## Wiener process and martingales

# Wiener process

#### Stochastic calculus course

The goal: price an option in the framework of Black and Scholes model.

- Very short: 4 weeks only.
- Mathematics is hard.
- Informal definitions and theorems.
- Problem solving and computer simulations.

## **Wiener process**

Here goes the plot!

#### **Stochastic process**

## **Definition**



Stochastic or random process is a collection of random variables indexed by time variable t.

Continuous time:  $(X_t, t \ge 0)$ .

Discrete time:  $(X_t, t \in \{0, 1, 2, 3, ...\})$ .

#### Notation remark:

- $(X_t, t \ge 0)$  or  $(X_t)$  the collection of random variables;
- $X_t$  one particular random variable.

#### **Wiener process**

## **Definition**



Stochastic process  $(W_t, t \ge 0)$  is called Wiener process or

#### Brownian motion if

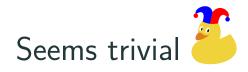
- 1.  $W_0 = 0$ .
- 2. Increments  $W_t W_s$  are normally distributed  $\mathcal{N}(0; t s)$ .
- 3. Increment  $W_t W_s$  is independent of the past values  $(W_u, u \leq s)$ .
- 4.  $\mathbb{P}(\text{trajectory of }(W_t) \text{ is continuous}) = 1.$

Tradition: when we consider two arbitrary moments of time, s and t, we usually assume  $s \leq t$ .

## **Divide and conquer**

The main trick to study properties:

Future value = Known value + Unpredictable change



$$W_t = W_s + (W_t - W_s)$$

#### **Conditional probability exercise**

$$\mathbb{P}(W_{10} > 2 \mid W_6 = 3) = \mathbb{P}(W_{10} - W_6 + W_6 > 2 \mid W_6 = 3) =$$

$$= \mathbb{P}(W_{10} - W_6 + 3 > 2 \mid W_6 = 3) = \mathbb{P}(W_{10} - W_6 > -1).$$

$$W_{10} - W_6 \sim \mathcal{N}(0; 4), \text{ hence } \frac{W_{10} - W_6 - 0}{\sqrt{4}} \sim \mathcal{N}(0; 1).$$

We will use standard normal distribution function

$$F(u) = \mathbb{P}(Z \leq u)$$
, where  $Z \sim \mathcal{N}(0; 1)$ .

$$\mathbb{P}(W_{10} - W_6 > -1) = \mathbb{P}\left(\frac{W_{10} - W_6}{2} > -\frac{1}{2}\right) =$$
$$= \mathbb{P}(Z > -0.5) = \mathbb{P}(Z < 0.5) = F(0.5) \approx 0.69.$$

#### More gentlemen's agreements

On slides we will follow these agreements:

- s and t denote two arbitrary time moments with  $0 \le s \le t$ .
- $(W_t)$  denotes a Wiener process.
- Z denotes a standard normal random variable,  $Z \sim \mathcal{N}(0; 1)$ .
- F(u) denotes the standard normal distribution function,  $F(u) = \mathbb{P}(Z \le u)$ .

#### Independence of increments: example

#### **Property**

Increment  $W_t - W_s$  is independent of the past values  $(W_u, u \leq s)$ .

 $W_6-W_4$  is independend of  $W_4$ ,  $W_3$ ,  $W_{2.5}$ ,  $W_1$ , ...

 $W_6-W_4$  is independent of  $W_4-W_3$ ,  $W_{2.5}-W_1$ .

The increments  $W_6-W_4$ ,  $W_4-W_3$ ,  $W_{2.5}-W_1$  are independent.

#### Independence of increments: full glory

If the time intervals  $[s_1,t_1]$ ,  $[s_2,t_2]$ , ...,  $[s_k,t_k]$  are non overlapping, Here will be a small picture

then the increments  $W(t_1)-W(s_1)$ ,  $W(t_2)-W(s_2)$ , …,  $W(t_k)-W(s_k)$  are independent.

Remark: the right border of an interval may touch the left border of the next one, but may not exceed it,  $t_j \leq s_{j+1}$ .

#### **Expectation and variance**

$$\mathbb{E}(W_t) = \mathbb{E}(W_t - W_0) = 0$$

$$Var(W_t) = Var(W_t - W_0) = t - 0 = t$$

For t > s:

$$Cov(W_s, W_t) = Cov(W_s, W_s + (W_t - W_s)) = Cov(W_s, W_s) = s$$

$$Cov(W_7, W_3) = 3.$$

#### Two friends

## **Definition**



Stochastic process  $(X_t, t \ge 0)$  that may be written as

$$X_t = aW_t + bt,$$

is called brownian motion with drift and scaling.

## **Definition**



Stochastic process  $(S_t, t \ge 0)$  that may be written as

$$S_t = S_0 \exp(aW_t + bt),$$

is called geometric brownian motion.



here will be the plots of BM with drift and geometric BM

## BM with drift and scaling

Exercise  $\red$ . Find  $\mathbb{E}(5W_t + 6t)$  and  $\mathrm{Var}(5W_t + 6t)$ .

$$\mathbb{E}(5W_t + 6t) = 0 + 6t = 6t$$

$$Var(5W_t + 6t) = Var(5W_t) = 25t$$

#### Frequently used expected values

#### Expected values of exponents:

- $\mathbb{E}(\exp(aZ)) = \exp(a^2/2)$  for  $Z \sim \mathcal{N}(0; 1)$ .
- $\mathbb{E}(\exp(aW_t)) = \exp(a^2t/2)$  for Wiener process  $W_t$ .

# How these are obtained?



$$\mathbb{E}(\exp(aZ)) = \int_{-\infty}^{+\infty} \exp(az)f(z) dz =$$

$$= \int_{-\infty}^{+\infty} \exp(az) \frac{1}{\sqrt{2\pi}} \exp(-z^2/2) dz$$

#### **Moment generating function**

# Definition 🕹

The moment generating function (MGF) of a random variable X is defined as

$$M_X(a) = \mathbb{E}(\exp(aX)).$$

- $M_Z(a) = \exp(a^2/2)$  for a normal  $Z \sim \mathcal{N}(0; 1)$ .
- $M_{W_t}(a) = \exp(a^2t/2)$  for a Wiener process  $W_t$ .

## Why may we need MGF?

$$M'(u) = \frac{d}{du}\mathbb{E}(\exp(uX)) = \mathbb{E}(X\exp(uX))$$

$$M'(0) = \mathbb{E}(X)$$



MGF is a funny way to calculate expected value!

$$M''(0) = \mathbb{E}(X^2)$$

$$M'''(0) = \mathbb{E}(X^3)$$

$$M^{(k)}(0) = \mathbb{E}(X^k)$$

#### Wiener process: summary

- Stochastic process with normal and independent increments.
- Wiener process with drift and geometric Wiener process.
- Moment generating function.

# **Conditional expectation**

#### **Conditional expectation: short plan**

- Modeling information using sigma-algebras;
- Properties of conditional expected value;
- Conditional variance.

#### **Modeling information**

John knows the value of X.

Maria knows the value of X and Y.

Maria knows more!

How to model this mathematically?

#### Sigma-algebra

#### Informal definition



Sigma-algebra ( $\sigma$ -algebra) generated by random variables X and Y is the collection of all events that can be stated in terms of these random variables.

Notation:  $\sigma(X, Y)$ .

#### **Example**

The sigma-algebra  $\sigma(X,Y)$  contains the events  $\{X<5\}$ ,  $\{X>2Y\}$ ,  $\{\sin Y>\cos X\}$ , ...

## **Modeling information**

John knows the value of X,  $\mathcal{F}_J = \sigma(X)$ .

Maria knows the value of X and Y,  $\mathcal{F}_M = \sigma(X,Y)$ 

Maria knows more:  $\mathcal{F}_J \subset \mathcal{F}_M$ .

#### Measurability

#### **Definition**



The random variable Z is measurable with respect to  $\sigma$ -algebra  $\mathcal F$  if  $\sigma(Z)\subset \mathcal F.$ 

Information in  $\mathcal{F}$  is sufficient to calculate the value of Z.

#### Informal theorem



The random variable Z is measurable with respect to  $\sigma(X,Y)$  if and only if Z is a deterministic function of X and Y.

#### **Best prediction**

#### Informal definition



The best prediction of a random variable Y given  $\sigma$ -algebra  $\mathcal{F}$  is called conditional expected value  $\mathbb{E}(Z \mid \mathcal{F})$ .

## Difference of $\mathbb{E}(Z \mid \mathcal{F})$ and $\mathbb{E}(Z)$

If I know X and Y then my best prediction of Z may depend on X and Y.

In general:  $\mathbb{E}(Z \mid \mathcal{F})$  is a random variable.

#### **Notation**

- $\mathbb{E}(Z \mid \mathcal{F})$ : for a general  $\sigma$ -algebra  $\mathcal{F}$ ;
- $\mathbb{E}(Z \mid \sigma(X, Y))$  or  $\mathbb{E}(Z \mid X, Y)$ : for  $\sigma$ -algebra generated by X and Y.

#### When we may omit conditioning?

- If Z is independend of X and Y then  $\mathbb{E}(Z \mid X, Y) = \mathbb{E}(Z)$ : If I know nothing useful about Z then I can drop my information.
- $\mathbb{E}(\mathbb{E}(Z\mid\mathcal{F}))=\mathbb{E}(Z)$ : The average of best guess is the average of predicted variable.

#### The case of known variable

If Z is known (measurable with respect to  $\mathcal{F}$ ), then we may treat Z like a constant:

$$\mathbb{E}(Z \mid \mathcal{F}) = Z;$$

$$\mathbb{E}(2\exp(5W_t) \mid W_t) = 2\exp(5W_t);$$

$$\mathbb{E}(2ZR + Z^2 \mid \mathcal{F}) = 2Z\mathbb{E}(R \mid \mathcal{F}) + Z^2.$$

#### **Conditional variance**

## **Definition**



The conditional variance  $Var(Z \mid \mathcal{F})$  is the conditional expected value of the squared error of the best prediction,

$$\operatorname{Var}(Z \mid \mathcal{F}) = \mathbb{E}(\Delta^2 \mid \mathcal{F}), \text{ where } \Delta = Z - \mathbb{E}(Z \mid \mathcal{F}).$$

#### **Theorem**



$$\operatorname{Var}(Z \mid \mathcal{F}) = \mathbb{E}(Z^2 \mid \mathcal{F}) - (\mathbb{E}(Z \mid \mathcal{F}))^2.$$

#### **Properties of conditional variance**

- Irrelevant information may be omitted: If Z is independent of  $\mathcal{F}$  then  $\mathbb{E}(Z\mid\mathcal{F})=\mathbb{E}(Z)$  and  $\mathrm{Var}(Z\mid\mathcal{F})=\mathrm{Var}(Z).$
- If Z is known (measurable with respect to  $\mathcal{F}$ ), then we may treat Z like a constant:

$$Var(2\exp(5W_t) \mid W_t) = 0;$$

$$Var(Z^3 + 3ZR \mid \mathcal{F}) = 0 + (3Z)^2 Var(R \mid \mathcal{F}).$$

#### **Conditioning: summary**

- Sigma-algebra  $\sigma(X,Y)$  is the collection of all events that can be stated using X and Y.
- Conditional expected value  $\mathbb{E}(Z\mid X,Y)$  is the best prediction of Z using X and Y.
- Conditional variance  $\mathrm{Var}(Z\mid X,Y)$  is the conditional expected value of the squared error of the best prediction.

# Martingales

#### Martingales: short plan

- Filtration models the information acquisition.
- Definition of a martingale.
- Examples of martingales.

#### **Filtration**

The  $\sigma$ -algebra  $\mathcal{F}_t$  describes all the information available at time t.

## **Definition**



The family of sigma-algebras  $(\mathcal{F}_t, t \geq 0)$  is called filtration if it grows in time,  $\mathcal{F}_s \subset \mathcal{F}_t$  for  $s \leq t$ .

Reminder: Sigma-algebra  $\mathcal{F}_t$  is the collection of events.

#### **Natural filtration**

#### **Definition**



The filtration  $(\mathcal{F}_t, t \geq 0)$  is called a natural filtration of a process  $(X_t, t \geq 0)$  if at time t you have only the information about past values of the process,

$$\mathcal{F}_t = \sigma(X_u, u \in [0; t]).$$

#### **Examples**

Let  $(\mathcal{F}_t)$  be a natural filtration of a Wiener process  $(W_t)$ .

$$\{W_2 < 5\} \in \mathcal{F}_2, \{W_2 > W_5\} \in \mathcal{F}_6,$$

$$\{W_2 < 5\} \not\in \mathcal{F}_1, \{W_2 > W_5\} \not\in \mathcal{F}_2.$$

## Martingale

#### **Definition**



Consider a filtration  $(\mathcal{F}_t, t \geq 0)$  and a process  $(M_t, t \geq 0)$ .

If the best prediction of the future value  $M_t$  of a process is its current value  $M_s$  for  $s \leq t$ ,

$$\mathbb{E}(M_t \mid \mathcal{F}_s) = M_s,$$

then  $(M_t)$  is called a martingale with respect to the filtration  $(\mathcal{F}_t)$ .

Usually we consider natural filtration  $(\mathcal{F}_t)$  of the process  $(M_t)$ .

#### Simple examples

#### Constant process:

If  $M_t = 777$  for all t then  $\mathbb{E}(M_t \mid \mathcal{F}_s) = 777 = M_s$ .

#### Wiener process:

$$\mathbb{E}(W_t \mid \mathcal{F}_s) = \mathbb{E}(W_s + (W_t - W_s) \mid \mathcal{F}_s) = W_s + \mathbb{E}(W_t - W_s) = W_s.$$

## More examples

#### **Theorem**



The process  $Z_t = W_t^2 - t$  is a martingale.

## Proof 2



$$\mathbb{E}(W_t^2 - t \mid \mathcal{F}_s) = \mathbb{E}((W_s + (W_t - W_s))^2 \mid \mathcal{F}_s) - t =$$

$$= \mathbb{E}(W_s^2 + (W_t - W_s)^2 + 2W_s(W_t - W_s) \mid \mathcal{F}_s) - t =$$

$$= W_s^2 + \mathbb{E}((W_t - W_s)^2 \mid \mathcal{F}_s) + 2W_s\mathbb{E}(W_t - W_s \mid \mathcal{F}_s) - t =$$

$$= W_s^2 + \mathbb{E}((W_t - W_s)^2) + 2W_s\mathbb{E}(W_t - W_s) - t =$$

$$= W_s^2 + (t - s) + 2W_s \cdot 0 - t = W_s^2 - s$$

#### More examples

#### **Theorem**



The process  $Z_t = \exp(aW_t - a^2t/2)$  is a martingale for every constant a.

This martingale is very useful in Black and Scholes model.

## Martingales in discrete time

#### **Theorem**



Consider a filtration  $(\mathcal{F}_t, t \in \{0, 1, 2, ...\})$  and a process  $(M_t, t \in \{0, 1, 2, ...\})$ .

In discrete time the condition

$$\mathbb{E}(M_t \mid \mathcal{F}_s) = M_s \text{ for all } s \leq t$$

is completely equivalent to

$$\mathbb{E}(M_{t+1} \mid \mathcal{F}_t) = M_t.$$

#### **Random walk**

Consider independent and identically distributed  $Z_1$ ,  $Z_2$ , ... with  $\mathbb{E}(Z_t)=0$ . The cumulative sum

$$S_t = Z_1 + Z_2 + \ldots + Z_t$$
, with  $S_0 = 0$ 

is called a random walk.

#### **Theorem**



The random walk process is a martingale.

$$\mathcal{F}_t = \sigma(Z_1, Z_2, Z_3, \dots, Z_t)$$

$$\mathbb{E}(S_{t+1} \mid \mathcal{F}_t) = \mathbb{E}(S_t + Z_{t+1} \mid \mathcal{F}_t) = S_t + \mathbb{E}(Z_{t+1}) = S_t.$$

#### **Martingales: summary**

- Filtration models the information acquisition.
- The best prediction of a martingale is its current value.
- Martingales related to Wiener process, random walk.