

Introduction to Mathematical Finance

Solution sheet 7

Solution 7.1 Assume without loss of generality that $\Omega = \{1, 2\}^T$ and $Y_k((x_1, \dots, x_T)) = 1 + y_{x_k}$, where $y_1 := d$ and $y_2 := u$.

- (a) **First method.** Any measure $Q \approx P$ on \mathcal{F}_T can be described by its transition probabilities $q_{x_1}, q_{x_1, x_2}, \dots, q_{x_1, \dots, x_T}$, where $x_1, \dots, x_k \in \{1, 2\}$ and

$$\begin{aligned} q_{x_1} &:= Q[Y_1 = 1 + y_{x_1}], \\ q_{x_1, \dots, x_k} &:= Q[Y_k = 1 + y_{x_k} \mid Y_1 = 1 + y_{x_1}, \dots, Y_{k-1} = 1 + y_{x_{k-1}}], \quad k = 2, \dots, T. \end{aligned} \quad (1)$$

Since \widehat{S}^1 is finitely valued (and hence bounded) and adapted, it is a Q -martingale if and only if for all $k = 0, \dots, T-1$, we have

$$E_Q \left[\widehat{S}_{k+1}^0 \mid \mathcal{F}_k \right] = \widehat{S}_k^0 \quad Q\text{-a.s.} \quad (2)$$

By definition of \widehat{S}^0 and since it is strictly positive, the martingale property (2) holds if and only if

$$E_Q \left[\widehat{S}_k^0 \frac{1+r}{Y_{k+1}} \mid \mathcal{F}_k \right] = \widehat{S}_k^0 \quad Q\text{-a.s.},$$

which holds if and only if

$$E_Q \left[\frac{1+r}{Y_{k+1}} \mid \mathcal{F}_k \right] = 1 \quad Q\text{-a.s.}$$

Note that we do not know a priori whether the Y_k are independent under Q . Since $\mathcal{F}_0 = \{\emptyset, \Omega\}$, $\mathcal{F}_k = \sigma(Y_1, \dots, Y_k)$ for $k = 1, \dots, T$, and Y_k only takes two values, \widehat{S}^0 is a Q -martingale if and only if $E_Q \left[\frac{1+r}{Y_1} \right] = 1$ and for all $k \in \{1, \dots, T-1\}$ and all $x_1, \dots, x_k \in \{1, 2\}$ we have

$$E_Q \left[\frac{1+r}{Y_{k+1}} \mid Y_1 = 1 + y_{x_1}, \dots, Y_k = 1 + y_{x_k} \right] = 1. \quad (3)$$

Now, we have

$$\begin{aligned} E_Q \left[\frac{1+r}{Y_1} \right] = 1 &\Leftrightarrow \frac{1+r}{1+d}q_1 + \frac{1+r}{1+u}(1-q_1) = 1 \\ &\Leftrightarrow \left(\frac{1+u}{1+d} - 1 \right) q_1 = \frac{1+u}{1+r} - 1 \\ &\Leftrightarrow \frac{u-d}{1+d}q_1 = \frac{u-r}{1+r} \\ &\Leftrightarrow q_1 = \frac{1+d}{1+r} \frac{u-r}{u-d}. \end{aligned} \quad (4)$$

Similarly, for all $k \in \{1, \dots, T-1\}$ and all $x_1, \dots, x_k \in \{1, 2\}$, we have

$$E_Q \left[\frac{1+r}{Y_{k+1}} \mid Y_1 = 1 + y_{x_1}, \dots, Y_k = 1 + y_{x_k} \right] = 1 \Leftrightarrow q_{x_1, \dots, x_k, 1} = \frac{1+d}{1+r} \frac{u-r}{u-d}.$$

Note that $q_{x_1, \dots, x_k, 1}$ does not depend on x_1, \dots, x_k and equals q_1 . Hence, we may conclude that there exists a unique equivalent martingale measure Q^{**} for \widehat{S}^0 , under which Y_1, \dots, Y_k are i.i.d. and we have

$$Q^{**}[Y = 1 + d] = \frac{1 + d u - r}{1 + r u - d} =: q_1^{**} \quad \text{and} \quad Q^{**}[Y = 1 + u] = \frac{1 + u r - d}{1 + r u - d} =: q_2^{**}.$$

Second method: For $k = 1, \dots, T$, set $\widehat{Y}_k = \frac{1+r}{Y_k}$. Then $\widehat{S}_k^1 = 1^k = 1$ and $\widehat{S}_k^0 = \prod_{j=1}^k \widehat{Y}_j$ for $k = 0, \dots, T$, where the \widehat{Y}_k are independent under P and take the two values $1 + \hat{u}$ and $1 + \hat{d}$ with probability p_1 and p_2 , respectively, where $\hat{u} = \frac{r-d}{1+d} > 0$ and $\hat{d} = \frac{r-u}{1+u} < 0$. In conclusion, $(\widehat{S}^0, \widehat{S}^1)$ can be viewed as a binomial model with $\hat{u} > \hat{r} = 0 > \hat{d}$.

Variante (i): Recalling that \hat{u} corresponds to p_1 and \hat{d} to p_2 , it follows that the unique equivalent martingale measure Q^{**} for \widehat{S}^1 is given by

$$Q^{**}[\{(x_1, \dots, x_T)\}] := \prod_{k=1}^T q_{x_k}^{**}, \quad x_1, \dots, x_T \in \{1, 2\}, \quad (5)$$

where

$$\begin{aligned} q_1^{**} &= \frac{\hat{r} - \hat{d}}{\hat{u} - \hat{d}} = \frac{\frac{u-r}{1+u}}{\frac{r-d}{1+d} - \frac{r-u}{1+u}} = \frac{1 + d u - r}{1 + r u - d} \\ q_2^{**} &= \frac{\hat{u} - \hat{r}}{\hat{u} - \hat{d}} = \frac{\frac{r-d}{1+d}}{\frac{r-d}{1+d} - \frac{r-u}{1+u}} = \frac{1 + u r - d}{1 + r u - d}. \end{aligned} \quad (6)$$

Variante (ii): We are looking for the unique strictly positive Q^* -martingale $Z^{Q^{**}; Q^*}$ starting at 1 such that

$$\frac{\widetilde{S}^0}{\widetilde{S}^1} Z^{Q^{**}; Q^*} \text{ is a } Q^*\text{-martingale.}$$

But we already know that $\frac{\widetilde{S}^1}{\widetilde{S}^0}$ is a Q^* -martingale and strictly positive. Hence by uniqueness,

$$Z^{Q^{**}; Q^*} = \frac{\widetilde{S}^1}{\widetilde{S}^0} = S^1.$$

(b) The unique equivalent martingale measure Q^* for $S^1 = \frac{\widetilde{S}^1}{\widetilde{S}^0}$ on \mathcal{F}_T is given by

$$Q^*[\{(x_1, \dots, x_T)\}] := \prod_{j=1}^T q_{x_j}^*, \quad (7)$$

where $q_1^* = \frac{u-r}{u-d}$ and $q_2^* = \frac{r-d}{u-d}$. By part (a), we have

$$\frac{q_1^{**}}{q_1^*} = \frac{1+d}{1+r} \quad \text{and} \quad \frac{q_2^{**}}{q_2^*} = \frac{1+u}{1+r}. \quad (8)$$

Thus, since Ω is finite, we may deduce that for all $(x_1, \dots, x_T) \in \{1, 2\}^T$ we have

$$\begin{aligned} \frac{dQ^{**}}{dQ^*}((x_1, \dots, x_T)) &= \frac{Q^{**}[\{(x_1, \dots, x_T)\}]}{Q^*[\{(x_1, \dots, x_T)\}]} = \prod_{k=1}^T \frac{q_{x_k}^{**}}{q_{x_k}^*} \\ &= \frac{\prod_{k=1}^T (1 + y_{x_k})}{(1+r)^T} = \frac{\prod_{k=1}^T Y_k((x_1, \dots, x_T))}{(1+r)^k} \\ &= \frac{\widetilde{S}_T^1((x_1, \dots, x_T))}{\widetilde{S}_T^0((x_1, \dots, x_T))}. \end{aligned} \quad (9)$$

- (c) Denote by Z the density process of Q^{**} with respect to Q^* . Let $k \in \{1, \dots, T\}$ and $\tilde{H} \in L_+^0(\mathcal{F}_T)$. Since $Z_T = \frac{\tilde{S}_T^1}{S_T^1} = S_T^1$ by part (b) and S^1 is a Q^* -martingale, it follows by the definition of a density process that for $k = 0, \dots, T$ we have

$$Z_k = E_{Q^*}[Z_T | \mathcal{F}_k] = E_{Q^*}[S_T^1 | \mathcal{F}_k] = S_k^1 = \frac{\tilde{S}_k^1}{\tilde{S}_k^0} \quad Q^*\text{-a.s.} \quad (10)$$

Thus, by the Bayes formula we get for $k = 0, \dots, T$

$$\tilde{S}_k^1 E_{Q^{**}} \left[\frac{\tilde{H}}{\tilde{S}_T^1} \middle| \mathcal{F}_k \right] = \frac{\tilde{S}_k^1}{Z_k} E_{Q^*} \left[\frac{Z_T \tilde{H}}{\tilde{S}_T^1} \middle| \mathcal{F}_k \right] = \tilde{S}_k^0 E_{Q^*} \left[\frac{\tilde{H}}{\tilde{S}_T^0} \middle| \mathcal{F}_k \right]. \quad (11)$$

Solution 7.2 We use the following notations:

$$S_0^1 = 80, \quad \tilde{S}_1^1 = \begin{cases} 80(1 + y_1) & \text{with probability } p_1 = 0.2 \\ 80(1 + y_2) & p_2 = 0.3 \\ 80(1 + y_3) & p_3 = 0.5 \end{cases},$$

with $y_1 = \frac{1}{2}, y_2 = \frac{1}{8}, y_3 = -\frac{1}{4}$. Let $q_i = \mathbb{Q}[\{\tilde{S}_1^1 = s_0(1 + y_i)\}]$.

- (a) We work on the (finite) path space Ω and use the filtration generated by the price process. S^1 is therefore adapted and integrable, it is then a martingale under \mathbb{Q} if and only if $\mathbb{E}_{\mathbb{Q}}[S_1^1] = S_0^1$.

$$\begin{aligned} \mathbb{E}_{\mathbb{Q}}[S_1^1] = S_0^1 &\Leftrightarrow \mathbb{E}_{\mathbb{Q}}[\tilde{S}_1^1] = S_0^1(1 + r) \\ \mathbb{Q} \text{ is an EMM} &\Leftrightarrow \begin{cases} 120q_1 + 90q_2 + 60q_3 = 80 \cdot 1.05 \\ q_1 + q_2 + q_3 = 1 \\ 0 < q_1, q_2, q_3 < 1 \end{cases} \\ &\Leftrightarrow \begin{cases} q_2 = 0.8 - 2q_1 \\ q_3 = 0.2 + q_1 \\ q_1 \in (0, 0.4) \end{cases} \end{aligned}$$

The set of all equivalent martingale measures is given by

$$P_e(S) = \{\mathbb{Q}_\lambda = (\lambda, 0.8 - 2\lambda, 0.2 + \lambda) \mid \lambda \in (0, 0.4)\}.$$

The set of all arbitrage-free prices is given by (we can use the risk-neutral valuation formula)

$$\begin{aligned} C &= \{c_\lambda := \mathbb{E}_{\mathbb{Q}_\lambda} \left[\frac{(\tilde{S}_1^1 - 80)^+}{1 + r} \right] \mid \lambda \in (0, 0.4)\} \\ &= \{c_\lambda = \frac{1}{1.05} (40 \cdot \lambda + 10(0.8 - 2\lambda)) \mid \lambda \in (0, 0.4)\} \\ &= \{c_\lambda = \frac{1}{1.05} (20\lambda + 8) \mid \lambda \in (0, 0.4)\}. \end{aligned}$$

This set is the whole open interval (7.619, 15.238).

- (b) $\tilde{H} \in L_+^0$ can be replicated if there exists an admissible self-financing strategy $\varphi = (\varphi^0, \vartheta)$ such that

$$\varphi_1^0(1 + r) + \vartheta_1 \tilde{S}_1^1 = \tilde{H}.$$

Let \tilde{H}^{y_i} be the value of the payoff if $\tilde{S}_1^1 = s_0(1 + y_i)$, we are looking for non trivial solutions of the following system:

$$\begin{bmatrix} 1.05 & 120 \\ 1.05 & 90 \\ 1.05 & 60 \end{bmatrix} \cdot \begin{bmatrix} \varphi_1^0 \\ \vartheta_1 \end{bmatrix} = \begin{bmatrix} \tilde{H}^{y_1} \\ \tilde{H}^{y_2} \\ \tilde{H}^{y_3} \end{bmatrix}.$$

This system admits non trivial solutions if and only if

$$\det \begin{bmatrix} 1.05 & 120 & \tilde{H}^{y_1} \\ 1.05 & 90 & \tilde{H}^{y_2} \\ 1.05 & 60 & \tilde{H}^{y_3} \end{bmatrix} = 0$$

In order to be attainable a contingent claim must satisfy

$$-\tilde{H}^{y_1} + 2\tilde{H}^{y_2} - \tilde{H}^{y_3} = 0$$

(c) It's easy to check that our call option doesn't satisfy the previous equation.

Solution 7.3

(a) Define the processes by

$$M_k = \sum_{j=1}^k (X_j - E[X_j | \mathcal{F}_{j-1}])$$

and

$$A_k = \sum_{j=1}^k (E[X_j | \mathcal{F}_{j-1}] - X_{j-1}).$$

These processes are well-defined since X is integrable. Note that both processes are integrable, M is adapted, and A is predictable. Furthermore,

$$E[\Delta M_{k+1} | \mathcal{F}_k] = E[X_{k+1} - E[X_{k+1} | \mathcal{F}_k] | \mathcal{F}_k] = 0 \quad P\text{-a.s.}$$

Hence, M is a martingale.

(b) Suppose that also M' and A' satisfy $X = X_0 + M' + A' = X_0 + M + A$. Then $M' - M = A - A'$. Denote this process by Y and observe that it is both a martingale and predictable. Thus,

$$Y_{k-1} = E[Y_k | \mathcal{F}_{k-1}] = Y_k \quad P\text{-a.s.},$$

hence Y is constant. Since $M_0 = M'_0 = 0$ P -a.s., $Y = 0$ P -a.s., showing that the decomposition is almost surely unique.

(c) Suppose X is a supermartingale. Then, for every $k \in \mathbb{N}_0$,

$$0 \geq E[\Delta X_{k+1} | \mathcal{F}_k] = E[\Delta M_{k+1} + \Delta A_{k+1} | \mathcal{F}_k] = \Delta A_{k+1} \quad P\text{-a.s.},$$

where we use that M is a martingale and A is predictable. This shows that $A_{k+1} \leq A_k$ P -a.s.

For the converse, suppose $\Delta A_{k+1} \leq 0$ P -a.s. Then,

$$E[\Delta X_{k+1} | \mathcal{F}_k] = E[\Delta M_{k+1} + \Delta A_{k+1} | \mathcal{F}_k] \leq E[\Delta M_{k+1} | \mathcal{F}_k] = 0 \quad P\text{-a.s.}$$

for all $k \in \mathbb{N}_0$. We thus conclude that X is a supermartingale.

Solution 7.4

```

1 def trinomial_price(maturity, spot, strike, rate, vol, steps_number, payoff_fct
   =None, graph_name=None):
2     """Compute the trinomial price. Draw graph if graph_name is given.
3     """
4     deltaT = maturity / steps_number
5     discount_factor = exp(-rate * deltaT)

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6 up = exp(vol * sqrt(2*deltaT))
7 down = 1 / up
8 denominator = exp(vol * sqrt(deltaT/2)) - exp(-vol * sqrt(deltaT/2))
9 proba_up = ((exp(rate * deltaT/2) - exp(-vol * sqrt(deltaT/2))) / denominator
10 ) ** 2
11 proba_down = ((exp(vol * sqrt(deltaT/2)) - exp(rate * deltaT/2)) / denominator
12 ) ** 2
13 proba_middle = 1 - proba_up - proba_down
14 steps = range(steps_number)
15 spot_prices = [spot * up ** i for i in reversed(steps[1:])] + [spot] + [spot
16 * down ** i for i in steps[1:]]
17 option_prices = [payoff_fct(spot_price, strike) for spot_price in spot_prices
18 ]
19
20 # The following two list are only needed to display the graph:
21 spot_prices_history = [spot_prices]
22 option_prices_history = [option_prices]
23
24 def next_option_price(spot_price, price_up, price_midlle, price_down):
25     option_price = (discount_factor *(proba_up * price_up +
26                                     proba_middle * price_midlle +
27                                     proba_down * price_down))
28
29 while len(option_prices) > 1:
30     option_prices = [next_option_price(spot_prices[i], *option_prices[i-1:i+2])
31                     for i in range(1, len(option_prices)-1)]
32     spot_prices = spot_prices[1:-1]
33     spot_prices_history.insert(0, spot_prices)
34     option_prices_history.insert(0, option_prices)
35
36 if graph_name:
37     create_graph(graph_name, spot_prices_history, option_prices_history)
38
39 return option_prices[0]

```