

# Monte Carlo methods for American options

**Ahmed Kebaier**

[kebaier@math.univ-paris13.fr](mailto:kebaier@math.univ-paris13.fr)

**HEC, Paris**

# Outline

## 1 Pricing American Options

# The Market model

- Let  $(W_t)_{t \geq 0}$  be a brownian motion under  $\mathbb{P}^*$  the risk neutral probability measure
- Under  $\mathbb{P}^*$  the Black-Scholes model is solution to

$$dS_t = rS_t dt + \sigma S_t dW_t, \quad \text{where } r \geq 0, \sigma > 0 \text{ and } S_0 > 0$$

with explicit solution given by

$$S_t = S_0 \exp \left( \left( r - \frac{\sigma^2}{2} \right) T + \sigma W_T \right)$$

- Let  $(\mathcal{F}_t)_{t \geq 0}$  denotes the natural filtration associated to  $W$ .

## European option setting

- A European option with maturity  $T$  and with payoff  $f(S_T)$  may be exercised only at the expiration date  $T$ .
- The price of this European option at time  $t \in [0, T]$  is equal to the value at time  $t$  of the associated hedging portfolio given by

$$V_t = H_t^0 e^{rt} + H_t S_t.$$

# European option setting

- A European option with maturity  $T$  and with payoff  $f(S_T)$  may be exercised only at the expiration date  $T$ .
- The price of this European option at time  $t \in [0, T]$  is equal to the value at time  $t$  of the associated hedging portfolio given by

$$V_t = H_t^0 e^{rt} + H_t S_t.$$

- Under  $\mathbb{P}^*$ , the process  $(\tilde{S}_t)_{0 \leq t \leq T}$  given by  $\tilde{S}_t = e^{-rt} S_t$  is an  $\mathcal{F}_t$ -martingale.

## European option setting

- A European option with maturity  $T$  and with payoff  $f(S_T)$  may be exercised only at the expiration date  $T$ .
- The price of this European option at time  $t \in [0, T]$  is equal to the value at time  $t$  of the associated hedging portfolio given by

$$V_t = H_t^0 e^{rt} + H_t S_t.$$

- Under  $\mathbb{P}^*$ , the process  $(\tilde{S}_t)_{0 \leq t \leq T}$  given by  $\tilde{S}_t = e^{-rt} S_t$  is an  $\mathcal{F}_t$ -martingale.
- The portfolio  $V$  is self-financing then

$$\tilde{V}_t = V_0 + \int_0^t H_s d\tilde{S}_s$$

# European option setting

- A European option with maturity  $T$  and with payoff  $f(S_T)$  may be exercised only at the expiration date  $T$ .
- The price of this European option at time  $t \in [0, T]$  is equal to the value at time  $t$  of the associated hedging portfolio given by

$$V_t = H_t^0 e^{rt} + H_t S_t.$$

- Under  $\mathbb{P}^*$ , the process  $(\tilde{S}_t)_{0 \leq t \leq T}$  given by  $\tilde{S}_t = e^{-rt} S_t$  is an  $\mathcal{F}_t$ -martingale.
- The portfolio  $V$  is self-financing then

$$\tilde{V}_t = V_0 + \int_0^t H_s d\tilde{S}_s$$

- Under  $\mathbb{P}^*$ , the process  $(\tilde{V}_t)_{0 \leq t \leq T}$  is also an  $\mathcal{F}_t$ -martingale.

# European option setting

- A European option with maturity  $T$  and with payoff  $f(S_T)$  may be exercised only at the expiration date  $T$ .
- The price of this European option at time  $t \in [0, T]$  is equal to the value at time  $t$  of the associated hedging portfolio given by

$$V_t = H_t^0 e^{rt} + H_t S_t.$$

- Under  $\mathbb{P}^*$ , the process  $(\tilde{S}_t)_{0 \leq t \leq T}$  given by  $\tilde{S}_t = e^{-rt} S_t$  is an  $\mathcal{F}_t$ -martingale.
- The portfolio  $V$  is self-financing then

$$\tilde{V}_t = V_0 + \int_0^t H_s d\tilde{S}_s$$

- Under  $\mathbb{P}^*$ , the process  $(\tilde{V}_t)_{0 \leq t \leq T}$  is also an  $\mathcal{F}_t$ -martingale.
- Then

$$V_t = e^{-r(T-t)} \mathbb{E}^*(f(S_T) | \mathcal{F}_t).$$



# Bermudan options

- Let us consider  $0 = t_0 < t_1 < \dots < t_n = T$  a discrete subdivision of  $[0, T]$ .

# Bermudan options

- Let us consider  $0 = t_0 < t_1 < \dots < t_n = T$  a discrete subdivision of  $[0, T]$ .

# Bermudan options

- Let us consider  $0 = t_0 < t_1 < \dots < t_n = T$  a discrete subdivision of  $[0, T]$ .
  - A Bermudan option gives the right to the buyer to exercise at any date  $t_0, \dots, t_n$  and pays  $f(S_{t_k})$  at time  $t_k$ .
  - Let  $(\tilde{V}_t)_{0 \leq t \leq T}$  denote the associated hedging portfolio.
- Then,

# Bermudan options

- Let us consider  $0 = t_0 < t_1 < \dots < t_n = T$  a discrete subdivision of  $[0, T]$ .
- A Bermudan option gives the right to the buyer to exercise at any date  $t_0, \dots, t_n$  and pays  $f(S_{t_k})$  at time  $t_k$ .
- Let  $(\tilde{V}_t)_{0 \leq t \leq T}$  denote the associated hedging portfolio. Then,
  - at date  $T = t_n$  we have  $V_{t_n} = f(S_{t_n})$

# Bermudan options

- Let us consider  $0 = t_0 < t_1 < \dots < t_n = T$  a discrete subdivision of  $[0, T]$ .
- A Bermudan option gives the right to the buyer to exercise at any date  $t_0, \dots, t_n$  and pays  $f(S_{t_k})$  at time  $t_k$ .
- Let  $(\tilde{V}_t)_{0 \leq t \leq T}$  denote the associated hedging portfolio.

Then,

- at date  $T = t_n$  we have  $V_{t_n} = f(S_{t_n})$
- at date  $T = t_{n-1}$  we have

$$\begin{aligned} V_{t_{n-1}} &= \max \left( f(S_{t_{n-1}}) ; e^{-r(t_n - t_{n-1})} \mathbb{E}^* [f(S_{t_n}) | \mathcal{F}_{t_{n-1}}] \right) \\ &= \max \left( f(S_{t_{n-1}}) ; e^{-r(t_n - t_{n-1})} \mathbb{E}^* [V_{t_n} | \mathcal{F}_{t_{n-1}}] \right) \end{aligned}$$

# Bermudan options

- Let us consider  $0 = t_0 < t_1 < \dots < t_n = T$  a discrete subdivision of  $[0, T]$ .
- A Bermudan option gives the right to the buyer to exercise at any date  $t_0, \dots, t_n$  and pays  $f(S_{t_k})$  at time  $t_k$ .
- Let  $(\tilde{V}_t)_{0 \leq t \leq T}$  denote the associated hedging portfolio.

Then,

- at date  $T = t_n$  we have  $V_{t_n} = f(S_{t_n})$
- at date  $T = t_{n-1}$  we have

$$\begin{aligned} V_{t_{n-1}} &= \max \left( f(S_{t_{n-1}}) ; e^{-r(t_n - t_{n-1})} \mathbb{E}^* [f(S_{t_n}) | \mathcal{F}_{t_{n-1}}] \right) \\ &= \max \left( f(S_{t_{n-1}}) ; e^{-r(t_n - t_{n-1})} \mathbb{E}^* [V_{t_n} | \mathcal{F}_{t_{n-1}}] \right) \end{aligned}$$

- In the same way we get  $\forall k \in \{0, \dots, n-1\}$

$$\begin{cases} V_T &= f(S_T) \\ V_{t_k} &= \max \left( f(S_{t_k}) ; e^{-r(t_{k+1} - t_k)} \mathbb{E}^* [V_{t_{k+1}} | \mathcal{F}_{t_k}] \right), \end{cases}$$

# Remarks

- Note that as the process  $(S_t)_{0 \leq t \leq T}$  is Markovian then

$$\mathbb{E}^* [V_{t_{k+1}} | \mathcal{F}_{t_k}] = \mathbb{E}^* [V_{t_{k+1}} | S_{t_k}].$$

- The price of a Bermudan option is more expensive than the price of an European option.
- For  $f(x) = (x - K)_+$  we have

$$e^{-r(t_{k+1} - t_k)} \mathbb{E}^* [V_{t_{k+1}} | \mathcal{F}_{t_k}] \geq f(S_{t_k})$$

# Remarks

- Note that as the process  $(S_t)_{0 \leq t \leq T}$  is Markovian then

$$\mathbb{E}^* [V_{t_{k+1}} | \mathcal{F}_{t_k}] = \mathbb{E}^* [V_{t_{k+1}} | S_{t_k}].$$

- The price of a Bermudan option is more expensive than the price of an European option.
- For  $f(x) = (x - K)_+$  we have

$$e^{-r(t_{k+1} - t_k)} \mathbb{E}^* [V_{t_{k+1}} | \mathcal{F}_{t_k}] \geq f(S_{t_k})$$

- If we let  $n \rightarrow \infty$  than the price of the Bermudan option tends to the price of the American option.



## Theorem

- Now if we consider  $\forall k \in \{0, \dots, n-1\}$

$$\begin{cases} V_T &= f(S_T) \\ V_{t_k} &= \max \left( f(S_{t_k}); e^{-r(t_{k+1}-t_k)} \mathbb{E}^*[V_{t_{k+1}} | \mathcal{F}_{t_k}] \right), \end{cases}$$

Then,

$$V_{t_k} = \sup_{\tau \in \{t_k, \dots, t_n\}} e^{-r(\tau-t_k)} \mathbb{E}^*(f(S_\tau) | \mathcal{S}_{t_k})$$

- The stopping time

$$\tau_k^* = \inf \left\{ t_i \in \{t_k, \dots, t_n\} \mid f(S_{t_i}) \geq e^{-r(t_{i+1}-t_i)} \mathbb{E}^*[f(S_{t_{i+1}}) | \mathcal{S}_{t_k}] \right\}$$

satisfies

$$V_{t_k} = e^{-r(\tau_k^*-t_k)} \mathbb{E}^*[f(S_{\tau_k^*}) | \mathcal{S}_{t_k}].$$

# Longstaff Schwarz algorithm

- We consider the sequence  $(\tau_k^*)_{0 \leq k \leq n}$  where

$$\tau_k^* = \inf \left\{ t_i \in \{t_k, \dots, t_n\} \mid f(S_{t_i}) \geq e^{-r(t_{i+1}-t_i)} \mathbb{E}^* [f(S_{t_{i+1}}) | \mathcal{S}_{t_k}] \right\}$$

- The sequence  $(\tau_k^*)_{0 \leq k \leq n}$  satisfies a dynamic programming principle

$$\begin{cases} \tau_n^* &= T \\ \tau_k^* &= t_k \mathbf{1}_{B_k} + \tau_{k+1}^* \mathbf{1}_{B_k^c}, \quad \text{for } 0 \leq k \leq n-1 \end{cases}$$

where

$$B_k = \left\{ f(S_{t_k}) \geq \mathbb{E}^* [e^{-r(\tau_{k+1}^* - t_k)} f(S_{\tau_{k+1}^*}) | \mathcal{S}_{t_k}] \right\}$$

# Longstaff Schwarz algorithm

- We consider the sequence  $(\tau_k^*)_{0 \leq k \leq n}$  where

$$\tau_k^* = \inf \left\{ t_i \in \{t_k, \dots, t_n\} \mid f(S_{t_i}) \geq e^{-r(t_{i+1}-t_i)} \mathbb{E}^* [f(S_{t_{i+1}}) | \mathcal{S}_{t_k}] \right\}$$

- The sequence  $(\tau_k^*)_{0 \leq k \leq n}$  satisfies a dynamic programming principle

$$\begin{cases} \tau_n^* &= T \\ \tau_k^* &= t_k \mathbf{1}_{B_k} + \tau_{k+1}^* \mathbf{1}_{B_k^c}, \quad \text{for } 0 \leq k \leq n-1 \end{cases}$$

where

$$B_k = \left\{ f(S_{t_k}) \geq \mathbb{E}^* [e^{-r(\tau_{k+1}^* - t_k)} f(S_{\tau_{k+1}^*}) | \mathcal{S}_{t_k}] \right\}$$

- How to approximate  $\mathbb{E}^* [e^{-r(\tau_{k+1}^* - t_k)} f(S_{\tau_{k+1}^*}) | \mathcal{S}_{t_k}]$  ?

# Regression

- We write  $\mathbb{E}^*[e^{-r(\tau_{k+1}^* - t_k)} f(S_{\tau_{k+1}^*}) | \mathcal{S}_{t_k}] = \sum_{\ell \geq 1} \alpha_{k,\ell} P_\ell(S_{t_k})$  where  $(P_\ell)_{\ell \geq 1}$  is a basis function.
- Using the definition of the conditional expectation, the sequence  $(\alpha_{k,\ell})_{\ell \geq 1}$  is the sequence that minimizes the distance

$$\mathbb{E}^* \left[ \left( e^{-r(\tau_{k+1}^* - t_k)} f(S_{\tau_{k+1}^*}) - \sum_{\ell \geq 1} \alpha_{k,\ell} P_\ell(S_{t_k}) \right)^2 \right]$$

- In practice we need to truncate the sum  $\sum_{\ell \geq 1} \alpha_{k,\ell} P_\ell(S_{t_k})$  and approximate it by

$$\sum_{\ell=1}^L \alpha_{k,\ell} P_\ell(S_{t_k}), \text{ where } L > 1.$$

# Longstaff Schwarz algorithm

- 1 Simulate  $(S_{t_0}^j, \dots, S_{t_n}^j)_{1 \leq j \leq M}$   $M$  copies of  $(S_{t_0}, \dots, S_{t_n})$
- 2 For all  $1 \leq j \leq M$  we set  $\tau_{j,n} = t_n = T$
- 3 Then compute the sequence  $(\alpha_{k,\ell}^j)_{1 \leq \ell \leq L}$  that minimizes

$$\frac{1}{M} \sum_{j=1}^M \left[ \left( e^{-r(\tau_{j,k+1} - t_k)} f(S_{\tau_{j,k+1}}^j) - \sum_{\ell=1}^L \alpha_{k,\ell} P_{\ell}(S_{t_k}^j) \right)^2 \right]$$

- 4 For all  $j \in \{1, \dots, M\}$  we define

$$\tau_{j,k} = t_k \mathbf{1}_{A_{j,k}} + \tau_{j,k+1} \mathbf{1}_{A_{j,k}^c}, \quad \text{for } 0 \leq k \leq n-1$$

where

$$A_{j,k} = \left\{ f(S_{t_k}^j) \geq \sum_{\ell=1}^L \alpha_{k,\ell}^j P_{\ell}(S_{t_k}^j) \right\}$$

# Price approximation

- For  $k = 0$  we the price of the Bermudan option is approximated by

$$\frac{1}{M} \sum_{j=1}^M e^{-r\tau_{j,0}} f(S_{\tau_{j,0}}^j)$$

- The Longstaff Schwarz algorithm converges in  $L^2$  as  $L \rightarrow \infty$  and for fixed  $L$  converges almost surely as  $M \rightarrow \infty$ .

# Computing the coordinates $(\alpha_{k,l})_{1 \leq l \leq L}$

## A Basic approach

For a fixed time step  $t_k$ , this approach consists on simply solving the following system

$$\begin{pmatrix} P_1(S_{t_k}^1) & \cdots & P_L(S_{t_k}^1) \\ \vdots & \vdots & \vdots \\ P_1(S_{t_k}^M) & \cdots & P_L(S_{t_k}^M) \end{pmatrix} \begin{pmatrix} \alpha_{k,1} \\ \vdots \\ \alpha_{k,L} \end{pmatrix} = \begin{pmatrix} e^{-r(\tau_{1,k+1}-t_k)} f(S_{\tau_{1,k+1}}^1) \\ \vdots \\ e^{-r(\tau_{M,k+1}-t_k)} f(S_{\tau_{M,k+1}}^M) \end{pmatrix}$$

- Advantage: easy to implement
- Drawback: not a high accuracy.

# Computing the coordinates $(\alpha_{k,l})_{1 \leq l \leq L}$

## An optimal approach

- For a fixed time step  $t_k$  we aim at computing the sequence  $(\alpha_{k,l}^j)_{1 \leq l \leq L}$  that minimizes

$$\sum_{j=1}^M \left[ \left( e^{-r(\tau_{j,k+1}-t_k)} f(S_{\tau_{j,k+1}}^j) - \sum_{\ell=1}^L \alpha_{k,\ell} P_{\ell}(S_{t_k}^j) \right)^2 \right]$$

- we differentiate the above quantity with respect to  $\alpha_{k,\ell_0}$  and solve

$$\sum_{j=1}^M \left( e^{-r(\tau_{j,k+1}-t_k)} f(S_{\tau_{j,k+1}}^j) - \sum_{\ell=1}^L \alpha_{k,\ell} P_{\ell}(S_{t_k}^j) \right) P_{\ell_0}(S_{t_k}^j) = 0$$



# An optimal approach

- This is equivalent to solve

$$\sum_{\ell=1}^L \left( \sum_{j=1}^M P_{\ell}(S_{t_k}^j) P_{\ell_0}(S_{t_k}^j) \right) \alpha_{k,\ell} = \sum_{j=1}^M e^{-r(\tau_{j,k+1}-t_k)} f(S_{\tau_{j,k+1}}^j) P_{\ell_0}(S_{t_k}^j)$$

- Let  $H_{\ell_0,\ell} = \sum_{j=1}^M P_{\ell}(S_{t_k}^j) P_{\ell_0}(S_{t_k}^j)$  we need to solve

$$\sum_{\ell=1}^L H_{\ell_0,\ell} \alpha_{k,\ell} = \sum_{j=1}^M e^{-r(\tau_{j,k+1}-t_k)} f(S_{\tau_{j,k+1}}^j) P_{\ell_0}(S_{t_k}^j)$$

- We can write this in a matrix equation given by

$$H \alpha_k = \sum_{j=1}^M e^{-r(\tau_{j,k+1}-t_k)} f(S_{\tau_{j,k+1}}^j) P(S_{t_k}^j)$$

where

$$\alpha_k = (\alpha_{k,1}, \dots, \alpha_{k,L}) \text{ and } P(S_{t_k}^j) = (P_1(S_{t_k}^j), \dots, P_L(S_{t_k}^j))$$