

APPLICATION OF FBSDE IN OPTIMAL INVESTMENT PROBLEM

Beso Chikvinidze Revaz Tevzadze

Abstract. The wealth maximization problem for random utility defined on the half-line is considered. For the solution of this problem the system of Forward Backward Stochastic Differential Equations is derived.

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We consider a financial market model, where the dynamics of asset prices is described by the continuous semimartingale S defined on the complete probability space (Ω, \mathcal{F}, P) with continuous filtration $F = (F_t, t \in [0, T])$, where $\mathcal{F} = F_T$ and $T < \infty$. We work with discounted terms, i.e. the bond is assumed to be a constant.

Suppose that

$$S_t = M_t + \int_0^t \lambda_s d\langle M \rangle_s, \quad \int_0^t \lambda_s^2 d\langle M \rangle_s < \infty$$

for all t P -a.s., where M is a continuous local martingale and λ is a predictable process.

Let $\hat{U} = \hat{U}(x, \omega) : \mathbb{R}_+ \times \Omega \rightarrow \mathbb{R}$ be a random utility function, such that \hat{U} is measurable and continuously differentiable, increasing, strictly concave P -a.s.. Let also be given the non-random utility function $U = U(x) : \mathbb{R}_+ \rightarrow \mathbb{R}$, tree-times continuously differentiable, increasing, strictly concave and $U'(\infty) = 0$, $U'(0) = \infty$.

We denote by Π^x the set of admissible strategies with initial capital x which is now defined by

$$\Pi^x := \{ \pi \text{ is predictable, } P\left(\int_0^T |\pi_s|^2 d\langle M \rangle_s < \infty\right) = 1 \}.$$

The associated wealth process is given by $X_t^\pi := x + \int_0^t \pi_s X_s^\pi dS_s$, $t \in [0, T]$. We consider the maximization problem

$$\max_{\pi \in \Pi^x} E\hat{U}(X_T^\pi). \tag{1}$$

We furthermore need to impose the following assumptions on \hat{U} .

(B) There exists the maximizer (π^*, X^*) of problem (1), i.e. $E\hat{U}(X_T^{\pi^*}) = \max_{\pi \in \Pi^x} E\hat{U}(X_T^\pi)$, $X^* = X^{\pi^*}$, and $\hat{U}'(X^*)$ is the density of the martingale measure.

Theorem 1. *Let condition B) be satisfied. Then*

$$\pi_t^* = -\frac{U'_t(X_t^*)}{X_t^* U''_t(X_t^*)} \left(\lambda_t + \frac{\psi_t}{P_t} \right),$$

where the triple (P, ψ, L) , $L \perp M$ is the solution of BSDE

$$dP_t = P_t \left[\left(\lambda_t + \frac{\psi_t}{P_t} \right)^2 - \frac{1}{2} \frac{U_t'''(X_t^*) U_t'(X_t^*)}{U_t''(X_t^*)^2} \left(\lambda_t + \frac{\psi_t}{P_t} \right)^2 \right] d\langle M \rangle_t \\ + \psi_t dM_t + dL_t, \quad P_T = \frac{\hat{U}'(X_T^*)}{U'(X_T^*)}.$$

Proof. It is easy to verify that $X_t^* \left(\int_0^t h_u (\lambda_u - \pi_u^*) d\langle M \rangle_u + \int_0^t h_u dM_u \right)$ is a local martingale for each bounded, predictable h w.r.t martingale measure. This means that $E[X_T^* \hat{U}'(X_T^*) | F_t] \left(\int_0^t h_u (\lambda_u - \pi_u^*) d\langle M \rangle_u + \int_0^t h_u dM_u \right)$ is a local martingale.

Let $P_t = \frac{E[X_T^* \hat{U}'(X_T^*) | F_t]}{X_t^* U'(X_t^*)}$. Then by the Ito formula we have decomposition

$$P_t = P_0 + \int_0^t \alpha_u d\langle M \rangle_u + \int_0^t \psi_u dM_u + L_t. \quad (2)$$

Using the Ito formula for $F(X_t^*, P_t, H_t) = P_t X_t^* U'(X_t^*) H_t$, where $H_t = H_0 + \int_0^t h_u (\lambda_u - \pi_u^*) d\langle M \rangle_u + \int_0^t h_u dM_u$, we obtain

$$\begin{aligned} & dF(X_t^*, P_t, H_t) \\ &= [P_t(X_t^* U''(X_t^*) + U'(X_t^*)) H_t \pi_t^* X_t^* \lambda_t + X_t^* U'(X_t^*) H_t \alpha_t \\ &+ P_t X_t^* U'(X_t^*) (\lambda_t - \pi_t^*) h_t + P_t (U''(X_t^*) + \frac{1}{2} X_t^* U'''(X_t^*)) H_t (\pi_t^*)^2 (X_t^*)^2 \\ &\quad + U'(X_t^*) \psi_t h_t X_t^* + (X_t^* U''(X_t^*) + U'(X_t^*)) H_t \psi_t \pi_t^* X_t^* \\ &\quad + (X_t^* U''(X_t^*) + U'(X_t^*)) P_t h_t \pi_t^* X_t^*] d\langle M \rangle_t + \text{local martingale} \\ &= \{ H_t [P_t (X_t^* U''(X_t^*) + U'(X_t^*)) \pi_t^* X_t^* \lambda_t + X_t^* U'(X_t^*) \alpha_t \\ &\quad + P_t (U''(X_t^*) + \frac{1}{2} X_t^* U'''(X_t^*)) (\pi_t^*)^2 (X_t^*)^2 + (X_t^* U''(X_t^*) + U'(X_t^*)) \psi_t \pi_t^* X_t^*] \\ &\quad + h_t [U'(X_t^*) \psi_t X_t^* \\ &\quad + P_t X_t^* U'(X_t^*) (\lambda_t - \pi_t^*) + (X_t^* U''(X_t^*) + U'(X_t^*)) P_t \pi_t^* X_t^*] \} d\langle M \rangle_t \\ &\quad + \text{local martingale} \end{aligned}$$

Equalizing coefficients of H and h to zero we get

$$\begin{aligned} & P_t (X_t^* U''(X_t^*) + U'(X_t^*)) \pi_t^* \lambda_t + U'(X_t^*) \alpha_t \\ &+ P_t (U''(X_t^*) + \frac{1}{2} X_t^* U'''(X_t^*)) (\pi_t^*)^2 X_t^* + (X_t^* U''(X_t^*) + U'(X_t^*)) \psi_t \pi_t^* = 0 \end{aligned}$$

and

$$P_t U'(X_t^*) (\lambda_t - \pi_t^*) + U'(X_t^*) \psi_t + (X_t^* U''(X_t^*) + U'(X_t^*)) P_t \pi_t^* = 0.$$

Hence

$$\alpha_t = -\frac{(P_t\lambda_t + \psi_t)(X_t^*U''(X_t^*) + U'(X_t^*))}{U'(X_t^*)}\pi_t^* \quad (3)$$

$$-\frac{P_t(U''(X_t^*) + \frac{1}{2}X_t^*U'''(X_t^*))X_t^*}{U'(X_t^*)}|\pi_t^*|^2, \quad (4)$$

$$\pi_t^* = -\frac{U'(X_t^*)}{X_t^*U''(X_t^*)}\left(\lambda_t + \frac{\psi_t}{P_t}\right).$$

Plugging π^* into (3) yields

$$\begin{aligned} \alpha_t &= \left(\lambda_t + \frac{\psi_t}{P_t}\right)P_t\left[\frac{(X_t^*U''(X_t^*) + U'(X_t^*))}{U'(X_t^*)}\frac{U'(X_t^*)}{X_t^*U''(X_t^*)}\right. \\ &\quad \left.-\frac{U''(X_t^*) + \frac{1}{2}X_t^*U'''(X_t^*)}{U'(X_t^*)}\frac{U'(X_t^*)^2}{X_t^*U''(X_t^*)^2}\right] \\ &= \left(\lambda_t + \frac{\psi_t}{P_t}\right)P_t\left(1 - \frac{1}{2}\frac{U'''(X_t^*)U'(X_t^*)}{U''(X_t^*)^2}\right), \end{aligned}$$

Therefore we get

$$dP_t = P_t\left(\lambda_t + \frac{\psi_t}{P_t}\right)^2\left(1 - \frac{1}{2}\frac{U'''(X_t^*)U'(X_t^*)}{U''(X_t^*)^2}\right)d\langle M \rangle_t + \psi_t dM_t + dL_t, \quad P_T = \frac{\hat{U}'(X_T^*)}{U'(X_T^*)}.$$

□

Corollary 1. *The quadruple (X^*, Y, Z, N) , where $Y_t = \ln P_t$, $Z_t = \frac{\psi_t}{P_t}$, $N_t = \int_0^t \frac{1}{P_s} dL_s$, is the solution of the Forward Backward Stochastic Differential Equation (FBSDE)*

$$\begin{aligned} dX_t^* &= -\frac{U'(X_t^*)(\lambda_t + Z_t)}{U''(X_t^*)}dS_t, \quad X_0^* = x, \\ dY_t &= \left[(\lambda_t + Z_t)^2 - \frac{1}{2}\frac{U'''(X_t^*)U'(X_t^*)}{U''(X_t^*)^2}(\lambda_t + Z_t)^2\right]d\langle M \rangle_t \\ &\quad - \frac{1}{2}Z_t^2d\langle M \rangle_t - \frac{1}{2}d\langle N \rangle_t + Z_t dM_t + dN_t, \quad N \perp M, \\ Y_T &= \ln \frac{\hat{U}'(X_T^*)}{U'(X_T^*)}. \end{aligned}$$

Remark 1. The FBSDE for the case $\hat{U}(x) = U(x + H)$ was considered in [1] and [2]. We can show, that (X^*, Y) , where $Y_t = (U')^{-1}(P_t U'(X_t)) - X_t^*$, is the solution of FBSDE (3.10) of [1]

R E F E R E N C E S

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Author(s) address(es):

Beso Chikvinidze
Institute of Cybernetics
Georgian Technical University
S. Euli str. 5, 0186 Tbilisi, Georgia
E-mail: beso.chiqvinidze@gmail.com

Revaz Tevzadze
Institute of Cybernetics
Georgian Technical University
S. Euli str. 5, 0186 Tbilisi, Georgia
E-mail: rtevzadze@gmail.com