MTH 9831. LECTURE 10

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ABSTRACT. This lecture continues the discussion of Asian options and then turns to American options.

- 1. Asian options (continued: reduction of dimension through the change of numeraire)
- 2. American options. Part 1: perpetual American put.

1. Asian options

Recall the BSM framework:

$$dS(t) = rS(t)dt + \sigma S(t)d\widetilde{B}(t), \quad 0 \le t \le T.$$

$$dD(t) = -rD(t)dt, \quad D(0) = 1, \quad \text{i.e. } D(t) = e^{-rt}, \quad 0 \le t \le T.$$

Here \widetilde{B} is a standard BM with respect to the risk-neutral measure $\widetilde{\mathbb{P}}$. Previously we derived the following equation:

(1)
$$d(D(t)S(t)) = \sigma D(t)S(t)d\widetilde{B}(t), \quad 0 \le t \le T.$$

The time t value of an option with the payoff V(T) is

$$V(t) = \widetilde{\mathbb{E}}(e^{-r(T-t)}V(T) \mid \mathcal{F}(t)), \quad 0 \le t \le T,$$

and

$$e^{-rt}V(t) = \widetilde{\mathbb{E}}(e^{-rT}V(T) \mid \mathcal{F}(t)), \quad 0 \le t \le T,$$

is a $\widetilde{\mathbb{P}}$ -martingale. We assume that $r > 0^1$ and $V(T) = (\frac{1}{c} \int_{T-c}^T S(t) dt - K)_+$ where c is a fixed constant, $0 < c \le T$. The pricing will be done in 2 steps.

Step 1: Create a portfolio process X(t) such that

(2)
$$X(T) = \frac{1}{c} \int_{T-c}^{T} S(t)dt - K.$$

Step 2: Use 1 share of underlying as a unit, reduce the dimension, and set up a simpler PDE and boundary conditions than those in Lecture 9.

Step 1. Let $X(t) = \gamma(t)S(t) + (X(t) - \gamma(t)S(t))$, where $\gamma(t)$ is non-random and differentiable. Our goal is to find $\gamma(t)$, $0 \le t \le T$, such that X(T) matches (2).

First of all, recall that under the self-financing condition

$$dX(t) = \gamma(t)dS(t) + r(X(t) - \gamma(t)S(t))dt.$$

Discounting, we get the equation

$$d(D(t)X(t)) = -rD(t)X(t)dt + D(t)dX(t) = D(t)\gamma(t)dS(t) - rD(t)\gamma(t)S(t)dt.$$

Expressing² $D(t)\gamma(t)dS(t)$ as

$$d(D(t)\gamma(t)S(t)) - S(t)d(D(t)\gamma(t)) = d(D(t)\gamma(t)S(t)) + rD(t)S(t)\gamma(t)dt - D(t)S(t)d\gamma(t)$$

and canceling $\pm rD(t)\gamma(t)S(t)dt$ terms we arrive at

(3)
$$d(D(t)X(t)) = d(D(t)\gamma(t)S(t)) - D(t)S(t)d\gamma(t).$$

¹See Exercise 7.8 in the textbook for the case r = 0.

 $^{^2}D(t)\gamma(t)$ is a regular process, so the usual product rule applies.

Observe for the future use that, (3) and (1) imply

(4)
$$d(X(t)D(t)) = \gamma(t)\sigma D(t)S(t)d\widetilde{B}(t).$$

Next we integrate (3) from T-c to T and put $D(t)=e^{-rt}$:

$$e^{-rT}X(T) - e^{-r(T-t)}X(T-c) = e^{-rT}\gamma(T)S(T) - e^{-r(T-c)}\gamma(T-c)S(T-c) - \int_{T-c}^{T} e^{-rt}S(t)\gamma'(t)dt;$$

$$X(T) = e^{rc}X(T-c) + \gamma(T)S(T) - e^{rc}\gamma(T-c)S(T-c) - \frac{1}{c} \int_{T-c}^{T} S(t)e^{-r(t-T)}c\gamma'(t)dt.$$

To get a match with the expression in (2) we should set

(5)
$$-c\gamma'(t)e^{-r(T-t)} = 1, \quad \text{for all } t \in [T-c, T];$$

(6)
$$e^{rc}X(T-c) + \gamma(T)S(T) - e^{rc}\gamma(T-c)S(T-c) = -K.$$

Solving (5) we get

(7)
$$\gamma(t) = \gamma(T-c) - \frac{1}{rc}(e^{-r(T-t)} - e^{-rc}), \ t \in [T-c, T].$$

At time T-c, we know X(T-c) and S(T-c), but we can not deterministically control S(T). Thus, the only way (6) can hold is when $\gamma(T)=0$ and (6) becomes

(8)
$$e^{rc}X(T-c) - e^{rc}\gamma(T-c)S(T-c) = -K.$$

The condition $\gamma(T) = 0$ and (7) lead to

$$0 = \gamma(T) = \gamma(T - c) - \frac{1}{rc}(1 - e^{-rc}),$$
$$\gamma(T - c) = \frac{1}{rc}(1 - e^{-rc}),$$

and from (8),

$$X(T-c) = \frac{1}{rc}(1 - e^{-rc})S(T-c) - Ke^{-rc}.$$

The last equation corresponds to the following position at time T-c:

- long $\frac{1}{rc}(1-e^{-rc})$ shares of stock;
- short Ke^{-rT} shares of MMA (the cost of 1 share of MMA at time T-c is $e^{r(T-c)}$).

What should we do at time 0 to ensure that at time T-c? we shall have this position? If at time 0 we purchase $\frac{1}{rc}(1-e^{-rc})$ shares by borrowing $e^{-rT}K$ from a bank and just simply hold this portfolio up to time T-c, then we shall get what we want. Such portfolio at time 0 costs

$$X(0) = \frac{1}{rc}(1 - e^{-rc})S(0) - e^{-rT}K.$$

Our hedging strategy is as follows: at time 0 start with X(0) in cash and keep over time the portfolio that has

(1.1)
$$\gamma(t) = \begin{cases} \frac{1}{rc} (1 - e^{-rc}), & \text{if } 0 \le t \le T - c; \\ \frac{1}{rc} (1 - e^{-r(T-t)}), & \text{if } T - c \le t \le T. \end{cases}$$

shares of stock borrowing/depositing money from/to MMA as needed. Then at time T, we have (2).

Step 2. Now back to pricing. The payoff is $V(T) = (X(T))_+$ where X(T) is given by (2). Under $\widetilde{\mathbb{P}}$, the price is

$$V(t) = e^{-r(T-t)}\widetilde{\mathbb{E}}((X(T))_{+}|\mathcal{F}(t)).$$

Our tool is "change of numeraire". Define $Y(t) = \frac{X(t)}{S(t)}$. This means that we use one unit of stock as numeraire. Recall (1) and (4):

$$\begin{split} d(D(t)X(t)) &= \gamma(t)\sigma D(t)S(t)d\widetilde{B}(t) = \frac{\gamma(t)\sigma}{Y(t)}D(t)X(t)d\widetilde{B}(t);\\ d(D(t)S(t)) &= \sigma D(t)S(t)d\widetilde{B}(t). \end{split}$$

By Theorem 3.3 of Lecture 7, we have that

$$dY(t) = Y(t) \left(\frac{\gamma(t)\sigma}{Y(t)} - \sigma \right) d\widetilde{B}^S(t) = \sigma(\gamma(t) - Y(t)) d\widetilde{B}^S(t),$$

where $\widetilde{B}^{S}(t) = \widetilde{B}(t) - \sigma t$ is a BM under $\widetilde{\mathbb{P}}^{S}$, and

$$\widetilde{\mathbb{P}}^{S}(A) = \frac{1}{S(0)} \int_{A} D(T)S(T)d\widetilde{\mathbb{P}}, \quad A \in \mathcal{F}(T).$$

Recall that Radon-Nikodým derivative process

$$Z(t) = \frac{S(t)D(t)}{S(0)}, \quad 0 \le t \le T.$$

Since Y(t), $0 \le t \le T$, satisfies $dY(t) = \sigma(\gamma(t) - Y(t))d\widetilde{B}^S(t)$, it is a Markov process and a martingale under $\widetilde{\mathbb{P}}^S$. We have

$$\begin{split} V(t) &= e^{rt} \widetilde{\mathbb{E}} \left(e^{-rT} (X(T))_{+} \mid \mathcal{F}(t) \right) \\ &= e^{rt} \widetilde{\mathbb{E}} \left(\underbrace{e^{-rT} S(T)}_{=S(0)Z(T)} \left(\frac{e^{-rT} X(T)}{e^{-rT} S(T)} \right)_{+} \mid \mathcal{F}(t) \right) \\ &= e^{rt} S(0) \widetilde{\mathbb{E}} (Z(T) (Y(T))_{+} \mid \mathcal{F}(t)) \\ &\stackrel{\text{Lec. 5, Lem. 5.4}}{=} e^{rt} S(0) Z(t) \widetilde{\mathbb{E}}^{S} ((Y(T))_{+} \mid \mathcal{F}(t)) \\ &= S(t) \widetilde{\mathbb{E}}^{S} ((Y(T))_{+} \mid \mathcal{F}(t)). \end{split}$$

Since Y is a Markov process, there is g(t, y) such that

$$g(t, Y(t)) = \widetilde{\mathbb{E}}^S((Y(T))_+ \mid \mathcal{F}(t)), \quad 0 \le t \le T.$$

In particular,

$$g(T, Y(T)) = \widetilde{\mathbb{E}}^{S}((Y(T))_{+} \mid \mathcal{F}(T)) = (Y(T))_{+}.$$

and

(9)
$$g(T,y) = y_+, \quad y \in \mathbb{R}.$$

(Recall that $Y(T) = \frac{X(T)}{S(T)}$, where X(T) can be positive or negative.) As usual, by the tower property, g(t, Y(t)) is a martingale under $\widetilde{\mathbb{P}}$. Compute

$$dg(t, Y(t)) = \left(g_t(t, Y(t)) + \frac{1}{2}\sigma^2(\gamma(t) - Y(t))^2 g_{yy}(t, Y(t))\right) dt + \sigma(\gamma(t) - Y(t))g_y(t, Y(t)) d\tilde{B}^S(t).$$

We conclude that g(t, y) has to solve

(10)
$$g_t(t,y) + \frac{1}{2}\sigma^2(\gamma(t) - y)^2 g_{yy}(t,y) = 0, \quad 0 \le t < T, \quad y \in \mathbb{R}.$$

We have so far one boundary condition (9), we need conditions at infinity to ensure uniqueness. If $y \to -\infty$, then it means that Y(t) is very negative, so the chances that $(Y(T))_+ > 0$ are going to 0. Thus, in the limit $\widetilde{\mathbb{E}}^S((Y(T))_+ | \mathcal{F}(t)) = 0$, and

(11)
$$\lim_{y \to -\infty} g(t, y) = 0, \quad 0 \le t \le T.$$

If $y \to +\infty$, then the chances that $(Y(T))_+ > 0$ are going to 1, and the limit of $\widetilde{\mathbb{E}}^S((Y(T))_+|\mathcal{F}(t))$ as $y \to \infty$ is approximately the same as that of $\widetilde{\mathbb{E}}^S(Y(T)|\mathcal{F}(t)) = Y(t)$. This gives

(12)
$$\lim_{y \to \infty} (g(t, y) - y) = 0, \quad 0 \le t \le T.$$

We have shown the following result.

Theorem 1.1 (Večeř). For $0 \le t \le T$, the price V(t) at time t of the Asian call option paying $\left(\frac{1}{c}\int_{T-c}^{T}S(t)dt-K\right)_{\perp}$ at time T is given by

$$V(t) = S(t)g\left(t, \frac{X(t)}{S(t)}\right),$$

where g(t, y) satisfies (10) with boundary conditions (9), (11), (12), and X(t) is a portfolio value process with $\gamma(t)$ given by (1.1).

Remark 1.2. Exactly the same steps can be used to price a discretely sampled Asian call with payoff

$$V(T) = \left(\frac{1}{m} \sum_{i=1}^{m} S(t_i) - K\right)_{+}, \quad 0 < t_1 < \dots < t_m = T.$$

See pp. 329–330 of the textbook.

2. Perpetual American put

We shall need the following notions and facts:

(a) Stopping time; first passage time.

Fact: the first passage time of a continuous process is stopping time, i.e.

$$\forall m \in \mathbb{R}, \quad \tau_m := \inf\{t \ge 0 : \ X(t) = m\}$$

is a stopping time with respect to the natural filtration of X. Proof is a required reading: see Shreve II, p. 342

- (b) Optional stopping theorem, Shreve II, p. 342.
- (c) Laplace transform of the first passage time for the drifted Brownian Motion (see HW2 or Shreve II, Theorem 8.3.2). More precisely, let m > 0, $\mu \in \mathbb{R}$, $X(t) = B(t) + \mu t$, and $\tau_m = \inf\{t \geq 0 : X(t) = m\}$. Then $\forall \lambda > 0$,

(13)
$$\mathbb{E}e^{-\lambda\tau_m} = e^{m\mu - m\sqrt{\mu^2 + 2\lambda}}$$

Assume the BSM model: under $\widetilde{\mathbb{P}}$ (risk-neutral)

$$dS(t) = rS(t)dt + \sigma S(t)d\widetilde{B}(t), \quad r, \sigma > 0 \text{ (constants)}.$$

Consider a perpetual American put with strike K. Denote by $v_*(x)$ its price when S(0) = x. How to determine $v_*(x)$?

Notice that the decision to exercise the put at time t can only depend on the information up to time t. This tells us that the exercise time should be a stopping time. When is it optimal to exercise? When the risk-neutral expectation of the present value of the payoff is maximal, i.e. when τ (the exercise time) is such that $\widetilde{\mathbb{E}}(e^{-\tau r}(K-S(\tau)))$ is maximized.³ Maximized over which variable(s)? Over all possible stopping times. Thus, we define

$$v_*(x) := \max_{\tau \in \mathcal{T}} \widetilde{\mathbb{E}} \left(e^{-\tau r} (K - S(\tau)) \right),$$

where \mathcal{T} is the set of all stopping times relative to the natural filtration generated by $(\tilde{B}(t))_{t\leq 0}$. This formula does not seem to help much, since we do not know how to maximize over all stopping times. How does one even start describing all possible stopping times?

³We can drop the subscript +, because the option will not be exercises unless $K - S(\tau) > 0$.

This is where the "perpetuity" simplifies the problem. Since the option never expires, the decision to exercise should not depend on time to expiration (it is always ∞), it can only depend on the path. By the Markov property, if we are given that S(t)=x, the future and the past of the process S relative to time t are independent. Then it seems reasonable to seek the optimal exercise time as the first time when the price falls to some level $L_* < K$. Then our problem is reduced to finding this optimal level L_* and setting the optimal exercise time τ_* to be the first time when S hits level L_* .

Let
$$\tau_L = \min\{t \le 0 : S(t) = L\}, \ 0 < L < K, \text{ and if } S(0) \ge L$$

$$v_L(S(0)) := \widetilde{\mathbb{E}}(e^{-\tau_L r}(K - S(\tau_L))).$$

But $S(\tau_L) = L$, and for 0 < L < K and $S(0) \ge L$,

$$v_L(S(0)) = (K - L) \widetilde{\mathbb{E}} e^{-\tau_L r}.$$

If S(0) < L, then at time 0, the stock is already below L, and the option should be exercised immediately. Thus

$$v_L(S(0)) = \begin{cases} K - S(0), & \text{if } S(0) < L; \\ (K - L)\widetilde{\mathbb{E}}e^{-\tau_L r}, & \text{if } S(0) \ge L. \end{cases}$$

When S(0) = L, $\tau_L = 0$, the function $v_L(S(0))$ is continuous at S(0) = L, and

$$v_L(S(0)) = \begin{cases} K - S(0), & \text{if } S(0) \le L; \\ (K - L)\widetilde{\mathbb{E}}e^{-\tau_L r}, & \text{if } S(0) \ge L. \end{cases}$$

Lemma 2.1.

$$v_L(x) = \begin{cases} K - x, & \text{if } x \in [0, L]; \\ (K - L) \left(\frac{x}{L}\right)^{-\frac{2r}{\sigma^2}}, & \text{if } x \ge L. \end{cases}$$

Proof. We have already discussed the case $S(0) = x \in [0, L]$. Assume now that x > L, i.e. S(0) > L. Then τ_L is the first time when

$$S(t) = xe^{\sigma \tilde{B}(t) + (r - \frac{1}{2}\sigma^2)t}$$

reaches level L. Hence, S(t) = L iff

$$\begin{split} \sigma \widetilde{B}(t) + \left(r - \frac{1}{2}\sigma^2\right)t &= \log\frac{L}{x};\\ \underline{\widetilde{B}}(t) + \frac{r - \frac{1}{2}\sigma^2}{\sigma}t &= \underbrace{\frac{1}{\sigma}\log\frac{L}{x}}_{\text{drifted BM}}. \end{split}$$
 level m , but $m < 0$ as $x > L$

This does not fit into the assumptions of (13). The change of sign will help.

$$\underbrace{-\widetilde{B}(t)}_{\text{still BM}} + \underbrace{\left(\frac{1}{2}\sigma - \frac{r}{\sigma}\right)t}_{\mu \in \mathbb{R}} = \underbrace{\frac{1}{\sigma}\log\frac{x}{L}}_{=m>0}.$$

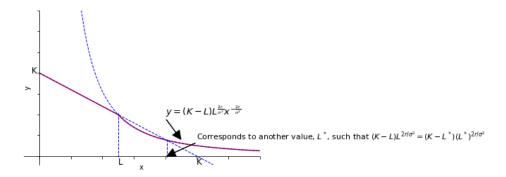
Using (13) with $\lambda = r$, $\mu = \frac{1}{2}\sigma - \frac{r}{\sigma}$, and $m = \frac{1}{\sigma}\log\frac{x}{L}$, we get

$$\begin{split} \widetilde{\mathbb{E}}e^{-\tau_L r} &= e^{m\mu - m\sqrt{\mu^2 + 2r}}; \\ \mu^2 + 2r &= \frac{\sigma^2}{4} - r + \frac{r^2}{\sigma^2} + 2r = \left(\frac{\sigma}{2} + \frac{r}{\sigma}\right)^2; \\ m\mu - m\sqrt{\mu^2 + 2r} &= \left(\frac{1}{\sigma}\log\frac{x}{L}\right)\left(\frac{1}{2}\sigma - \frac{r}{\sigma} - \frac{\sigma}{2} - \frac{r}{\sigma}\right) = -\frac{2r}{\sigma^2}\log\frac{x}{L}; \\ \widetilde{\mathbb{E}}e^{-\tau_L r} &= \left(\frac{x}{L}\right)^{-\frac{2r}{\sigma^2}}. \end{split}$$

Now we have to find out which level L^* is optimal, i.e. for which L the function $v_L(x)$ is the largest. It could be that for different x, we get different L^* , but, as we shall see, this does not happen.

$$v_L(x) = \left\{ \begin{array}{ll} K-x, & \text{if } x \in [0,L]; \\ (K-L)L^{\frac{2r}{\sigma^2}}x^{-\frac{2r}{\sigma^2}}, & \text{if } x \geq L. \end{array} \right.$$

Set $g(L) = (K - L)L^{2r/\sigma^2}$ and maximize it over $L \in [0, K]$.



$$g'(L) = \frac{2r}{\sigma^2} K L^{\frac{2r}{\sigma^2} - 1} - \left(\frac{2r}{\sigma^2} + 1\right) L^{\frac{2r}{\sigma^2}}$$

$$= L^{\frac{2r}{\sigma^2} - 1} \left(\frac{2r}{\sigma^2} K - \frac{2r + \sigma^2}{\sigma^2} L\right) = 0;$$

$$L_* = \frac{2rK}{2r + \sigma^2} \in (0, K) \quad (r, \sigma > 0).$$

Since g(0) = g(K) = 0, $g(L) \ge 0$, and this is the only critical point, it is the point where g(L) attains the absolute maximum on [0, K]. Putting this L_* in the equation for v_L , we get

$$\begin{split} v_{L_*}(x) &= \left\{ \begin{array}{ll} K-x, & \text{if } x \in [0,L_*]; \\ (K-L_*) \left(\frac{x}{L_*}\right)^{-\frac{2r}{\sigma^2}}, & \text{if } x \geq L_*. \end{array} \right. \\ v'_{L_*}(x) &= \left\{ \begin{array}{ll} -1, & \text{if } x \in [0,L_*]; \\ -(K-L_*) \frac{2r}{\sigma^2} \frac{1}{L_*} \left(\frac{x}{L_*}\right)^{-\frac{2r}{\sigma^2}-1}, & \text{if } x \geq L_*. \end{array} \right. \\ v'_{L_*}(L_{*+}) &= -\frac{(K-L_*)}{L_*} \frac{2r}{\sigma^2} = -\frac{2rK}{L_*\sigma^2} + \frac{2r}{\sigma^2} = -1. \end{split}$$

Conclusion: $v_{L_*}(x)$ is a continuously differentiable function on $[0, \infty)$, and $v'_{L_*}(L_*) = -1$. It is obvious that $v''_{L_*}(x)$ can not be continuous and will have a jump at $x = L_*$.

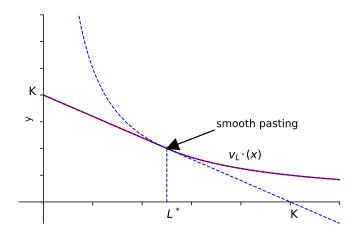
$$v_{L_*}''(x) = \begin{cases} 0 & x \in [0, L_*] \\ (K - L_*) \frac{2r}{\sigma^2} (\frac{2r}{\sigma^2} + 1) \frac{1}{L_*^2} (\frac{x}{L_*})^{-\frac{2r}{\sigma^2} - 2} & x \ge L_* \end{cases}$$

$$v_{L_*}''(L_{*+}) > 0$$

For $x > L_*$, it can be verified by substitution that

$$(14) rv_{L_*}(x) - rxv'_{L_*}(x) - \frac{1}{2}\sigma^2 x^2 v''_{L_*}(x) = (K - L_*) \left(r + \frac{2r^2}{\sigma^2} - \frac{r(2r + \sigma^2)}{\sigma^2}\right) \left(\frac{x}{L_*}\right)^{-\frac{2r}{\sigma^2}} = 0.$$
 For $x \in [0, L_*]$,

(15)
$$rv_{L_*}(x) - rxv'_{L_*}(x) - \frac{1}{2}\sigma^2 x^2 v''_{L_*}(x) = r(K - x) + rx = rK.$$



This implies that $v_{L_*}(x)$ satisfies "linear complementarity conditions"

$$(16) v(x) \ge (K - x)_+;$$

(17)
$$rv(x) - rxv'(x) - \frac{1}{2}\sigma^2 x^2 v''(x) \ge 0, \quad \forall x \ge 0;$$

(18) and for each
$$x \ge 0$$
, equality holds in either (16) or (17).

At the point $x=L_*$ we can use either $v''_{L_*}(L_{*+})$ or $v''_{L_*}(L_{*-})$ in (17), the inequality still holds. It turns out that $v_{L_*}(x)$ is the only bounded continuously differentiable function on $[0,\infty)$ $(C^1([0,\infty))$ that satisfies (16)–(18). (This is Exercise 8.3 in Shreve II.)

To reiterate: analytically, the perpetual put price is the only bounded $C^1([0,\infty)$ function which satisfies (16)–(18).