

SOLUTIONS TO EXAM FOR STOCHASTIC MODELS IN DISCRETE TIME 3.75 ECTS

Master's program of Financial Mathematics
October 30, 2009, 9.00 – 13.00

Max number of points: 30.

Halmstad University grading bounds: 12p \Rightarrow grade 3, 18p \Rightarrow grade 4, 24p \Rightarrow grade 5.

ECTS bounds: 12p \Rightarrow grade E, 15p \Rightarrow grade D, 18p \Rightarrow grade C, 21p \Rightarrow grade B, 24p \Rightarrow grade A.

Allowed aids: Summary of formulae attached to the exam, calculator and dictionary.

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1. Let $X = \{(X_n, \mathcal{F}_n) : n \geq 0\}$ be a stochastic sequence with $E(|X_0|) < \infty$. Prove that if X is a martingale transformation, then X is a local martingale. (4p)

Solution: (See pp 98 in *Essentials of Stochastic Finance. Facts, Models, Theory.* by A.N. Shiryaev.) \square

2. Assume $\{X_n\}$ is an $MA(2)$ process with parameters $b_0 = b_1 = b_2 = \sigma_\epsilon^2 = 1$. Calculate

(a) the variance $D(X_n)$ (3p)

(b) the probability $P(X_n > X_{n+1} + 1)$ (4p)

Solution:

(a) $D(X_n) = D(1 + \epsilon_{n-1} + \epsilon_{n-2} + \epsilon_n) = D(\epsilon_{n-1}) + D(\epsilon_{n-2}) + D(\epsilon_n) = 3$.

(b) $P(X_n > X_{n+1} + 1) = 1 - P(X_n - X_{n+1} \leq 1)$ where $X_n - X_{n+1}$ is normally distributed with expectation $\mu = E(X_n - X_{n+1}) = 0$ and variance $\sigma^2 = D(X_n - X_{n+1}) = D(1 + \epsilon_n + \epsilon_{n-1} + \epsilon_{n-2} - (1 + \epsilon_{n+1} + \epsilon_n + \epsilon_{n-1})) = D(\epsilon_{n-2} - \epsilon_{n+1}) = 2$. Thus $P(X_n > X_{n+1} + 1) = 1 - \Phi\left(\frac{1-0}{\sqrt{2}}\right) = 1 - \Phi(0.7071) = 0.2389$. \square

3. Let $\{\xi_k\}_{k=0}^\infty$ be a classical random walk starting in 0 at time 0.

(a) Prove that $\{\xi_k\}$ is non-stationary. (4p)

(b) Calculate the probability that the random walk returns to zero at time n . (5p)

Solution:

(a) If the process is not weakly stationary it can not be stationary in any sense. To be weakly stationary we have to have $E(\xi_k) = m$ and $C(\xi_k, \xi_{k+h}) = R(h)$. Let us write $\xi_k = \sum_{i=1}^k X_i$ where $\{X_i\}$ are independent and distributed $P(X_i = -1) = P(X_i = 1) = \frac{1}{2}$. Since $E(X_i) = 0$ we have that

$D(X_i) = E(X_i^2) = (-1)^2 P(X_i = -1) + 1^2 P(X_i = 1) = 1$ and $C(\xi_k, \xi_{k+1}) = C(\sum_{i=1}^k X_i, \sum_{i=1}^{k+1} X_i) = C(\sum_{i=1}^k X_i, \sum_{i=1}^k X_i) + C(\sum_{i=1}^k X_i, \sum_{i=k+1}^{k+1} X_i) = \sum_{i=1}^k D(X_i) + 0 = k$. Thus the covariance function of $\{\xi_k\}$ can not be a function of h (the time distance between the variables), and consequently $\{\xi_k\}$ can not be stationary.

- (b) Since the random walk is $\xi_n = \sum_{i=1}^n X_i$ where $\{X_i\}$ are independent and distributed $P(X_i = -1) = P(X_i = 1) = \frac{1}{2}$, the variables X_i can be written as $2Y_i - 1$ where $\{Y_i\}$ are independent and Bernoulli distributed with $\frac{1}{2}$, i.e. $P(Y_i = 0) = P(Y_i = 1) = \frac{1}{2}$ and $\sum_{i=1}^n Y_i \in \text{Bin}(n, \frac{1}{2})$. Now, $P(\xi_n = 0) = P(\sum_{i=1}^n (2Y_i - 1) = 0) = P(2(\sum_{i=1}^n Y_i) - n = 0) = P(\sum_{i=1}^n Y_i = \frac{n}{2}) = \begin{cases} \binom{n}{n/2} (\frac{1}{2})^{n/2} (1 - \frac{1}{2})^{n-n/2} & \text{if } n \text{ is even} \\ 0 & \text{if } n \text{ is odd} \end{cases} = \begin{cases} \binom{n}{n/2} 2^{-n} & \text{if } n \text{ is even} \\ 0 & \text{if } n \text{ is odd} \end{cases} \quad \square$

4. (*Cusum procedure*) Let $\{X_k\}$ be a sequence of independent identically distributed random variables with zero mean, and let the sequence $\{S_n\}$ be defined by

$$S_n = \max_{k=1,2,\dots,n} \sum_{j=k}^n X_j$$

- (a) Derive a recursive representation of $\{S_n\}$, i.e. find a relationship f such that, with $S_1 = X_1$, we have $S_n = f(X_n, S_{n-1})$ for $n = 2, 3, \dots$ (3p)
- (b) Prove that $\{S_n\}$ is a submartingale with respect to the flow $\mathcal{F}_n = \sigma(X_1, X_2, \dots, X_n)$. (2p)

Solution:

$$\begin{aligned} \text{(a)} \quad S_n &= \max_{k=1,2,\dots,n} \sum_{j=k}^n X_j \\ &= \max(\sum_{j=1}^n X_j, \sum_{j=2}^n X_j, \dots, \sum_{j=n-1}^n X_j, X_n) \\ &= \max(X_n + \sum_{j=1}^{n-1} X_j, X_n + \sum_{j=2}^{n-1} X_j, \dots, X_n + X_{n-1}, X_n + 0) \\ &= X_n + \max(\sum_{j=1}^{n-1} X_j, \sum_{j=2}^{n-1} X_j, \dots, X_{n-1}, 0) \\ &= X_n + \max\left(0, \max(\sum_{j=1}^{n-1} X_j, \sum_{j=2}^{n-1} X_j, \dots, X_{n-1})\right) \\ &= X_n + \max(0, S_{n-1}) \end{aligned}$$

- (b) Assuming that $E(|X_n|)$ is finite $= m$, we have that $E(|S_n|) = E(|\max_{1 \leq k \leq n} \sum_{j=k}^n X_j|) \leq E(\max_{1 \leq k \leq n} \sum_{j=k}^n |X_j|) \leq E(\sum_{j=1}^n |X_j|) = mn < \infty$. Of course $S_n = \max_{1 \leq k \leq n} \sum_{j=k}^n X_j \in \mathcal{F}_n = \sigma(X_1, X_2, \dots, X_n)$. Finally, since $\{X_k\}$ are independent and have zero mean, we have that $E(X_{n+1} | \mathcal{F}_n) = 0$, and therefore that $E(S_{n+1} | \mathcal{F}_n) = E(X_{n+1} + \max(0, S_n) | \mathcal{F}_n) = E(X_{n+1} | \mathcal{F}_n) + E(\max(0, S_n) | \mathcal{F}_n) = 0 + \max(0, S_n) \geq S_n$. \square

5. Assume that $\{h_t\}$ is a stochastic volatility process of order 1 with parameters a_0, a_1 such that $0 < a_1 < 1$ and $c = 1$. Prove that the fourth moment of h_t is bounded from above by $3e^2$. (5p)

Solution: In the stochastic volatility model of order 1 with $c = 1$ we have $h_t = \sigma_t \epsilon_t$ where $\sigma_t^2 = e^{\Delta_t}$ and $\Delta_t = a_0 + a_1 \Delta_{t-1} + \delta_t$. Then, due to stationarity $E(h_t^4) = m_4$ where

$$\begin{aligned}
 m_4 &= E(\sigma_t^4 \epsilon_t^4) \\
 &= E(\sigma_t^4) E(\epsilon_t^4) \\
 &= 3E(e^{2\Delta_t}) \\
 &= 3E(e^{2(a_0 + a_1 \Delta_{t-1} + \delta_t)}) \\
 &= 3e^{2a_0} E(e^{2a_1 \Delta_{t-1}}) E(e^{2\delta_t})
 \end{aligned} \tag{1}$$

Now let $Y_t = e^{2\Delta_t}$. Then $m_4 = 3E(Y_t)$. Further, $f(y) = y^{a_1}$ is a concave function since $0 < a_1 < 1$. Thus $E(f(Y_{t-1})) = E(e^{2a_1 \Delta_{t-1}}) \leq (E(e^{2\Delta_{t-1}}))^{a_1} = (E(f(Y_{t-1})))^{a_1}$, and $m_4 \leq 3e^{2a_0} E(e^{2\delta_t}) (E(e^{2\Delta_{t-1}}))^{a_1}$. Due to stationarity we get $\frac{1}{3}m_4 = E(Y_t) = E(Y_{t-1})$ (from equation (1) above) and therefore $m_4 \leq 3e^{2a_0} E(e^{2\delta_t}) (\frac{1}{3})^{a_1} m_4^{a_1}$. Solving with respect to m_4 we get $m_4^{1-a_1} \leq 3^{1-a_1} e^{2a_0} E(e^{2\delta_t})$. Since $\delta_t \in N(0, 1)$ we have

$$\begin{aligned}
 E(e^{2\delta_t}) &= \int_{\mathbb{R}} e^{2x} \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx \\
 &= \int_{\mathbb{R}} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(x^2 - 4x + 4 - 4)} dx \\
 &= \int_{\mathbb{R}} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(x-2)^2 + 2} dx \quad \{u = x - 2\} \\
 &= e^2 \int_{\mathbb{R}} \frac{1}{\sqrt{2\pi}} e^{-u^2/2} du \\
 &= e^2
 \end{aligned}$$

and therefore

$$m_4 \leq (3^{1-a_1} e^{2a_0} e^2)^{1/(1-a_1)} \leq 3e^{2(a_0+1)/(1-a_1)}$$

Finally $a_0 > 0$ and $0 < a_1 < 1$ so $a_0 + 1 > 1$ and $1 - a_1 \leq 1$ and therefore

$$E(h_t^4) = m_4 \leq 3e^2$$

□