

# Chapter VII.

## Optimal stopping in mathematical finance

### 25. The American option

#### 25.1. Infinite horizon

1. According to theory of modern finance (see e.g. [197]) the arbitrage-free price of the *American put option* with *infinite horizon* (perpetual option) is given by

$$V(x) = \sup_{\tau} \mathbb{E}_x(e^{-r\tau}(K - X_{\tau})^+) \quad (25.1.1)$$

where the supremum is taken over all stopping times  $\tau$  of the geometric Brownian motion  $X = (X_t)_{t \geq 0}$  solving

$$dX_t = rX_t dt + \sigma X_t dB_t \quad (25.1.2)$$

with  $X_0 = x > 0$  under  $\mathbb{P}_x$ . We recall that  $B = (B_t)_{t \geq 0}$  is a standard Brownian motion process started at zero,  $r > 0$  is the interest rate,  $K > 0$  is the strike (exercise) price, and  $\sigma > 0$  is the volatility coefficient.

The equation (25.1.2) under  $\mathbb{P}_x$  has a unique (strong) solution given by

$$X_t = x \exp(\sigma B_t + (r - \sigma^2/2)t) \quad (25.1.3)$$

for  $t \geq 0$  and  $x > 0$ . The process  $X$  is strong Markov (diffusion) with the infinitesimal generator given by

$$\mathbb{L}_X = rx \frac{\partial}{\partial x} + \frac{\sigma^2}{2} x^2 \frac{\partial^2}{\partial x^2}. \quad (25.1.4)$$

The aim of this subsection is to compute the arbitrage-free price  $V$  from (25.1.1) and to determine the optimal exercise time  $\tau_*$  (at which the supremum in (25.1.1) is attained).

**2.** The optimal stopping problem (25.1.1) will be solved in two steps. In the first step we will make a guess for the solution. In the second step we will verify that the guessed solution is correct (Theorem 25.1).

From (25.1.1) and (25.1.3) we see that the closer  $X$  gets to 0 the less likely that the gain will increase upon continuation. This suggests that there exists a point  $b \in (0, K)$  such that the stopping time

$$\tau_b = \inf \{t \geq 0 : X_t \leq b\} \quad (25.1.5)$$

is optimal in the problem (25.1.1). [In (25.1.5) we use the standard convention that  $\inf(\emptyset) = \infty$  (see Remark 25.2 below).]

Standard arguments based on the strong Markov property (cf. Chapter III) lead to the following *free-boundary problem* for the *unknown value function*  $V$  and the *unknown point*  $b$ :

$$\mathbb{L}_X V = rV \quad \text{for } x > b, \quad (25.1.6)$$

$$V(x) = (K - x)^+ \quad \text{for } x = b, \quad (25.1.7)$$

$$V'(x) = -1 \quad \text{for } x = b \quad (\text{smooth fit}), \quad (25.1.8)$$

$$V(x) > (K - x)^+ \quad \text{for } x > b, \quad (25.1.9)$$

$$V(x) = (K - x)^+ \quad \text{for } 0 < x < b. \quad (25.1.10)$$

**3.** To solve the free-boundary problem note that the equation (25.1.6) using (25.1.4) reads

$$Dx^2V'' + rxV' - rV = 0 \quad (25.1.11)$$

where we set  $D = \sigma^2/2$ . One may now recognize (25.1.11) as the Cauchy–Euler equation. Let us thus seek a solution in the form

$$V(x) = x^p. \quad (25.1.12)$$

Inserting (25.1.12) into (25.1.11) we get

$$p^2 - \left(1 - \frac{r}{D}\right)p - \frac{r}{D} = 0. \quad (25.1.13)$$

The quadratic equation (25.1.13) has two roots,  $p_1 = 1$  and  $p_2 = -r/D$ . Thus the general solution of (25.1.11) can be written as

$$V(x) = C_1 x + C_2 x^{-r/D} \quad (25.1.14)$$

where  $C_1$  and  $C_2$  are undetermined constants. From the fact that  $V(x) \leq K$  for all  $x > 0$ , we see that  $C_1$  must be zero. Thus (25.1.7) and (25.1.8) become

two algebraic equations in two unknowns  $C_2$  and  $b$  (free-boundary). Solving this system one gets

$$C_2 = \frac{D}{r} \left( \frac{K}{1 + D/r} \right)^{1+r/D}, \quad (25.1.15)$$

$$b = \frac{K}{1 + D/r}. \quad (25.1.16)$$

Inserting (25.1.15) into (25.1.14) upon using that  $C_1 = 0$  we conclude that

$$V(x) = \begin{cases} \frac{D}{r} \left( \frac{K}{1+D/r} \right)^{1+r/D} x^{-r/D} & \text{if } x \in [b, \infty), \\ K - x & \text{if } x \in (0, b]. \end{cases} \quad (25.1.17)$$

Note that  $V$  is  $C^2$  on  $(0, b) \cup (b, \infty)$  but only  $C^1$  at  $b$ . Note also that  $V$  is convex on  $(0, \infty)$ .

4. In this way we have arrived at the two conclusions of the following theorem.

**Theorem 25.1.** *The arbitrage-free price  $V$  from (25.1.1) is given explicitly by (25.1.17) above. The stopping time  $\tau_b$  from (25.1.5) with  $b$  given by (25.1.16) above is optimal in the problem (25.1.1).*

*Proof.* To distinguish the two functions let us denote the value function from (25.1.1) by  $V_*(x)$  for  $x > 0$ . We need to prove that  $V_*(x) = V(x)$  for all  $x > 0$  where  $V(x)$  is given by (25.1.17) above.

1°. The properties of  $V$  stated following (25.1.17) above show that Itô's formula (page 67) can be applied to  $e^{-rt}V(X_t)$  in its standard form (cf. Subsection 3.5). This gives

$$\begin{aligned} e^{-rt}V(X_t) &= V(x) + \int_0^t e^{-rs}(\mathbb{L}_X V - rV)(X_s)I(X_s \neq b) ds \\ &\quad + \int_0^t e^{-rs} \sigma X_s V'(X_s) dB_s. \end{aligned} \quad (25.1.18)$$

Setting  $G(x) = (K - x)^+$  we see that  $(\mathbb{L}_X G - rG)(x) = -rK < 0$  so that together with (25.1.6) we have

$$(\mathbb{L}_X V - rV) \leq 0 \quad (25.1.19)$$

everywhere on  $(0, \infty)$  but  $b$ . Since  $\mathbb{P}_x(X_s = b) = 0$  for all  $s$  and all  $x$ , we see that (25.1.7), (25.1.9)–(25.1.10) and (25.1.18)–(25.1.19) imply that

$$e^{-rt}(K - X_t)^+ \leq e^{-rt}V(X_t) \leq V(x) + M_t \quad (25.1.20)$$

where  $M = (M_t)_{t \geq 0}$  is a continuous local martingale given by

$$M_t = \int_0^t e^{-rs} \sigma X_s V'(X_s) dB_s. \quad (25.1.21)$$

(Using that  $|V'(x)| \leq 1$  for all  $x > 0$  it is easily verified by standard means that  $M$  is a martingale.)

Let  $(\tau_n)_{n \geq 1}$  be a localization sequence of (bounded) stopping times for  $M$  (for example  $\tau_n \equiv n$  will do). Then for every stopping time  $\tau$  of  $X$  we have by (25.1.20) above

$$e^{-r(\tau \wedge \tau_n)}(K - X_{\tau \wedge \tau_n})^+ \leq V(x) + M_{\tau \wedge \tau_n} \quad (25.1.22)$$

for all  $n \geq 1$ . Taking the  $\mathbb{P}_x$ -expectation, using the optional sampling theorem (page 60) to conclude that  $\mathbb{E}_x M_{\tau \wedge \tau_n} = 0$  for all  $n$ , and letting  $n \rightarrow \infty$ , we find by Fatou's lemma that

$$\mathbb{E}_x(e^{-r\tau}(K - X_\tau)^+) \leq V(x). \quad (25.1.23)$$

Taking the supremum over all stopping times  $\tau$  of  $X$  we find that  $V_*(x) \leq V(x)$  for all  $x > 0$ .

2°. To prove the reverse inequality (equality) we observe from (25.1.18) upon using (25.1.6) (and the optional sampling theorem as above) that

$$\mathbb{E}_x\left(e^{-r(\tau_b \wedge \tau_n)}V(X_{\tau_b \wedge \tau_n})\right) = V(x) \quad (25.1.24)$$

for all  $n \geq 1$ . Letting  $n \rightarrow \infty$  and using that  $e^{-r\tau_b}V(X_{\tau_b}) = e^{-r\tau_b}(K - X_{\tau_b})^+$  (both expressions being 0 when  $\tau_b = \infty$ ), we find by the dominated convergence theorem that

$$\mathbb{E}_x(e^{-r\tau_b}(K - X_{\tau_b})^+) = V(x). \quad (25.1.25)$$

This shows that  $\tau_b$  is optimal in (25.1.1). Thus  $V_*(x) = V(x)$  for all  $x > 0$  and the proof is complete.  $\square$

**Remark 25.2.** It is evident from the definition of  $\tau_b$  in (25.1.5) and the explicit representation of  $X$  in (25.1.3) that  $\tau_b$  is not always finite. Using the well-known Doob formula (see e.g. [197, Chap. VIII, § 2a, (51)])

$$\mathbb{P}\left(\sup_{t \geq 0}(B_t - \alpha t) \geq \beta\right) = e^{-2\alpha\beta} \quad (25.1.26)$$

for  $\alpha > 0$  and  $\beta > 0$ , it is straightforwardly verified that

$$\mathbb{P}_x(\tau_b < \infty) = \begin{cases} 1 & \text{if } r \leq D \text{ or } x \in (0, b], \\ \left(\frac{b}{x}\right)^{(r/D)-1} & \text{if } r > D \text{ and } x \in (b, \infty) \end{cases} \quad (25.1.27)$$

for  $x > 0$ .