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More (alternative) theory

Continuous Time Finance

The Martingale Approach to Optimal Investment Theory

Ch 20

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essential ingredient is
completeness of the market

Contents

- • Decoupling the wealth profile from the portfolio choice.
- Lagrange relaxation. *(seen before, but will be explained again)*
- Solving the general wealth problem.
- Example: Log utility.
- Example: The numeraire portfolio.

Problem Formulation

Standard model with internal filtration (see pp. 360, 368)

$$dS_t = D(S_t)\alpha_t dt + D(S_t)\sigma_t dW_t,$$

$$dB_t = rB_t dt.$$

Assumptions:

- Drift and diffusion terms are allowed to be arbitrary adapted processes.
- The market is complete. ("N=E")
- We have a given initial wealth x_0

Problem:

$$\max_{h \in \mathcal{H}} E^P [\Phi(X_T)]$$

(only terminal wealth)

where

$$\mathcal{H} = \{\text{self financing portfolios}\}$$

given the initial wealth $X_0 = x_0$.

Some observations

- In a complete market, there is a unique martingale measure Q .
- Every claim Z satisfying the budget constraint

$$e^{-rT} E^Q [Z] = x_0,$$

is attainable by an $h \in \mathcal{H}$ and vice versa

$\uparrow$ self-financing

- We can thus write our problem as

Note: E^P

$$\max_Z E^P [\Phi(Z)]$$

subject to the constraint

Note: E^Q

$$e^{-rT} E^Q [Z] = x_0.$$

- We can forget the wealth dynamics! *for the*

time being, see step 2 below)

$h_T' S_T = Z$ and
 $x_0 = h_0' S_0 = e^{-rT} E^Q [h_T' S_T] = e^{-rT} E^Q [Z]$
 and the value process V^h of a self-financing portfolio process is a Q -martingale.

Basic Ideas

Our problem was

$$\max_Z E^P [\Phi(Z)]$$

subject to

$$e^{-rT} E^Q [Z] = x_0.$$

Idea I:

We can **decouple** the optimal portfolio problem into:

1. Finding the optimal wealth profile $\hat{Z}$.
2. Given $\hat{Z}$, find the replicating portfolio. *(here the dynamics come in!)*

Idea II:

- Rewrite the constraint under the measure $\underline{P}$ *(instead of Q).*
- Use Lagrangian techniques to relax the constraint.

Lagrange formulation

Recall

Problem:

$$\max_Z E^P [\Phi(Z)]$$

$z \in \mathcal{F}_T$

subject to

$$e^{-rT} E^P [L_T Z] = x_0.$$

Now: constraint in terms of measure P !

Here L is the likelihood process, i.e.

$$L_t = \frac{dQ}{dP}, \quad \text{on } \mathcal{F}_t, \quad 0 \leq t \leq T, \quad \text{and}$$

recall $E^Q z = E^P [L_T z]$

The Lagrangian of the problem is

$$\mathcal{L} = E^P [\Phi(Z)] + \lambda \{x_0 - e^{-rT} E^P [L_T Z]\}$$

i.e.

$$\mathcal{L} = E^P [\Phi(Z) - \lambda e^{-rT} L_T Z] + \lambda x_0$$

note: both expectations under P !

The optimal wealth profile

Given enough convexity and regularity we now expect, given the dual variable λ , to find the optimal Z by maximizing

$$\mathcal{L} = E^P [\Phi(Z) - \lambda e^{-rT} L_T Z] + \lambda x_0$$

over unconstrained Z , i.e. to maximize *the Lebesgue integral*

$$\int_{\Omega} \{ \Phi(Z(\omega)) - \lambda e^{-rT} L_T(\omega) Z(\omega) \} dP(\omega)$$

This is a trivial problem! *(if you look at it the right way)*
We can simply maximize $Z(\omega)$ for each ω separately.

$$\max_z \{ \Phi(z) - \lambda e^{-rT} L_T z \}, \text{ with } L_T = L_T(\omega), \text{ } z = Z(\omega)$$

under the integral

The optimal wealth profile

Our problem: (repeated from previous slide)

$$\max_z \{ \Phi(z) - \lambda e^{-rT} L_T z \}$$

First order condition

$$\Phi'(z) = \lambda e^{-rT} L_T$$

The optimal Z is thus given by

$$\hat{Z} = G(\lambda e^{-rT} L_T)$$

$\hat{Z}$ depends on λ !

where

$$G(y) = [\Phi']^{-1}(y).$$

*(if Φ' has
inverse with
 y in its domain)*

The dual variable λ is determined by the constraint

$$e^{-rT} E^P [L_T \hat{Z}] = x_0.$$

implicitly
↓

*you have to solve λ
from this equation, $\hat{Z}$ depends on λ ,
and hope/prove that a unique
solution exists*

start of lecture 13b

Example – log utility

Assume that

Then *inverse of Φ' is* $\Phi(x) = \ln(x)$, $\Phi'(x) = \frac{1}{x} \Rightarrow$
 $g(y) = \frac{1}{y}$, *for all $y > 0$*

Thus

$$\hat{Z} = G(\lambda e^{-rT} L_T) = \frac{1}{\lambda} e^{rT} L_T^{-1}$$

Finally λ is determined by

$$e^{-rT} E^P [L_T \hat{Z}] = x_0.$$

i.e.

$$e^{-rT} E^P \left[L_T \frac{1}{\lambda} e^{rT} L_T^{-1} \right] = x_0.$$

so $\lambda = x_0^{-1}$ and

$$\hat{Z} = x_0 e^{rT} L_T^{-1},$$

*to be interpreted as optimal
wealth at time T , given
the budget constraint.*

The optimal wealth process

- We have computed the optimal **terminal** wealth profile

$$\hat{Z} = \hat{X}_T = x_0 e^{rT} L_T^{-1} \quad (1)$$

- What does the optimal wealth **process** $\hat{X}_t$ look like?

We have (why?) *(discounted traded assets are Q-martingales)*

$$\hat{X}_t = e^{-r(T-t)} E^Q \left[\hat{X}_T \mid \mathcal{F}_t \right] \quad (2)$$

so we obtain *from (1) and (2):*

$$\hat{X}_t = x_0 e^{rt} E^Q \left[L_T^{-1} \mid \mathcal{F}_t \right]$$

But L^{-1} is a Q-martingale (why?) so we obtain

abstract theory, $L_T^{-1} = dP/dQ$ on $\mathcal{F}_T$

Q is now the base measure

$$\hat{X}_t = x_0 e^{rt} L_t^{-1}.$$

The Optimal Portfolio

- We have computed the optimal wealth process: $\hat{X}_t$
once we know L_t .
- How do we compute the optimal portfolio?

Assume for simplicity that we have a standard Black-Scholes model *(complete model!)*

$$dS_t = \mu S_t dt + \sigma S_t dW_t,$$

$$dB_t = r B_t dt$$

Recall that

$$\hat{X}_t = x_0 e^{rt} L_t^{-1}.$$

L_t^{-1} satisfies (Girsanov theory) an equation like $dL_t^{-1} = L_t^{-1} \psi_t dW_t^Q$ for some ψ_t *Q-Wiener*

W is P -Wiener

$= L_t^{-1} (\dots) dt + L_t^{-1} \psi_t dW_t^Q$, as we know $dW^Q = (\dots) dt + dW_t$.

Basic Program

$\hat{X}_t = X_0 e^{rt} L_t^{-1}$,
you "know" L_t

1. Use Ito and the formula for $\hat{X}_t$ to compute $d\hat{X}_t$ like

$$d\hat{X}_t = \hat{X}_t(\quad)dt + \hat{X}_t\beta_t dW_t \quad (\text{find } \beta_t \text{ later})$$

where we do not care about $(\quad)$.

2. Recall that (for some $\hat{u}_t$, portfolio weight) involves dW_t
see p.382

$$d\hat{X}_t = \hat{X}_t \left\{ (1 - \hat{u}_t) \frac{dB_t}{B_t} + \hat{u}_t \frac{dS_t}{S_t} \right\}$$

which we write as

$$d\hat{X}_t = \hat{X}_t \{ \quad \} dt + \hat{X}_t \hat{u}_t \sigma dW_t$$

3. We can identify $\hat{u}$ as

$$\hat{u}_t = \frac{\beta_t}{\sigma}$$

(if we know β_t ,
see further down)

We recall

$$(1) \quad \hat{X}_t = x_0 e^{rt} L_t^{-1}.$$

We also recall that

$(L_t \text{ is } \frac{dQ}{dP} \text{ on } \mathcal{F}_t)$

$$dL_t = L_t \varphi dW_t,$$

where

$$\varphi = \frac{r - \mu}{\sigma}$$

From this we have $(\text{Itô for } L_t^{-1})$

$$(2) \quad dL_t^{-1} = \varphi^2 L_t^{-1} dt - L_t^{-1} \varphi dW_t = -\varphi L_t^{-1} dW_t$$

and we obtain from (1) and (2), do the calculus!,

$$d\hat{X}_t = \hat{X}_t \{ \quad \} dt - \hat{X}_t \varphi dW_t \rightarrow \beta_t = -\varphi$$

Result: The optimal portfolio is given by

$$\hat{u}_t = \frac{\beta_t}{\sigma}$$

$$\Rightarrow \hat{u}_t = \frac{\mu - r}{\sigma^2}$$

(which we have seen as Market price of risk)

Note that $\hat{u}$ is a “myopic” portfolio in the sense that it does not depend on the time horizon T .

ψ_t on p. 382 is then $\psi_t = -\varphi_t$

(many lectures ago)

(Another example)

A Digression: The Numeraire Portfolio

Standard approach:

- Choose a fixed numeraire (portfolio) N .
- Find the corresponding martingale measure, i.e. find Q^N s.t.

$$\frac{B}{N}, \quad \text{and} \quad \frac{S}{N}$$

are Q^N -martingales.

Alternative approach: (swap the two steps above)

- Choose a fixed measure $Q \sim P$.
- Find numeraire N such that $Q = Q^N$:

$\frac{X_t}{N_t}$ is Q^N -martingale
if X_t is value
of traded asset

Special case:

- Set $Q = P$, our choice
- Find numeraire N such that $Q^N = P$ i.e. such that

$$\frac{B}{N}, \quad \text{and} \quad \frac{S}{N}$$

$\frac{P}{N}$
are Q^N -martingales under the **objective** measure P .

- This N is called the **numeraire portfolio**.

Specialize further:

Log utility and the numeraire portfolio

Definition:

The **growth optimal portfolio** (GOP) is the portfolio which is optimal for log utility (for arbitrary terminal date T).

Theorem:

Assume that X is GOP. Then X is the numeraire portfolio.

Proof:

We have to show that the process

$$Y_t = \frac{S_t}{X_t}$$

is a P martingale.

We have (see p. 381)

$$\frac{S_t}{X_t} = x_0^{-1} e^{-rt} S_t L_t$$

which is a P martingale,

since $x_0^{-1} e^{-rt} S_t$ is a Q martingale. Use "Bayes" (Additional exercise 3 = exercise C.9 in the book)

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