G-Doob-Meyer Decomposition and Its Applications in Bid-Ask Pricing for Derivatives under Knightian Uncertainty

Wei Chen

School of Economics, Shandong University, Jinan 250100, China

Correspondence should be addressed to Wei Chen; weichen@sdu.edu.cn

Received 20 April 2015; Revised 23 July 2015; Accepted 9 August 2015

Academic Editor: Jafar Biazar

The target of this paper is to establish the bid-ask pricing framework for the American contingent claims against risky assets with G-asset price systems on the financial market under Knightian uncertainty. First, we prove G-Dooby-Meyer decomposition for G-supermartingale. Furthermore, we consider bid-ask pricing American contingent claims under Knightian uncertainty, by using G-Dooby-Meyer decomposition; we construct dynamic superhedge strategies for the optimal stopping problem and prove that the value functions of the optimal stopping problems are the bid and ask prices of the American contingent claims under Knightian uncertainty. Finally, we consider a free boundary problem, prove the strong solution existence of the free boundary problem, and derive that the value function of the optimal stopping problem is equivalent to the strong solution to the free boundary problem.

1. Introduction

The earliest and one of the most penetrating analyses on the pricing of the American option is by McKean [1]. There the problem of pricing the American option is transformed into a Stefan or free boundary problem. Solving the latter, McKean writes the American option price explicitly up to knowing a certain function, the optimal stopping boundary.

Bensoussan [2] presents a rigorous treatment for American contingent claims that can be exercised at any time before or at maturity. He adapts the Black and Scholes [3] methodology of duplicating the cash flow from such a claim to this situation by skillfully managing a self-financing portfolio that contains only the basic instruments of the market, that is, the stocks and the bond, and that entails no arbitrage opportunities before exercise. Bensoussan shows that the pricing of such claims is indeed possible and characterized the exercise time by means of an appropriate optimal stopping problem. In the study of the latter, Bensoussan employs the so-called “penalization method,” which forces rather stringent boundedness and regularity conditions on the payoff from the contingent claim.

From the theory of optimal stopping, it is well known that the value process of the optimal stopping problem can be characterized as the smallest supermartingale majorant to the stopping reward. Based on the Doob-Meyer decomposition for the supermartingale, a “martingale” treatment of the optimal stopping problem is used for handling pricing of the American option by Karatzas [4] and El Karou and Karatzas [5, 6].

The Doob decomposition theorem was proved by and is named for Doob [7]. The analogous theorem in the continuous time case is the Doob-Meyer decomposition theorem proved by Meyer in [8, 9]. For the pricing American option problem in incomplete market, Kramkov [10] constructs the optional decomposition of supermartingale with respect to a family of equivalent local martingale measures. He calls such a representation optional because, in contrast to the Doob-Meyer decomposition, it generally exists only with an adapted (optional) process C. He applies this decomposition to the problem of hedging European and American style contingent claims in the setting of incomplete security markets. Using the optional decomposition, Frey [11] considers construction of superreplication strategies via optimal stopping which is similar to the optimal stopping problem that arises in the pricing of American-type derivatives on a family of probability space with equivalent local martingale measures.

For the realistic financial market, the asset price in the future is uncertain, the probability distribution of the asset
price in the future is unknown, which is called Knightian uncertainty [12]. The probability distribution of the nature state in the future is unknown; investors have uncertain subjective belief, which makes their consumption and portfolio choice decisions uncertain and leads the uncertain asset price in the future. Pricing contingent claims against such assets under Knightian uncertainty is an open problem. Peng in [13, 14] constructs G framework which is an analysis tool for nonlinear system and is applied in pricing European contingent claims under volatility uncertainty [15, 16].

The target of this paper is to establish the bid-ask pricing framework for the American contingent claims against risky assets with G-asset price systems (see [17]) on the financial market under Knightian uncertainty. Firstly, on sublinear expectation space, by using potential theory and sublinear expectation theory we construct G-Doob-Meyer decomposition for G-supermartingale, that is, a right continuous G-supermartingale could be decomposed as a G-martingale and a right continuous increasing process and the decomposition is unique. Second, we define bid and ask prices of the American contingent claim against the assets with G-asset price systems and apply the G-Doob-Meyer decomposition to prove that the bid and ask prices of American contingent claims under Knightian uncertainty could be described by the optimal stopping problem. Finally, we present a free boundary problem, and by using the penalization technique (see [18]) we derive that if there exists strong supersolution to the free boundary problem, then the strong solution to the free boundary problem exists. And by using truncation and regularisation technique, we prove that the strong solution to the free boundary problem is the value function of the optimal stopping problem which is corresponding with pricing problem of the American contingent claim under Knightian uncertainty.

The rest of this paper is organized as follows. In Section 2, we give preliminaries for the sublinear expectation theory. In Section 3 we prove G-Doob-Meyer decomposition for G-supermartingale. In Section 4, using G-Doob-Meyer decomposition, we construct dynamic superhedge strategies for the optimal stopping problem and prove that the solution of the optimal stopping problem is the bid and ask prices of the American contingent claims under Knightian uncertainty. In Section 5, we consider a free boundary problem, prove the strong solution existence of the free boundary problem, and derive that the solution of the optimal stopping problem is equivalent to the strong solution to the free boundary problem.

2. Preliminaries

Let $\Omega$ be a given set and let $\mathcal{H}$ be a linear space of real valued functions defined on $\Omega$ containing constants. The space $\mathcal{H}$ is also called the space of random variables.

**Definition 1.** A sublinear expectation $\tilde{E}$ is a functional $\tilde{E}: \mathcal{H} \to \mathbb{R}$ satisfying

(i) monotonicity:

$$\tilde{E}[X] \geq \tilde{E}[Y] \quad \text{if } X \geq Y,$$

(ii) constant preserving:

$$\tilde{E}[c] = c \quad \text{for } c \in \mathbb{R},$$

(iii) subadditivity: for each $X, Y \in \mathcal{H},$

$$\tilde{E}[X + Y] \leq \tilde{E}[X] + \tilde{E}[Y],$$

(iv) positive homogeneity:

$$\tilde{E}[\lambda X] = \lambda \tilde{E}[X] \quad \text{for } \lambda \geq 0.$$

The triple $(\Omega, \mathcal{H}, \tilde{E})$ is called a sublinear expectation space.

In this section, we mainly consider the following type of sublinear expectation spaces $(\Omega, \mathcal{H}, \tilde{E})$: if $X_1, X_2, \ldots, X_n \in \mathcal{H}$ then $\varphi(X_1, X_2, \ldots, X_n) \in \mathcal{H}$ for $\varphi \in C_{b, \text{Lip}}(\mathbb{R}^n)$, where $C_{b, \text{Lip}}(\mathbb{R}^n)$ denotes the linear space of functions $\varphi$ satisfying

$$|\varphi(x) - \varphi(y)| \leq C (1 + |x|^m + |y|^m) |x - y|$$

for $x, y \in \mathbb{R}$, some $C > 0$, $m \in \mathbb{N}$ is depending on $\varphi$.

For each fixed $p \geq 1$, we take $\mathcal{H}_{p} = \{X \in \mathcal{H}, \tilde{E}[|X|^p] = 0\}$ as our null space and denote $\mathcal{H}/\mathcal{H}_{p}$ as the quotient space. We set $\|X\|_p := (\tilde{E}[|X|^p])^{1/p}$ and extend $\mathcal{H}/\mathcal{H}_{p}$ to its completion $\overline{\mathcal{H}}_p$ under $\|\cdot\|_p$. Under $\|\cdot\|_p$, the sublinear expectation $\tilde{E}$ can be continuously extended to the Banach space $(\overline{\mathcal{H}}_p, \|\cdot\|_p)$. Without loss generality, we denote the Banach space $(\overline{\mathcal{H}}_p, \|\cdot\|_p)$ as $L^p_0(\Omega, \mathcal{H}, \tilde{E})$. For the G-frame work, we refer to [13, 14].

In this paper we assume that $\mu, \overline{\mu}, \mathcal{P}, \overline{\mathcal{P}}$ are positive constants such that $\mu \leq \overline{\mu}$ and $\mathcal{P} \leq \overline{\mathcal{P}}$.

**Definition 2.** Let $X_1$ and $X_2$ be two random variables in a sublinear expectation space $(\Omega, \mathcal{H}, \tilde{E})$; $X_1$ and $X_2$ are called identically distributed, denoted by $X_1 \overset{d}{=} X_2$ if

$$\tilde{E}[\varphi(X_1)] = \tilde{E}[\varphi(X_2)] \quad \forall \varphi \in C_{b, \text{Lip}}(\mathbb{R}^n).$$

**Definition 3.** In a sublinear expectation space $(\Omega, \mathcal{H}, \tilde{E})$, a random variable $Y$ is said to be independent of another random variable $X$, if

$$\tilde{E}[\varphi (X, Y)] = \tilde{E}\left[\tilde{E}[\varphi (X, Y)] |_{x=X}\right],$$

**Definition 4 (G-normal distribution).** A random variable $X$ on a sublinear expectation space $(\Omega, \mathcal{H}, \tilde{E})$ is called G-normal distributed if

$$aX + b\overline{X} = \sqrt{a^2 + b^2} X \quad \text{for } a, b \geq 0,$$

where $\overline{X}$ is an independent copy of $X$. 

We denote by $S(d)$ the collection of all $d \times d$ symmetric matrices. Let $X$ be G-normal distributed random vectors on $(\Omega, \mathcal{H}, \hat{E})$; we define the following sublinear function:

$$G(A) := \frac{1}{2} \hat{E}[\langle AX, X \rangle], \quad A \in S(d).$$

(9)

**Remark 5.** For a random variable $X$ on the sublinear space $(\Omega, \mathcal{H}, \hat{E})$, there are four typical parameters to character $X$:

$$\overline{\mu}_X = \hat{E}X, \quad \underline{\mu}_X = -\hat{E}[-X], \quad \overline{\sigma}_X^2 = \hat{E}X^2, \quad \underline{\sigma}_X^2 = -\hat{E}[-X^2],$$

(10)

where $[\underline{\mu}_X, \overline{\mu}_X]$ and $[\underline{\sigma}_X^2, \overline{\sigma}_X^2]$ describe the uncertainty of the mean and the variance of $X$, respectively.

It is easy to check that if $X$ is G-normal distributed, then

$$\overline{\mu}_X = \hat{E}X = \underline{\mu}_X = -\hat{E}[-X] = 0,$$

(11)

and we denote the G-normal distribution as $N(\{0\}, [\underline{\sigma}_X^2, \overline{\sigma}_X^2])$.

If $X$ is maximally distributed, then

$$\overline{\sigma}_X^2 = \hat{E}X^2 = \underline{\sigma}_X^2 = -\hat{E}[-X^2] = 0,$$

(12)

and we denote the maximal distribution (see [14]) as $N([\underline{\mu}_X, \overline{\mu}_X], \{0\})$.

Let $\mathcal{F}$ as Borel field subsets of $\Omega$. We are given a family $\{\mathcal{F}_t\}_{t \in R_+}$ of Borel subfields of $\mathcal{F}$, such that

$$\mathcal{F}_s \subset \mathcal{F}_t, \quad s < t.$$  

(13)

**Definition 6.** We call $(X_t)_{t \in R}$ a $d$-dimensional stochastic process on a sublinear expectation space $(\Omega, \mathcal{H}, \hat{E}, \mathcal{F}, \{\mathcal{F}_t\}_{t \in R_+})$, if, for each $t \in R$, $X_t$ is a $d$-dimensional random vector in $\mathcal{H}$.

**Definition 7.** Let $(X_t)_{t \in R}$ and $(Y_t)_{t \in R}$ be $d$-dimensional stochastic processes defined on a sublinear expectation space $(\Omega, \mathcal{H}, \hat{E}, \mathcal{F}, \{\mathcal{F}_t\}_{t \in R_+})$, for each $t = (t_1, t_2, \ldots, t_n) \in \mathcal{F}$;

$$F_{t}^X[\varphi] := \hat{E}[\varphi(X_t)], \quad \forall \varphi \in C_{lip}(R^{md}).$$

(14)

is called the finite dimensional distribution of $X_t$, $X$ and $Y$ are said to be identically distributed, that is, $X \overset{d}{=} Y$, if

$$F_{t}^X[\varphi] = F_{t}^Y[\varphi], \quad \forall \varphi \in C_{lip}(R^{md}),$$

(15)

where $\mathcal{T} := \{t = (t_1, t_2, \ldots, t_n) : \forall n \in N, \ t_1 \in R, \ t_i \neq t_j, \ 0 \leq i, j \leq n, \ i \neq j\}$.

**Definition 8.** A process $(B_t)_{t \geq 0}$ on the sublinear expectation space $(\Omega, \mathcal{H}, \hat{E}, \mathcal{F}, \{\mathcal{F}_t\}_{t \in R_+})$ is called a G-Brownian motion if the following properties are satisfied:

(i) $B_0(\omega) = 0$;

(ii) For each $t, s > 0$, the increment $B_{t+s} - B_t$ is G-normal distributed by $N(\{0\}, [\underline{\sigma}_s^2, \overline{\sigma}_s^2])$ and is independent of $(B_{t_1}, B_{t_2}, \ldots, B_{t_n})$, for each $n \in N$ and $t_1, t_2, \ldots, t_n \in (0, t)$.

From now on, the stochastic processes we will consider in the rest of this paper are all in the sublinear space $(\Omega, \mathcal{H}, \hat{E}, \mathcal{F}, \{\mathcal{F}_t\}_{t \in R_+})$.

3. **G-Doob-Meyer Decomposition for G-Supermartingale**

**Definition 9.** A G-supermartingale (resp., G-submartingale) is a real valued process $\{X_t\}$, well adapted to the $\mathcal{F}_t$ family, such that

(i) $\hat{E}[|X_t|] < \infty, \ \forall t \in R_+$,

(ii) $\hat{E}[X_{t+s} | \mathcal{F}_s] \leq$ (resp.$\geq$) $X_s$

$$\forall t \in R_+, \ \forall s \in R_+.$$  

(16)

If equality holds in (ii), the process is a G-martingale.

We will consider right continuous G-supermartingales; then if $\{X_t\}$ is right continuous G-supermartingale, (ii) in (16) holds with $\mathcal{F}_t$ replaced by $\mathcal{F}_{t_+}$.

**Definition 10.** Let $A$ be an event in $\mathcal{F}_{t_+}$; one defines capacity of $A$ as

$$c(A) = \hat{E}[I_A],$$

(17)

where $I_A$ is indicator function of event $A$.

**Definition 11.** Process $X_t$ and $Y_t$ are adapted to the filtration $\mathcal{F}_t$. One calls $Y_t$ equivalent to $X_t$, if and only if

$$c(Y_t \neq X_t) = 0.$$  

(18)

For a right continuous G-supermartingale $\{X_t\}$ with $\hat{E}[X_t]$ is right continuous function of $t$; we can find a right continuous G-supermartingale $\{Y_t\}$ equivalent to $\{X_t\}$ by defining
Without loss generality, we denote \( \mathcal{F}_t = \mathcal{F}_{t+} \).

**Definition 12.** For a positive constant \( T \), one defines stop time \( \tau \) in \([0, T]\) as a positive, random variable \( \tau(\omega) \) such that \( \{ \tau \leq T \} \in \mathcal{F}_T \).

In [19, 20], authors discuss the definition of stop time and its related theory in G frame work.

Let \( \{X_t\} \) be a right continuous G-supermartingale, denote \( X_\infty \) as the last element of the process \( X_t \), and then the process \( \{X_t\}_{0 \leq t \leq \infty} \) is a G-supermartingale.

**Definition 13.** A right continuous increasing process is a well adapted stochastic process \( \{A_t\} \) such that

(i) \( A_0 = 0 \) a.s,

(ii) for almost every \( \omega \), the function \( t \rightarrow A_t(\omega) \) is positive, increasing, and right continuous. Let \( A_\infty(\omega) := \lim_{t \rightarrow \infty} A_t(\omega) \); one will say that the right continuous increasing process is integrable if \( \tilde{E}[A_\infty] < \infty \).

**Definition 14.** An increasing process \( A \) is called natural if for every bounded, right continuous G-martingale \( \{M_t\}_{0 \leq t \leq \infty} \) we have

\[
\tilde{E} \left[ \int_{[0, t]} M_t dA_t \right] = \tilde{E} \left[ \int_{[0, t]} M_t dA_t \right],
\]

for every \( 0 < t < \infty \).

**Lemma 15.** If \( A \) is an increasing process and \( \{M_t\}_{0 \leq t \leq \infty} \) is bounded, right continuous G-martingale, then

\[
\tilde{E} \left[ M_t A_t \right] = \tilde{E} \left[ \int_{[0, t]} M_t dA_t \right].
\]

In particular, condition (20) in Definition 14 is equivalent to

\[
\tilde{E} \left[ M_t A_t \right] = \tilde{E} \left[ \int_{[0, t]} M_t dA_t \right].
\]

**Proof.** For a partition \( \Pi = \{t_0, t_1, \ldots, t_n\} \) of \([0, t]\), with \( 0 = t_0 \leq t_1 \leq \cdots \leq t_n = t \), we define

\[
M_t^{\Pi} = \sum_{k=1}^{n} M_{t_k} I_{[t_{k-1}, t_k)} (s).
\]

Since \( M \) is G-martingale

\[
\tilde{E} \left[ \int_{[0, t]} M_t dA_t \right] = \tilde{E} \left[ \sum_{k=1}^{n} M_{t_k} (A_{t_k} - A_{t_{k-1}}) \right]
= \tilde{E} \left[ \sum_{k=1}^{n} M_{t_k} A_{t_k} - \sum_{k=1}^{n-1} M_{t_{k+1}} A_{t_k} \right]
= \tilde{E} \left[ M_t A_t - \sum_{k=1}^{n-1} (M_{t_k} - M_{t_{k+1}}) A_{t_k} \right]
= \tilde{E} \left[ M_t A_t - \sum_{k=1}^{n-1} (M_{t_k} - M_{t_{k+1}}) A_{t_k} \right]
= \tilde{E} \left[ M_t A_t \right],
\]

and we finish the proof of the Lemma.

**Definition 16.** A positive right continuous G-supermartingale \( \{Y_t\} \) with \( \lim_{t \rightarrow \infty} Y_t(\omega) = 0 \) is called a potential.

**Definition 17.** For \( a \in [0, \infty) \), a process \( \{X_t, t \in [0, a]\} \) is said to be uniformly integrable on \([0, a]\) if

\[
\sup_{t \in [0, a]} \tilde{E}[X_t I_{[X_t > x]}] \rightarrow 0, \quad \text{as } x \rightarrow 0.
\]

**Definition 18.** Let \( a \in [0, \infty) \), and let \( \{X_t\} \) be a right continuous process; we will say that it belongs to the class (GD) on this interval, if all the random variables \( X_T \) are uniformly integrable and \( \bar{T} \) is stop time bounded by \( a \). If \( \{X_t\} \) belongs to the class (GD) on every interval \([0, a]\), \( a < \infty \), it will be said to belong locally to the class (GD).

If \( \{A_t\} \) is an integrable right continuous, increasing process, then process \( \{-A_t\} \) is a negative G-supermartingale, and \( \{\tilde{E}[A_\infty | \mathcal{F}_s] - A_s\} \) is a potential of the class (GD), which we will call the potential generated by \( \{A_t\} \).

**Proposition 19.** (1) Any right continuous G-martingale \( \{X_t\} \) belongs locally to class (GD).

(2) Any right continuous G-supermartingale \( \{X_t\} \), which is bounded from above, belongs locally to class (GD).

(3) Any right continuous supermartingale \( \{X_t\} \), which belongs locally to class (GD) and is uniformly integrable, belongs to class (GD).

**Proof.** (1) If \( a < \infty \) and \( T \) is a stop time, \( T \leq a \), then G-martingale process \( \{X_t\} \) has \( X_T = \tilde{E}[X_a | \mathcal{F}_T] \). Hence

\[
\tilde{E} \left[ X_T I_{[X_T > x]} \right] \leq \tilde{E} \left[ X_a I_{[X_a > x]} \right],
\]

(26)
As $n \cdot c(|X_T| > n) \leq \mathbb{E}[|X_T|] \leq \mathbb{E}[|X_n|]$, we have $c(|X_T| > n) \to 0$ as $n \to \infty$; then $\mathbb{E}[|X_T|] \leq \mathbb{E}[|X_\omega|]$, from which we prove (1).

(2) If $a < \infty$ and $T$ is a stop time, $T \leq a$, then $G$-supermartingale process $\{X_t\}$ has $X_T \geq \mathbb{E}[X_T | \mathcal{F}_T]$. Suppose that $\{X_t\}$ is negative; then

$$
\mathbb{E}[-X_T I_{|X_T| < n}] \geq \mathbb{E}[-X_n I_{|X_n| < n}];
$$

we complete the proof of (2) by using similar argument in proof (1).

(3) $\{X_t\}$ is uniformly integrable; we set

$$
X_t = \mathbb{E}[X_\infty | \mathcal{F}_t] + (X_t - \mathbb{E}[X_\infty | \mathcal{F}_t]).
$$

The first part on the right-hand of the above equation $\mathbb{E}[X_\infty | \mathcal{F}_t]$ is a $G$-martingale and equivalent to a right continuous process, and from (1) we know that it belongs to class (GD). We denote the second part in the above equation $\{Y_t\}$; it is a positive right continuous $G$-supermartingale, and $\lim_{t \to \infty} Y_t(\omega) = 0$ a.s. Next we will prove that $\{Y_t\}$ belongs to class (GD). Since both $\inf(T, a)$ and $\sup(T, a)$ are stop times

$$
\mathbb{E}[Y_t I_{Y_t > n}] \leq \mathbb{E}[Y_{T^*} I_{T^* > n}] + \mathbb{E}[Y_t I_{|Y_t| > n}]
$$

(29)

Consider that $\lim_{n \to \infty} \mathbb{E}[Y_t] = 0$ and $\{Y_t\}$ locally belongs to (GD); that is, $\lim_{n \to \infty} \mathbb{E}[Y_t I_{T^* > n}] = 0$, which prove that

$$
\lim_{n \to \infty} \mathbb{E}[Y_t I_{Y_t > n}] = 0.
$$

(30)

We complete the proof.

\[\square\]

Lemma 20. Let $\{X_t\}$ be a right continuous $G$-supermartingale and $\{X^n_t\}$ a sequence of decomposed right continuous $G$-supermartingale:

$$
X^n_t = M^n_t - A^n_t,
$$

(31)

where $\{M^n_t\}$ is $G$-martingale and $\{A^n_t\}$ is right continuous increasing process. Suppose that, for each $t$, $X^n_t$ converge to $X_t$ in the $L^1_G(\Omega)$ topology, and $A^n_t$ are uniformly integrable in $n$. Then the decomposition problem is solvable for the $G$-supermartingale $\{X_t\}$; more precisely, there are a right continuous increasing process $\{A_t\}$ and a $G$-martingale $\{M_t\}$, such that $X_t = M_t - A_t$.

Proof. We denote by $w$ the weak topology $w(L^1_G(\Omega), L^\infty_G(\Omega))$; a sequence of integrable random variables $f_n$ converges to a random variable $f$ in the $w$-topology, if and only if $f$ is integrable, and

$$
\lim_{n \to \infty} \mathbb{E}[f_n g] = \mathbb{E}[f g], \forall g \in L^\infty_G(\Omega).
$$

(32)

Since $A^n_t$ are uniformly integrable in $n$, by the properties of the sublinear expectation $\mathbb{E}[]$ there exists a $w$-convergent subsequence $A^n_k$ converging in the $w$-topology to the random variables $A'_t$, for all rational values of $t$. To simplify the notations, we will use $A^n_t$ converging to $A'_t$ in the $w$-topology for all rational values of $t$. An integrable random variable $f$ is $\mathcal{F}_t$-measurable if and only if it is orthogonal to all bounded random variables $g$ such that $\mathbb{E}[g | \mathcal{F}_t] = 0$; it follows that $A'_t$ is $\mathcal{F}_t$-measurable. For $s < t, s$ and $t$ rational,

$$
\mathbb{E}\left[(A^n_t - A^n_s) I_B\right] \geq 0,
$$

(33)

where $B$ denote any $\mathcal{F}_s$ set.

As $X^n_t$ converge to $X_t$ in $L^1_G(\Omega)$ topology, which is in a stronger topology than $w$, the $M^n_t$ converge to random variables $M'_t$ for $t$ rational, and the process $\{M'_t\}$ is $G$-martingale; then there is a right continuous $G$-martingale $\{M_t\}$, defined for all values of $t$, such that $c(M_t \neq M'_t) = 0$ for each rational $t$. We define $A_t = X_t + M_t$; $\{A_t\}$ is a right continuous increasing process or at least becomes so after a modification on a set of measure zero. We complete the proof.

\[\square\]

Lemma 21. Let $\{X_t\}$ be a potential and belong to class (GD). One considers the measurable, positive, and well-adapted processes $H = \{H_t\}$ with the property that the right continuous increasing processes

$$
A(H) = \left\{ A_t (H, \omega) = \left\{ \int_0^t H_s (\omega) ds \right\} \right\}
$$

(34)

are integrable, and the potentials $Y(H) = \{Y_t (H, \omega)\}$ they generate are majorized by $X_t$. Then, for each $t$, the random variables $A_t (H)$ of all such processes $A(H)$ are uniformly integrable.

Proof. It is sufficient to prove that the $A_{\infty}(H)$ are uniformly integrable.

(1) First we assume that $X_t$ is bounded by some positive constant $C$; then $\mathbb{E}[A^2_{\infty}(H)] \leq 2C^2$, and the uniform integrability follows.

We have that

$$
A^2_{\infty}(H, \omega) = 2 \int_0^\infty A_{\infty}(H, \omega) - A_u (H, \omega) dA_u (H, \omega)
$$

(35)

$$
= 2 \int_0^\infty A_{\infty}(H, \omega) dH_u (\omega) du.
$$

By using the subadditive property of the sublinear expectation $\mathbb{E}$, we derive that

$$
\mathbb{E} [A^2_{\infty}(H, \omega)] = \mathbb{E} [\mathbb{E} [A^2_{\infty}(H, \omega) | \mathcal{F}_t]]
$$

$$
\leq 2 \mathbb{E} \left[ \int_0^\infty H_u \mathbb{E} [A_{\infty}(H, \omega) - A_u (H, \omega) | \mathcal{F}_u] du \right]
$$

(36)

$$
= 2 \mathbb{E} \left[ \int_0^\infty H_u Y_u (H) du \right] \leq 2C \mathbb{E} \left[ \int_0^\infty H_u du \right]
$$

$$
= 2C \mathbb{E} [Y_0 (H)] \leq 2C^2.
$$
(2) In order to prove the general case, it will be enough to prove that any \( H \) such that \( A(H^\epsilon) \) is equal to a sum \( H^F + H_\epsilon \), where (i) \( A(H^\epsilon) \) generates a potential bounded by \( c \), and (ii) \( \hat{E}[A_{\infty}(H_\epsilon)] \) is smaller than some number \( \epsilon_\epsilon \), independent of \( H \), such that \( \epsilon_\epsilon \to 0 \) as \( c \to 0 \). Define
\[
H_\epsilon^c(\omega) = H_\epsilon(\omega) I_{(X_\omega(t) \epsilon [0,c])},
\]
(37)
\[
H_\epsilon = H_\epsilon - H_\epsilon^c.
\]
Set
\[
T^c(\omega) = \inf \{ t : \text{such that } X_t(\omega) \geq c \},
\]
(38)
as \( c \) goes to infinity \( \lim_{c \to \infty} T^c(\omega) = \infty \); therefore \( X^c \to 0 \), and class (GD) property implies that \( \hat{E}[X^c] \to 0 \). \( T^c \) is a stop time, and \( I_{(X(\omega) \epsilon [0,c])} = 1 \) before time \( T^c \). Hence
\[
\hat{E}[A_{\infty}(H_\epsilon)] = \hat{E}\left[ \int_0^\infty H_u \left( 1 - I_{(X_u(\omega) \epsilon [0,c])} \right) \right] du
\leq \hat{E}\left[ \int_0^\infty H_u du \right]
= \hat{E}[A_{\infty}(H) - A_{T^c}(H)]
= \hat{E}\left[ \hat{E}[A_{\infty}(H) - A_{T^c}(H) | \mathcal{F}_t] \right]
= \hat{E}\left[ Y_{T^c}(H) \right] \leq \hat{E}[X^c(H)] \leq \epsilon_\epsilon,
\]
for large enough \( c \), from which we prove (ii). We will prove (i); first we prove that \( Y(H^\epsilon) \) is bounded by \( c \):
\[
Y_t(H^\epsilon) = \hat{E}\left[ A_{\infty}(H^\epsilon) - A_t^c(H^\epsilon) | \mathcal{F}_t \right]
= \hat{E}\left[ \int_t^\infty H_u I_{(X_u(\omega) \epsilon [0,c])} du | \mathcal{F}_t \right]
\leq \hat{E}\left[ \int_t^\infty H_u I_{(X_u(\omega) \epsilon [0,c])} du | \mathcal{F}_t \right]
= \hat{E}\left[ \int_t^{S^c(\omega)} H_u I_{(X_u(\omega) \epsilon [0,c])} du | \mathcal{F}_t \right]
= \hat{E}[Y_{S^c} | \mathcal{F}_t] \leq c,
\]
where we set
\[
S^c(\omega) = \inf \{ t : \text{such that } X_t(\omega) \leq c \}
\]
(41)
and use
\[
\int_S^{S^c(\omega)} H_u I_{(X_u(\omega) \epsilon [0,c])} du = 0.
\]
(42)
Inequality (40) holds for each \( t \), for every rational \( t \) and for every \( t \) in consideration of the right continuity, which complete the proof.

**Lemma 22.** Let \( \{X_t\} \) be a potential and belong to class (GD), \( k \) is a positive number, define \( Y_t = \hat{E}[X_{t+k} | \mathcal{F}_t] \), and then \( \{Y_t\} \) is a \( G \)-supermartingale. Denote by \( \{p_kX_t\} \) a right continuous version of \( \{Y_t\} \); then \( \{p_kX_t\} \) is potential.

Use the same notations as in Lemma 21. Let \( k \) be a positive number, and \( H_{k,\omega}(x) = (X_\omega(x) - p_kX_\omega(x))/k \). The process \( H_{k,\omega} \) verifies the assumptions of Lemma 21, and their potentials increase to \( \{X_t\} \) as \( k \to 0 \).

**Proof.** If \( t < u \)
\[
\hat{E}\left[ \int_t^u X_s - p_kX_s | \mathcal{F}_t \right] \leq \hat{E}\left[ \int_t^u X_s | \mathcal{F}_t \right] - \hat{E}\left[ \int_t^u p_kX_s | \mathcal{F}_t \right]
\]
(43)
and, by the subadditive property of the sublinear expectation \( \hat{E} \), we derive that
\[
\hat{E}\left[ \int_t^u X_s - p_kX_s | \mathcal{F}_t \right] - \hat{E}\left[ \int_t^u p_kX_s | \mathcal{F}_t \right]
\geq \hat{E}\left[ \int_t^u X_s | \mathcal{F}_t \right] - \hat{E}\left[ \int_t^u p_kX_s | \mathcal{F}_t \right]
\]
(44)
Hence, we derive that for any \( u, t \) such that \( u > t \)
\[
\hat{E}\left[ \int_t^u X_s - p_kX_s | \mathcal{F}_t \right] \geq 0.
\]
(46)
If there exists \( s_0 \geq 0 \) such that \((1/k)|X_{s_0} - p_kX_{s_0}| < 0 \), the right continuous of \( \{X_t\} \) implies that there exists \( \delta > 0 \) such that \((1/k)|X_{s_0} - p_kX_{s_0}| < 0 \) on the interval \([s_0, s_0 + \delta] \). Thus
\[
\hat{E}\left[ \int_t^{s_0+\delta} X_s - p_kX_s | \mathcal{F}_t \right] < 0,
\]
(47)
which is contradiction; we prove that \((X_t(\omega) - p_kX_t(\omega))/k\) is a positive, measurable, and well-adapted process.
Since \( \{X_t\} \) is right continuous G-supermartingale
\[
\lim_{t \downarrow s} X_s = X_t,
\]
\[
\lim_{k \downarrow 0} Y_t( H_k) = \hat{B}_t.
\]
The uncertain drift \( b_t \) can be written as
\[
b_t = \int_0^t \mu_u \, du, \quad (59)
\]
we finish the proof. \( \Box \)

From Lemmas 20, 21, and 22 we can prove the following theorem.

**Theorem 23.** A potential \( \{X_t\} \) belongs to class (GD) if and only if it is generated by some integrable right continuous increasing process.

**Theorem 24 (G-Doob-Meyer’s decomposition).** (1) \( \{X_t\} \) is a right continuous G-supermartingale if and only if it belongs to class (GD) on every finite interval. More precisely, \( \{X_t\} \) is then equal to the difference of a G-martingal \( M_t \) and a right continuous increasing process \( A_t \):
\[
X_t = M_t - A_t, \quad (49)
\]
(2) If the right continuous increasing process \( A \) is natural, the decomposition is unique.

**Proof.** (1) The necessity is obvious. We will prove the sufficiency; we choose a positive number \( a \) and define
\[
X'_t(\omega) := X_t(\omega), \quad t \in [0, a],
\]
\[
X''_t(\omega) := X_a(\omega), \quad t > a;
\]
the \( \{X'_t\} \) is a right continuous G-supermartingale of the class (GD), and by Theorem 23 there exists the following decomposition
\[
X'_t = M'_t - A'_t, \quad (51)
\]
where \( \{M'_t\} \) is a G-martingal and \( \{A'_t\} \) is a right continuous increasing process.

Let \( a \to \infty \), as in Lemma 22 the expression of \( Y_t( H_k) \) that \( A'_t \) depend only on the values of \( X'_t \) on intervals \([0, t+\varepsilon]\), with \( \varepsilon \) small enough. As \( a \to \infty \), they do not vary anymore once \( a \) has reached values greater than \( t \), as again Lemma 20; we finish the proof of the Theorem.

(2) Assume that \( X \) admits both decompositions:
\[
X_t = M'_t - A'_t = M''_t - A''_t, \quad (52)
\]
where \( M'_t \) and \( M''_t \) are G-martingale and \( A'_t, A''_t \) are natural increasing process. We define
\[
\{C_t := A'_t - A''_t = M'_t - M''_t \}. \quad (53)
\]
Then \( \{C_t\} \) is a G-martingale, and, for every bounded and right continuous G-martingale \( \{\xi_t\} \), from Lemma 15 we have
\[
\hat{E} [\xi_t (A'_t - A''_t)] = \hat{E} \left( \int_{[0,t]} \xi_s \, dC_s \right)
\]
\[
= \lim_{n \to \infty} \sum_{k=1}^n \xi_{t_k} (C_{t_k} - C_{t_{k-1}}),
\]
where \( \Pi_n = \{t_0^{(n)}, \ldots, t_{m}^{(n)}\}, \ n \geq 1 \) is a sequence of partitions of \([0, t]\) with max\( t_i^{(n)} - t_{i-1}^{(n)} \to 0 \) converging to zero as \( n \to \infty \). Since \( \xi \) and \( C \) are both G-martingale, we have
\[
\hat{E} [\xi_{t_k} (C_{t_k}^{(n)} - C_{t_{k-1}}^{(n)})] = 0,
\]
and thus \( \hat{E} [\xi_{t_k} (A'_t - A''_t)] = 0 \).

For an arbitrary bounded random variable \( \xi \), we can select \( \{\xi_t\} \) to be a right continuous equivalent process of \( \hat{E}[\xi | \mathcal{F}_t] \), and we obtain that \( \hat{E}[\xi (A'_t - A''_t)] = 0 \). We set \( \xi = 1_{A'_t \neq A''_t} \); therefore \( c(A'_t \neq A''_t) = 0 \).

By Theorem 24 and G-martingale decomposition theorem in [14, 21], we have the following G-Doob-Meyer theorem.

**Theorem 25.** \( \{X_t\} \) is a right continuous G-supermartingale; there exists a right continuous increasing process \( A_t \) and adapted process \( \eta_t \), such that
\[
X_t = \int_0^t \eta_s \, dB_s - A_t, \quad (56)
\]
where \( B_t \) is G-Brownian motion.

### 4. Superhedging Strategies and Optimal Stopping

#### 4.1. Financial Model and G-Asset Price System

We consider a financial market with a nonrisky asset (bond) and a risky asset (stock) continuously trading in market. The price \( P(t) \) of the bond is given by
\[
dP(t) = rP(t) \, dt \quad P(0) = 1,
\]
where \( r \) is the short interest rate; we assume a constant nonnegative short interest rate. We assume the risk asset with the G-asset price system \( (\mathcal{S}_t, \mathcal{F}_t, \hat{P}) \) (see [17]) on sublinear expectation space \((\Omega, \mathcal{F}, \hat{E}, \hat{P}, (\mathcal{F}_t))\) under Knightian uncertainty, for given \( t \in [0, T] \) and \( x \in \mathbb{R} \)
\[
dS_t^x = S_t^x \, dB_t + S_t^x \left( \mu_t \, dt + \partial B_t \right),
\]
where \( B_t \) is the generalized G-Brownian motion. The uncertain volatility is described by the G-Brownian motion \( \hat{B}_t \). The uncertain drift \( \mu_t \) can be rewritten as
\[
\mu_t = \int_0^t \mu_u \, du,
\]
where \( \mu_t \) is the asset return rate [22]. Then the uncertain risk premium of the G-asset price system
\[
\theta_t = \mu_t - r,
\]
is uncertain and distributed by \( N([\mu-r, \mu-r], \{0\}) \) [22], where \( r \) is the interest rate of the bond.

Define
\[
\tilde{B}_t := B_t - rt = b_t + \tilde{B}_t - rt;
\]
we have the following G-Girsanov theorem (presented in [17, 23]).

**Theorem 26** (G-Girsanov theorem). Assume that \((B_t)_{t \geq 0}\) is generalized G-Brownian motion on \((\Omega, \mathcal{F}, \mathcal{E}, \mathcal{G})\), and \(\tilde{B}_t\) is defined by (61); there exists G-expectation space \((\Omega, \mathcal{G}, \mathcal{H}, \mathcal{E})\) such that \(\tilde{B}_t\) is G-Brownian motion under the G-expectation \(\mathcal{G}\), and
\[
\tilde{E} \left[ \tilde{B}_t^2 \right] = \mathcal{G} \left[ \tilde{B}_t^2 \right],
\]
\[
-\tilde{E} \left[ \tilde{B}_t^2 \right] = -\mathcal{G} \left[ \tilde{B}_t^2 \right].
\]

By the G-Girsanov theorem, the G-asset price system (58) of the risky asset can be rewritten on \((\Omega, \mathcal{F}, \mathcal{E}, \mathcal{G})\) as follows:
\[
dS^x_u = S^x_u (r dt + d\tilde{B}_t),
\]
then by G-Ito formula we have
\[
S^x_u = x \exp \left( r (u - t) + \tilde{B}_u - \frac{1}{2} \left( \langle \tilde{B} \rangle_u - \langle \tilde{B} \rangle_t \right) \right),
\]
4.2. Construction of Superreplication Strategies via Optimal Stopping. We consider the following class of contingent claims.

**Definition 27.** One defines a class of contingent claims with the nonnegative payoff \( \xi \in L_\mathcal{G}^+(\Omega_T) \) having the following form:
\[
\xi = f \left( S^{x,x}_T \right)
\]
for some function \( f: \Omega \to R \) such that the process
\[
f_u := f \left( S^{x,x}_u \right)
\]
is bounded below and càdlàg.

We consider a contingent claim \( \xi \) with payoff defined in Definition 27 written on the stock \( S_t \) with maturity \( T \). We give definitions of superhedging (resp., subhedging) strategy and ask (resp., bid) price of the claim \( \xi \).

**Definition 28.** (1) A self-financing superstrategy (resp. substrategy) is a vector process \((Y, \pi, C)\) (resp., \((-Y, \pi, C)\)) such that \( Y \) is the wealth process, \( \pi \) is the portfolio process, and \( C \) is the cumulative consumption process, such that
\[
dY_t = rY_t dt + \pi_t d\tilde{B}_t - dC_t,
\]
(67)
\[
\text{resp. } -dY_t = -rY_t dt + \pi_t d\tilde{B}_t - dC_t,
\]
where \( C \) is an increasing, right continuous process with \( C_0 = 0 \). The superstrategy (resp., substrategy) is called feasible if the constraint of nonnegative wealth holds
\[
Y_t \geq 0, \quad t \in [0, T].
\]
(68)
(2) A superhedging (resp. subhedging) strategy against the contingent claim \( \xi \) is a feasible self-financing superstrategy \((Y, \pi, C)\) (resp., substrategy \((-Y, \pi, C)\)) such that \( Y_T = \xi \) (resp., \(-Y_T = -\xi\)). We denote by \( \mathcal{H}(\xi) \) (resp., \( \mathcal{H}'(-\xi) \)) the class of superhedging (resp., subhedging) strategies against \( \xi \), and if \( \mathcal{H}(\xi) \) (resp., \( \mathcal{H}'(-\xi) \)) is nonempty, \( \xi \) is called superhedgable (resp., subhedgable).

(3) The ask-price \( X(t) \) at time \( t \) of the superhedgable claim \( \xi \) is defined as
\[
X(t) = \inf \{ x \geq 0 : \exists (Y_t, \pi_t, C_t) \in \mathcal{H}(\xi) \quad \text{such that} \quad Y_t = x \},
\]
(69)
and bid-price \( X'(t) \) at time \( t \) of the subhedgable claim \( \xi \) is defined as
\[
X'(t) = \sup \{ x \geq 0 : \exists (-Y_t, \pi_t, C_t) \in \mathcal{H}'(-\xi) \quad \text{such that} \quad -Y_t = -x \}.
\]
(70)
Under uncertainty, the market is incomplete and the superhedging (resp., subhedging) strategy of the claim is not unique. The definition of the ask-price \( X(t) \) implies that the ask-price \( X(t) \) is the minimum amount of risk for the buyer to superhedging the claim; then it is coherent measure of risk of all superstrategies against the claim for the buyer. The coherent risk measure of all superstrategies against the claim can be regarded as the sublinear expectation of the claim; we have the following representation of bid-ask price of the claim via optimal stopping (Theorem 31).

Let \( \mathcal{G}_t \) be a filtration on G-expectation space \((\Omega, \mathcal{G}, \mathcal{H}, \mathcal{E}, (\mathcal{G}_t)_{t \geq 0})\), and \( \tau_1 \) and \( \tau_2 \) be \( \mathcal{G}_t \)-stopping times such that \( \tau_1 \leq \tau_2 \) a.s. We denote by \( \mathcal{G}_{\tau_1, \tau_2} \) the set of all finite \( \mathcal{G}_t \)-stopping times \( \tau \) with \( \tau_1 \leq \tau \leq \tau_2 \).

For given \( t \in [0, T] \) and \( x \in \mathbb{R}_+ \), we define the function \( V^{\inf}_{x} : [0, T] \times \Omega \to \mathbb{R} \) as the value function of the following optimal-stopping problem:
\[
V^{\inf}_{x}(t, S_t) := \sup_{\tau \in \mathcal{G}_{\tau_1, \tau_2}} \mathcal{E} \left[ f \left( S_{\tau} \right) \right].
\]
Proposition 29. Consider two stopping times \( \tau \leq \hat{\tau} \) on filtration \( \mathcal{F} \). Let \( (f_t)_{t \geq 0} \) denote some adapted and RCLL-stochastic process, which is bounded below. Then we have for two points \( s, t \in [0, \hat{\tau}] \) and \( s < t \)
\[
\sup_{\tau \in \mathcal{F}_{t, \hat{\tau}}} \{ E^G_T [ f_{\tau} ] \} = E^G_s \left[ \sup_{\tau \in \mathcal{F}_{s, t}} \{ E^G_T [ f_{\tau} ] \} \right].
\] (72)

Proof. By the consistent property of the conditional G-expectation, for \( \tau \in \mathcal{F}_{s, t}, s, t \in [0, \hat{\tau}], \) and \( s < t \)
\[
E^G_s \left[ E^G_{t} [ f_{\tau} ] \right] = E^G_s \left[ \sup_{\tau \in \mathcal{F}_{s, t}} \{ E^G_T [ f_{\tau} ] \} \right];
\] (73)

thus we have
\[
\sup_{\tau \in \mathcal{F}_{s, t}} \{ E^G_T [ f_{\tau} ] \} \leq E^G_s \left[ \sup_{\tau \in \mathcal{F}_{s, t}} \{ E^G_T [ f_{\tau} ] \} \right].
\] (74)

There exists a sequence \( \{ \tau_n \} \rightarrow \tau^* \in [\tau, \hat{\tau}] \) as \( n \rightarrow \infty \), such that
\[
\lim_{n \rightarrow \infty} E^G_{\tau_n} [ f_{\tau_n} ] - E^G_{t} [ f_{\tau} ] = \sup_{\tau \in \mathcal{F}_{s, t}} \{ E^G_T [ f_{\tau} ] \};
\] (75)

notice that
\[
E^G_s \left[ \sup_{\tau \in \mathcal{F}_{s, t}} \{ E^G_T [ f_{\tau} ] \} \right] = E^G_s \left[ E^G_{t} [ f_{\tau} ] \right] = E^G_{t} [ f_{\tau} ];
\] (76)

we prove the Proposition. \( \square \)

Proposition 30. The process \( V_{\text{Am}}(t, S_t)_{t \geq 0} \) is a G-supermartingale in \( (\Omega, \mathcal{F}, P, \mathcal{F}, \mathcal{F}_t) \).

Proof. By Proposition 29, for \( 0 \leq s \leq t \leq T \)
\[
E^G_s \left[ \sup_{\tau \in \mathcal{F}_t, \hat{\tau}} \{ E^G_{\tau} [ f (S) ] \} \right] = \sup_{\tau \in \mathcal{F}_t, \hat{\tau}} E^G_s \left[ f (S) \right].
\] (77)

Since \( \mathcal{F}_t, \hat{\tau} \subseteq \mathcal{F}_s, \hat{\tau} \), we have
\[
\sup_{\tau \in \mathcal{F}_t, \hat{\tau}} E^G_s \left[ f (S) \right] \leq \sup_{\tau \in \mathcal{F}_s, \hat{\tau}} E^G_s \left[ f (S) \right].
\] (78)

Thus, we derive that
\[
E^G_s \left[ \sup_{\tau \in \mathcal{F}_t, \hat{\tau}} E^G_{\tau} [ f (S) ] \right] \leq \sup_{\tau \in \mathcal{F}_s, \hat{\tau}} E^G_s \left[ f (S) \right].
\] (79)

We prove the Proposition. \( \square \)

Theorem 31. Assume that the financial market under uncertainty consists of the bond which has the price process satisfying (57) and risky assets with the price processes as the G-asset price systems (58) and can trade freely; the contingent claim \( \xi \) which is written on the risky assets with the maturity \( T > 0 \) has the class of the payoff defined in Definition 27, and the function \( V_{\text{Am}}(t, S_t) \) is defined in (71). Then there exists a superhedging (resp., subhedging) strategy for \( \xi \), such that the process \( V = (V_t)_{0 \leq t \leq T} \) defined by
\[
V_t := e^{-\rho(T-t)} V_{\text{Am}}(t, S_t),
\] (80)
is the ask (resp., bid) price process against \( \xi \).

Proof. The value function for the optimal stop time \( V_{\text{Am}}(t, S_t) \) is a G-supermartingale; it is easily to check that \( e^{-\rho t} V_t \) is G-supermartingale. By G-Doob-Meyer decomposition
\[
e^{-\rho t} V_t = M_t - C_t,
\] (81)
where \( M_t \) is a G-martingale and \( C_t \) is an increasing process with \( C_0 = 0 \). By G-martingale representation theorem [14, 21]
\[
M_t = E^G \{ M_T \} + \int_0^T \eta_s d\tilde{B}_s - K_t,
\] (82)
where \( \eta_t \in H^1_{\mathcal{G}}(0, T) \), \( -K_t \) is a G-martingale, and \( C_t \) is an increasing process with \( C_0 = 0 \). From the above equation, we have
\[
e^{-\rho t} V_t = E^G \{ M_T \} + \int_0^T \eta_t d\tilde{B}_s - (K_t + C_t),
\] (83)
hence \( (V_t, e^{\int \eta_t ds}, \int_0^T e^{\int \eta_t ds} d(C_t - K_t) ds) \) is a superhedging strategy. Assume that \( (Y_t, \pi_t, C_t) \) is a superhedging strategy against \( \xi \); then
\[
e^{-\rho t} Y_t = e^{-\rho T} \xi - \int_t^T \pi_s d\tilde{B}_s + C_t,
\] (84)
Taking conditional G-expectation on both sides of (84) and noticing that the process \( C_t \) is an increasing process with \( C_0 = 0 \), we derive
\[
e^{-\rho t} Y_t \geq E^G_t \left[ e^{-\rho T} \xi_t \right],
\] (85)
which implies that
\[
Y_t \geq E^G_t \left[ e^{-\rho(T-t)} \xi_t \right] \geq E^G_t \left[ e^{-\rho(T-t)} \sup_{\tau \in \mathcal{F}_t} \{ f_{\tau} \} \right]
\] (86)
\[
\geq e^{-\rho(T-t)} \sup_{\tau \in \mathcal{F}_t} E^G_t \left[ f_{\tau} \right] \geq e^{-\rho(T-t)} \sup_{\tau \in \mathcal{F}_t} E^G_t \left[ f_{\tau} \right] = V_t
\]
from which we prove that \( V_t = e^{-\rho(T-t)} V_{\text{Am}}(t, S_t) \) is the ask price against the claim \( \xi \) at time \( t \). Similarly we can prove that \( -e^{-\rho(T-t)} \sup_{\tau \in \mathcal{F}_t} E^G_t \left[ -f_{\tau} \right] \) is the bid price against the claim \( \xi \) at time \( t \). \( \square \)
5. Free Boundary and Optimal Stopping Problems

For given \( t \in [0, T] \), \( x \in \mathbb{R}^d \), and \( d = 1 \), the G-asset price system (58) of the risky asset can be rewritten as follows:

\[
\begin{align*}
\frac{dS_t^x}{S_t^x} &= S_t^x \left( rd t + dB_t \right), \\
S_0^x &= x.
\end{align*}
\]  

(87)

We define the following deterministic function:

\[
u^a(t, x) := e^{-r(T-t)} \mathcal{V}^{Am}(t, S_t^x),
\]  

(88)

where

\[
\mathcal{V}^{Am}(t, S_t^x) = \sup_{ s \in \mathcal{F} \cap [t, T]} \left[ f(S_s) \right].
\]  

(89)

From Theorem 31, the price of an American option with expiry date \( T \) and payoff function \( f \) is the value function of the optimal stopping problem:

\[
u^a(t, x) := e^{-r(T-t)} \sup_{ s \in \mathcal{F} \cap [t, T]} \left[ f(S_s) \right].
\]  

(90)

We define operator \( L \) as follows:

\[
Lu = G(D^2 u) + r Du + \partial_t u,
\]  

(91)

where \( G(\cdot) \) is the sublinear function defined by (9). We consider the free boundary problem

\[
\begin{align*}
\mathcal{L}u := & \max \left\{ Lu - ru, f - u \right\} = 0, \quad \text{in } [0, T] \times \mathbb{R}, \\
u(T, \cdot) = f(T, \cdot), \quad \text{in } \mathbb{R}.
\end{align*}
\]  

(92)

Denote

\[
\delta_T := [0, T] \times \mathbb{R},
\]  

(93)

for \( p \geq 1 \)

\[
\delta^p(\delta_T) := \left\{ u \in L^p(\delta_T) : D^p u, Du, \partial_t u \in L^p(\delta_T) \right\}.
\]  

(94)

And, for any compact subset \( D \) of \( \delta_T \), we denote \( \delta^p(\delta_T) \) as the space of functions \( u \) in \( \delta^p(\delta_T) \).

Definition 32. A function \( u \in \delta^1_{log}(\delta_T) \cap C(\mathbb{R} \times [0, T]) \) is a strong solution of problem (92) if \( \mathcal{L}u = 0 \) almost everywhere in \( \delta_T \) and it attains the final datum pointwise. A function \( u \in \delta^1_{log}(\delta_T) \cap C(\mathbb{R} \times [0, T]) \) is a strong supersolution of problem (92) if \( \mathcal{L}u \leq 0 \).

We will prove the following existence results.

Theorem 33. If there exists a strong supersolution \( \overline{u} \) of problem (92) then there also exists a strong solution \( u \) of (92) such that \( u \leq \overline{u} \) in \( \delta_T \). Moreover \( u \in \delta^p_{loc}(\delta_T) \) for any \( p \geq 1 \) and consequently, by the embedding theorem we have \( u \in C^{1,\alpha}_{\delta_T} \) for any \( \alpha \in [0, 1] \).

Theorem 34. Let \( u \) be a strong solution to the free boundary problem (92) such that

\[
|u(t, x)| \leq Ce^{|x|^2}, \quad (t, x) \in \delta_T
\]  

(95)

form some constants \( C, \lambda \) with \( \lambda \) sufficiently small so that

\[
E^G \left[ \exp \left( \lambda \sup_{t \leq T} |u_{\delta_T}^x(t, x)|^2 \right) \right] < \infty
\]  

(96)

holds. Then we have

\[
u(t, x) = e^{-r(T-t)} \sup_{ s \in \mathcal{F} \cap [t, T]} \left[ f(S_s) \right];
\]  

(97)

that is, the solution of the free boundary problem is the value function of the optimal stopping problem. In particular such a solution is unique.

5.1. Proof of Theorem 34. We employ a truncation and regularization technique to exploit the weak interior regularity properties of \( u \); for \( R > 0 \) we set for \( \mathbb{R}^d \) and \( |x| < R \), and, for \( x \in \mathbb{R} \) denoting by \( \tau_x \) the first exit time of \( S_x^\epsilon \) from \( B_R \), it is easy check that \( E^G[\tau_x] \) is finite. As a first step we prove the following result: for every \( (t, x) \in [0, T] \times B_R \) and \( \tau \in \mathcal{F}_{1, T} \) such that \( \tau \in [t, \tau_x] \), it holds that

\[
u(t, x) = E^G \left[ u(t, S_{\tau}^x) \right] - E^G \left[ \int_t^{\tau} Lu(s, S_s^x) \, ds \right].
\]  

(98)

For fixed, positive, and small enough \( \epsilon \), we consider a function \( u^{\epsilon, R} \) on \( \mathbb{R} \) with compact support and such that \( u^{\epsilon, R} = u \) on \( |t, T - \epsilon| \times B_R \). Moreover we denote by \( (u^{\epsilon, R, n})_{n \in \mathbb{N}} \) a regularizing sequence obtained by convolution of \( u^{\epsilon, R} \) with the usual mollifiers; then for any \( n \geq 1 \) we have \( u^{\epsilon, R, n} \in \delta^p(\mathbb{R}^d) \) and

\[
\lim_{n \to \infty} \left\| u^{\epsilon, R, n} - u^{\epsilon, R} \right\|_{L^p([t, T - \epsilon] \times B_R)} = 0.
\]  

(99)

By G-Itô formula we have

\[
\begin{align*}
\frac{d}{ds} u^{\epsilon, R, n}(\tau, S^{\epsilon, R, n}_s) &= u^{\epsilon, R, n}(t, x) + \frac{1}{2} \int_t^{\tau} D^2 u^{\epsilon, R, n} \, dB_s \\
&\quad + \int_t^{\tau} r Du^{\epsilon, R, n} \, ds + \int_t^{\tau} \partial_t u^{\epsilon, R, n} \, ds + \int_t^{\tau} Du^{\epsilon, R, n} \, dB_s,
\end{align*}
\]  

(100)

which implies that

\[
E^G \left[ u^{\epsilon, R, n}(\tau, S^{\epsilon, R, n}_\tau) \right] = u^{\epsilon, R, n}(t, x) + \int_t^{\tau} Lu^{\epsilon, R, n} \, ds.
\]  

(101)

We have

\[
\lim_{n \to \infty} u^{\epsilon, R, n}(t, x) = u^{\epsilon, R}(t, x)
\]  

(102)

and, by dominated convergence,

\[
\lim_{n \to \infty} E^G \left[ u^{\epsilon, R, n}(\tau, S^{\epsilon, R, n}_\tau) \right] = E^G \left[ u^{\epsilon, R}(\tau, S^{\epsilon, R}_\tau) \right].
\]  

(103)
We have
\[
|E^G\left[ \int_t^T L^e_{t,R,n}(s,S_t^x) \, ds \right] | - E^G\left[ \int_t^T L^e_R(s,S_t^x) \, ds \right] \leq E^G\left[ \int_t^T \left| L^e_{t,R,n}(s,S_t^x) - L^e_R(s,S_t^x) \right| \, ds \right] ;
\]
by sublinear expectation representation theorem (see [14]) there exists a family of probability space Q, such that
\[
E^G\left[ \int_t^T \left| L^e_{t,R,n}(s,S_t^x) - L^e_R(s,S_t^x) \right| \, ds \right] = \text{ess sup}_{P \in Q} \left[ \int_t^T \left| L^e_{t,R,n}(s,S_t^x) - L^e_R(s,S_t^x) \right| \, ds \right].
\]
Since \( \tau \leq \tau_R \)
\[
\text{ess sup}_{P \in Q} \int_t^T \left| L^e_{t,R,n}(s,S_t^x) - L^e_R(s,S_t^x) \right| \, ds \leq \text{ess sup}_{P \in Q} \int_t^T \left| L^e_{t,R}(s,y) - L^e_R(s,y) \right| \, ds,
\]
where \( \Gamma_p(t,x,\cdot,\cdot) \in L^q([t,T] \times B_2) \), for some \( q > 1 \), is the transition density of the solution of
\[
dx_{t,s}^x = x_{t,s}^1 \, (r \, ds + \sigma_{s,p} \, dW_{s,p}),
\]
where \( W_{t,p} \) is Wiener process in probability space \( (\Omega_1,P,\mathcal{F}_t,\mathbb{P}_t) \) and \( \sigma_{s,p} \) is adapted process such that \( \sigma_{s,p} \in [\sigma,\overline{\sigma}] \). By Hölder inequality, we have \((1/p + 1/q = 1)\)
\[
\int_t^T \int_{B_2} \left| L^e_{t,R}(s,y) - L^e_R(s,y) \right| \, ds \, dy \leq \left\| L^e_{t,R}(s,y) \right\|_{L^q([t,T] \times B_2)} \left\| \Gamma_p(t,x,s,y) \right\|_{L^p([t,T] \times B_2)},
\]
and then we obtain that
\[
\lim_{n \to \infty} E^G\left[ \int_t^T L^e_{t,R,n}(s,S_t^x) \, ds \right] = E^G\left[ \int_t^T L^e_R(s,S_t^x) \, ds \right] ;
\]
This concludes the proof of (98), since \( u^e_R = u_0(t,T - \epsilon) \times B_2 \) and \( \epsilon > 0 \) is arbitrary.

Since \( Lu \leq 0 \), we have for any \( \tau \in \mathcal{F}_{t,T} \)
\[
E^G\int_t^T L u(s,S_t^x) \, ds \leq 0;
\]
we infer from (98) that
\[
\tau \left( u(t,x) \geq E^G\left[ \left. u \left( \tau \land \tau_R, S_{t,R}, S_{\tau,R}^x \right) \right| \tau \right] \right)
\]
Next we pass to the limit as \( R \to +\infty \): we have
\[
\lim_{R \to +\infty} \tau \land \tau_R = \tau,
\]
and by the growth assumption (95)
\[
\left| u \left( \tau \land \tau_R, S_{t,R}, S_{\tau,R}^x \right) \right| \leq C \exp (\lambda \sup_{t \in [T] \times B_2} \left| S_{t,R}^x \right|) .
\]
As \( R \to +\infty \)
\[
u(t,x) \geq E^G[u(t,x) \land \lambda] \geq E^G[f(t,x) \land \lambda].
\]
This shows that
\[
u(t,x) \geq \sup_{\tau \in \mathcal{F}_{t,T}} E^G[f(t,x) \land \lambda].
\]
We conclude the proof by putting
\[
\tau_0 = \inf \left\{ s \in [T] \mid u(s,S_t^x) = f(s,S_t^x) \right\} .
\]
Since \( Lu = 0 \) a.e., where \( u > \phi \), it holds
\[
E^G\left[ \int_t^T \left( f(t,x) \right) \, ds \right] = 0
\]
and from (98) we derive that
\[
u(t,x) = E^G[u(t,x) \land \lambda] .
\]
Repeating the previous argument to pass to the limit in \( R \), we obtain
\[
u(t,x) = E^G[u(t,x) \land \lambda] ;
\]
Therefore, we finish the proof.

5.2. Free Boundary Problem. Here we consider the free boundary problem on a bounded cylinder. We denote the bounded cylinders as the form \([0,T] \times H_n\), where \((H_n)\) is an increasing covering of \( R^d \) (\( d = 1 \)). We will prove the existence of a strong solution to problem
\[
\max \{ Lu, f - u \} = 0, \quad \text{in } H(T) := [0,T] \times H,
\]
\[
u|_{\partial_p H(T)} = f,
\]
where \( H \) is a bounded domain of \( R^d \) and
\[
\partial_p H(T) := \partial H(T) \setminus \{(T) \times H\}
\]
is the parabolic boundary of \( H(T) \).

We assume the following condition on the payoff function.
Assumption 35. The payoff function \( \xi = f(S_{t,x}) \) has the following assumption expressed by the sublinear function:

\[
-G(-D^2 f) \geq c \quad \text{in } H,
\]  

(122)

where \( G(\cdot) \) is the sublinear function defined by (9).

Theorem 36. One assumes assumption 5.1 holds. Problem (120) has a strong solution \( u \in \mathcal{S}_{\text{loc}}^p(H(T)) \cap C(H(T)) \). Moreover \( u \in \mathcal{S}_{\text{loc}}^p(H(T)) \) for any \( p > 1 \).

Proof. The proof is based on a standard penalization technique (see [18]). We consider a family \( \{ \beta(\epsilon) \}_{\epsilon \in [0,1]} \) of smooth functions such that, for any \( \epsilon \), function \( \beta(\epsilon) \) is increasing bounded on \( R \) and has bounded first order derivative, such that

\[
\beta(\epsilon)(s) \leq \epsilon, \quad s > 0,
\]

\[
\lim_{\epsilon \to 0} \beta(\epsilon)(s) = -\infty, \quad s < 0.
\]

(123)

We denote by \( f^\delta \) the regularization of \( f \) and consider the following penalized and regularized problem and denote the solution as \( u_{\epsilon,\delta} \)

\[
Lu = \beta(\epsilon)(u - f^\delta), \quad \text{in } H(T),
\]

\[
u_{\delta,H(T)} = f^\delta;
\]

Lions [24], Krylov [25], and Nisio [26] prove that problem (124) has a unique viscosity solution \( u_{\epsilon,\delta} \in C^{\alpha,\alpha}(H(T)) \cap C(H(T)) \) with \( \alpha \in [0,1] \).

Next, we firstly prove the uniform boundedness of the penalization term:

\[
|\beta(\epsilon)(u_{\epsilon,\delta} - f^\delta)| \leq c, \quad \text{in } H(T),
\]

(125)

with \( c \) independent of \( \epsilon \) and \( \delta \).

By construction \( \beta(\epsilon) \leq \epsilon \), it suffices to prove the lower bound in (125). By continuity, \( \beta(\epsilon)(u_{\epsilon,\delta} - f^\delta) \) has a minimum \( \xi \) in \( H(T) \) and we may suppose

\[
\beta(\epsilon)(u_{\epsilon,\delta}(\xi) - f^\delta(\xi)) \leq 0;
\]

otherwise we prove the lower bound. If \( \xi \in \partial_p H(T) \) then

\[
\beta(\epsilon)(u_{\epsilon,\delta}(\xi) - f^\delta(\xi)) = \beta(\epsilon)(0) = 0.
\]

(126)

(127)

On the other hand, if \( \xi \in H(T) \), then we recall that \( \beta(\epsilon) \) is increasing and consequently \( u_{\epsilon,\delta} - f^\delta \) also has a (negative) minimum in \( \xi \). Thus, we have

\[
Lu_{\epsilon,\delta}(\xi) - Lf^\delta(\xi) \geq 0 \geq u_{\epsilon,\delta}(\xi) - f^\delta(\xi).
\]

(128)

By Assumption 35 on \( f \), we have that \( Lf^\delta(\xi) \) is bounded uniformly in \( \delta \). Therefore, by (128), we deduce

\[
\beta(\epsilon)(u_{\epsilon,\delta}(\xi) - f^\delta(\xi)) = Lu_{\epsilon,\delta}(\xi) \geq Lf^\delta(\xi) \geq c,
\]

(129)

where \( c \) is a constant independent of \( \epsilon,\delta \) and this proves (125).

Secondly, we use the \( \delta^p \) interior estimate combined with (125), to infer that, for every compact subset \( D \) in \( H(T) \) and \( p \geq 1 \), the norm \( \|u_{\epsilon,\delta}\|_{\delta^p(D)} \) is bounded uniformly in \( \epsilon \) and \( \delta \).

It follows that \( (u_{\epsilon,\delta}) \) converges as \( \epsilon,\delta \to 0 \) weakly in \( \delta^p \) on compact subsets of \( H(T) \) to a function \( u \).

Moreover

\[
\lim_{\epsilon,\delta \to 0} \beta(\epsilon)(u_{\epsilon,\delta} - f^\delta) \leq 0,
\]

(130)

so that \( Lu \leq f \) a.e. in \( H(T) \). On the other hand, \( Lu = f \) a.e. in set \( \{u > f\} \).

Finally, it is straightforward to verify that \( u \in C(H(T)) \) and assumes the initial-boundary conditions, by using standard arguments based on the maximum principle and barrier functions.

\( \square \)

Proof of Theorem 33. The proof of Theorem 33 about the existence theorem for the free boundary problem on unbounded domains is similar to [27] by using Theorem 36 about the existence theorem for the free boundary problem on the regular bounded cylindrical domain.

\( \square \)

Conflict of Interests

The author declares that there is no conflict of interests regarding the publication of this paper.

References


