

# Chapter 13: Continuous-Time Markets

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# Securities Market Model

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# Notation

- Money market account has price  $R$  with  $dR/R = r dt$ .
- $n$  locally risky assets with dividend-reinvested prices  $S_i$ .
- $\mu =$  vector of  $n$  stochastic processes  $\mu_i$
- $\sigma = n \times k$  matrix of stochastic processes
- $B =$  vector of  $k$  independent Brownian motions.  $k \geq n$ .
- Assume no redundant assets, meaning  $\sigma$  has rank  $n$ .

# Asset Price Dynamics

- Assume, for each risky asset  $i$ ,

$$\frac{dS_{it}}{S_{it}} = \mu_{it} dt + \sum_{j=1}^k \sigma_{ijt} dB_{jt}$$

- Stacking the asset returns,

$$dS/S \stackrel{\text{def}}{=} \begin{pmatrix} dS_{1t}/S_{1t} \\ \vdots \\ dS_{nt}/S_{nt} \end{pmatrix} = \mu_t dt + \sigma_t dB_t$$

# Covariance Matrix of Returns

- Drop the  $t$  subscript for simplicity. We have

$$\begin{aligned}\left(\frac{dS_i}{S_i}\right) \left(\frac{dS_\ell}{S_\ell}\right) &= \left(\sum_{j=1}^k \sigma_{ij} dB_j\right) \left(\sum_{j=1}^k \sigma_{\ell j} dB_j\right) \\ &= \sum_{j=1}^k \sigma_{ij} \sigma_{\ell j} dt\end{aligned}$$

- Stacking the returns:

$$\begin{aligned}(dS/S) \left(\frac{dS}{S}\right)' &= (\sigma dB)(\sigma dB)' \\ &= \sigma (dB)(dB)' \sigma' = \sigma \sigma' dt = \Sigma dt\end{aligned}$$

for  $\Sigma = \sigma \sigma'$ .

# Intertemporal Budget Constraint

- Let  $\phi_i$  denote the amount of the consumption good invested in risky asset  $i$ .
- Let  $W$  = wealth,  $C$  = consumption,  $Y$  = labor income.
- The intertemporal budget constraint is

$$dW = (Y - C) dt + \theta' dS + (W - \theta' S) r dt$$

where  $\theta = (\theta_1, \dots, \theta_n)'$  denotes share holdings.

- Setting  $\phi_i = \theta_i S_i \Rightarrow$

$$dW = (Y - C) dt + \phi' (dS/S) + (W - \phi' \iota) r dt$$

- Equivalently,

$$dW = (Y - C) dt + rW dt + \phi'(dS/S - r\iota) dt$$

- Equivalently,

$$dW = (Y - C) dt + rW dt + \phi'(\mu - r\iota) dt + \phi' \sigma dB$$

## In Terms of Fractions of Wealth Invested

- Assuming  $W > 0$ , we can define  $\pi = \phi/W$  and write the intertemporal budget constraint as

$$dW = (Y - C) dt + rW dt + W\pi'(\mu - r\iota) dt + W\pi'\sigma dB$$

- Equivalently,

$$\frac{dW}{W} = \frac{Y - C}{W} dt + r dt + \pi'(\mu - r\iota) dt + \pi'\sigma dB$$

- If  $Y = C$ , the wealth process is said to be self financing.

# First Optimization Problem

- Horizon  $T$ . No intermediate consumption ( $C = 0$ ). No labor income ( $Y = 0$ ). Log utility for terminal wealth.  $W_0$  given.
- $\max E[\log(W_T)]$  over portfolio processes  $\pi$  subject to

$$\frac{dW}{W} = r dt + \pi'(\mu - r\mathbf{1}) dt + \pi' \sigma dB$$

- Solve the wealth equation like we solved for GBM before (take logs, integrate, then exponentiate). We get

$$W_T = W_0 \exp \left( \int_0^T \left( r_t + \pi_t'(\mu_t - r_t) - \frac{1}{2} \pi_t' \Sigma_t \pi_t \right) dt + \int_0^T \pi_t' \sigma_t dB_t \right)$$

- So,  $E[\log W_T]$  is

$$\log W_0 + E \left[ \int_0^T \left( r_t + \pi_t'(\mu_t - r_t) - \frac{1}{2} \pi_t' \Sigma_t \pi_t \right) dt + \int_0^T \pi_t' \sigma_t dB_t \right]$$



- Use iterated expectations to get

$$\log W_0 + E_T \left[ \int_0^T E_t \left[ r_t + \pi_t'(\mu_t - r_t) - \frac{1}{2} \pi_t' \Sigma_t \pi_t \right] dt + \int_0^T E_t[\pi_t' \sigma_t dB_t] \right]$$

- Actually need a technical condition for this:

$$E \int_0^T \pi_t' \Sigma_t \pi_t dt < \infty$$

which implies a local martingale is a martingale.

- Conclusion is: choose  $\pi_t$  to maximize

$$\pi_t'(\mu_t - r_t) - \frac{1}{2} \pi_t' \Sigma_t \pi_t$$

- Implies

$$\pi_t^* = \Sigma_t^{-1}(\mu_t - r_t)$$

# SDF Processes

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# Definition of SDF Processes

- Define a stochastic process  $M$  to be an SDF process if
  - $M_0 = 1$
  - $M_t > 0$  for all  $t$  with probability 1
  - $MR$  is a local martingale, where  $R$  denotes the price of the money market account,
  - $MS_i$  is a local martingale, for  $i = 1, \dots, n$ , where the  $S_i$  are the dividend-reinvested asset prices.
- 'Local martingale' means zero drift (no  $dt$  part).

# Characterization of SDF Processes

- We can show: A stochastic process  $M > 0$  with  $M_0 = 1$  is an SDF process if and only if  $E[dM/M] = -r dt$  and

$$(\mu - r) dt = -(dS/S) \left( \frac{dM}{M} \right)$$

- Use  $MR = \text{local martingale}$  to get  $E[dM/M] = -r dt$ .
- Use  $MS_i = \text{local martingale}$  for each  $i$  to get displayed equation.

# No Uncertainty or Risk Neutrality

- SDF process is

$$M_t = e^{-rt}$$

if  $r$  is constant or

$$M_t = e^{-\int_0^t r_s ds}$$

if  $r$  varies over time.

- So,

$$\frac{dM}{M} = -r dt$$

- With risk aversion, it is only true that the drift of  $dM/M$  is  $-r$  which we express as  $E[dM/M] = -r dt$

# Single Period Model

- The condition  $E[dM/M] = -r dt$  parallels a single period model. Set  $M_0 = 1$  and  $M_1 = \tilde{m}$ . Then,
  - $\Delta M/M_0 = (\tilde{m} - 1)/1$
  - $E[\Delta M/M_0] = 1/R_f - 1 = (1 - R_f)/R_f = -r_f/R_f$
- The condition

$$(\mu - r_t) dt = -(dS/S) \left( \frac{dM}{M} \right)$$

parallels

$$(\forall i) \quad E[\tilde{R}_i] - R_f = -R_f \text{cov}(\tilde{R}_i, \tilde{m})$$

# Prices of Risk

- Start with  $M$  being an Itô process with drift of  $dM/M$  being  $-r$ .  
This means

$$\frac{dM_t}{M_t} = -r_t dt - \lambda'_t dB_t$$

for some  $\lambda$  process.

- The choice of  $-\lambda$  instead of  $+\lambda$  is arbitrary but convenient.
- Then,

$$(dS/S) \left( \frac{dM}{M} \right) = -\sigma(dB)(dB)'\lambda = \sigma\lambda dt$$

- So,

$$(\mu - r) dt = -(dS/S) \left( \frac{dM}{M} \right) \Rightarrow \mu - r = \sigma\lambda$$

- $\lambda$  called price of risk process.

# Projections of SDF Processes

- One solution  $\lambda$  of the equation  $\sigma\lambda = \mu - r\iota$  is

$$\lambda_p \stackrel{\text{def}}{=} \sigma'(\sigma\sigma')^{-1}(\mu - r\iota) = \sigma'\Sigma^{-1}(\mu - r\iota)$$

- For this solution,

$$\begin{aligned}\lambda_p' dB &= (\mu - r\iota)'\Sigma^{-1}\sigma dB \\ &= \pi'\sigma dB\end{aligned}$$

for  $\pi = \Sigma^{-1}(\mu - r\iota)$  (the log-optimal portfolio). Thus, it is spanned by the assets.

- Every solution  $\lambda$  of the equation  $\sigma\lambda = \mu - r\iota$  is of the form

$$\lambda = \lambda_p + \zeta$$

where  $\zeta$  is orthogonal to the assets in the sense that  $\sigma\zeta = 0$ .



# Valuation

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- For an asset with price process  $P$  and dividend process  $D$ ,

$$P_t = E_t \left[ \int_t^u \frac{M_\tau}{M_t} D_\tau d\tau + \frac{M_u}{M_t} P_u \right]$$

for any SDF process  $M$  (subject to a local martingale being a martingale).

- Ruling out bubbles, we can take  $u$  to infinity.
- Likewise, for any  $(W, C)$  satisfying the intertemporal budget constraint (assuming a local martingale is a martingale),

$$W_t = E_t \left[ \int_t^u \frac{M_\tau}{M_t} (C_\tau - Y_\tau) d\tau + \frac{M_u}{M_t} W_u \right]$$

- Ruling out Ponzi schemes, we can take  $u$  to infinity.

# Complete Markets

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# How Many Assets do we Need?

- Assume the Brownian motions are the only sources of uncertainty.
- Then the market is complete if the rank of  $\sigma$  is  $k$  (as many non-redundant assets as there are Brownian motions).
- We are assuming for simplicity that there are no redundant assets (rank  $\sigma$  is  $n$ ), so completeness is equivalent to  $\sigma$  being square and nonsingular.

# Why Completeness?

- Martingale representation theorem: with Brownian uncertainty, every martingale  $Y$  is spanned by the Brownian motions meaning  $dY = \gamma' dB$ .
- When  $\sigma$  is square and nonsingular, we can set  $\pi = \sigma^{-1}\gamma$  to get  $dY = \pi'\sigma dB$  w, which is the stochastic part of a portfolio return.

# Uniqueness of the SDF Process

- When markets are complete, there is a unique solution of  $\sigma\lambda = \mu - r\iota$  given by  $\lambda = \sigma^{-1}(\mu - r)$ .
- So, there is a unique SDF process

## Second Optimization Problem

- Complete markets, finite horizon, continuous consumption, no labor income. Consumption process must satisfy

$$W_0 = E \int_0^T M_t C_t dt$$

- max

$$E \int_0^T e^{-\delta t} u(C_t) dt$$

subject to the above constraint.

- Lagrangean:

$$E \int_0^T \{e^{-\delta t} u(C_t) - \gamma M_T C_t\} dt$$

- Maximize pointwise. FOC is

$$u'(C_t) = \gamma M_t$$