## Pricing American options on exponential Levy processes

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**Abstract.** to be added

## 1. Introduction and Motivation

To be added.

## 2. the formulation of the algorithm and implementation

We are assuming that the dynamics of the prices of the underlying risky security  $\{S_t\}_{0 \le t \le T}$  follows a process of the form

$$S_t = e^{X_t}$$

where  $X_t$  follows a Levy process with  $X_0 = \ln S_0$ .

Let us denote the characteristic function of  $X_t$  by  $\phi_t(v) = E[e^{iv \cdot X_t}], x \in \mathbb{R}$ .

We shall focus on the Bermuda put option with M periods. The the value for call option can be derived by the parity equation. The algorithm can be used for the other popular options where exercise payoff has a simple expression in term of underlying asset.

The exercise value at any time t before maturity is

$$G(S_t) = \begin{cases} (\alpha K - \beta S_t)^+ & S_t \le K \\ 0 & S_t > K \end{cases}$$

where K is the strike price and  $\alpha, \beta$  are non negative parameters. Let  $\Delta = T/M, \, k = \ln K$ . Let

$$\phi(v) = E[e^{ivX_{\Delta}}].$$

be the characteristic function. We are going to write  $S_{j\Delta}$  and  $X_{j\Delta}$  as  $S_j$  and  $X_j$  for convenience. So

$$S_{j+1} = e^{X_{j+1}} = e^{X_j} e^{X_{j+1} - X_j} \sim S_j e^Z$$

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where  $Z = X_{j+1} - X_j$  has the density q(z) (without depending on j). We shall scale the  $X_j$  by

$$X_i = \sigma Y_i + \mu$$

where

$$\sigma = var(X_j), \quad \mu = E[X_j]$$

and

$$Z = X_{i+1} - X_i = \sigma(Y_{i+1} - Y_i) = \sigma W$$

where the distribution of  $W = Y_{j+1} - Y_j$  is independent on j. Define

$$C_{j}^{s}(y) = C_{j}(e^{\sigma y + \mu}), \quad V_{j}^{s}(y) = V_{j}(e^{\sigma y + \mu}), \quad G^{s}(y) = G(e^{\sigma y + \mu})$$

Let  $f_W(w)$  be the density function of W and  $f_Z(z)$  be the density of Z, we have

$$f_W(w) = \sigma f_Z(\sigma w)$$

SO

$$e^{r\delta}C_j^s(y) = e^{r\delta}C_j(e^{\sigma y + \mu}) = E[V_{j+1}(e^{\sigma y + \mu}e^{\sigma W})]$$
$$= \int_{-\infty}^{\infty} V_{j+1}(e^{\sigma y + \mu + \sigma w})f_W(w)dw$$
$$= \int_{-\infty}^{\infty} V_{j+1}^s(y+w)f_W(w)dw$$

Assume that

$$|f_W(w)|_{[L,R]} \approx F_0/2 + \sum_{k=1}^{N-1} F_k cos(k\pi \frac{x-L}{R-L})$$

where

$$F_k = \frac{2}{R-L} Re(\int_L^R f_W(w) e^{ik\pi(w-L)/(R-L)} dw)$$

$$\approx \frac{2}{R-L} Re(\phi_W(\frac{k\pi}{R-L}) e^{\pi kL/(R-L)})$$

where

$$\phi_W(t) = \int_{-\infty}^{\infty} e^{itw} f_W(w) dw$$

Let

$$b_0 = -\infty < b_1 < \dots b_k < cdots < b_{M+1} = R < \infty$$

be a partition of  $(-\infty, \infty)$  and the  $V_{j+1}$  is equal to the payoff at

$$V_{j+1}^s(y) = G^s(y), \quad -\infty < y \le b_1$$

and  $V_{j+1}$  can be approximated by d-degree polynomial on each interval  $[b_k, b_{k+1})$  for  $(1 \le k \le M)$ :

$$V_{j+1}^{s}(y) = \sum_{h=0}^{d} c_{k,h} (y - b_k)^h, \quad b_k \le y < b_{k+1},$$

so

$$e^{r\delta}C_{j}^{s}(y) = \int_{-\infty}^{b_{1}-y} G^{s}(w+y)f_{W}(w)dw$$

$$+ \sum_{k=1}^{M} \int_{b_{k}-y}^{b_{k+1}-y} V_{j+1}^{s}(y+w)f_{W}(w)dw$$

$$+ \int_{b_{M+1}-y}^{\infty} V_{j+1}^{s}(w+y)f_{W}(w)dw$$

$$:= I + II + III$$

To estimate I, let

$$y_1 = \min(L, b_1 - y), \quad y_2 = \min(\max(L, b - y_1), R), y_3 = \max(R, b_1 - y)$$

we discuss I in three cases. We assume that  $G^s(y)$  is an decreasing function, i.e. a put-style option.

1.  $b_1 - y \le L$ .  $y_1 = b_1 - y$  and  $y_2 = L$ . It is clear that we have

$$0 \le I - \int_{L}^{y_2} G^s(y+w) f_W(w) dw \le G^s(-\infty) F_W(L) \tag{1}$$

2.  $L \le b_1 - y \le R$ ,  $y_1 = L$  and  $y_2 = b_1 - y$ ,

$$I = \int_{L}^{y_2} G^{s}(y+w) f_{W}(w) dw + \int_{-\infty}^{L} G^{s}(y+w) f_{W}(w) dw$$

It is clear

$$0 \le \int_{-\infty}^{L} G^{s}(y+w) f_{W}(w) dw \le G^{s}(-\infty) F_{W}(L)$$

so the equation (1) holds as in the case 1.

3.  $b_1 - y > R$ ,  $y_1 = L$  and  $y_2 = R$ , and

$$I = \left( \int_{-\infty}^{L} + \int_{L}^{y_2} + \int_{R}^{b_1 - y} \right) G^s(y + w) f_W(w) dw$$

It is straightforward to show

$$0 \le \int_{R}^{b_1 - y} G^s(y + w) f_W(w) dw \le G^s(-\infty) (1 - F_W(R))$$

So we have

$$0 \le I - \int_{L}^{y_2} G^s(y+w) f_W(w) dw \le G^s(-\infty) (1 - P(L \le W \le R))$$
 (2)

and

$$0 \le III \le V_{j+1}^s(R)P(W \ge R - y)$$

Let

$$low(y, k) = min(max(b_k - y, L), R), \quad up(y, k) = min(max(b_{k+1} - y, L), R),$$

then

$$0 \leq \int_{b_{k}-y}^{b_{k+1}-y} V_{j+1}^{s}(y+w) f_{W}(w) dw - \int_{low(y,k)}^{up(y,k)} V_{j+1}^{s}(y+w) f_{W}(w) dw$$
  
$$\leq V_{j+1}^{s}(b_{k}) (1 - P(L \leq W \leq R))$$

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