

# Some Continuous-time Limits

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# Continuous time

Level evolution:

$$dV_t = V_t \mu_t^V dt + V_t \sigma_t^V \cdot dW_t$$

Logarithm evolution:

$$d\widehat{V}_t = \hat{\mu}_t^V dt + \sigma_t^V \cdot dW_t$$

where  $\hat{\mu}_t^V = \mu_t^V - \frac{1}{2}|\sigma_t^V|^2$ .

# Discrete-time approximation

$$\begin{aligned}\widehat{V}_t &= \frac{1}{1-\rho} \log \left[ (1-\beta_\epsilon) \exp \left[ (1-\rho) \widehat{C}_t \right] + \beta_\epsilon \exp \left[ (1-\rho) \widehat{R}_t \right] \right] \\ \widehat{R}_t &= \frac{1}{1-\gamma} \log \mathbb{E} \left[ \exp \left( (1-\gamma) \widehat{V}_{t+\epsilon} \right) \mid \mathfrak{A}_t \right]\end{aligned}$$

where  $\beta_\epsilon = \exp(-\delta\epsilon)$ . Rewrite this as:

$$\begin{aligned}\frac{1}{1-\rho} \log \left[ (1-\beta_\epsilon) \exp \left[ (1-\rho) \left( \widehat{C}_t - \widehat{V}_t \right) \right] + \beta_\epsilon \exp \left[ (1-\rho) \left( \widehat{R}_t - \widehat{V}_t \right) \right] \right] \\ = 0 \\ \widehat{R}_t - \widehat{V}_t = \frac{1}{1-\gamma} \log \mathbb{E} \left( \exp \left[ (1-\gamma) \left( \widehat{V}_{t+\epsilon} - \widehat{V}_t \right) \right] \mid \mathfrak{A}_t \right)\end{aligned}$$

## Discrete-time approximation cont.

$$\begin{aligned}\frac{d}{d\epsilon} \left( \widehat{R}_t - \widehat{V}_t \right) \Big|_{\epsilon=0} &= \frac{d}{d\epsilon} \frac{1}{1-\gamma} \log \mathbb{E} \left( \exp \left[ (1-\gamma) \left( \widehat{V}_{t+\epsilon} - \widehat{V}_t \right) \right] \mid \mathfrak{A}_t \right) \Big|_{\epsilon=0} \\ &= \hat{\mu}_t^V + \frac{(1-\gamma)}{2} |\sigma_t^V|^2 \\ &= \mu_t^V - \frac{\gamma}{2} |\sigma_t^V|^2\end{aligned}$$

by local log normality.

# Discrete-time approximation cont.

Differentiate with respect to  $\epsilon$ :

$$\begin{aligned} & \frac{1}{1-\rho} \log \left[ (1-\beta_\epsilon) \exp \left[ (1-\rho) \left( \widehat{C}_t - \widehat{V}_t \right) \right] + \beta_\epsilon \exp \left[ (1-\rho) \left( \widehat{R}_t - \widehat{V}_t \right) \right] \right] \\ &= 0 \end{aligned}$$

$$\begin{aligned} & \frac{\delta \left[ \left( \frac{C_t}{V_t} \right)^{1-\rho} - 1 \right]}{1-\rho} + \frac{d}{d\epsilon} \left( \widehat{R}_t - \widehat{V}_t \right) \Big|_{\epsilon=0} \\ &= \frac{\delta \left[ \left( \frac{C_t}{V_t} \right)^{1-\rho} - 1 \right]}{1-\rho} + \mu_t^V - \frac{\gamma}{2} |\sigma_t^V|^2 \\ &= 0. \end{aligned}$$

Duffie-Epstein refer to  $\gamma$  as a “**variance multiplier**.”

# Robustness and subjective beliefs

Girsanov theory:

- ▷ Consider  $H$  processes with the same dimension as the underlying Brownian motion.
- ▷ Form a **positive martingales** to induce changes in the probability measure:

$$dM_t = M_t H_t \cdot dW_t$$

- ▷ Observe that

$$d \log M_t = -\frac{1}{2} |H_t|^2 dt + H_t \cdot dW_t$$

- ▷ Under the  $H$  change of measure:

$$dW_t = H_t dt + d\tilde{W}_t^H$$

where  $\tilde{W}^H$  is a Brownian motion.

# Robustness and subjective beliefs

Recall two relations:

$$\begin{aligned}d\log M_t &= -\frac{1}{2} |H_t|^2 dt + H_t \cdot dW_t \\dW_t &= H_t dt + d\tilde{W}_t^H\end{aligned}$$

Observations:

- ▷ Change of measure induces a **local mean** or drift  $H$  process in the Brownian motion
- ▷ Under the change of measure, the drift in  $\log M$  is  $\frac{1}{2} |H_t|^2$  - **local relative entropy**

# A heuristic digression

Discrete approximation:

$$\log M_{t+\epsilon} - \log M_t = -\frac{\epsilon}{2} |H_t|^2 + H_t \cdot (W_{t+\epsilon} - W_t)$$

Let  $w$  be a realized  $W_{t+\epsilon} - W_t$  and  $h$  be a realization  $H_t$ . Then  $\log M_{t+\epsilon} - \log M_t$  contributes  $-\frac{\epsilon}{2} h' h + h \cdot w$  to the log-likelihood. The standard normal density for  $(W_{t+\epsilon} - W_t)$  contributes  $-\frac{1}{\epsilon} w' w$ . Put together, we have a log-likelihood:

$$-\frac{\epsilon}{2} h' h + h \cdot w - \frac{1}{2\epsilon} w' w = -\frac{1}{2\epsilon} (w - \epsilon h)' (w - \epsilon h)$$

The altered density has mean  $\epsilon h$ , which is the realized value of  $H_t$ .



# Robust adjustment in cont. time

- ▷ **Replace** the value function drift  $\hat{\mu}_t^V$  with

$$\hat{\mu}_t^V + \sigma_t^V \cdot H_t$$

- ▷ **Solve:**

$$\min_{H_t} \hat{\mu}_t^V + \sigma_t^V \cdot H_t + \frac{\xi_m}{2} H_t \cdot H_t$$

- ▷ **Minimizing  $H_t$ :**

$$H_t^* = -\frac{1}{\xi_m} \sigma_t^V$$

- ▷ **Minimized objective**

$$\hat{\mu}_t^V - \frac{1}{\xi_m} |\sigma_t^V|^2$$

Set  $\gamma - 1 = \frac{1}{2\xi_m}$  to connect to our previous analysis.

# Smooth ambiguity/prior robustness 1

- ▷ Suppose that

$$d\widehat{V}_t = \hat{\mu}_t^V(\theta)dt + \sigma_t^V dW_t$$

where  $\theta$  is not observed.

- ▷ **Replace:**

$$\frac{\delta \left[ \left( \frac{C_t}{\widehat{V}_t} \right)^{1-\rho} - 1 \right]}{1-\rho} + \hat{\mu}_t^V - \frac{1}{2\xi_m} |\sigma_t^V|^2 = 0$$

with

$$\frac{\delta \left[ \left( \frac{C_t}{\widehat{V}_t} \right)^{1-\rho} - 1 \right]}{1-\rho} - \frac{1}{\xi_p} \log \mathbb{E} \left( \exp \left[ -\xi_p \hat{\mu}_t^V(\theta) \right] \mid \mathfrak{A}_t \right) - \frac{1}{2\xi_m} |\sigma_t^V|^2 = 0$$

Hansen-Maio (PNAS)

## Smooth ambiguity/prior robustness 2

$$\begin{aligned} & -\frac{1}{\xi_p} \log \mathbb{E} \left( \exp \left[ -\xi_p \hat{\mu}_t^V(\theta) \right] \mid \mathfrak{A}_t \right) \\ & = \min_{N_t(\theta)} \mathbb{E} \left[ N_t(\theta) \hat{\mu}_t^V(\theta) \mid \mathfrak{A}_t \right] + \xi_p \mathbb{E} \left[ N_t(\theta) \log N_t(\theta) \mid \mathfrak{A}_t \right] \end{aligned}$$

where

$$N_t^*(\theta) = \frac{\exp \left[ -\xi_p \hat{\mu}_t^V(\theta) \right]}{\mathbb{E} \left( \exp \left[ -\xi_p \hat{\mu}_t^V(\theta) \right] \mid \mathfrak{A}_t \right)}$$

Exponential tilting

# Value functions in a Markov world

- ▷ Markov state dynamics

$$dX_t = \mu_x(X_t) + \sigma_x(X_t)dW_t$$

- ▷ Suppose that

$$\widehat{V}_t = f(X_t)$$

- ▷ Then

$$\hat{\mu}_t^V = \frac{\partial f}{\partial x}(X_t)\mu_x(X_t) + \frac{1}{2}\text{trace}\left[\sigma_x(X_t)'\frac{\partial^2 f}{\partial x\partial x'}(X_t)\sigma_x(X_t)\right]$$

$$\sigma_t^V = \sigma_x(X_t)'\left[\frac{\partial f}{\partial x}(X_t)\right]'$$

$$|\sigma_t^V|^2 = \left[\frac{\partial f}{\partial x}(X_t)\right]\sigma_x(X_t)\sigma_x(X_t)'\left[\frac{\partial f}{\partial x}(X_t)\right]'$$

Note that  $\frac{\partial f}{\partial x}$  is  $n$  (# of state variables) by 1 and  $\frac{\partial^2 f}{\partial x\partial x'}$  is  $n$  by  $n$ .

# Stochastic discount factor evolution

Abstract from ambiguity (see Hansen-Miao, forthcoming ET for the inclusion of ambiguity)

▷ **Local evolution:**

$$dS_t = S_t \mu_t^S dt + S_t \sigma_t^S \cdot dW_t$$

$$d \log S_t = \hat{\mu}_t^S dt - \frac{1}{2} |\sigma_t^S|^2 dt + \sigma_t^S \cdot dW_t$$

$$dC_t = C_t \mu_t^C dt + C_t \sigma_t^C \cdot dW_t$$

$$d \log C_t = \hat{\mu}_t^C dt + \sigma_t^C \cdot dW_t$$

where

$$\hat{\mu}_t^S = \mu_t^S - \frac{1}{2} |\sigma_t^S|^2$$

$$\hat{\mu}_t^C = \mu_t^C - \frac{1}{2} |\sigma_t^C|^2$$

▷  $-\mu_t^S$  **local risk-free rate**;  $-\sigma_t^S$  -vector of **local shock prices**

# Evolution of logarithms

- ▷ Evolution over a period of length  $\epsilon$  :

$$\begin{aligned}\log S_{t+\epsilon} - \log S_t &= -\epsilon\delta - \rho \left[ \widehat{C}_{t+\epsilon} - \widehat{C}_t \right] \\ &\quad + (1 - \gamma) \left[ \widehat{V}_{t+\epsilon} - \widehat{R}_t \right] + (\rho - 1) \left[ \widehat{V}_{t+\epsilon} - \widehat{R}_t \right]\end{aligned}$$

- ▷ Instantaneous counterparts

$$\begin{aligned}\hat{\mu}_t^S &= -\delta - \rho\hat{\mu}_t^C - \frac{(1 - \gamma)^2}{2} |\sigma_t^V|^2 - \frac{(1 - \gamma)(\rho - 1)}{2} |\sigma_t^V|^2 \\ &= -\delta - \rho\hat{\mu}_t^C + \frac{(\gamma - 1)(\rho - \gamma)}{2} |\sigma_t^V|^2 \\ \sigma_t^S &= -\rho\sigma_t^C + (1 - \gamma)\sigma_t^V + (\rho - 1)\sigma_t^V \\ &= -\rho\sigma_t^C + (\rho - \gamma)\sigma_t^V\end{aligned}$$

- ▷  $\mu_t^S = \hat{\mu}_t^S + \frac{1}{2} |\sigma_t^S|^2$

# Elaboration

Consider:

$$(1 - \gamma) \left[ \widehat{V}_{t+\epsilon} - \widehat{R}_t \right]$$

Recall that

$$\mathbb{E} \left( \exp \left[ (1 - \gamma) \left[ \widehat{V}_{t+\epsilon} - \widehat{R}_t \right] \right] \mid \mathfrak{A}_t \right) = 1.$$

Thus

$$\lim_{\epsilon \downarrow 0} \frac{(1 - \gamma)}{\epsilon} \mathbb{E} \left( \widehat{V}_{t+\epsilon} - \widehat{R}_t \mid \mathfrak{A}_t \right) + \frac{(1 - \gamma)^2}{2} |\sigma_t^V|^2 = 0$$

and

$$\lim_{\epsilon \downarrow 0} \frac{(\rho - 1)}{\epsilon} \mathbb{E} \left( \widehat{V}_{t+\epsilon} - \widehat{R}_t \mid \mathfrak{A}_t \right) + \frac{(1 - \gamma)(\rho - 1)}{2} (|\sigma_t^V|)^2 = 0.$$