

## 8 APPENDIX

The investor's prior distribution over the initial value of  $\beta$  is assumed to be Gaussian, and  $\beta$  is assumed to follow equation (8). Because  $de(t)/e(t)$  and  $dp(t)$  are following a joint Brownian motion; and all the parameters in equation (4), (8), and (9) are linear functions of the unobservable state variable  $\beta$ , the distribution function of  $\beta$  at time  $t$ ,  $F_t(x) = P(\beta \leq x | F_t^I)$ , is Gaussian given the investor's information structure at time  $t$ ,  $F_t^I$ , which is generated by the joint processes  $I(t) = (e(t), P(t))$ . Let  $b_t \equiv E(\beta | F_t^I)$ ; then  $b$  will be the optimal estimate of  $\beta$  from  $I(t)$ . The knowledge of the variance  $v_t \equiv E((\beta - b_t)(\beta - b_t)^\top | F_t^I)$ , which is an  $n \times n$  matrix, gives us a measure of filtering error and the investor's level of uncertainty. We call the process for  $I(t) = (e(t), P(t))$  the investor's signal used to form expectations of  $\beta$ . Let  $Z_1 = (Z_1(t), F_t)$  and  $Z_2 = (Z_2(t), F_t)$  be two mutually independent Wiener processes, where  $Z_1(t) = [Z_{\beta_1}(t), Z_{\beta_2}(t)]$  and  $Z_2(t) = [Z_{r_d}(t), Z_{r_f}(t), Z_e(t)]$ . We write the processes for signals as

$$dI = (I_0(I, t) + I_1(I, t)\beta)dt + \sum_{i=1}^2 C_i(I, t)dZ_i,$$

where  $I_0(I, t)$  is an  $(2 + 1) \times 1$  vector with  $\bar{\mu}$  in the first row and  $A_0$  in the remaining rows, and  $I_1(I, t)$  and  $C_1(I, t)$  and  $C_2(I, t)$  and matrices of dimension  $((2 + 1) \times 2)$  and  $((2 + 1) \times 2)$  and  $(3 \times 3)$ , with  $L - \bar{L}$  and  $A_1$  forming the rows of the first and zero matrix of the second

and  $\sigma_e$  and  $\sigma_p$  forming the rows of the third. Rewrite the process for  $\beta$  as

$$d\beta = (B_0(I, t) + B_1(I, t)\beta)dt + \sum_{i=1}^2 c_i(I, t)dZ_i.$$

Define  $\Gamma(I, t) = (c \circ C)(I, t)$  as the covariance between the signals and the state variables,  $\Sigma(I, t) = (c \circ c)(I, t)$  as the variance-covariance matrix of the state variables, and  $\Delta(I, t) = (C \circ C)(I, t)$  as the variance-covariance matrix of the signal processes where  $c \circ c = c_1 c_1^\top + c_2 c_2^\top$ ,  $c \circ C = c_1 C_1^\top + c_2 C_2^\top$ ,  $C \circ C = C_1 C_1^\top + C_2 C_2^\top$ . A direct extension of Theorem 12.3 in Liptser and Shiriyayev (2001) to vectors of measurement and transition equations yields

$$\begin{aligned} db(t) &= [B_0(I, t) + B_1(I, t)b] dt + [v(t)I_1(I, t)^\top + \Gamma(I, t)] \Delta(I, t)^{-1} \\ &\quad \times [dI - (I_0(I, t) + I_1(I, t)b)dt], \end{aligned}$$

$$\begin{aligned} dv(t) &= B_1(I, t)v(t) + v(t)B_1(I, t)^\top + \Sigma(I, t) - \\ &\quad [v(t)I_1(I, t)^\top + \Gamma(I, t)] \Delta(I, t)^{-1} [v(t)I_1(I, t)^\top + \Gamma(I, t)]^\top, \end{aligned}$$

where

$$\begin{aligned} d\widehat{W}(t) &= C_2(I, t)^{-1} \{dI - E(dI|F_t^I)\} \\ &= C_2(I, t)^{-1} \{dI - [(I_0(I, t) + I_1(I, t)b)dt]\}. \end{aligned}$$

Then we can write the process  $I(t)$  as

$$\begin{pmatrix} dr_d(t) \\ dr_f(t) \\ \frac{de(t)}{e(t)} \end{pmatrix} = \left( \begin{pmatrix} a_d(b_d - r_d(t)) \\ a_f(b_f - r_f(t)) \\ \bar{\mu}_e \end{pmatrix} + \begin{pmatrix} 0 & 0 \\ 0 & 0 \\ r_d(t) - b_d & r_f(t) - b_f \end{pmatrix} \begin{pmatrix} \beta_1(t) \\ \beta_2(t) \end{pmatrix} \right) dt \\ + \begin{pmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} dZ_{\beta_1}(t) \\ dZ_{\beta_2}(t) \end{pmatrix} + \begin{pmatrix} \sigma_{r_d} & 0 & 0 \\ 0 & \sigma_{r_f} & 0 \\ 0 & 0 & \sigma_e(t) \end{pmatrix} \begin{pmatrix} dZ_d(t) \\ dZ_f(t) \\ dZ_e(t) \end{pmatrix},$$

and  $\beta$  as

$$\begin{pmatrix} d\beta_1(t) \\ d\beta_2(t) \end{pmatrix} = \begin{pmatrix} \vartheta_1(\bar{\beta}_1 - \beta_1(t)) \\ \vartheta_2(\bar{\beta}_2 - \beta_2(t)) \end{pmatrix} dt + \begin{pmatrix} \sigma_{\beta_1} & 0 \\ 0 & \sigma_{\beta_2} \end{pmatrix} \begin{pmatrix} dZ_{\beta_1}(t) \\ dZ_{\beta_2}(t) \end{pmatrix} \\ + \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} dZ_d(t) \\ dZ_f(t) \\ dZ_e(t) \end{pmatrix}$$

The process  $b$  and  $v$  will be the following equation

$$\begin{pmatrix} db_1(t) \\ db_2(t) \end{pmatrix} = \begin{pmatrix} \vartheta_1(\bar{b}_1 - b_1(t)) \\ \vartheta_2(\bar{b}_2 - b_2(t)) \end{pmatrix} dt + \left[ \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} + \begin{pmatrix} v_{11}(t) & v_{21}(t) \\ v_{12}(t) & v_{22}(t) \end{pmatrix} \begin{pmatrix} 0 & 0 & r_d(t) - b_d \\ 0 & 0 & r_f(t) - b_f \end{pmatrix} \right] \begin{bmatrix} \sigma_{r_d}^2 & 0 & 0 \\ 0 & \sigma_{r_f}^2 & 0 \\ 0 & 0 & \sigma_e^2(t) \end{bmatrix}^{-1} \left[ \begin{pmatrix} dr_d(t) \\ dr_f(t) \\ \frac{de(t)}{e(t)} \end{pmatrix} - \begin{pmatrix} a_d(b_d - r_d(t)) \\ a_f(b_f - r_f(t)) \\ \bar{\mu}_e \end{pmatrix} + \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} b_1(t) \\ b_2(t) \end{pmatrix} \right] dt$$

$$\begin{pmatrix} dv_{11}(t) & dv_{21}(t) \\ dv_{12}(t) & dv_{22}(t) \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} + \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} + \begin{pmatrix} \sigma_{\beta_1}^2 & 0 \\ 0 & \sigma_{\beta_2}^2 \end{pmatrix} - \left[ \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} + \begin{pmatrix} v_{11}(t) & v_{21}(t) \\ v_{12}(t) & v_{22}(t) \end{pmatrix} \begin{pmatrix} 0 & 0 & r_d(t) - b_d \\ 0 & 0 & r_f(t) - b_f \end{pmatrix} \right] \times \begin{bmatrix} \sigma_{r_d}^2 & 0 & 0 \\ 0 & \sigma_{r_f}^2 & 0 \\ 0 & 0 & \sigma_e^2(t) \end{bmatrix}^{-1} \times \left[ \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} + \begin{pmatrix} v_{11}(t) & v_{21}(t) \\ v_{12}(t) & v_{22}(t) \end{pmatrix} \begin{pmatrix} 0 & 0 & r_d(t) - b_d \\ 0 & 0 & r_f(t) - b_f \end{pmatrix} \right]^\top$$

where

$$\begin{pmatrix} d\widehat{Z}_d(t) \\ d\widehat{Z}_f(t) \\ d\widehat{Z}_e(t) \end{pmatrix} = \begin{pmatrix} \sigma_{r_d} & 0 & 0 \\ 0 & \sigma_{r_f} & 0 \\ 0 & 0 & \sigma_e(t) \end{pmatrix}^{-1} \left[ \begin{pmatrix} dr_d(t) \\ dr_f(t) \\ \frac{de(t)}{e(t)} \end{pmatrix} - \begin{pmatrix} a_d(b_d - r_d(t)) \\ a_f(b_f - r_f(t)) \\ \bar{\mu}_e \end{pmatrix} + \begin{pmatrix} 0 & 0 \\ 0 & 0 \\ r_d(t) - b_d & r_f(t) - b_f \end{pmatrix} \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} \right] dt .$$