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In [8]: import numpy as np
import matplotlib.pyplot as plt
from scipy.stats import norm
from scipy import stats
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In [3]: # 1. American Option Pricing using Binomial Tree Model
def american_option_price(S, K, T, r, sigma, N, option_type='call'):
    dt = T / N
    u = np.exp(sigma * np.sqrt(dt))
    d = 1 / u
    p = (np.exp(r * dt) - d) / (u - d)

    # Initialize asset prices at maturity
    ST = np.array([S * u**j * d**(N-j) for j in range(N+1)])

    # Initialize option values at maturity
    if option_type == 'call':
        C = np.maximum(ST - K, 0)
    else: # put
        C = np.maximum(K - ST, 0)

    # Iterate backwards
    for i in range(N-1, -1, -1):
        ST = ST[:-1] / u # Asset prices one step earlier
        C_hold = np.exp(-r * dt) * (p * C[1:] + (1-p) * C[:-1])
        if option_type == 'call':
            C = np.maximum(C_hold, ST - K)
        else: # put
            C = np.maximum(C_hold, K - ST)

    return C[0]
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In [4]: # 2. Analysis of American Option Prices
S0 = 100 # Current stock price
K = 100  # Strike price
T = 1    # Time to expiration (in years)
r = 0.05 # Risk-free rate
sigma = 0.2 # Volatility

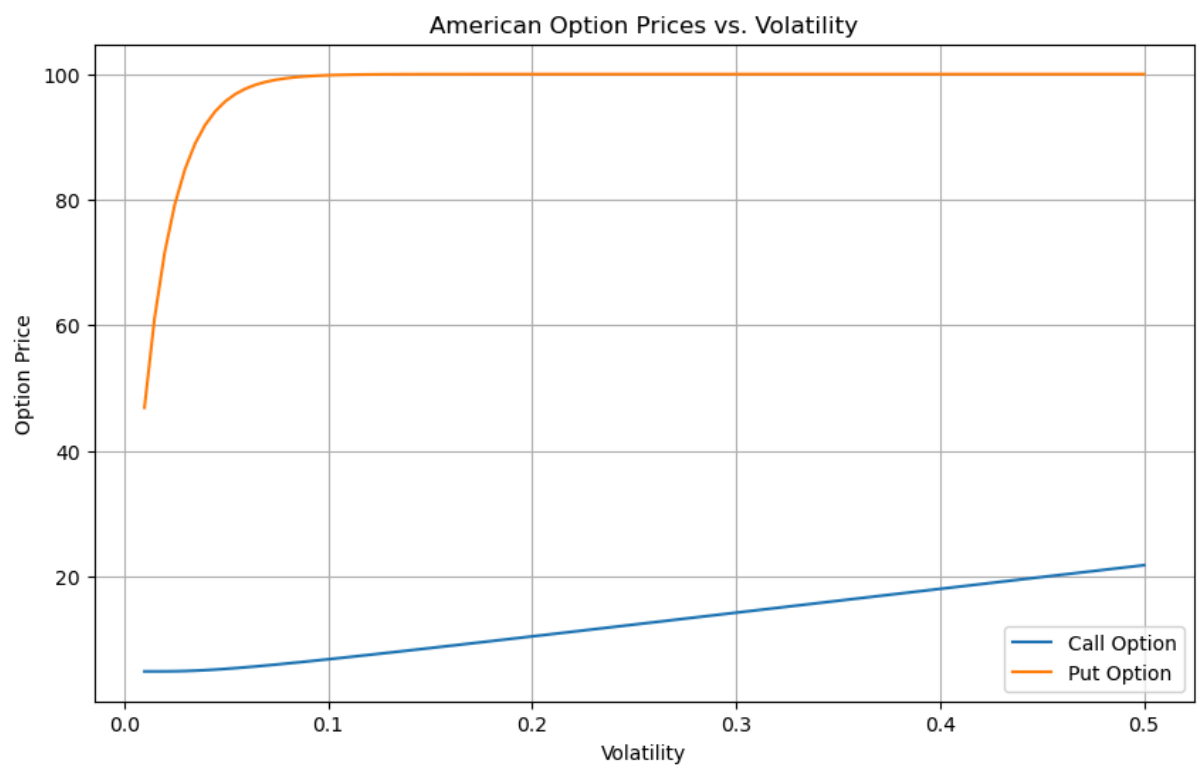
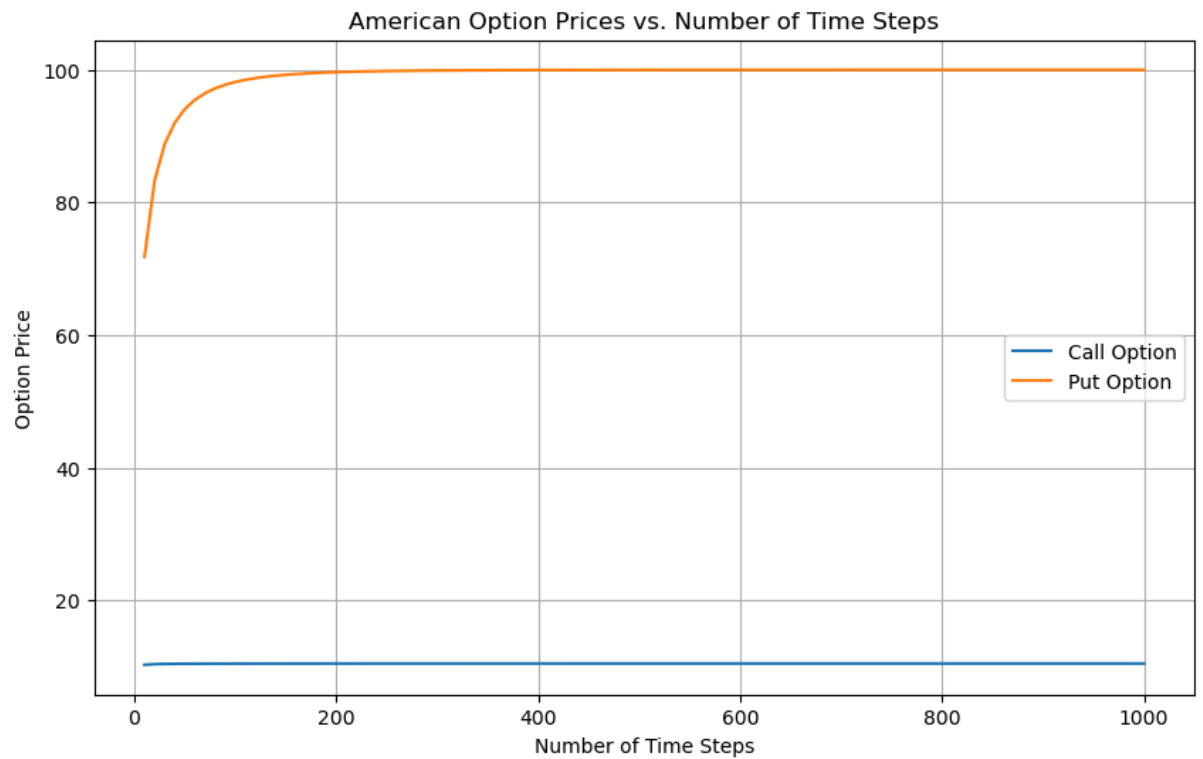
# Analyze price with respect to number of time steps
N_values = range(10, 1001, 10)
call_prices = [american_option_price(S0, K, T, r, sigma, N) for N in N_values]
put_prices = [american_option_price(S0, K, T, r, sigma, N, 'put') for N in N_values]

plt.figure(figsize=(10, 6))
plt.plot(N_values, call_prices, label='Call Option')
plt.plot(N_values, put_prices, label='Put Option')
plt.xlabel('Number of Time Steps')
plt.ylabel('Option Price')
plt.title('American Option Prices vs. Number of Time Steps')
plt.legend()
plt.grid(True)
plt.show()

# Analyze price with respect to volatility
sigma_values = np.linspace(0.01, 0.5, 100)
call_prices_sigma = [american_option_price(S0, K, T, r, sigma, 1000) for sigma in sigma_values]
put_prices_sigma = [american_option_price(S0, K, T, r, sigma, 1000, 'put') for sigma in sigma_values]

plt.figure(figsize=(10, 6))
plt.plot(sigma_values, call_prices_sigma, label='Call Option')
plt.plot(sigma_values, put_prices_sigma, label='Put Option')
plt.xlabel('Volatility')
plt.ylabel('Option Price')
plt.title('American Option Prices vs. Volatility')
plt.legend()
plt.grid(True)
plt.show()

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In [5]: # 3. Ornstein-Uhlenbeck Process
def simulate_ou_process(n, N, Delta, theta, sigma, R0, mu):
    times = np.arange(n+1) * Delta
    X = np.zeros((N, n+1))
    for i in range(1, n+1):
        x = X[:, i-1]
        m = x * np.exp(-theta * Delta)
        v = (sigma**2 / (2*theta)) * (1 - np.exp(-2*theta * Delta))
        X[:, i] = np.random.normal(m, np.sqrt(v), N)

    Rt = np.exp(-theta * times) * R0 + (1 - np.exp(-theta * times)) * mu
    return times, Rt

# Simulate OU process
n = 10000
N = 10
Delta = 10 / n
theta = 0.5
sigma = 0.02
R0 = 0.1
mu = 0.05

times, Rt = simulate_ou_process(n, N, Delta, theta, sigma, R0, mu)

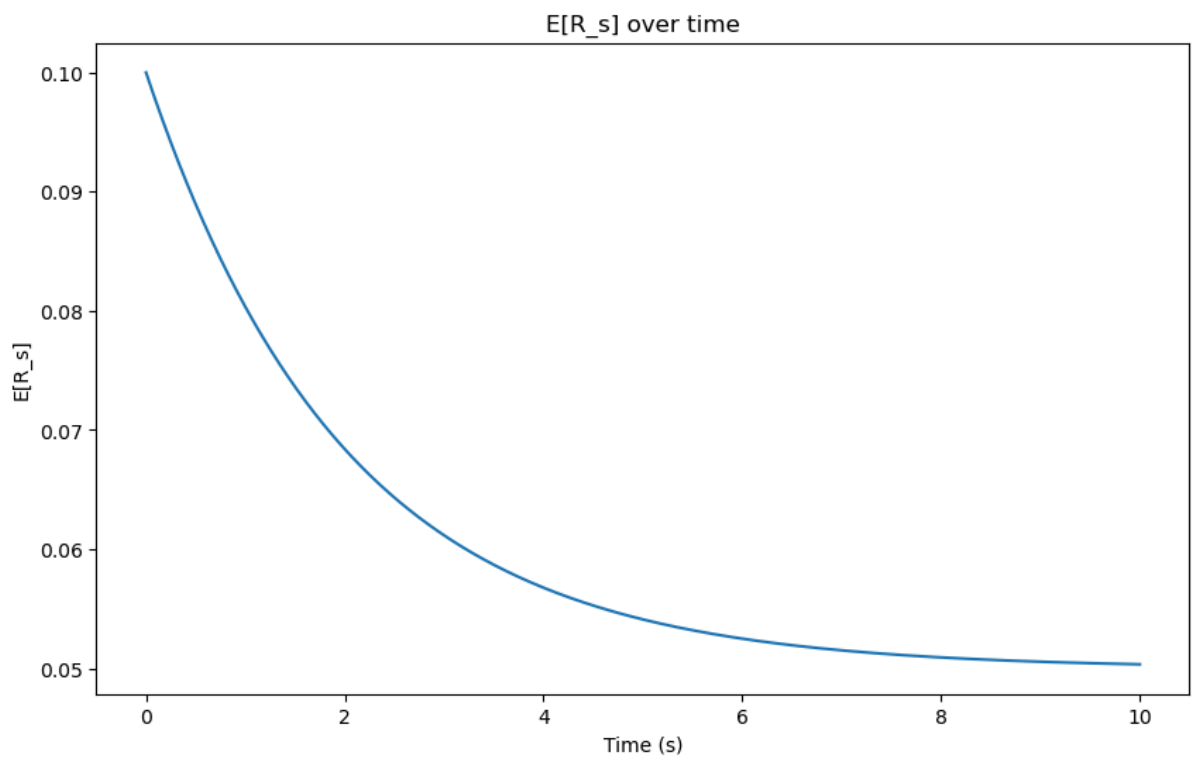
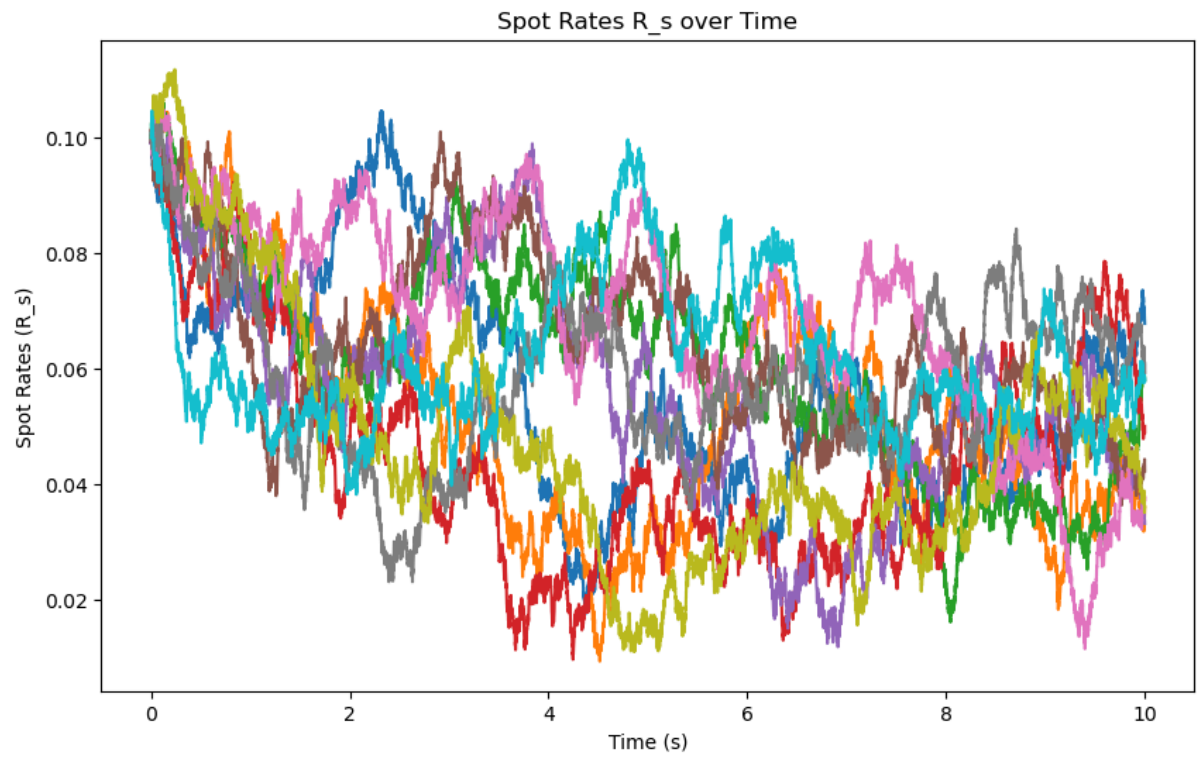
plt.figure(figsize=(10, 6))
for i in range(N):
    plt.plot(times, Rt[i, :])
plt.xlabel('Time (s)')
plt.ylabel('Spot Rates (R_s)')
plt.title('Spot Rates R_s over Time')
plt.show()

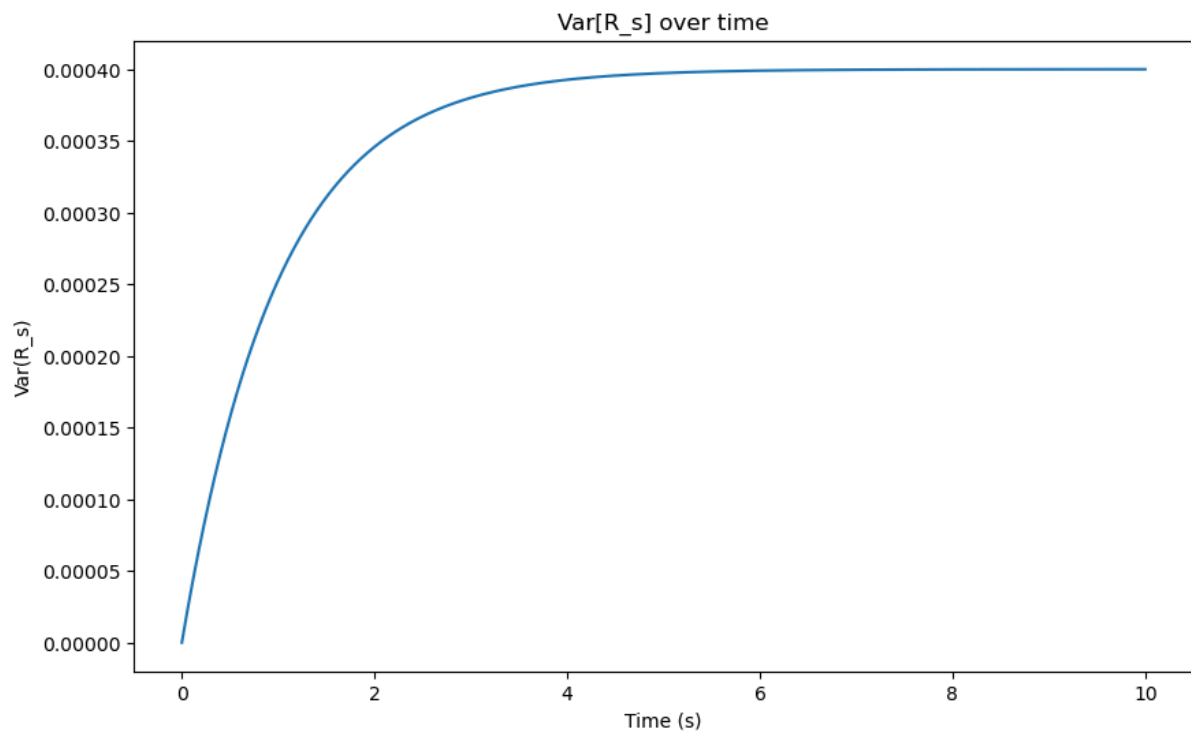
# Plot expectation and variance
expRt = np.exp(-theta * times) * R0 + (1 - np.exp(-theta * times)) * mu
varRt = (sigma**2 / (2*theta)) * (1 - np.exp(-2*theta * times))

plt.figure(figsize=(10, 6))
plt.plot(times, expRt)
plt.xlabel('Time (s)')
plt.ylabel('E[R_s]')
plt.title('E[R_s] over time')
plt.show()

plt.figure(figsize=(10, 6))
plt.plot(times, varRt)
plt.xlabel('Time (s)')
plt.ylabel('Var(R_s)')
plt.title('Var[R_s] over time')
plt.show()

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In [10]: # 4. Vasicek Model
def simulate_vasicek(n, N, Delta, theta, sigma, R0, mu):
    times, Rt = simulate_ou_process(n, N, Delta, theta, sigma, R0, mu)
    Q = np.exp(-Delta * np.cumsum(Rt, axis=1))
    return times, Q

times, Q = simulate_vasicek(n, N, Delta, theta, sigma, R0, mu)

plt.figure(figsize=(10, 6))
for i in range(N):
    plt.plot(times, Q[i, :])
plt.xlabel('Time (t)')
plt.ylabel('Bond price (Qt)')
plt.title('Bond price Qt over time t')
plt.show()

# Distribution of Qt
n = 1000
N = 1000
Delta = 10 / n
times, Q = simulate_vasicek(n, N, Delta, theta, sigma, R0, mu)

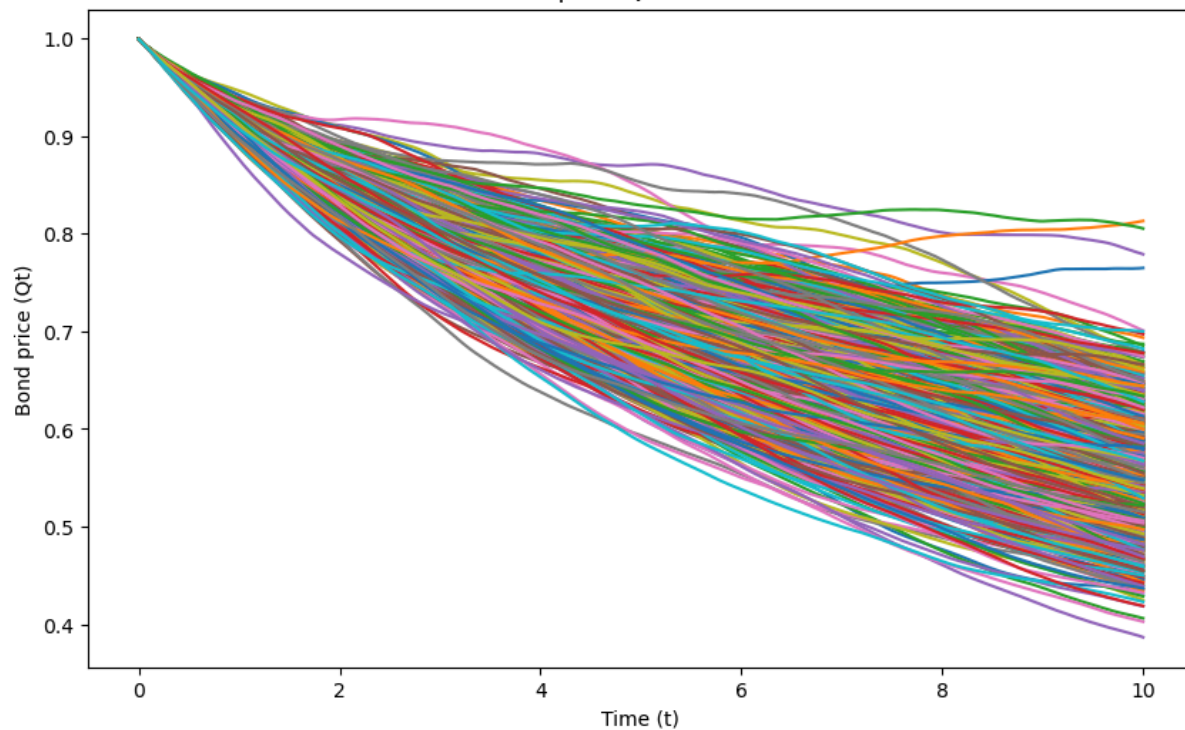
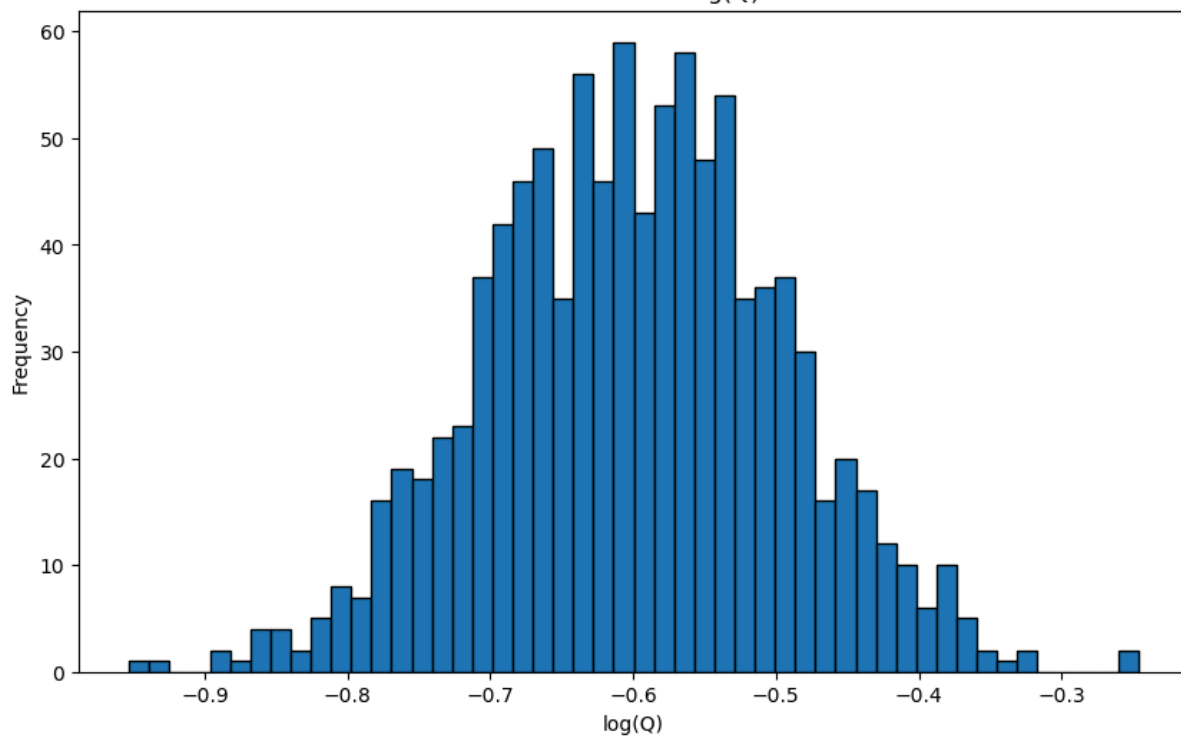
logQ = np.log(Q[:, -1])

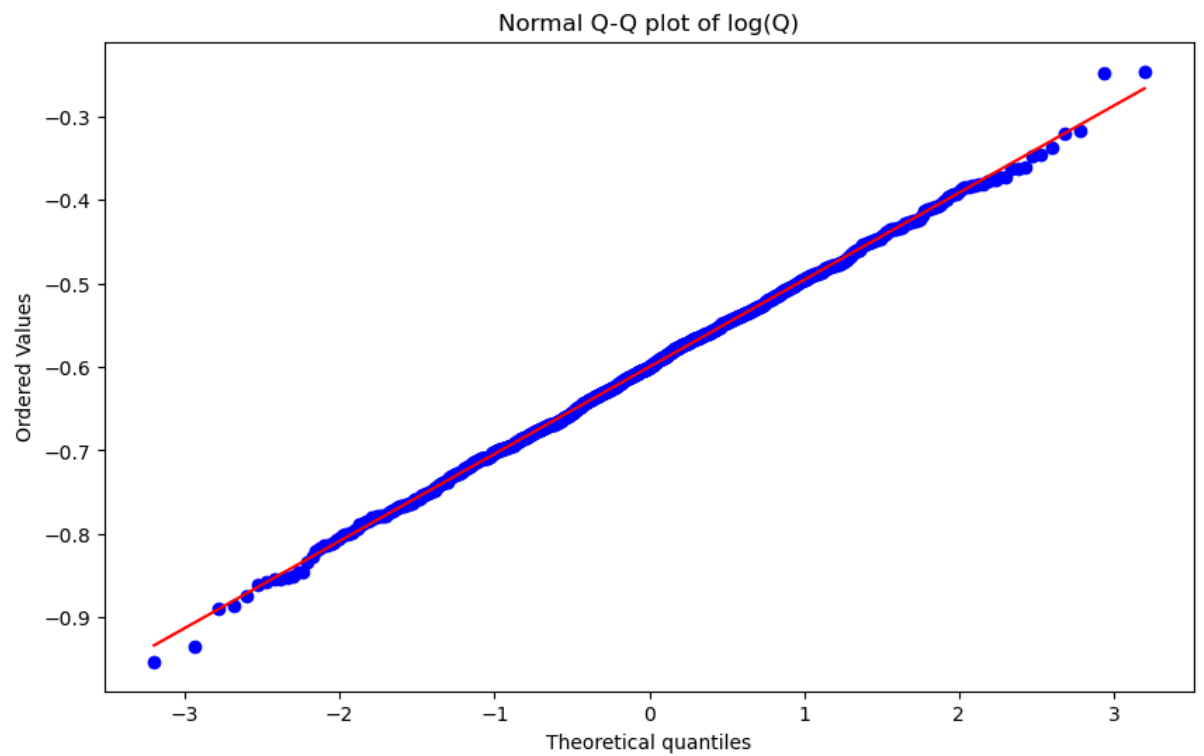
plt.figure(figsize=(10, 6))
plt.hist(logQ, bins=50, edgecolor='black')
plt.xlabel('log(Q)')
plt.ylabel('Frequency')
plt.title('Distribution of log(Q)')
plt.show()

plt.figure(figsize=(10, 6))
stats.probplot(logQ, dist="norm", plot=plt)
plt.title('Normal Q-Q plot of log(Q)')
plt.show()
plt.close()

print(f"Mean of log(Q): {np.mean(logQ)}")
print(f"Variance of log(Q): {np.var(logQ)}")

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Bond price Q_t over time t Distribution of $\log(Q)$ 



Mean of $\log(Q)$: -0.6000891301263563

Variance of $\log(Q)$: 0.010807831141708977

In []: