

# Monte-Carlo Method in American Option Pricing

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## Motivation

Financial contracts in mind that allow to exercise early before expiry:

- Early surrender of life insurance contracts
- Formosa bond: callable

How to price / hedge (Computing sensitivities)?

Pricing contracts with early exercise feature is very hard, if not impossible. Analytical (closed-form) solutions are rare, should be numerically solved.

In a complete and arbitrage-free market

## Pricing Formula of European Option

$$E_0 = E^Q\{d_{0,T} g_T(X_T)\}$$

where

- $g_t$  — nonnegative payoff function
- $(X_t)_{0 \leq t \leq T}$  — underlying stochastic process
- $d_{s,t}$  — nonnegative  $\mathcal{F}((X_u)_{s \leq u \leq t})$ -measurable discount factors satisfying  $d_{0,t} = d_{0,s} \cdot d_{s,t}$  for  $s < t$

## Pricing Formula of American Option

$$V_0 = \sup_{\tau \in \mathcal{T}([0, T])} E^Q \{d_{0, \tau} g_{\tau}(X_{\tau})\}$$

## Stopping Time

A stopping time  $\tau \in \mathcal{T}([0, T])$  is a measurable function of  $(X_t)_{0 \leq t \leq T}$  with values in  $[0, T]$  and has the property that  $\forall r \in [0, T]$ , the event  $\{\tau \leq r\}$  is contained in the  $\sigma$ -algebra  $\mathcal{F}_r = \mathcal{F}((X_t)_{0 \leq t \leq r})$ .

## Reduction From Continuous Time to Discrete Time

Set  $f_\tau \equiv d_{0,\tau} g_\tau$ , the discrete time formula is

$$V_0 = \sup_{\tau \in \mathcal{T}(0, \dots, T)} E^Q \{f_\tau(X_\tau)\}$$

- $\{X_0, X_1, \dots, X_T\}$  — underlying discrete time stochastic process
- $\mathcal{T}(0, \dots, T)$  — all stopping times with value in  $0, \dots, T$

## The Aim

The optimal stopping time  $\tau^*$  such that

$$V_0 = \sup_{\tau \in \mathcal{T}(0, \dots, T)} E^Q \{f_\tau(X_\tau)\} = E \{f_{\tau^*}(X_{\tau^*})\}$$

## Essential Notions

$$q_t(x) = \sup_{\tau \in \mathcal{T}(t+1, \dots, T)} \mathbb{E}\{f_\tau(X_\tau) | X_t = x\} \text{ as } 0 \leq t < T; \quad q_T(x) = 0$$

$$v_t(x) = \sup_{\tau \in \mathcal{T}(t, t+1, \dots, T)} \mathbb{E}\{f_\tau(X_\tau) | X_t = x\}$$

$$\tau_t^* = \inf\{s \geq t + 1 : q_s(X_s) \leq f_s(X_s)\}, \quad t \in \{-1, 0, \dots, T-1\}$$

## Interpretation

- $q_t(x)$ : Given  $X_t = x$ , the value of the option at  $t$  without selling.
- $v_t(x)$ : Given  $X_t = x$ , the value we get in the mean if we sell the option optimally after  $t - 1$ .

## Theorem

The optimal stopping time  $\tau^*$  is

$$\tau^* = \tau_{-1}^* = \inf\{s \in \{0, 1, \dots, T\} : q_s(X_s) \leq f_s(X_s)\}$$

N.B. Books about American option usually state this result without proof; those with proofs are incomplete at best. Complete proof is given in Kohler M. "A Review on Regression-Based Monte Carlo Method for Pricing American Options".

Representations of  $q_t(x)$ 

$$\begin{aligned}
 q_t(x) &= E\{f_{\tau_t^*}(X_{\tau_t^*})|X_t = x\} \\
 &= E\{\max(f_{t+1}(X_{t+1}), q_{t+1}(X_{t+1}))|X_t = x\} \\
 &= E\{\Theta_{t+1, t+w+1}^{(w)}|X_t = x\}
 \end{aligned}$$

where

$$\begin{aligned}
 \Theta_{t+1, t+w+1}^{(w)} &= \\
 &\sum_{s=t+1}^{t+w+1} f_s(X_s) \cdot 1_{\{f_{t+1}(X_{t+1}) < q_{t+1}(X_{t+1}), \dots, f_{s-1}(X_{s-1}) < q_{s-1}(X_{s-1}), f_s(X_s) \geq q_s(X_s)\}} \\
 &+ q_{t+w+1}(X_{t+w+1}) \cdot 1_{\{f_{t+1}(X_{t+1}) < q_{t+1}(X_{t+1}), \dots, f_{t+w+1}(X_{t+w+1}) < q_{t+w+1}(X_{t+w+1})\}}
 \end{aligned}$$

The exact value of  $q_t(x)$  is hard to compute because of the conditional expectation.  
 Insight: use regression estimates to approximately compute!



## Excursion: Variance Reduction by Control Variate

If we want to compute  $E(X)$ , suppose we can find another r.v.  $Y$  'close' to  $X$  with known  $E(Y)$ . Then r.v.

$$Z = X + E(Y) - Y$$

satisfies  $E(Z) = E(X) + E(Y) - E(Y) = E(X)$ . In this context,  $Y$  is the control variate.