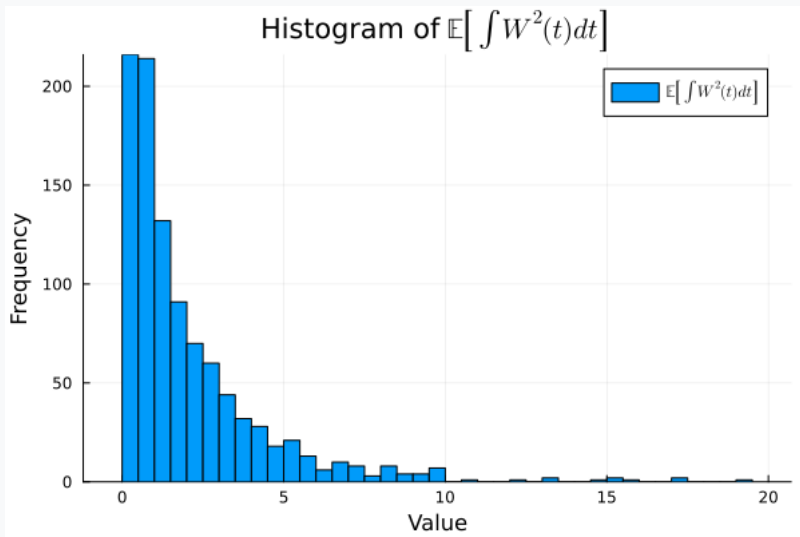


Figure: 2e

## Question 2f

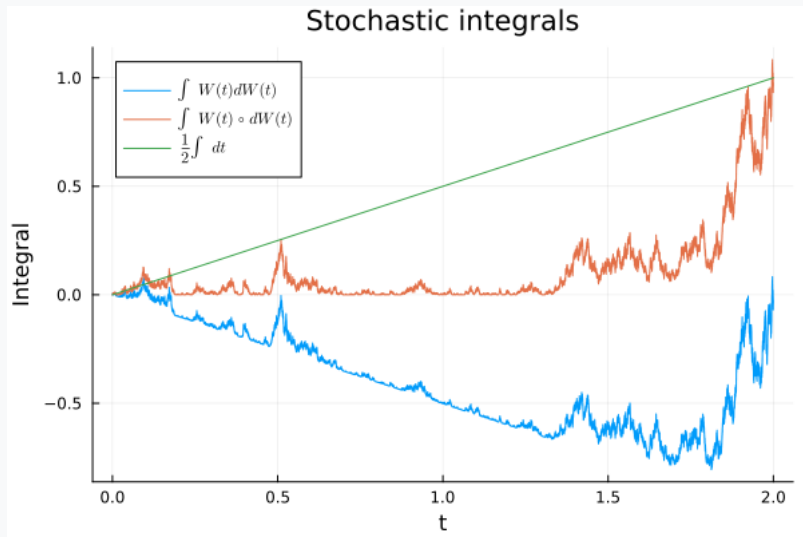


Figure: 2f

# Question 3: Weak and Strong Order of Convergence for EM Method

**SDE:**

$$dX(t) = \mu X(t)dt + \sigma X(t)dW(t), \quad X(0) = 3, \quad \mu = 2, \quad \sigma = 0.10$$

- **Weak order of convergence equal to 1:** to show that

$$|E[X_1] - E[X(1)]| = C\Delta t$$

$$|X_0 e^{\mu*1} - \frac{1}{numPaths} \sum_{j=1}^{numPaths} X_j(1)| = C\Delta t$$

- **Strong order of convergence equal to 0.5:** to show that

$$E|X_1 - X(1)| = C\Delta t^{0.5}$$

$$\frac{1}{numPaths} \sum_{j=1}^{numPaths} |X_0 e^{\mu*1 + 0.5*\sigma^2*1} - X_j(1)| = C\Delta t^{0.5}$$

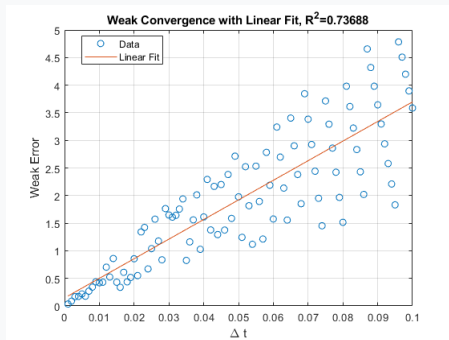
where  $X_1$  is the exact solution and  $X(1)$  is the estimated solution.

# Weak Order of Convergence Results

**Weak Order of Convergence** focuses on the **expected values** of the numerical solution compared to the exact solution

$$|E[X_1] - E[X(1)]| = C\Delta t$$

- A plot of the error versus  $\Delta t$  for the weak order of convergence of 1.

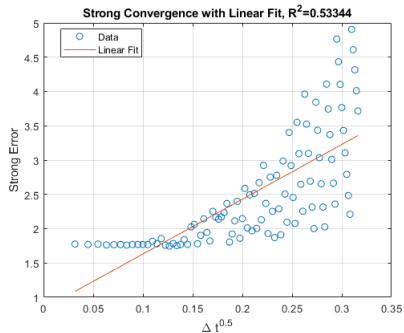


# Strong Order of Convergence Results

**Strong Order of Convergence to 0.5** is concerned with the **pathwise accuracy**, evaluating how closely the numerical solution follows individual realizations of the exact solution.

$$E|X_1 - X(1)| = C\Delta t^{0.5}$$

- A plot of the error versus  $\Delta t^{0.5}$  for the strong order of convergence of 0.5.



## Question 4

We consider the following SDE:

$$dX(t) = \mu X(t)dt + \sigma X(t)dW(t), \quad X(0) = 3, \quad \mu = 2, \quad \sigma = 0.10$$

- Simulate this stochastic process over the interval  $[0, 20]$  using an implicit method of the form:

$$X_{n+1} = X_n + (1 - \theta)\Delta t f(X_n) + \theta\Delta t f(X_{n+1}) + \sqrt{\Delta t}\alpha_n g(X_n)$$

- We use  $\theta = 0.5$  and  $\alpha_n = \sigma X_n$ .
- We plot the results for  $\Delta t = 0.01$  and  $\Delta t = 0.001$ .

## Question 4 Continued

We plot the explicit and implicit methods for  $\Delta t = 0.01$  and  $\Delta t = 0.001$ :

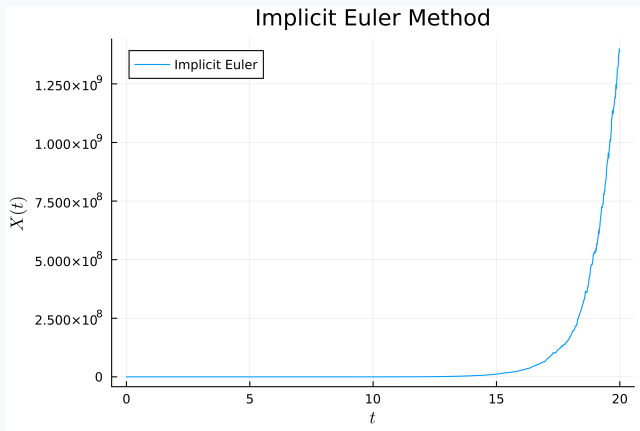


Figure: Implicit Euler Method

# Question 4 Continued

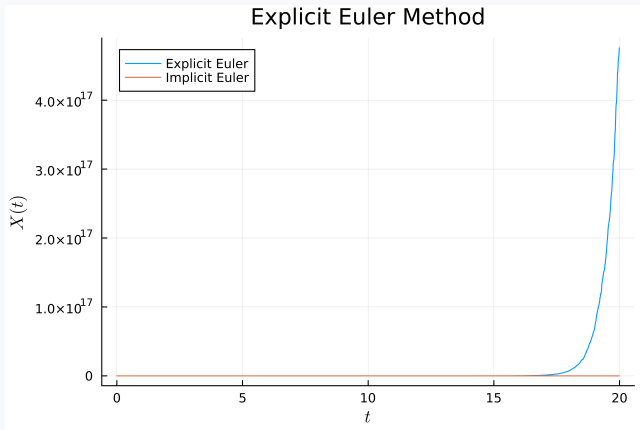


Figure: Explicit Euler Method



## Question 4 Continued

We determine the values of  $\mu$  and  $\sigma$  for which the SDE is mean-square stable:

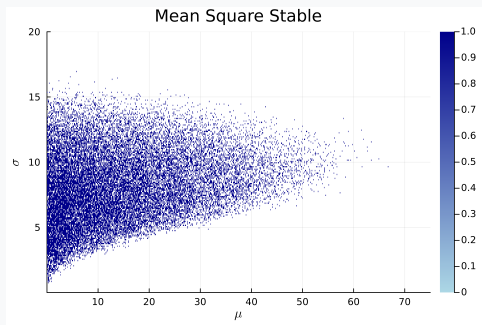


Figure: Mean-Square Stability

## Question 4 Continued

We theoretically calculate the values of  $\mu$  and  $\sigma$  for which the SDE is mean-square stable:

$$\begin{aligned}dX(t) &= \mu X(t)dt + \sigma X(t)dW(t) \\ &= \mu X(t)dt + \sigma X(t)\sqrt{dt}Z(t)\end{aligned}$$

where  $Z(t) \sim N(0, 1)$ .

We find that the values of  $\mu$  and  $\sigma$  for which the SDE is mean-square stable are calculated as:

$$\begin{aligned}\mathbb{E}[(dX(t))^2] &= \mathbb{E}[(\mu X(t)dt + \sigma X(t)\sqrt{dt}Z(t))^2] \\ &= \mu^2 X(t)^2 dt^2 + \sigma^2 X(t)^2 dt\end{aligned}$$

which implies that  $\mu^2 dt^2 + \sigma^2 dt \leq 0$ .

## Question 4 Continued

We also find that the asymptotic stability of the SDE is determined by the drift term  $\mu$  and the diffusion term  $\sigma$ .

We find that the SDE is mean-square stable in the range calculated:

$$\lim_{n \rightarrow \infty} |Y_n| = 0, \text{ with probability one, for any } X_0$$

and the region is similar to the region of mean-square stability.

## Question 5

An algorithm to simulate exit times for the SDE:

- 1: Choose a step size  $\Delta t$
- 2: Choose some paths,  $M$
- 3: **for**  $s = 1$  to  $M$  **do**
- 4:     Set  $t_n = 0$  and  $X_n = X_0$
- 5:     **while**  $X_n > a$  and  $X_n < b$  **do**
- 6:         Compute a  $N(0, 1)$  sample  $\xi_n$
- 7:         Replace  $X_n$  by  $X_n + \mu X_n \Delta t + \sigma X_n \xi_n$
- 8:         Replace  $t_n$  by  $t_n + \Delta t$
- 9:     **end while**
- 10:    Set  $T_s^{exit} = t_n - 1/2\Delta t$
- 11: **end for**
- 12: Set  $a_M = \frac{1}{M} \sum_{s=1}^M T_s^{exit}$
- 13: Set  $b_M^2 = \frac{1}{M-1} \sum_{s=1}^M (T_s^{exit} - a_M)^2$

## Question 5 Continued

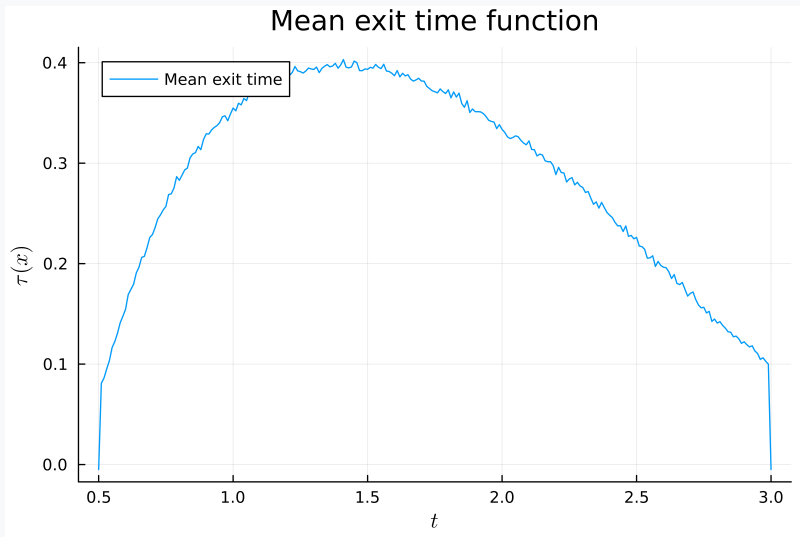


Figure: 5

## Question 5 Continued

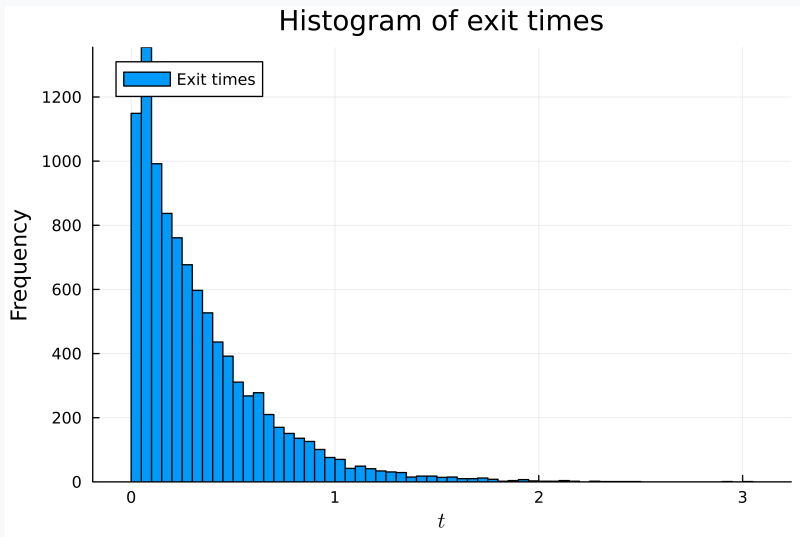


Figure: 5

Questions?