

Stochastic Calculus for Financial

$[t-s/\epsilon]$

$$\frac{1}{\sqrt{\epsilon}} \sum_{i=1}^{\lfloor t-s/\epsilon \rfloor} X_i \rightarrow N(0, \sigma^2)$$

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$$C = \frac{1}{n}, n \rightarrow \infty$$

$$\left[\sum_{i=1}^{\lfloor t-s \rfloor} X_i \right] \xrightarrow{\text{a.s.}} N(0, \sigma^2)$$

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Fubini

$$\int \int f(x,y) dx dy = \int_x \int_y f(x,y) dy dx$$

Martingale
Brownian Motion < Markov process

Stochastic Calculus

Ito Calculus - Black-Scholes PDE

Girsanov Thm

Risk-Neutral Pricing theory

Financial application

Martingale Representation Thm.

Fundamental thm of asset pricing.

relation between arbitrage opportunity & existence neutral of risk neutral measure

1^o Probability measure.

sample space

Given (Ω, \mathcal{F}) - a prob measure $P: \mathcal{F} \rightarrow [0, 1]$.
σ-algebra.

$$\textcircled{1} \quad P(\Omega) = 1$$

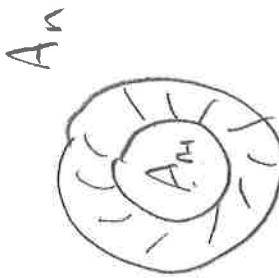
$$\textcircled{2} \quad \text{Countable-additivity: } \{A_n\} \subseteq \mathcal{F}, A_n \cap A_m = \emptyset, \\ P\left(\bigcup_{n=1}^{\infty} A_n\right) = \sum_{n=1}^{\infty} P(A_n).$$

Ex 1: $A_1 \subseteq A_2 \subseteq A_3 \subseteq \dots \subseteq A_n \subseteq \dots$, then

$$P\left(\bigcup_{n=1}^{\infty} A_n\right) = \lim_{n \rightarrow \infty} P(A_n)$$

$$\begin{aligned} B_1 &= A_1, & B_2 &= A_2 \setminus A_1, & B_n &= A_n \setminus A_{n-1}, \dots \\ \bigcup_{n=1}^{\infty} B_n &= \bigcup_{n=1}^{\infty} A_n, & \{B_n\} &\text{ are pairwise disjoint.} \end{aligned}$$

$$\begin{aligned} P\left(\bigcup_{n=1}^{\infty} A_n\right) &= P\left(\bigcup_{n=1}^{\infty} B_n\right) \\ &= \sum_{n=1}^{\infty} P(B_n) = \lim_{N \rightarrow \infty} \sum_{n=1}^N P(A_n \setminus A_{n-1}) \\ &= \lim_{N \rightarrow \infty} \sum_{n=1}^N \underbrace{(P(A_n) - P(A_{n-1}))}_{(P(A_n) - P(A_{n-1}))} \\ &= \lim_{N \rightarrow \infty} P(A_N). \end{aligned}$$



Subadditive of \mathbb{P} : $\{A_n\} \in \mathcal{F}$,

$$\begin{aligned} \textcircled{1} \quad \mathbb{P}\left(\bigcup_{i=1}^n A_i\right) &\leq \sum_{i=1}^n \mathbb{P}(A_i) & \checkmark \\ \textcircled{2} \quad \mathbb{P}\left(\bigcup_{i=1}^{\infty} A_i\right) &\leq \sum_{i=1}^{\infty} \mathbb{P}(A_i) \end{aligned}$$

$$\begin{aligned} \textcircled{1} \quad \mathbb{E}\left(\bigwedge_{i=1}^n A_i\right) &\leq \mathbb{E}\left(\bigwedge_{i=1}^n \mathbb{1}_{A_i}\right) \quad (\text{You check!}) \\ &= \mathbb{P}\left(\bigcap_{i=1}^n A_i\right) \end{aligned}$$

$$\begin{aligned} \textcircled{2} \quad B_1 = A_1, \quad B_2 = A_1 \cup A_2, \quad \dots, \quad B_n = \bigcup_{i=1}^n A_i, \quad \dots \\ \mathbb{P}\left(\bigcup_{n=1}^{\infty} A_n\right) &= \mathbb{P}\left(\bigcup_{n=1}^{\infty} B_n\right) = \lim_{n \rightarrow \infty} \mathbb{P}(B_n) = \lim_{n \rightarrow \infty} \frac{\mathbb{P}\left(\bigcap_{i=1}^n A_i\right)}{\sum_{i=1}^n \mathbb{P}(A_i)} \\ &\leq \lim_{n \rightarrow \infty} \frac{\sum_{i=1}^n \mathbb{P}(A_i)}{\sum_{i=1}^{\infty} \mathbb{P}(A_i)} = \frac{\sum_{i=1}^{\infty} \mathbb{P}(A_i)}{\sum_{i=1}^{\infty} \mathbb{P}(A_i)}. \end{aligned}$$

2°

Change of measure

Given (Ω, \mathcal{Q}, P) .

Given $Z \geq 0, \mathbb{E}(Z) = 1$

$$(*) \quad Q(A) = \int_A Z(\omega) dP(\omega) = \mathbb{E}(Z \mathbf{1}_A), \quad A \in \mathcal{Q}.$$

You can check that Q is a probability measure on (Ω, \mathcal{Q}) .

$(*)$ is called change of measure formula.

$$\begin{array}{ccc} (*) & \longrightarrow & Q \\ P & & \end{array}$$

$Y \sim N(\mu, 1)$, $X = Y - \mu \sim N(0, 1)$ under \mathbb{P}

$$\mathbb{P}(Y \leq t) = \int_{-\infty}^t \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(y-\mu)^2} dy, \quad \mathbb{P}(X \leq t) = N(t)$$

$$Z = e^{-\mu X - \frac{1}{2}\mu^2} > 0, \quad \mathbb{E}(Z) = 1$$

$$Q(A) = \mathbb{E}(Z \mathbb{1}_A) \quad \mathbb{E}^Q(\cdot)$$

[Q]: What's the distribution of Y under Q ??

$$\begin{aligned} Q(Y \leq t) &= \mathbb{E}(Z \mathbb{1}_{\{Y \leq t\}}) \\ &= \mathbb{E}\left(e^{-\mu X - \frac{1}{2}\mu^2} \mathbb{1}_{\{X \leq t - \mu\}}\right) \end{aligned}$$

$$\begin{aligned}
 &= \int_{-\infty}^R e^{-\mu x - \frac{1}{2}\sigma^2} \mathbb{1}_{\{X \leq t\}} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x^2} dx \\
 &= \int_{-\infty}^{t+\mu} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(x+\mu)^2} dx \\
 &\stackrel{z=x+\mu}{=} \int_{-\infty}^t \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2} dz = N(t)
 \end{aligned}$$

$$\begin{array}{ccc}
 \mathbb{P} & \xrightarrow{(x)} & \mathbb{Q} \\
 Y \sim N(\mu, \sigma^2) & \implies & N(0, 1)
 \end{array}$$

$$\boxed{dS_t} = S_t [\underbrace{\mu(t, S_t) dt + \underbrace{S_t \sigma(t, S_t) dW_t}_{\text{drift}}}_{\text{Volatility}}]$$

$$Y = \frac{\mu + \frac{\sigma^2}{2}}{\text{drift}} \text{ vol}$$

3° R.V. & Expectation:

Ex1: $\Omega = [0,1]$, \mathbb{P} -Lebesgue measure.

$$\mathcal{Q} = \mathcal{B}([0,1])$$

$$P([a,b]) = b-a$$

$$X(\omega) = \begin{cases} 1 & \text{if } \omega \in \mathbb{Q}^c \\ 0 & \text{if } \omega \in \mathbb{Q} - \text{rational.} \end{cases}$$

$$\{\omega\} = [\omega, \omega]$$

Show X is a r.v. and compute $E(X)$.

$$\left\{ \omega : X(\omega) \leq a \right\} \in \mathcal{Q}$$

$$\left\{ \omega : X(\omega) = 0 \right\} \in \mathcal{Q} \quad \cup$$

$$\mathbb{Q} \cap [0,1].$$

In HW, $\{0\} \in \mathcal{B}$

$$\mathbb{Q} \cap [0,1] = \left(\bigcup_{\omega \in \mathbb{Q} \cap [0,1]} \{\omega\} \right) \in \mathcal{B}([0,1])$$

Countable.

$$\mathbb{E}(X) = \sum_{\omega \in \mathbb{Q}^c} \mathbb{P}(\omega) = \mathbb{P}([0,1] \cap \mathbb{Q}^c)$$

$$1 - \frac{\mathbb{P}([0,1] \cap \mathbb{Q})}{\mathbb{P}([0,1])} = 1 - \frac{\mathbb{P}(\{\omega\})}{\mathbb{P}([0,1])} = 1 - 0 = 1$$

$$\mathbb{E}(X) = \int_{\Omega} X(\omega) dP(\omega)$$

Ex 2

$$\mathbb{E}(X) = \int_0^\infty P(X \geq t) dt$$

Show

$$= \int_0^\infty \left[\int_{\Omega} \mathbb{1}_{\{X \geq t\}} dP(\omega) \right] dt$$

$$= \int_{\Omega} \left[\int_0^\infty \frac{\mathbb{1}_{\{X \geq t\}}}{\int_{\Omega} \mathbb{1}_{\{X(\omega) \geq t\}} dP(\omega)} dt \right] dP(\omega)$$

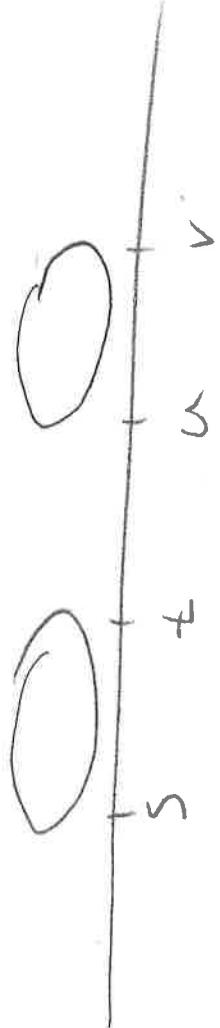
$$= \int_{\Omega} \left[\int_0^\infty \frac{\mathbb{1}_{\{X(\omega) \geq t\}}}{t} dt \right] dP(\omega) = \int_0^\infty \left[\int_{\Omega} \frac{\mathbb{1}_{\{X(\omega) \geq t\}}}{t} dP(\omega) \right] dt = \int_0^\infty \frac{X(\omega)}{t} dt = X(\omega)$$

4° Brownian Motion:

Recall: A stochastic process $(B_t)_{t \geq 0}$ is a B.M.

- ① $B_0 = 0$ (a.s.)
- ② $t \mapsto B_t(\omega)$ is cont (a.s.).
- ③ (Stationary increments): $0 \leq t < t+h$,
 $B_{t+h} - B_t \sim N(0, h)$
- ④ (independent increments), $0 \leq s < t \leq u < v$

$$B_t - B_s \perp\!\!\!\perp B_v - B_u$$



Ex 1: If B is a B.M.

$$\textcircled{1} \text{ then } X_t = \frac{1}{\sqrt{n}} B_{nt}, t \geq 0 \text{ is B.M.}$$

(You check)

$$\textcircled{2} \quad Y_t = \begin{cases} \pm \frac{1}{\sqrt{t}} B_{\frac{1}{t}}, & \text{if } t > 0 \\ 0 & \text{if } t = 0 \end{cases}$$

is a BM

(You check)

Def: A stochastic process X is called Gaussian if for

for $t_1 < t_2 < \dots < t_k$, $(X_{t_1}, X_{t_2}, \dots, X_{t_k})$ is jointly normal distributed.

Fact: finite dimensional distribution of a Gaussian process X is characterized by

$$m(t) = \mathbb{E}(X_t)$$

$$\rho(s,t) = \mathbb{E}((X_t - m(t))(X_{s-t} - m(s)))$$

$$X = B, \quad m(t) = \mathbb{E}(B_t) = 0 \quad B_t \sim N(0, t)$$
$$\rho(s,t) = \mathbb{E}(B_s B_t) = \min(s, t)$$

Conclusion: A process B is a B.M. iff

B is a Gaussian process, with

$$m(t) = 0$$

$$\rho(s,t) = s \wedge t$$