

MATH4210: Financial Mathematics Tutorial 5

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9 October, 2024

Convergence of r.v.s

Let $(\Omega, \mathcal{F}, \mathbb{P})$ be a probability space. X and $\{X_n\}$ are \mathbb{R} valued (sequence of) r.v.s.

Definition (Convergence almost surely)

Denote by $X_n \rightarrow X$ a.s. (almost surely) if

$$\mathbb{P}[\{\omega \in \Omega : \lim_{n \rightarrow \infty} X_n(\omega) = X(\omega)\}] = 1$$

Definition (Convergence in Probability)

Denote by $X_n \rightarrow X$ in probability if for any $\rho > 0$

$$\underbrace{\lim_{n \rightarrow \infty} \mathbb{P}[\{\omega \in \Omega : |X_n(\omega) - X(\omega)| \geq \rho\}]}_{} = 0$$

Convergence of r.v.s

Proposition

$X_n \rightarrow X$ a.s. implies $X_n \rightarrow X$ in probability.

Proposition

$X_n \rightarrow X$ in probability implies there exists a subsequence of X_n converging to X a.s..

$$X_{n_k} \rightarrow X \text{ a.s.}$$

$k \rightarrow \infty$

Definition (Convergence in Law (in Distribution))

Let F_n and F be the c.d.f. of X_n and X for all $n \in \mathbb{N}$. $X_n \rightarrow X$ in Law (in Distribution) if

$$\lim_{n \rightarrow \infty} F_n(x) = F(x)$$

for any $x \in \mathbb{R}$ where F is continuous at x .

Proposition

$X_n \rightarrow X$ in probability implies $X_n \rightarrow X$ in Law.

Convergence of r.v.s.

(a) $\lim_{n \rightarrow \infty} E[|X_n - X|^2] = 0 \Rightarrow \lim_{n \rightarrow \infty} E[|X_n - X|] = 0$

fix $n \in \mathbb{N}$
 $E[|X_n - X|] = E[\sqrt{|X_n - X|^2}]$
 $f(x) = \sqrt{x}$ concave

Recall Jensen inequality
 if f is concave
 $E[f(x)] \leq f(E(x))$

Definition $\Rightarrow E[|X_n - X|] \leq \sqrt{E[|X_n - X|^2]} \Rightarrow 0$ as $n \rightarrow \infty$ \square

Given $p > 0$, denote by $X_n \rightarrow X$ in L^p if

$$\lim_{n \rightarrow \infty} \underbrace{E[|X_n - X|^p]}_{} = 0$$

Question

(a). Show that $X_n \rightarrow X$ in L^2 implies $X_n \rightarrow X$ in L^1 .

(b). Show that $X_n \rightarrow X$ in L^p implies $X_n \rightarrow X$ in probability.

(b) fix $p > 0$ fix $N \in \mathbb{N}$ markov inequality

$$P\{\{w \in \Omega : |X_n - X| \geq p\}\} = P\{\{w \in \Omega : |X_n - X|^p \geq p^p\}\} \quad E[|X| > M] \leq \frac{1}{M} E[|X|]$$

$$\leq \frac{1}{p^p} E[|X_n - X|^p] \Rightarrow 0 \text{ as } n \rightarrow \infty$$

$$\rightarrow 0 \text{ as } n \rightarrow \infty \quad \square$$

Brownian Motions

$$\text{IBP } \int u du = uv - \int v du$$

$$\partial_x u(t,x) = \int_{\mathbb{R}} f(y+x) P_z(y) dy$$

(a) Define $Z \sim N(0, T-t)$ $P_z = \text{pdf of } Z$

$$u(t,x) = E[f(z+x)] = \int_{\mathbb{R}} f(y+x) P_z(y) dy$$

$$\begin{aligned} &= \underbrace{\int_{\mathbb{R}} f(y+x) P_z(y) dy}_{\text{IBP}} \Big|_{-\infty}^{\infty} - \\ &\quad \underbrace{\int_{\mathbb{R}} P_z(y) f(y+x) dy}_{\text{IBP}} \end{aligned}$$

Question

Let $f : \mathbb{R} \rightarrow \mathbb{R}$ be such that for some constant $C > 0$, $f(x) < e^{C|x|}$ for all $x \in \mathbb{R}$. Define

$$u(t,x) = \mathbb{E}[f(B_T) | B_t = x] = \mathbb{E}[f(B_T - B_t + x)].$$

Show that

(a)

$$\partial_x u(t,x) = \mathbb{E}\left[\frac{B_T - B_t}{T-t} f(B_T - B_t + x)\right]$$

(b)

$$\partial_x^2 u(t,x) = \mathbb{E}\left[\frac{(B_T - B_t)^2 + (T-t)}{(T-t)^2} f(B_T - B_t + x)\right]$$

$$P_z(y) = \frac{1}{\sqrt{2\pi(T-t)}} e^{\frac{-y^2}{2(T-t)}} \cdot \frac{1}{(T-t)}$$

$$\partial_x u(t,x) = 0 + \int_{\mathbb{R}} \frac{1}{(T-t)} f(y+x) P_z(y) dy = E\left[\frac{f(z+x)}{T-t} \cdot z\right] = E\left[\frac{f(B_T - B_t + x)}{T-t} \cdot B_T - B_t\right]$$

Greeks of Option

$\sim \mathcal{N}(\mu, \sigma^2)$
P

$$\frac{\partial}{\partial s} C_E(t, S_t) = N(d_1) + S_t \cdot \frac{\partial}{\partial s} N(d_1) - e^{-r(T-t)} K \frac{\partial}{\partial s} N(d_2)$$

$$\begin{aligned}\frac{\partial}{\partial s} N(d_1) &= \frac{\partial}{\partial s} \int_{-\infty}^{d_1} f_Z(y) dy \quad Z \sim N(0, 1) \quad f_Z(y) : \text{pdf of } Z \\ &= f_Z(d_1) \frac{\partial}{\partial s} d_1\end{aligned}$$

Question

Consider the European call option price at time t :

$$C_E(t, S_t) = \underline{S_t} \underline{N(d_1)} - e^{-r(T-t)} K N(d_2)$$

where $d_1 = \frac{\ln(S_t/K) + (r + \sigma^2/2)(T-t)}{\sigma\sqrt{T-t}}$ and $d_2 = d_1 - \underbrace{\sigma\sqrt{T-t}}$. Compute

(a) Delta: $\Delta = \partial_s C_E(t, S_t)$.

(b) Gamma: $\Gamma = \partial_s^2 C_E(t, S_t)$.

$$\frac{\partial}{\partial s} d_1 = \frac{1}{\sigma\sqrt{T-t}} \frac{1}{S_t} = \frac{\partial}{\partial s} d_2$$

$$\begin{aligned}\frac{\partial}{\partial s} C_E(t, S_t) &= N(d_1) + \frac{1}{\sigma\sqrt{T-t}} (S_t f_Z(d_1) - e^{-r(T-t)} K f_Z(d_2)) \\ &= N(d_1)\end{aligned}$$