

Optimal Portfolio Strategy with Discounted Stochastic Cash Inflows

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ABSTRACT

This paper examines optimal portfolios with discounted stochastic cash inflows (SCI). The cash inflows are invested into a market that is characterized by inflation-linked bond, a stock and a cash account. It was assumed that inflation-linked bond, stock and the cash inflows are stochastic and follow a standard geometric Brownian motion. The variational form of Merton portfolio strategy was obtained by assuming that the investor chooses constant relative risk averse (CRRA) utility function. The inter-temporal hedging terms that offset any shock to the SCI were obtained. A closed form solution to our resulting non-linear partial differential equation was obtained.

Keywords: Optimal Portfolio; Stochastic Cash Inflows; Inflation-Linked Bond; Variational Form; Intertemporal Hedging Terms

1. Introduction

This paper consider the optimal portfolio strategies with valuing expected discounted cash inflows. It was assumed that the underlying assets and cash inflow process follow a standard geometric Brownian motion. The investment of the investor's SCI into a cash account, an inflation-linked bond and a stock was considered. The SCI and stock price were correlated with inflation and stock market risk.

In a related literature, [1] studied an optimal investment problem in a continuous-time framework where the interest rate follows the Cox-Ingersoll-Ross dynamics. They obtained a closed form solution for the optimal investment strategy under a complete market framework. They assumed that the investor chooses CRRA utility function. [2] considered the process of finding the optimal portfolio, optimal consumption and the efficient frontier for a small agent in an economy. They also considered financial market that composed of two sources of uncertainties: an m-dimensional Brownian motion and a continuous time Markov chain. They found theoretically, that the regimes of an economy have a significant impacts on the portfolio and consumption decision as well as on the efficient frontier of a small investor. [3] studied the present value and overcome the difficulty of independence by reversing the order of the cash flow. He found that similar recursive formulas for the present value is also applicable to the future value of the ex-

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pected returns. [4] considered the problem of constructing a portfolio of finitely many assets whose returns are described by a discrete joint distribution. They proposed a new portfolio optimization model involving stochastic dominance constraints on the portfolio return. [5] studied how portfolio composition changes with individual wealth. [6] considered the present value of expected future contribution into the pension fund. [7] investigated the stochastic dynamics of depository financial institution assets, liabilities and capital under the influence of macroeconomic factors. They adopted dynamic programming techniques in their optimization process. They provided an analysis of the economic aspects of the depository financial institution modeling. [8] studied a portfolio problem of a fund manager who wants to maximize the expected utility of his terminal wealth in a complete financial market. He found that the optimal portfolio is formed by three components: a speculative, an hedging, and a preference-free hedging component. He obtained a close form solution to the asset allocation problem. [9] considered the optimal portfolio selection problem with portfolio constraints. They derived the general utility function using the martingale approach. They found that using CRRA utility function, optimal policies can be obtained explicitly when there are minimum capital requirements. ([10,11]) studied the optimal portfolio management in the accumulation phase of a defined contribution pension scheme. They obtained portfolio values with hedging strategies for a pension plan member. In

this paper, we consider the optimal portfolio and investment strategies involving cash inflow valuation over time. We obtain explicitly, analytical solution to our resulting HJB equation. The risk-free rate is assume to be deterministic.

The remaining parts of the paper is structured as follows: In Section 2, we described the structure of the financial market; In Section 3, we consider the dynamics of the discounted present value of expected SCI process; In Section 4, we consider the wealth process of the investor; In Section 5, we consider the valuation of the discounted present value of expected SCI process of the investor as well as the sensitivity of the present value of the SCI; In Section 6, we consider the optimization program and optimal portfolio and optimal solution of the investor's wealth using CRRA utility function; Section 7 concludes the paper.

2. The Model

Let (Ω, F, P) be a probability space. Let $\mathsf{F}(F) = \{F_t : t \in [0, T]\}$, where

 $F_t = \sigma(S(s), I(s): s \le t)$. The Brownian motions

 $W(t) = (W^{I}(t), W^{S}(t))$ is a 2-dimensional process, defined on a given filtered probability space

 $(\Omega, F, F(F), P)$, $t \in [0, T]$, where P is the real world probability measure, t the time period, T the terminal time period, $W^{T}(t)$ is the Brownian motion with respect to source of uncertainty arising from inflation and $W^{S}(t)$ is the Brownian motion with respect to source of uncertainty arising from the stock market. It is assumed that the market is arbitrage-free, complete and continuously open between time period 0 and T.

Financial Model

The dynamics for the cash account with the price Q(t) at time t is given by

$$dQ(t) = rQ(t)dt,$$

$$Q(0) = 1;$$
(1)

where r is the short term interest rate. The stock price S(t) at time t is given by the dynamics:

$$dS(t) = S(t)(\mu dt + \sigma_1 \cdot dW(t)),$$

$$S(0) = s_0;$$
(2)

where, μ is the expected growth rate of stock price, $\sigma_1 = \left(\rho\sigma_s, \sigma_s\sqrt{1-\rho^2}\right)$ and $0 < \rho < 1$. σ_s is the volatility of stock. The price of the inflation-linked bond B(t, I(t)) is given by the dynamics:

$$dB(t, I(t)) = B(t, I(t))[(r + \sigma_I \theta_I)dt + \sigma_B \cdot dW(t)],$$

$$B(0, I(0)) = b;$$
(3)

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where, $\sigma_B = (\sigma_I, 0)$ is the volatility of inflation-linked bond, θ_I is the market price of inflation risk, I(t) is the inflation index at time t and has the dynamics:

$$\mathrm{d}I(t) = qI(t)\mathrm{d}t + \sigma_I I(t)\mathrm{d}W^I(t),$$

where q is the expected rate of inflation, which is the difference between nominal interest rate, r real interest rate \overline{r} and σ_{I} is the volatility of inflation index. Since the market is complete, we have that

$$\Sigma := \begin{pmatrix} \sigma_B \\ \sigma_1 \end{pmatrix} = \begin{pmatrix} \sigma_I & 0 \\ \rho \sigma_S & \sigma_S \sqrt{1 - \rho^2} \end{pmatrix}, \quad (4)$$

$$\lambda := \left(\sigma_I \theta_I, \mu - r\right)'. \tag{5}$$

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Therefore, the market price of market risk is given by

$$\boldsymbol{\theta} \coloneqq \boldsymbol{\Sigma}^{-1} \boldsymbol{\lambda} = \begin{pmatrix} \boldsymbol{\theta}_I \\ \boldsymbol{\theta}_S \end{pmatrix} = \begin{pmatrix} \boldsymbol{\theta}_I \\ \frac{\boldsymbol{\mu} - \boldsymbol{r} - \boldsymbol{\rho} \boldsymbol{\sigma}_S \boldsymbol{\theta}_I \\ \sigma_S \sqrt{1 - \boldsymbol{\rho}^2} \end{pmatrix}, \quad (6)$$

where, θ_s is the market price of stock market risk. The exponential process

$$Z(t) := \exp\left(-\theta' \cdot W(t) - \frac{1}{2} \|\theta\|^2 t\right), 0 \le t \le T, \text{ is assumed}$$

to be a martingale. We now define the state-price density function by

$$\Lambda(t) \coloneqq \frac{Z(t)}{Q(t)} = Z(t) \exp(-rt),$$

$$0 \le t \le T.$$

3. Dynamics of Stochastic Cash Inflows

The dynamics of the stochastic cash inflows with price process, D(t) is given by

$$dD(t) = D(t)(kdt + \sigma'_D \cdot dW(t)),$$

$$D(0) = D_0;$$
(7)

where, $\sigma_D = (\sigma_1^D, \sigma_2^D)'$ is the volatility of the cash inflows and k is the expected growth rate of the cash inflows. σ_1^D is the volatility arising from inflation and σ_2^D is the volatility arising from the stock market. **Figure 1** presents the simulated diffusion paths of (7).

Figure 1 was obtained by setting k = 0.099,

$$\sigma_1^D = 0.25$$
, $\sigma_2^D = 0.36$, $D_0 = 100$, $dt = \frac{t - t_0}{nt}$, $t = 10$,

 $t_0 = 0$ and nt = 1000.

Applying It \hat{o} lemma to (7), we obtain

$$D(t) = D_0 \exp\left(\left(k - \frac{1}{2} \|\sigma_D\|^2\right) t + \sigma'_D \cdot W(t)\right).$$
(8)



Figure 1. Simulated diffusion paths of the stochastic cash inflows.

4. The Wealth Process

Let X(t) be the wealth process and

 $\Delta(t) = (\Delta_I(t), \Delta_S(t))'$ be the portfolio value at time t, where $\Delta_I(t)$ is the portfolio value in inflation-linked bond and $\Delta_S(t)$ is the portfolio value in stock at time t. Then, $\Delta_0(t) = 1 - \Delta_I(t) - \Delta_S(t)$ is the portfolio value in cash account at time t. Therefore, the dynamics of the wealth process is given by

$$dX(t) = X(t)\Delta_{s}(t)\frac{dS(t)}{S(t)}$$

+X(t)\Delta_{I}(t)\frac{dB(t,I(t))}{B(t,I(t))}
+X(t)(1-\Delta_{I}(t)-\Delta_{s}(t))\frac{dQ(t)}{Q(t)}+D(t)dt, (9)

X(0) = x

Substituting (1), (2) and (3) into (9), we obtain

$$dX(t) = (X(t)(r + \lambda \cdot \Delta(t)) + D(t))dt$$
$$+ X(t)(\Sigma\Delta(t))' \cdot dW(t), \qquad (10)$$
$$X(0) = x.$$

5. The Discounted Value of SCI

In this section, we determine the value of expected discounted SCI.

Definition 1:

The discounted value of the expected future SCI is defined as

$$\Psi(t) = E_t\left(\int_t^T \frac{\Lambda(u)}{\Lambda(t)} D(t) du\right)$$
(11)

where, $E_t = E(\cdot | F_t)$ is the conditional expectation with respect to the Brownian filtration $\{F_t\}_{t>0}$ and

 $\Lambda(t) = Z(t) \exp(-rt)$ is the stochastic discount factor which adjusts for nominal interest rate and market price of risks for stock and inflation-linked bond.

Proposition 1:

Suppose $\Psi(t)$ is the discounted value of the expected future SCI, then

$$\Psi(t) = \frac{D(t) \left(\exp\left(\left(k - r - \sigma_D \cdot \theta \right) (T - t) \right) - 1 \right)}{k - r - \sigma_D \cdot \theta}.$$
 (12)

Proof. By definition 1, we have that

$$\Psi(t) = D(t)E_t \left(\int_t^T \frac{\Lambda(u)D(u)}{\Lambda(t)D(t)} du \right)$$
(13)

Applying change of variable on (13), we have

$$\Psi(t) = D(t) E\left(\int_0^{T-t} \frac{\Lambda(\tau)D(\tau)}{\Lambda(0)D(0)} \mathrm{d}\tau\right)$$
(14)

Applying parallelogram law and martingale principles on (14), we have

$$E\left(\frac{\Lambda(\tau)D(\tau)}{\Lambda(0)D(0)}\right) = \exp\left(\left(k - r - \sigma_D \cdot \theta\right)\tau\right).$$

Therefore,

$$\Psi(t) = D(t) \Big(\int_0^{T-t} \exp((k - r - \sigma_D \cdot \theta)\tau) d\tau \Big).$$
(15)

Integrating, we have

$$\Psi(t) = \frac{D(t) \left(\exp\left(\left(k - r - \sigma_D \cdot \theta \right) (T - t) \right) - 1 \right)}{k - r - \sigma_D \cdot \theta}.$$
 (16)

Figure 2 was obtained by setting $D_0 = 100$, k = 0.099, r = 0.04, $\sigma_1^D = 0.25$, $\sigma_2^D = 0.36$, $\mu = 0.09$, $\sigma_s = 0.4$, $\sigma_I = 0.3$, $\rho = 0.6$ and $\theta_I = 0.08$.

At t = 0, we obtain the present discounted value of future SCI to be

$$\Psi(0) = \Psi_0 = \frac{D_0 \left(\exp\left(\left(k - r - \sigma_D \cdot \theta \right) T \right) - 1 \right)}{k - r - \sigma_D \cdot \theta}.$$
 (17)

If
$$r + \sigma_D \cdot \theta > k$$
 and we allow $T \to \infty$ *i.e.*,

$$\lim_{T \to \infty} \Psi_0 = \lim_{T \to \infty} \left[\frac{D_0 \left(\exp\left(\left(k - r - \sigma_D \cdot \theta \right) T \right) - 1 \right)}{k - r - \sigma_D \cdot \theta} \right],$$
$$r + \sigma_D \cdot \theta > k$$
$$= \frac{D_0}{r + \sigma_D \cdot \theta - k}, r + \sigma_D \cdot \theta > k.$$

For a deterministic case, we have the present discounted value future deterministic cash inflows to be



Figure 2. The flow of the discounted value of SCI.

$$\Psi_0 = \frac{D_0\left(\exp\left((k-r)T\right) - 1\right)}{k-r}.$$
 (18)

Hence.

$$\lim_{T \to \infty} \Psi_0 = \lim_{T \to \infty} \left[\frac{D_0 \left(\exp\left((k-r)T\right) - 1 \right)}{k-r} \right], r > k$$
$$= \frac{D_0}{r-k}, r > k.$$

This shows that as $T \to \infty$, Ψ_0 converges to $\frac{D_0}{r + \sigma_D \cdot \theta - k}, \text{ provided } r + \sigma_D \cdot \theta > k.$

For the deterministic case, it shows that as $T \to \infty$, D

$$\Psi_0$$
 converges to $\frac{D_0}{r-k}$, provided $r > k$.

Figure 3 was obtained by setting $D_0 = 100$, k = 0.099, r = 0.04, $\sigma_1^D = 0.25$, $\sigma_2^D = 0.36$, $\mu = 0.09$, $\sigma_s = 0.4$, $\sigma_I = 0.3$, $\rho = 0.6$ and $\theta_I = 0.08$.

Figure 2 represents the flow of the discounted value of stochastic cash inflow in the investment at time t and Figure 3 represents the present value of discounted future SCI at time t.

We now consider the sensitivity analysis of Ψ_0 . Proposition 2 establishes this fact.

Proposition 2:

Let
$$\beta = k - r - \sigma_D \cdot \theta$$
, then
 $\frac{\partial \Psi_0}{\partial T} = D_0 \exp(\beta T).$

Proof. The results follow by taking the partial derivatives of Ψ_0 with respect to T, D_0 , r, σ_D and k, respectively.

$$\frac{\partial \Psi_0}{\partial D_0} = \frac{1}{\beta} \left(\frac{1}{D_0} \frac{\partial \Psi_0}{\partial T} - 1 \right);$$



Figure 3. The present discounted value of future SCI.

$$\frac{\partial \Psi_0}{\partial r} = \frac{1}{\beta^2} \left((1 - \beta T) \frac{\partial \Psi_0}{\partial T} - D_0 \right);$$

$$\frac{\partial \Psi_0}{\partial \sigma_D} = \theta \frac{\partial \Psi_0}{\partial r};$$

$$\frac{\partial \Psi_0}{\partial k} = \frac{1}{\beta^2} \left((\beta T - 1) \frac{\partial \Psi_0}{\partial T} + D_0 \right) = -\frac{\partial \Psi_0}{\partial r}.$$

Table 1 shows the sensitivity analysis of the discounted value of the SCI.

Proposition 3:

Suppose that Proposition 1 holds, then

$$d\Psi(t) = \Psi(t) ((r + \sigma_D \cdot \theta) dt + \sigma'_D \cdot dW(t)) -D(t) dt.$$
(19)

Proof. Finding the differential of both sides of (16) and then substitute (7), we have

$$d\Psi(t) = \frac{D(t)(\exp(\beta(T-t))-1)}{\beta} \times (kdt + \sigma'_D \cdot dW(t))$$

+ $\frac{D(t)}{\beta}(-\beta \exp(\beta(T-t)))dt$
= $\frac{D(t)}{\beta}((\exp(\beta(T-t))-1)(kdt + \sigma'_D \cdot dW(t)))$
- $\beta(\exp(\beta(T-t))-1)dt) - D(t)dt$
= $\frac{D(t)(\exp(\beta(T-t))-1)}{\beta}((r + \sigma_D \cdot \theta)dt + \sigma'_D \cdot dW(t))$
- $D(t)dt$
= $\Psi(t)((r + \sigma_D \cdot \theta)dt + \sigma'_D \cdot dW(t)) - D(t)dt.$

Let V(t) be a value process at time t. We defined

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Table 1. Simulation of the sensitivity analysis.

Т	$rac{\partial \Psi_{_0}}{\partial D_{_0}}$	$rac{\partial \Psi_{_0}}{\partial \sigma_{_1}^{^D}}$	$rac{\partial \Psi_{_0}}{\partial \sigma_{_2}^{^D}}$	$rac{\partial \Psi_{_0}}{\partial T}$	$\frac{\partial \Psi_{_0}}{\partial r}$	$rac{\partial \Psi_{_0}}{\partial heta_{_I}}$	$rac{\partial \Psi_{_0}}{\partial heta_{_S}}$	$rac{\partial \Psi_{_0}}{\partial k}$
1	1.0022	-16624	-20005	0.4359	-207850	-51962	-74826	207850
2	2.0087	-14857	-17875	0.8738	-185720	-46429	-66858	185720
3	3.0197	-13079	-15736	1.3136	-163490	-40872	-58855	163490
4	4.0350	-11293	-13587	1.7552	-141160	-35290	-50818	141160
5	5.0548	-9499	-11429	2.1988	-118740	-29685	-42746	118740
6	6.0790	-7697	-9261	2.6444	-96220	-24054	-34638	96220
7	7.1077	-5888	-7084	3.0918	-73600	-18400	-26495	73600
8	8.1408	-4070	-4897	3.5413	-50880	-12720	-18317	50880
9	9.1785	-2245	-2701	3.9926	-28060	-7016	-10103	28060
10	10.2207	-412	-495	4.4460	-5150	-1287	-1853	5150

V(t) as

$$V(t) \coloneqq X(t) + \Psi(t), \tag{20}$$

where, X(t) satisfy (10) and $\Psi(t)$ satisfy (19). **Proposition 4:**

Let V(t) satisfy (20), X(t) satisfy (10) and $\Psi(t)$ satisfy (19), then

$$dV(t) = (r(X(t) + \Psi) + \Psi(t)\sigma_D \cdot \theta + X(t)\lambda \cdot \Delta(t))dt \quad (21) + (X(t)\Sigma\Delta(t) + \Psi(t)\sigma_D)' \cdot dW(t).$$

Proof. Finding the differential of both sides of (20) and then substitute in (10) and (19), the result follows.

6. Optimization of the Value of Wealth Process

We define the general value function

$$J(t,v) = E\left[u\left(V\left(t, X\left(t\right), \Psi\left(t\right)\right)\right) \middle| X(t) = x, \Psi(t) = \Psi\right]$$

where u(V(t)) is the path of V(t). Define Π to be the set of all admissible portfolio strategy that are *F*-progressively measurable, that satisfy the integrability conditions $E\left[\int_{t}^{T} \Delta'(u)\Delta(u)du\right] < \infty$ and let U(V(t)) be a concave function in V(t) such that U(V(t)) satisfies the HJB equation

$$U(t,v) = \sup_{\Delta \in \Pi} E\left[u(V(T)) | X(t) = x, \Psi(t) = \Psi\right]$$
$$= \sup_{\Delta \in \Pi} J(t,v)$$

Therefore, by applying Itô lemma on (21), we obtain the following HJB equation:

$$U_t(t,v) + \max_{\Delta \in \Pi} HV = 0, \qquad (22)$$

subject to:

$$U(T,v)=\frac{v^{1-\gamma}}{1-\gamma},$$

where,

$$HV = r\Psi U_{\Psi} + rxU_{x} + \Psi \sigma_{D} \cdot \theta U_{\Psi} + x\lambda \cdot \Delta(t)U_{x}$$
$$+ x\Psi \sigma_{D} (\Sigma\Delta(t))U_{x\Psi} + \frac{1}{2}x^{2} (\Sigma\Delta(t))' \Sigma\Delta(t)U_{xx}$$
$$+ \frac{1}{2}\Psi^{2}\sigma_{D}'\sigma_{D}U_{\Psi\Psi}$$
(23)

Let U(t,v) be the solution of the HJB equation (22). Since the utility function is concave and the value function is smooth, *i.e.*, $U(t,V) \in C^{1,2}(\mathbb{R} \times [0,T])$, then (22) is well-defined. Hence, we have the following:

$$\frac{\partial HV}{\partial \Delta(t)} = \lambda U_x + x \left(\Sigma \Delta^*(t) \right)' \Sigma U_{xx} + \Psi \sigma_D \Sigma U_{X\Psi}$$
(24)
= 0

from (24), we have

$$\left(\Delta^{*}(t)\right)' = \left(\Sigma'\Sigma\right)^{-1} \left(\frac{-\lambda U_{x}}{xU_{xx}}\right) - \frac{\left(\Sigma'\right)^{-1} \sigma_{D} \Psi U_{x\Psi}}{xU_{xx}}.$$
 (25)

Substituting (25) into (22), we obtain the following HJB equation:

$$U_{t} + rxU_{x} + (r + \sigma_{D} \cdot \theta)\Psi U_{\Psi} + \frac{1}{2}\sigma_{D}^{\prime}\sigma_{D}\Psi^{2}U_{\Psi\Psi}$$

$$-\frac{\lambda^{\prime}(\Sigma^{2})^{-1}\lambda U_{x}^{2}}{2U_{xx}} - \frac{\sigma_{D}^{\prime}\sigma_{D}\Psi^{2}U_{x\Psi}^{2}}{2U_{xx}}$$

$$-\frac{\sigma_{D}^{\prime}(\Sigma^{-1})\lambda\Psi U_{x}U_{x\Psi}}{U_{xx}} = 0.$$
(26)

This is the resulting HJB equation of our problem.

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Proposition 5:

The solution of the HJB equation (26) is of the form

$$U(t,v) = \frac{\left(vA(t)\right)^{1-\gamma}}{1-\gamma}, \gamma \neq 0, \gamma > 0,$$

with

$$A(t) = \exp\left(\left(r + \frac{\theta'\theta}{2\gamma}\right)(T-t)\right),$$

$$A(T) = 1.$$
(27)

Proof. Finding the partial derivatives of

 $U(t,v) = \frac{(vA(t))^{1-\gamma}}{1-\gamma} \text{ with respect to } t, x, xx, \Psi,$

 $x\Psi$ and $\Psi\Psi$ and then substitute into (26), we obtain the following:

$$v^{1-\gamma}A(t)^{-\gamma}A'(t) + rv^{1-\gamma}A(t)^{1-\gamma} + \frac{\lambda'(\Sigma^{2})^{-1}\lambda v^{1-\gamma}A(t)^{1-\gamma}}{2\gamma} = 0.$$
(28)

From (28), we obtain

$$A(t) = \exp\left(\left(r + \frac{\theta'\theta}{2\gamma}\right)(T-t)\right),$$

$$A(0) = 1$$
(29)

Using (29), we obtain

$$U(t,v) = \frac{v^{1-\gamma}}{1-\gamma} \exp\left(\left(r + \frac{\theta'\theta}{2\gamma}\right)(1-\gamma)(T-t)\right),$$

$$U(T,v) = \frac{v^{1-\gamma}}{1-\gamma}.$$
(30)

Figure 4 was obtained by setting $D_0 = 100$, r = 0.04, k = 0.099, T = 10, $\sigma_1^D = 0.25$, $\sigma_2^D = 0.36$, $\rho = 0.6$, $\mu = 0.09$, $\sigma_s = 0.4$, $\sigma_I = 0.3$ and $\theta_I = 0.08$.

Figure 4 shows the expected value of utility of wealth at time $t \in [0,10]$ for different values of γ . Here, γ measures the level of risk the investor is willing to take. Observe that the smaller the value of γ , the higher the expected value of wealth, and vice versa. Therefore, when $\gamma = 0.07$, the expected value of utility of wealth, U(t,v) = U(10,v) = 2780.9, when $\gamma = 0.08$, U(10,v) = 2271.5, when $\gamma = 0.09$, U(10,v) = 1914.1and when $\gamma = 0.10$, U(10,v) = 1648.7. Hence, the lower the rate at which investors are risk averse, the more the wealth that will accrue to them, and vice versa, which is an intuitive result.

Proposition 6:

Suppose that U(t,V) is the solution of the HJB Equation (26), then the optimal portfolio in inflation-linked bond, stock and cash account are given by

$$\Delta_I^*(t) = \frac{\theta_I V^*(t)}{\gamma X^*(t)} - \frac{\sigma_I^D \Psi(t)}{\sigma_I X^*(t)}.$$
(31)

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Figure 4. The expected value of utility of wealth for different values of γ .

$$\Delta_{s}^{*}(t) = \frac{\left(\sigma_{I}(\mu - r) + \theta_{I}\sigma_{s}\rho\left(\sigma_{s}\sqrt{1 - \rho^{2}} - \sigma_{I}\right)\right)V^{*}(t)}{\gamma\left(1 - \rho^{2}\right)\sigma_{I}\sigma_{s}^{2}X^{*}(t)} \quad (32)$$
$$-\frac{\left(\sigma_{I}\sigma_{2}^{D} - \sigma_{s}\sigma_{I}^{D}\rho\right)\Psi(t)}{\sigma_{I}\sigma_{s}\sqrt{1 - \rho^{2}}X^{*}(t)}.$$

$$\begin{split} &\Delta_0^*(t) = 1 \\ &- \frac{V^*(t)}{X^*(t)} \Biggl(\frac{\theta_I}{\gamma} + \frac{\left(\sigma_I(\mu - r) + \theta_I \sigma_S \rho \left(\sigma_S \sqrt{1 - \rho^2} - \sigma_I\right)\right)}{\gamma \left(1 - \rho^2\right) \sigma_I \sigma_S^2} \Biggr) \\ &+ \frac{\Psi(t)}{X^*(t)} \Biggl(\frac{\sigma_I^D}{\sigma_I} + \frac{\left(\sigma_I \sigma_2^D - \sigma_S \sigma_I^D \rho\right)}{\sigma_I \sigma_S \sqrt{1 - \rho^2}} \Biggr). \end{split}$$

Proof. From (25), we have that

$$\Delta^{*}(t) = \frac{\left(\Sigma^{2}\right)^{-1} \lambda V^{*}(t)}{\gamma X^{*}(t)} - \frac{\Sigma^{-1} \sigma_{D} \Psi(t)}{X^{*}(t)}.$$

But,

$$\Sigma^{-1}\sigma_{D} = \begin{pmatrix} \frac{\sigma_{1}^{D}}{\sigma_{I}} \\ \frac{\sigma_{I}\sigma_{2}^{D} - \sigma_{S}\sigma_{1}^{D}\rho}{\sigma_{S}\sigma_{I}\sqrt{1 - \rho^{2}}} \end{pmatrix}$$
$$\Sigma^{-2}\lambda = \begin{pmatrix} \frac{\theta_{I}}{\sigma_{I}} \\ \frac{\sigma_{I}(\mu - r) + \sigma_{S}\theta_{I}\rho(\sigma_{S}\sqrt{1 - \rho^{2}} - \sigma_{I})}{\sigma_{S}^{2}\sigma_{I}(1 - \rho^{2})} \end{pmatrix}$$

Therefore,

$$\begin{split} \Delta^{*}(t) &= \begin{pmatrix} \Delta_{I}^{*}(t) \\ \Delta_{S}^{*}(t) \end{pmatrix} = \frac{V^{*}(t)}{\gamma X^{*}(t)} \begin{pmatrix} \frac{\theta_{I}}{\sigma_{I}} \\ \frac{\sigma_{I}(\mu-r) + \sigma_{S}\theta_{I}\rho\left(\sigma_{S}\sqrt{1-\rho^{2}} - \sigma_{I}\right)}{\sigma_{S}^{2}\sigma_{I}\left(1-\rho^{2}\right)} \end{pmatrix} - \frac{\Psi(t)}{X^{*}(t)} \begin{pmatrix} \frac{\sigma_{I}^{D}}{\sigma_{I}} \\ \frac{\sigma_{I}\sigma_{2}^{D} - \sigma_{S}\sigma_{I}^{D}\rho}{\sigma_{S}\sigma_{I}\sqrt{1-\rho^{2}}} \end{pmatrix} \\ &= \begin{pmatrix} \frac{\theta_{I}V^{*}(t)}{\sigma_{I}\gamma X^{*}(t)} \\ \frac{\left(\sigma_{I}(\mu-r) + \sigma_{S}\theta_{I}\rho\left(\sigma_{S}\sqrt{1-\rho^{2}} - \sigma_{I}\right)\right)V^{*}(t)}{\sigma_{S}^{2}\sigma_{I}\left(1-\rho^{2}\right)\gamma X^{*}(t)} \end{pmatrix} - \begin{pmatrix} \frac{\sigma_{I}^{D}\Psi(t)}{\sigma_{I}X^{*}(t)} \\ \frac{\left(\sigma_{I}\sigma_{2}^{D} - \sigma_{S}\sigma_{I}^{D}\rho\right)\Psi(t)}{\sigma_{S}\sigma_{I}\sqrt{1-\rho^{2}}X^{*}(t)} \end{pmatrix} \\ &= \begin{pmatrix} \frac{\theta_{I}V^{*}(t)}{\sigma_{I}\gamma X^{*}(t)} - \frac{\sigma_{I}^{D}\Psi(t)}{\sigma_{I}\gamma X^{*}(t)} \\ \frac{\left(\sigma_{I}(\mu-r) + \sigma_{S}\theta_{I}\rho\left(\sigma_{S}\sqrt{1-\rho^{2}} - \sigma_{I}\right)\right)V^{*}(t)}{\sigma_{S}^{2}\sigma_{I}\left(1-\rho^{2}\right)\gamma X^{*}(t)} - \frac{\left(\sigma_{I}\sigma_{2}^{D} - \sigma_{S}\sigma_{I}^{D}\rho\right)\Psi(t)}{\sigma_{S}\sigma_{I}\sqrt{1-\rho^{2}}X^{*}(t)} \end{pmatrix} \end{split}$$

It implies that

$$\Delta_I^*(t) = \frac{\theta_I V^*(t)}{\gamma X^*(t)} - \frac{\sigma_I^D \Psi(t)}{\sigma_I X^*(t)}.$$
(33)

$$\Delta_{s}^{*}(t) = \frac{\left(\sigma_{I}\left(\mu-r\right)+\theta_{I}\sigma_{s}\rho\left(\sigma_{s}\sqrt{1-\rho^{2}}-\sigma_{I}\right)\right)V^{*}(t)}{\gamma\left(1-\rho^{2}\right)\sigma_{I}\sigma_{s}^{2}X^{*}(t)} \quad (34)$$

$$-\frac{\left(\sigma_{I}\sigma_{2}^{D}-\sigma_{s}\sigma_{1}^{D}\rho\right)\Psi(t)}{\sigma_{I}\sigma_{s}\sqrt{1-\rho^{2}}X^{*}(t)} \cdot$$

$$\Delta_{0}^{*}(t) = 1 - \left(\frac{\theta_{I}V^{*}(t)}{\gamma X^{*}(t)}\right)$$

$$+\frac{\left(\sigma_{I}\left(\mu-r\right)+\theta_{I}\sigma_{s}\rho\left(\sigma_{s}\sqrt{1-\rho^{2}}-\sigma_{I}\right)\right)V^{*}(t)}{\gamma\left(1-\rho^{2}\right)\sigma_{I}\sigma_{s}^{2}X^{*}(t)}\right)$$

$$+\frac{\sigma_{1}^{D}\Psi(t)}{\sigma_{I}X^{*}(t)} + \frac{\left(\sigma_{I}\sigma_{2}^{D}-\sigma_{s}\sigma_{1}^{D}\rho\right)\Psi(t)}{\sigma_{I}\sigma_{s}\sqrt{1-\rho^{2}}X^{*}(t)} \cdot$$

$$(35)$$

From (33), the first term represents the classical portfolio strategy while the second term represents the intertemporal hedging strategy that offset any shock to the SCI at time t. From (34),

$$\frac{\left(\sigma_{I}\left(\mu-r\right)+\theta_{I}\sigma_{S}\rho\left(\sigma_{S}\sqrt{1-\rho^{2}}-\sigma_{I}\right)\right)V^{*}\left(t\right)}{\gamma\left(1-\rho^{2}\right)\sigma_{I}\sigma_{S}^{2}X^{*}\left(t\right)} \quad \text{represents}$$

the classical portfolio strategy at time t and

 $\frac{\left(\sigma_{I}\sigma_{2}^{D}-\sigma_{S}\sigma_{I}^{D}\rho\right)\Psi(t)}{\sigma_{I}\sigma_{S}\sqrt{1-\rho^{2}}X^{*}(t)}$ represents the intertemporal

hedging term that offset shock resulting from the SCI at time t. From (35), observe that these hedging terms can be transfer to cash account at time t or it can be reinvest in stock and in inflation-linked bond at time t.

Figure 5 was obtained by setting $D_0 = 100$, r = 0.04, k = 0.099, T = 10, $\sigma_1^D = 0.25$, $\sigma_2^D = 0.36$, $\rho = 0.6$, $\mu = 0.09$, $\sigma_s = 0.4$, $\sigma_I = 0.3$, $\theta_I = 0.08$ and $\gamma = 0.5$. The optimal portfolio value in inflation-linked bond at time t = 10 is obtained as 0.16 (or 16%).

Figure 6 was obtained by setting $D_0 = 100$, r = 0.04, k = 0.099, T = 10, $\sigma_1^D = 0.25$, $\sigma_2^D = 0.36$, $\rho = 0.6$, $\mu = 0.09$, $\sigma_s = 0.4$, $\sigma_t = 0.3$, $\theta_t = 0.08$ and $\gamma = 0.5$. The optimal portfolio value in stock at time t = 10 is obtained as 1.001562 (or 100.1562%).

Figure 7 was obtained by setting $D_0 = 100$, r = 0.04, k = 0.099, T = 10, $\sigma_1^D = 0.25$, $\sigma_2^D = 0.36$, $\rho = 0.6$, $\mu = 0.09$, $\sigma_s = 0.4$, $\sigma_I = 0.3$, $\theta_I = 0.08$ and $\gamma = 0.5$. The optimal portfolio value in cash account at time t = 10 is obtained as -0.161562 (or -16.1562%).

In **Figures 5-7**, we set $D_0 = 100$, r = 0.04, k = 0.099, T = 10, $\sigma_1^D = 0.25$, $\sigma_2^D = 0.36$, $\rho = 0.6$, $\mu = 0.09$, $\sigma_s = 0.4$, $\sigma_I = 0.3$, $\theta_I = 0.08$ and $\gamma = 0.10$. **Figure 5** shows the portfolio value in inflation-linked bond. We found that the optimal portfolio value in inflation-linked bond at time t = 10 is 0.16 (or 16%). **Figure 6** shows the portfolio value in stock. We found that the optimal portfolio value in t = 10 is 1.001562 (or 100.1562%). **Figure 7** shows the portfolio value in cash



Figure 5. Portfolio value in inflation-linked bond.



Figure 6. Portfolio value in stock.



Figure 7. Portfolio value in cash account.

account. We found that the optimal portfolio value in cash account at time t = 10 is -0.161562 (or -16.1562%).

7. Conclusion

The optimal portfolio strategy X with discounted stochastic cash inflows was considered. It was assumed that the cash

inflow, stock and inflation-linked bond are stochastic and follow a standard geometric Brownian motion. The sensitivity analysis of the present value of the discounted cash inflows was carried out in this paper and the results are presented in **Table 1**. Analytical solution to the resulting HJB equation was obtained. It was found that the smaller the value of γ (which measure the level of risk the investor is willing to take), the higher the expected value of wealth, and vice versa. The optimal portfolio values in stock, inflation-linked bond and cash account were obtained. The resulting optimal portfolio values in stock and inflation-linked bond were found to involve intertemporal hedging terms that offset any shock to the SCI.

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