

## Assignment 7 CME 241

### Problem 1

The only difference compared to the modeling done in class is that the utility function is:  $U(x) = \log(x)$  for the consumption

- Riskless Asset:  $dR_t = rR_t dt$
- Risky Asset:  $dS_t = \mu S_t dt + \sigma S_t dz_t$ ,  $\mu > r > 0$ ,  $\sigma > 0$
- Wealth at time  $t$ :  $W_t \geq 0$
- Fraction of  $W_t$  allocated to the risky asset:  $\pi(t, W_t) = \pi_t$
- " " " " the riskless asset:  $1 - \pi(t, W_t) = 1 - \pi_t$
- Wealth consumption per unit of time:  $C(t, W_t) = C_t \geq 0$
- Process for wealth  $W_t$ :

$$dW_t = [(\pi_t(\mu - r) + r)W_t - C_t]dt + \pi_t \sigma W_t dz_t$$

The goal is to determine  $(\pi_t, C_t) \forall t \in [0, T]$  to maximize:

$$E \left[ \int_t^T e^{-\rho(s-t)} \log(C_s) ds + e^{-\rho(T-t)} \epsilon^\delta \log(W_T) | W_t \right], \text{ where } \epsilon^\delta \text{ is the bequest}$$

with  $0 < \epsilon < 1$ ,  $\delta = 1$

$\rho \geq 0$  is the utility discount rate.

### MDP Formulation:

- > State at time  $t$ :  $(t, W_t)$
- > Action at time  $t$ :  $(\pi_t, C_t)$
- > Reward per unit of time at time  $t$ :  $U(C_t) = \log(C_t)$
- > Return at time  $t$ :  $\int_t^T e^{-\rho(T-t)} \log(C_s) ds$
- > Policy  $(t, W_t) \rightarrow (\pi_t, C_t)$  that maximizes the Expected Return

So, the optimal Value Function  $V^*(t, W_t)$  is:

$$V^*(t, W_t) = \max_{\pi, C} E_t \left[ \int_t^T e^{-\rho(T-t)} \log(C_s) ds + e^{-\rho(T-t)} \epsilon \log(W_T) \right]$$

which satisfies the following recursive formulation for  $0 \leq t < T$ :

$$V^*(t, W_t) = \max_{\pi, C} E_t \left[ \int_t^{t_1} e^{-\rho(t-t_1)} \log(C_s) ds + e^{-\rho(t-t_1)} V^*(t_1, W_{t_1}) \right]$$

$$\Rightarrow e^{-\rho t} V^*(t, W_t) = \max_{\pi, C} E_t \left[ \int_t^{t_1} e^{-\rho s} \log(C_s) ds + e^{-\rho t_1} V^*(t_1, W_{t_1}) \right]$$

In stochastic differential form:

$$\max_{\pi_t, c_t} E_t \left[ d(e^{pt} V^*(t, W_t)) + e^{pt} \log(c_t) dt \right] = 0$$

$$\Rightarrow \max_{\pi_t, c_t} E_t \left[ dV^*(t, W_t) + \log(c_t) dt \right] = \rho V^*(t, W_t) dt$$

Applying Itô's Lemma and using the facts  $E[dz] = 0$  and  $E[dz^2] = dt$  we obtain the HJB equation in PDE form:

$$\max_{\pi_t, c_t} \left[ \frac{\partial V^*}{\partial t} + \frac{\partial V^*}{\partial W_t} ((\pi_t(\mu - r) + r)W_t - c_t) + \frac{\partial^2 V^*}{\partial W_t^2} \frac{\pi_t^2 \sigma^2 W_t^2}{2} + \log(c_t) \right] = \rho V^*(t, W_t)$$

Let us write the above equation more succinctly as:

$$\max_{\pi_t, c_t} \Phi(t, W_t; \pi_t, c_t) = \rho V^*(t, W_t) \quad (1)$$

To find optimal  $\pi_t^*, c_t^*$  by taking partial derivatives of  $\Phi(t, W_t; \pi_t, c_t)$  wrt  $\pi_t$  and  $c_t$  and equate them to 0 (1st order conditions for  $\Phi$ )

$$\bullet \frac{\partial \Phi}{\partial \pi_t} = 0 \Rightarrow (\mu - r) \frac{\partial V^*}{\partial W_t} + \frac{\partial^2 V^*}{\partial W_t^2} \pi_t^* \sigma^2 W_t = 0$$

$$\Rightarrow \pi_t^* = \frac{-\frac{\partial V^*}{\partial W_t} (\mu - r)}{\frac{\partial^2 V^*}{\partial W_t^2} \sigma^2 W_t} \quad (2)$$

$$\bullet \frac{\partial \Phi}{\partial c_t} = 0 \Rightarrow -\frac{\partial V^*}{\partial W_t} + \frac{1}{c_t^*} = 0 \Rightarrow c_t^* = \frac{1}{\frac{\partial V^*}{\partial W_t}} \quad (3)$$

$$\text{So (1)} \xrightarrow{(2)} \frac{\partial V^*}{\partial t} + \frac{\partial V^*}{\partial W_t} (\mu - r) W_t \pi_t^* + \frac{\partial V^*}{\partial W_t} r W_t - \frac{\partial V^*}{\partial W_t} c_t^* + \frac{\partial^2 V^*}{\partial W_t^2} \frac{\sigma^2 W_t^2}{2} \pi_t^{*2} + \log(c_t^*) = \rho V^*(t, W_t)$$

$$\Rightarrow \frac{\partial V^*}{\partial t} - \frac{(\mu - r)^2}{2\sigma^2} \frac{\left(\frac{\partial V^*}{\partial W_t}\right)^2}{\frac{\partial^2 V^*}{\partial W_t^2}} + \frac{\partial V^*}{\partial W_t} r W_t - 1 - \log\left(\frac{\partial V^*}{\partial W_t}\right) = \rho V^*(t, W_t) \quad (4)$$

The boundary condition is  $V^*(T, W_T) = E \log(W_T)$



We surmise with a guess solution  $V^*(t, W_t) = f(t) + \frac{1}{p} \log(W_t)$

Then  $\bullet \frac{\partial V^*}{\partial t} = f'(t)$

$\bullet \frac{\partial V^*}{\partial W_t} = \frac{1}{p W_t}$

$\bullet \frac{\partial^2 V^*}{\partial W_t^2} = -\frac{1}{p W_t^2}$

Hence by substituting the guess solution in the PDE we get.

$$f'(t) - \frac{(\mu-r)^2}{2\sigma^2} \frac{1}{p^2 W_t^2} + \frac{1}{p} r W_t - 1 - \log\left(\frac{1}{p W_t}\right) = p f(t) + \log(W_t)$$

$$\Rightarrow f'(t) + \frac{(\mu-r)^2}{2p\sigma^2} + \frac{r}{p} - 1 + \log(p) + \log(W_t) = p f(t) + \log(W_t)$$

$$\Rightarrow f'(t) = p f(t) - \left[ \frac{(\mu-r)^2}{2p\sigma^2} + \frac{r}{p} + \log(p) - 1 \right]$$

Let  $\boxed{v = \frac{(\mu-r)^2}{2p\sigma^2} + \frac{r}{p} + \log(p) - 1}$

Then we get the following PDE.

$$f'(t) = p f(t) - v \Rightarrow \frac{df}{dt} = p f - v \Rightarrow \frac{df}{p f - v} = dt \Rightarrow \int \frac{df}{p f - v} = \int dt + C$$

$$\stackrel{p \neq 0}{\Rightarrow} \frac{1}{p} \log(p f - v) = t + C \Rightarrow \log(p f - v) = p(t + C)$$

$$\Rightarrow p f - v = e^{p(t+C)} \Rightarrow f = \frac{v + e^{p(t+C)}}{p}$$

C is found from the boundary condition:

$$V^*(T, W_T) = \varepsilon \log(W_T)$$

$$\Rightarrow f(T) + \frac{1}{p} \log(W_T) = \varepsilon \log(W_T)$$

$$\Rightarrow f(T) = \log(W_T^\varepsilon) - \log(W_T^{1/p})$$

$$\Rightarrow f(T) = \log(W_T^{\varepsilon - 1/p})$$

$$\Rightarrow \frac{v + e^{p(T+C)}}{p} = \log(W_T^{\varepsilon - 1/p}) \Rightarrow e^{pC} = \frac{p \log(W_T^{\varepsilon - 1/p}) - v}{e^{pT}}$$

$$\Rightarrow \boxed{C = \frac{1}{p} \log \left[ \frac{p \log(W_T^{\varepsilon - 1/p}) - v}{e^{pT}} \right]}$$

Hence the control problem is solved:

$$\bullet \dot{V}^*(t, W_t) = \frac{V + e^{p(t+G)}}{p} + \frac{1}{p} \log(W_t)$$

$$\text{where } \bullet V = \frac{(\mu-r)^2}{2p\sigma^2} + \frac{r}{p} + \log(p) - 1$$

$$\bullet G = \frac{1}{p} \log \left[ \frac{p \log(W_t^{E^{-1/p}}) - V}{E p^T} \right]$$

$$\bullet \pi^*(t, W_t) = \frac{-\frac{1}{pW_t}(\mu-r)}{-\frac{1}{pW_t^2}\sigma^2 W_t} \Rightarrow \pi^*(t, W_t) = \frac{\mu-r}{\sigma^2}$$

$$\bullet \dot{C}^*(t, W_t) = \frac{1}{\frac{1}{pW_t}} \Rightarrow \dot{C}^*(t, W_t) = pW_t$$