

# **Mandatory Bachelor Project**

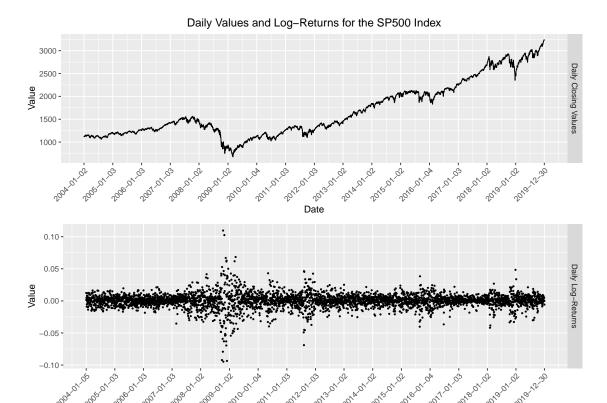
Math-Econ Programme

Financial Modeling with Continuous-Time Models

Theory and Applications

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**Figure 1:** The top plot shows daily closing values for the SP500 index between 2004/01/02-2019/12/30. Corresponding log-returns are represented by the black points in the lower plot. Source: <a href="https://finance.yahoo.com/quote/%5EGSPC/">https://finance.yahoo.com/quote/%5EGSPC/</a>.

# 1 Introduction

The financial market consists of a risky asset (a stock) and a risk-free asset (e.g., zero-coupon bond), both traded continuously up to some fixed time horizon T. As usual, the filtered probability space  $(\Omega, \mathcal{F}, (\mathcal{F}_t^W)_{0 \le t \le T}, P)$  is introduced to capture the information flow:  $\Omega$  contains all possible states of the market,  $\mathcal{F}$  is the corresponding  $\sigma$ -algebra, P is the physical measure, and  $(\mathcal{F}_t^W)$  is the  $\sigma$ -field generated by the Brownian motion  $(W_t)_{0 \le t \le T}$ , i.e.,

$$\mathcal{F}_t^W = \sigma \left\{ W_s \mid 0 \le s \le t \right\}. \tag{1}$$

In the remainder of this project,  $S_t$  denotes the value of the risky asset at time t, and the process is assumed to solve the stochastic differential equation (SDE)

$$dS_t = \mu(t, S_t)dt + \sigma(t, S_t)dW_t \tag{2}$$

$$S_0 = s > 0 \tag{3}$$

where  $\mu(\cdot,\cdot)$  and  $\sigma(\cdot,\cdot)$  are deterministic functions possibly depending on both time (t) and the state of the underlying  $(S_t)$ . Moreover,  $\sigma(t,s)>0$  for all  $(t,s)\in[0,T]\times\mathbb{R}^+$ , and both  $\mu(\cdot,\cdot)$  and  $\sigma(\cdot,\cdot)$  are sufficiently well-behaved such that a solution to the above SDE exists. We will use following terminology:  $\mu(\cdot,\cdot)$  and  $\sigma(\cdot,\cdot)$  are called the drift and volatility, respectively, of  $(S_t)$ , whereas  $dS_t$  is referred to as the dynamics of the process.

The risk-free asset's value through time is given by the process  $(B_t)_{0 \le t \le T}$  with dynamics

$$dB_t = rB_t dt (4)$$

$$B_0 = 1 \tag{5}$$

where r is constant. The risk-free asset is interpreted as the money account with short rate of interest r.

# **Questions: Model Assumptions**

The economic interpretation of the above setting is crucial for working with continuous-time models in practice. Answering the questions below will provide some intuition behind the model assumptions. I would recommend you to read chapters 2 and 4 in Björk (2009).

#### 1.1

Discuss the term *risk*. What is the difference between a risk-free and a risky asset? What would a locally risk-free asset be? How many sources of risk appear in this model setting? Is  $\mathcal{F}_t^W = \mathcal{F}_t^S$ ? What is the financial interpretation of the  $\mathcal{F}_t^S$ .

#### 1.2

Explain the economic interpretation of  $D_t := B_t/B_T$ . Is  $S_0 > 0$  a strong assumption? Illustrate  $B_t$ ,  $D_t$ ,  $S_t$  and  $D_tS_T$ . You can use the following pseudo-code for generating  $(S_t)$  (for fun: adjust n as well as the seed):

```
# Parameters

S <-  # Starting value of process

mu <- 0.05  # drift
sigma <- 0.2  # volatility

# Time grid
T <- 1  # End time
n <-  # Number of evaluations
dt <- T/(n)  # Equidistant time step

S_vec <- numeric(n)

for (i in 1:n){

Z <-  # Generate value from normal distribution with mean=0 and var=1
```

## 1.3

Is it reasonable to assume that stock prices are represented by continuous processes? Discuss the (dis)advantage of using continuous-time models for representing stock prices (e.g., discuss the trade off between numerical tractability and modeling reality/financial markets).

# 2 Geometric Brownian Motion

Assume that

$$S_t = S_0 e^{\left(\mu - \sigma^2/2\right)t + \sigma W_t} \tag{6}$$

where  $S_0 = s > 0$  is the (time-0) starting value of the process. The process in equation (6) is called a Geometric Brownian Motion (GBM) and is closely linked to equations (2)-(3), although we can't prove that until we get Ito's formula in our toolbox (next topic). The Geometric Brownian Motion is defined and discussed by Björk (2009) in chapter 5.

# **Questions: Geometric Brownian Motion**

# 2.1

Is  $(S_t)$  continuous? Illustrate  $(S_t)$  using pseudo-code from the box. Also compare this simulation technique with Euler's method and build-in function from sde-package respectively.

```
for (i in 1:n){
    Z <- # Generate value from normal dist. with mean=0 and var=1
    X <- # log(S_i/S_{i-1})
    S <- # Update S
    S_vec[i] <- S
}

# Or without the for-loop....
Z <- # Generate n values from multi dim. normal dist.
X <- cumsum(...) # Argument should be a n-dim vector
S <- # Generate GBM
```

## 2.2

Compute  $E(S_t)$  and  $V(S_t)$ . What is the distribution of  $S_t$ ? Also compute  $E(S_t | \mathcal{F}_s^S)$  for s < t (hint: realize that  $S_t = S_s \times S_t/S_s$ ). Is  $(S_t)$  a martingale under P?

#### 2.3

Let  $R_t := (S_t - S_{t-1})/S_t$  denote the *return* over one period. Determine the distribution of the associated log-return  $r_t := \log(S_t/S_{t-1})$ . How is  $R_t$  linked with  $r_t$ ? Explain the convenience of using log-returns when working with financial time-series.

#### 2.4

Would you prefer holding the risk-free asset or the risky-asset?

## 2.5

Produce figure 1. Are data a realization of a Geometric Brownian Motion (i.e., discuss whether model assumptions for GBM are violated)?

# 2.6

Derive the maximum likelihood estimator of  $\mu$  and  $\sigma$ . Estimate  $\mu$  and  $\sigma$  for the sample used to generate figure 1. Finally, discuss a significant shortfall for the maximum likelihood estimator of the drift  $\mu$  (e.g., you may support your arguments with data).

# 3 Itŏ Calculus

Itŏ calculus and in particular Itŏ's formula play a crucial role in mathematical finance. The following results are from chapter 4 in Björk (2009) (p. 49-62, read it!).

**Theorem 3.1 (Ito's formula)** Assume that the process  $X = (X_t)_{t \ge 0}$  has dynamics given by

$$dX_x = \mu_t dt + \sigma_t dW_t, \tag{7}$$

where  $\mu_t$  and  $\sigma_t$  are adapted processes, and let  $f: \mathbb{R} \times \mathbb{R} \to \mathbb{R}$  be  $C^{1,2}$ -function. Introduce a new process by  $Z_t = f(t, X_t)$ . Then  $Z = (Z_t)_{t \geq 0}$  has dynamics given by

$$dZ_{t} = df(t, X_{t}) = \left(f_{t}(t, X_{t}) + \mu_{t} f_{x}(t, X_{t}) + \frac{1}{2} \sigma_{t}^{2} f_{xx}(t, X_{t})\right) dt + \sigma_{t} f_{x}(t, X_{t}) dW_{t}, \quad (8)$$

where  $f_t$ ,  $f_x$  and  $f_{xx}$  denote the derivative with respect to time, the derivative with respective to space and the second-derivative with respective to space, respectively.

Another formulation of Ito's formula is stated in the following proposition — you may find this version useful!

**Proposition 3.2** With the assumptions as in 3.1,  $dZ_t$  is given by

$$dZ_t = f_t(t, X_t)dt + f_x(t, X_t)dX_t + \frac{1}{2}f_{xx}(t, X_t)(dX_t)^2,$$
(9)

where

$$(dt)^2 = 0,$$
  $dt \cdot dW_t = 0$  and  $(dW_t)^2 = dt.$ 

Finally, the multidimensional version of Ito's formula is given in definition 4.16 in Björk (2009).

# **Questions: Ito Calculus**

Itŏ

#### 3.1

State a condition such that the process  $(X_t)_{t\geq 0}$  with dynamics given by equation (7) becomes a P-martingale. What do we call X when  $\mu_t \geq 0$  P-almost surely? Finally, derive/state the condition such that  $Z_t = f(t, X_t)$  becomes a P-martingale.

#### 3.2

Compute the dynamics of  $(S_t)_{t\geq 0}$  where  $S_t$  is given by equation (6). Moreover, apply Ito's formula to compute  $E(S_t)$ .

From now on, we will specify  $(S_t)_{t\geq 0}$  by its stochastic differential and not equation (6).

#### 3.3

Compute the dynamics of  $(Z_t)_{t\geq 0}$  where  $Z_t = S_t/B_t$ . Hint: Use the multidimensional Ito formula from Björk (2009). Is the process a martingale?

# 4 Black-Scholes Model and Q-dynamics

The Black-Scholes model consists of two assets with dynamics given by

$$dB_t = rB_t dt, (10)$$

$$dS_t = \mu S_t dt + \sigma S_t dW_t, \tag{11}$$

where  $B_0=1$  and  $S_0=s>0$ , and  $r,\mu$  and  $\sigma>0$  are constants. This famous model was (and to some degree still is) used for pricing T-claims. Following definition 7.1 in Björk (2009), the stochastic variable  $\mathcal X$  is called a T-claim if it's  $\mathcal F_T^S$ -measurable. Here T is the date of maturity (or exercise date). A simple T-claim is on the form  $\mathcal X=\Phi(S_T)$ , where  $\Phi(\cdot)$  is the contract function.

The focal point is to price  $\mathcal X$  at any point in time t < T, i.e. determine the pricing function  $\pi(t;\mathcal X)$  for t < T with terminal condition  $\pi(T;\mathcal X) = \mathcal X$ . The Black-Scholes model is free of arbitrage as well as complete (we will discuss both concepts next time), and we will also assume that

$$\pi(t; \mathcal{X}) = F(t, S_t),$$

where F is a  $C^{1,2}$ -function.

Pricing in this type of models is conducted under a *risk-neutral* probability measure (also called equivalent martingale measure (EMM), see definition ) Q. In the Black-Scholes model, the pricing measure Q is unique — this follows directly from no-arbitrage and completeness of the model. Moreover, the discounted process  $(S_t/B_t)$  is a martingale under this measure. Theorem 10.5 and theorem 10.9 yield in Björk (2009) yield the existence of Q, whereas theorem 10.17 implies uniqueness of the pricing measure. Definition 10.11 and theorem 10.14 tell us that  $(S_t/B_t)$  is a martingale under this pricing measure.

In this section, we will show that the process the discounted process  $(S_t/B_t)$  has Q-dynamics given by

$$dS_t = rS_t dt + \sigma S_t dW_t^Q, \tag{12}$$

where  $W_t^Q$  is a Brownian motion under Q (not P!).

# **Questions: Black-Scholes Model**

## 4.1

According to theorem 11.3 (Girsanov) in Björk (2009), we can define a new probability Q on  $\mathcal{F}_T$  implying that

$$dW_t = \phi_t dt + dW_t^Q,$$

where  $W_t^Q$  is Q-Brownian motion. The process  $\phi_t$  is called the Girsanov kernel. Compute the Q-dynamics of  $(S_t/B_t)$ . Hint: Replace  $W_t$  with  $\phi_t dt + dW_t^Q$  and apply Ito's formula.

# 4.2

Choose  $\phi_t$  implying that the discounted process is a Q-martingale.

## 4.3

Verify that are the solution satisfies the assumptions in Girsavno's theorem. Conclude that Q (generated from  $\phi_t$ ) is the risk-neutral pricing measure.

#### 4.4

Compute the Q-dynamics of  $(S_t)$ .

# **4.5** (Black-Scholes Equation)

Compute the Q-dynamics of  $F(t,S_t)/B_t$  where  $F(t,S_t)$  is the pricing funcion of the T-claim  $\mathcal{X}$ . Derive the condition (PDE) such that  $F(t,S_t)/B_t$  becomes a Q-martingale. Finally, state the time-T boundary equation for the PDE.

- 5 Pricing Options in the Black-Scholes Model
- 6 TBA
- A) The Fundamental Theorem of Derivative Trading
- B) Stochastic Volatility

# References

Björk, T. (2009). Arbitrage Theory in Continuous Time (3 ed.). Oxford University Press.