Take Home Final

Stochastic Models for Finance and Insurance

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Problem 1 – Solution

Firstly we calculate the prices $X(t,\cdot)$ at time $t=\{1,2\}$ using binomial tree with the root $X_Y(0)=4$ and relation

$$X(t,H) = u \cdot X(t-1), \quad X(t,T) = d \cdot X(t-1)$$
 (1)

for u=2 and $d=\frac{1}{2}$:

$$X_Y(2, HH) = 16$$

$$X_Y(1, H) = 8$$

$$X_Y(0) = 4$$

$$X_Y(2, HT) = X_Y(2, TH) = 4$$

$$X_Y(2, TT) = 1$$

$$(2)$$

Then we calculate the probability measures

$$\mathbb{P}^{Y}(H) = \frac{1-d}{u-d} = \frac{1-\frac{1}{2}}{2-\frac{1}{2}} = \frac{1}{3},$$

$$\mathbb{P}^{Y}(T) = 1 - P^{Y}(H) = \frac{2}{3}.$$
(3)

Applying (3) to the following formula of the price of contingent claim V using Y as a reference asset for $t = \{0, 1\}$

$$V_Y(t) = V_Y(t+1, H) \cdot \mathbb{P}^Y(H) + V_Y(t+1, T) \cdot \mathbb{P}^Y(T)$$
 (4)

together with the results in (2) and facts

$$V(2) = \mathbb{1}(X_Y(2) \neq 4) \cdot Y(2), \tag{5}$$

$$Y_X(t) = \frac{1}{X_Y(t)},\tag{6}$$

we obtain the following binomial tree for the prices $V(t,\cdot)$, $t = \{0,1,2\}$:

$$V_{Y}(1,H) = \frac{1}{3}$$

$$V_{Y}(2,HH) = 1$$

$$V_{Y}(2,HT) = V(2,TH) = 0$$

$$V_{Y}(1,T) = \frac{2}{3}$$

$$V_{Y}(2,HT) = V(2,TH) = 0$$

$$V_{Y}(2,TT) = 1$$

$$(7)$$

Using the results in (7) and (2), the hedging positions are as follows:

$$\Delta^{X}(0) = \frac{V_{Y}(1, H) - V_{Y}(1, T)}{X_{Y}(1, H) - X_{Y}(1, T)} = \frac{\frac{1}{3} - \frac{2}{3}}{8 - 2} = -\frac{1}{18},$$

$$\Delta^{X}(1, H) = \frac{V_{Y}(2, HH) - V_{Y}(1, HT)}{X_{Y}(2, HH) - X_{Y}(2, HT)} = \frac{1 - 0}{16 - 4} = \frac{1}{12},$$

$$\Delta^{X}(1, T) = \frac{V_{Y}(2, TH) - V_{Y}(2, TT)}{X_{Y}(2, TH) - X_{Y}(2, TT)} = \frac{0 - 1}{4 - 1} = -\frac{1}{3}.$$
(8)

Problem 2 – Solution

At time T the contract's payoff is

$$V(T) = \mathbb{1}(L \le X_Y(T) \le U) \cdot Y(T) . \tag{9}$$

The price $V_Y(t)$ is a \mathbb{P}^Y martingale and thus it holds

$$V_Y(t) = \mathbb{E}_t^Y[\mathbb{1}(L \le X_Y(T) \le U)] = \mathbb{P}_t^Y[L \le X_Y(T) \le U] \tag{10}$$

in which the lower index t represents conditional expectation or probability, under condition $X_Y(t) = x$. Using the theory in sub-chapter 3.3 Price as an **Expectation** in the book Večeř [2011], we get

$$V_Y(t) = \mathbb{E}_t^Y [V_Y(T)] = \mathbb{E}_t^Y [f^Y(X_Y(T))]$$

= $\mathbb{E}_t^Y [\mathbb{1}(L \le X_Y(T) \le U) \cdot Y(T)] = \mathbb{E}_t^Y [\mathbb{1}(L \le X_Y(T) \le U)]$ (11)

Expressing $V^Y(t)$ in terms of price function u^Y , we obtain the following representation

$$u^{Y}(t,x) = V_{Y}(t) = \mathbb{E}_{t}^{Y}[f^{Y}(X_{Y}(T))]$$

$$= \mathbb{E}_{t}^{Y}[\mathbb{1}(L \leq X_{Y}(T) \leq U)]$$

$$= \mathbb{P}_{t}^{Y}[L \leq X_{Y}(T) \leq U]$$

$$= \mathbb{P}_{t}^{Y}[X_{Y}(T) \leq U] - \mathbb{P}_{t}^{Y}[X_{Y}(T) \leq L]$$

$$= 1 - \mathbb{P}_{t}^{Y}[X_{Y}(T) \geq U] - 1 + \mathbb{P}_{t}^{Y}[X_{Y}(T) \geq L]$$

$$= N(d_{L}-) - N(d_{U}-)$$
(12)

where in the last equation we use the relation

$$\mathbb{P}_{t}^{X}[X_{Y}(T) \ge K] = N\left(\frac{1}{\sigma\sqrt{T-t}} \cdot log\left(\frac{X_{Y}(T)}{K}\right) - \frac{1}{2}\sigma\sqrt{T-t}\right) =: N(d_{K}-) \quad (13)$$

in which $N(\cdot)$ represents a cumulative distribution function of standard normal distribution

$$N(x) = \int_{-\infty}^{x} \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}} dy \sim \mathcal{N}(0, 1) . \tag{14}$$

The hedging portfolio for this contract is of form

$$P(t) = \Delta^{X}(t) \cdot X(t) + \Delta^{Y}(t) \cdot Y(t)$$
(15)

with the following hedging positions

$$\Delta^{X}(t) = u_{x}^{Y}(t, X_{Y}(t))
= \frac{1}{X_{Y}(t)\sqrt{T-t}} \cdot (f(d_{L}-) - f(d_{U}-))
\Delta^{Y}(t) = u^{Y}(t, X_{Y}(t)) - u_{x}^{Y}(t, X_{Y}(t)) \cdot X_{Y}(t)
= N(d_{L}-) - N(d_{U}-) - \frac{1}{\sigma\sqrt{T-t}} \cdot (f(d_{L}-) - f(d_{U}-))$$
(16)

where $f(\cdot)$ is a density of standard normal distribution

$$f(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}, \quad x \in \mathbb{R} .$$
 (17)

Generally, based on the theorem 3.3. in Večeř [2011], the price function $u^Y(t,x)$ satisfies the partial differential equation

$$u_t^Y(t,x) + \frac{1}{2}\sigma^2 \cdot x^2 \cdot u_{xx}^2(t,x) = 0 , \qquad (18)$$

where for this time the lower index t represents a partial derivative. In our case this was verified by the software Wolfram Mathematica, see **Attachment A.1**.

Bibliography

J. Večeř. Stochastic Finance: A Numeraire Approach. Matfyzpress, 2011. ISBN 1-439-81252-7.

A. Attachments

A.1 Verification of Partial Differential Equation

■ Partial differential equation - verification

▼
$$ln[1]:= d[K_{-}] := 1/(\sigma * \sqrt{T - t}) * Log[x / K] - 1/2 * \sigma * \sqrt{T - t}$$
 (*value corresponding to denotation d- *)

▼ In[2]:= CumFunc =
$$\int_{-\infty}^{c} \frac{1}{\sqrt{2 \pi}} e^{\frac{-y^2}{2}}$$

 $dy \ (*definition \ of \ N(.) \ - \ cummulative \ d.f. \ of \ standard \ normal \ distribution*)$

▼ (*derivative by x of N(.) is density of N(0,1)*)

▼ In[3]:= DerivationofN[x_] =
$$1/\sqrt{2\pi}$$
 * e $\frac{-(x^2)^2}{2}$

V Out[3]=
$$\frac{e^{-\frac{x^2}{2}}}{\sqrt{2 \pi}}$$

$$\Psi_{\text{Out}[4]=} = \frac{e^{-\frac{1}{2} \left(-\frac{1}{2} \sqrt{-t+T} \ \sigma + \frac{\log \left[\frac{x}{\tau}\right]}{\sqrt{-t+T} \ \sigma}\right)^2 \left(\frac{\sigma}{4 \sqrt{-t+T}} + \frac{\log \left[\frac{x}{t}\right]}{2 \ (-t+T)^{\frac{3}{2} \sigma}\right)}}{\sqrt{2 \ \pi}} - \frac{e^{-\frac{1}{2} \left(-\frac{1}{2} \sqrt{-t+T} \ \sigma + \frac{\log \left[\frac{x}{\tau}\right]}{\sqrt{-t+T} \ \sigma}\right)^2 \left(\frac{\sigma}{4 \sqrt{-t+T}} + \frac{\log \left[\frac{x}{\tau}\right]}{2 \ (-t+T)^{\frac{3}{2} \sigma}\right)}}{\sqrt{2 \ \pi}} - \frac{e^{-\frac{1}{2} \left(-\frac{1}{2} \sqrt{-t+T} \ \sigma + \frac{\log \left[\frac{x}{\tau}\right]}{\sqrt{-t+T} \ \sigma}\right)^2 \left(\frac{\sigma}{4 \sqrt{-t+T}} + \frac{\log \left[\frac{x}{\tau}\right]}{2 \ (-t+T)^{\frac{3}{2} \sigma}\right)}}{\sqrt{2 \ \pi}} - \frac{e^{-\frac{1}{2} \left(-\frac{1}{2} \sqrt{-t+T} \ \sigma + \frac{\log \left[\frac{x}{\tau}\right]}{\sqrt{-t+T} \ \sigma}\right)^2 \left(\frac{\sigma}{4 \sqrt{-t+T}} + \frac{\log \left[\frac{x}{\tau}\right]}{2 \ (-t+T)^{\frac{3}{2} \sigma}\right)}}{\sqrt{2 \ \pi}} - \frac{e^{-\frac{1}{2} \left(-\frac{1}{2} \sqrt{-t+T} \ \sigma + \frac{\log \left[\frac{x}{\tau}\right]}{\sqrt{-t+T} \ \sigma}\right)^2 \left(\frac{\sigma}{4 \sqrt{-t+T}} + \frac{\log \left[\frac{x}{\tau}\right]}{2 \ (-t+T)^{\frac{3}{2} \sigma}}\right)}}{\sqrt{2 \ \pi}}$$

▼ (*first derivative of u^Y(t,x) by x*)

$$\blacktriangledown \ \, \mathsf{In}[S] = \ \, \mathsf{uDerByx} = \mathsf{DerivationofN}[\mathsf{d}[\mathsf{L}]] \, \star \, \mathsf{D}[\mathsf{d}[\mathsf{L}]] \, \star \, \mathsf{D}[\mathsf{d}[\mathsf{L}]] \, \star \, \mathsf{DerivationofN}[\mathsf{d}[\mathsf{U}]] \, \star \, \mathsf{D}[\mathsf{d}[\mathsf{U}]] \, \star \, \mathsf{D}[\mathsf{U}]] \, \star \, \mathsf{D}[\mathsf{d}[\mathsf{U}]] \, \star \, \mathsf{D}[\mathsf{U}] \, \mathsf{D}[\mathsf{U}] \, \mathsf{D}[\mathsf{U}] \, \mathsf{D}[\mathsf{U}]] \, \star \, \mathsf{D}[\mathsf{U}] \, \mathsf{D}[\mathsf{U}] \, \mathsf{D}[\mathsf{U}] \, \mathsf{D}[\mathsf{U}]] \, \mathsf{D}[\mathsf{U}] \, \mathsf{D}[\mathsf{U}] \, \mathsf{D}[\mathsf{U}] \, \mathsf{D}[\mathsf{U}]] \, \mathsf{D}[\mathsf{U}] \, \mathsf{D}[\mathsf{U}]] \, \mathsf{D}[\mathsf$$

$$\Psi \text{ Out}[5] = \frac{ e^{-\frac{1}{2} \left(-\frac{1}{2} \sqrt{-t+T} \ \sigma + \frac{\text{Log} \left[\frac{t}{\lambda}\right]}{\sqrt{-t+T} \ \sigma}\right)^2}}{\sqrt{2 \, \pi} \ \sqrt{-t+T} \ \sqrt{-t+T} \ x \ \sigma} - \frac{ e^{-\frac{1}{2} \left(-\frac{1}{2} \sqrt{-t+T} \ \sigma + \frac{\text{Log} \left[\frac{t}{\lambda}\right]}{\sqrt{-t+T} \ \sigma}\right)^2}}{\sqrt{2 \, \pi} \ \sqrt{-t+T} \ x \ \sigma}$$

▼ (*second derivative of u^Y(t,x) by x*)

▼ In[6]:= uDerByxx = D[uDerByx, x]

$$\begin{array}{l} \Psi \text{ Out} \text{(G)=} & -\frac{e^{-\frac{1}{2}\left(-\frac{1}{2}\sqrt{-t+T}\ \sigma + \frac{\text{Log}\left[\frac{x}{4}\right]}{\sqrt{-t+T}\ \sigma}\right)^2}}{\sqrt{2\ \pi}\ \sqrt{-t+T}\ x^2\ \sigma} + \frac{e^{-\frac{1}{2}\left(-\frac{1}{2}\sqrt{-t+T}\ \sigma + \frac{\text{Log}\left[\frac{x}{4}\right]}{\sqrt{-t+T}\ \sigma}\right)^2}}{\sqrt{2\ \pi}\ \sqrt{-t+T}\ x^2\ \sigma} - \\ & \frac{e^{-\frac{1}{2}\left(-\frac{1}{2}\sqrt{-t+T}\ \sigma + \frac{\text{Log}\left[\frac{x}{4}\right]}{\sqrt{-t+T}\ \sigma}\right)^2}\left(-\frac{1}{2}\sqrt{-t+T}\ \sigma + \frac{\text{Log}\left[\frac{x}{4}\right]}{\sqrt{-t+T}\ \sigma}\right)}{\sqrt{2\ \pi}\ \left(-t+T\right)\ x^2\ \sigma^2} + \frac{e^{-\frac{1}{2}\left(-\frac{1}{2}\sqrt{-t+T}\ \sigma + \frac{\text{Log}\left[\frac{x}{4}\right]}{\sqrt{-t+T}\ \sigma}\right)^2}\left(-\frac{1}{2}\sqrt{-t+T}\ \sigma + \frac{\text{Log}\left[\frac{x}{4}\right]}{\sqrt{-t+T}\ \sigma}\right)}{\sqrt{2\ \pi}\ \left(-t+T\right)\ x^2\ \sigma^2} \end{array}$$

▼ |n[8]| FullSimplify [uDerByt + 1 / 2 * σ ^2 * x^2 * uDerByxx] (*partial differential equation = 0 --> IT'S GOOD!*)

▼ Out[8]= 0