## Stochastic Processes and Stochastic Calculus - 9 Complete and Incomplete Market Models

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## Overview

1 Self-financing portfolio

2 Complete markets

3 Extensions of Black-Scholes

4 Incomplete market models

5 Back in discrete times

## Self-financing portfolio

Consider a market consisting in $d+1$ assets with prices

$$
\left(S_{t}^{0}, S_{t}^{1}, \ldots S_{t}^{d}\right)_{t \geq 0}
$$

A portfolio is self-financing if its value changes only because the asset prices change.
No money is withdrawn or inserted after the initial forming of the portfolio.

- A portfolio strategy $\left(H_{t}^{0}, H_{t}\right)_{t \geq 0}$ is an $(d+1)$-dim adapted process
- The corresponding value process is

$$
V_{t}=\sum_{i=0}^{d} H_{t}^{i} S_{t}^{i}=H_{t}^{0} S_{t}^{0}+H_{t} \cdot S_{t}
$$

- A portfolio is self-financing if

$$
\begin{aligned}
& \Delta V_{n}=H_{n}^{0} \Delta S_{n}^{0}+H_{n} \cdot \Delta S_{n} \quad(\text { discrete time }) \\
& \mathrm{d} V_{t}=H_{t}^{0} \mathrm{~d} S_{t}^{0}+H_{t} \cdot \mathrm{~d} S_{t} \quad(\text { continuous time })
\end{aligned}
$$

## Self-financing portfolio

In terms of discounted prices:

$$
\mathrm{d} \tilde{V}_{t}=H_{t} \cdot \mathrm{~d} \tilde{S}_{t}
$$

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## Proposition

For any adapted process $H_{t}=\left(H_{t}^{1}, \ldots, H_{t}^{d}\right)_{t \geq 0}$ and any initial value $V_{0}=x$, there exists a unique adapted process $\left(H_{t}^{0}\right)_{t \geq 0}$ such that the strategy $\left(H_{t}^{0}, H_{t}\right)_{t \geq 0}$ is self-financing.

## Proof.

$$
\begin{gathered}
\tilde{V}_{t}=x+\int_{0}^{t} H_{s} \cdot \mathrm{~d} \tilde{S}_{s}=H_{t} \cdot \tilde{S}_{t}+H_{t}^{0} \\
H_{t}^{0}=x+\int_{0}^{t} H_{s} \cdot \mathrm{~d} \tilde{S}_{s}-H_{t} \cdot \tilde{S}_{t}
\end{gathered}
$$

## Complete markets

## Definition

An $\mathcal{F}_{T}$-measurable random variable $X$ is an attainable claim if there exists a self-financing portfolio worth $X$ at time $T$.

## Definition

A market is complete if every contingent claim is attainable.

## Theorem

Assume that the market is arbitrage-free. Then, the following two statements are equivalent:

- the market is complete
- the martingale probability is unique.


## Complete market models

- It is theoretically possible to perfectly hedge contingent claims.
- Gives a unique no-arbitrage price.
- Allows us to derive a simple theory of pricing and hedging.
- Is a rather restrictive assumption.


## Black-Scholes model

Assuming a constant volatility B-S model gives a unique no-arbitrage price of an option

$$
\mathrm{d} S_{t}=S_{t}\left(\mu \mathrm{~d} t+\sigma \mathrm{d} B_{t}\right)
$$

The pricing formula depends only on one non-observable parameter: $\sigma$

$$
C\left(t, S_{t}\right)=x N\left(d_{1}\right)-K e^{-r(T-t)} N\left(d_{2}\right)
$$

where

$$
d_{1,2}=\frac{\log \left(S_{t} / K\right)+\left(r \pm \sigma^{2} / 2\right)(T-t)}{\sigma \sqrt{T-t}}
$$

In practice two methods are used to evaluate $\sigma$.

## Black-Scholes model

1 The historical method: Since

$$
S_{T}=S_{0} \exp \left[\sigma B_{T}-\left(\mu-\frac{\sigma^{2}}{2}\right) T\right]
$$

the random variables

$$
\log \left(\frac{S_{T}}{S_{0}}\right), \log \left(\frac{S_{2 T}}{S_{T}}\right), \ldots, \log \left(\frac{S_{N T}}{S_{(N-1) T}}\right)
$$

are independent Gaussian distributed with variance $\sigma^{2} T$.
Estimate $\sigma$ using asset prices observed in the past.
2 The implied method: we recover $\sigma$ by inversion of the Black-Scholes formula using quoted options.
No explicit formulas! Numerical methods need to be used.

## Volatility smile

Options based on the same underlying but with different strike and expiration time yield different implied volatilities.


Strike price

## Time-dependent volatility models

$$
\mathrm{d} S_{t}=S_{t}\left(\mu(t) \mathrm{d} t+\sigma(t) \mathrm{d} B_{t}\right)
$$

Similar formulas as in the Black-Scholes model replacing

$$
\begin{gathered}
\sigma^{2}(T-t) \rightsquigarrow \int_{t}^{T} \sigma^{2}(s) \mathrm{d} s \\
S_{t}=S_{0} \exp \left(\int_{0}^{t}\left(\mu(s)-\frac{\sigma^{2}(s)}{2}\right) \mathrm{d} s+\int_{0}^{t} \sigma(s) \mathrm{d} B_{s}\right)
\end{gathered}
$$

Does not avoid the volatility smile!

## Local volatility models

The volatility depends on the time and the stock price:

$$
\mathrm{d} S_{t}=S_{t}\left(\mu\left(t, S_{t}\right) \mathrm{d} t+\sigma\left(t, S_{t}\right) \mathrm{d} B_{t}\right)
$$

Note that $\mathcal{F}_{t}^{S}=\mathcal{F}_{t}^{B^{*}}$. The market is still complete.
For each $X \in L^{2}\left(\mathcal{F} T B, \mathbb{P}^{*}\right)$, there exists a replicating portfolio

$$
V_{t}=e^{-r(T-t)} \mathbb{E}^{*}\left[X \mid \mathcal{F}_{t}\right]=F\left(t, S_{t}\right), \quad H_{t}=\frac{\partial F}{\partial x}\left(t, S_{t}\right)
$$

## Need for more realistic models...

- In local volatility models, $\sigma$ is perfectly correlated with the stock price.
- Empirical studies reveal that the previous models can not capture heavy tails and asymmetries present in log-returns in practice.
- The real market is incomplete.


## Stochastic volatility models

Model volatility as a random process driven by its own source of randomness.

It is consistent with the highly variable and unpredictable nature of volatility.

Let $B_{t}^{1}, B_{t}^{2}$ be two independent Brownian motions.

$$
\left\{\begin{array}{l}
\mathrm{d} S_{t}=S_{t}\left(\mu_{t} \mathrm{~d} t+\sigma_{t} \mathrm{~d} B_{t}^{1}\right) \\
\mathrm{d} \sigma_{t}=\alpha\left(t, \sigma_{t}\right) \mathrm{d} t+\beta\left(t, \sigma_{t}\right) \mathrm{d} B_{t}^{2}
\end{array}\right.
$$

## Stochastic volatility models

Let $B_{t}:=\left(B_{t}^{1}, B_{t}^{2}\right)$ and $\mathcal{F}_{t}=\mathcal{F}_{t}^{B}$.
Girsanov theorem: $\left(B_{t}-\int_{0}^{t} H_{s} \mathrm{~d} s\right)_{t}$ is a 2-dim $\mathbb{P}^{*}$-Brownian motion

$$
\frac{d \mathbb{P}^{*}}{d \mathbb{P}^{2}}=\exp \left(\int_{0}^{T} H_{s} \mathrm{~d} B_{s}-\frac{1}{2} \int_{0}^{T}\left\|H_{s}\right\|_{2}^{2} \mathrm{~d} s\right)
$$

i.e.

$$
\widehat{B}_{t}^{1}:=B_{t}^{1}-\int_{0}^{t} H_{s}^{1} \mathrm{~d} s \quad \text { and } \quad \widehat{B}_{t}^{2}:=B_{t}^{2}-\int_{0}^{t} H_{s}^{2} \mathrm{~d} s
$$

are two independent Brownian motions w.r.t. $\mathbb{P}^{*}$.

## Stochastic volatility models

If

$$
H_{t}^{1}=-\frac{\mu_{t}-r}{\sigma_{t}}
$$

then

$$
\mathrm{d} S_{t}=S_{t}\left(r \mathrm{~d} t+\sigma_{t} \mathrm{~d} \widehat{B}_{t}^{1}\right)
$$

which means that the discounted price is a $\mathbb{P}^{*}$-martingale

$$
\mathrm{d} \tilde{S}_{t}=\tilde{S}_{t} \sigma_{t} \mathrm{~d} \widehat{B}_{t}^{1}
$$

There is no restriction on the process $H_{t}^{2}$.
Consequently, there are many probability measures under which the traded asset is a martingale.

## Stochastic volatility models

Note that $\mathcal{F}_{t}^{S} \supsetneq \mathcal{F}_{t}^{B^{1}}$. The model is not complete!
Let $X \in L^{2}\left(\Omega, \mathcal{F}_{T}, \mathbb{P}^{*}\right)$. By martingale representation theorem:

$$
\tilde{X}=X_{0}+\int_{0}^{T} K_{s}^{1} \mathrm{~d} \widehat{B}_{s}^{1}+\int_{0}^{T} K_{s}^{2} \mathrm{~d} \widehat{B}_{s}^{2}
$$

for some processes $K_{t}^{1}, K_{t}^{2}$. Hence

$$
\tilde{X}=X_{0}+\int_{0}^{T} \frac{K_{s}^{1}}{\sigma_{s} \tilde{S}_{s}} \mathrm{~d} \tilde{S}_{s}+\int_{0}^{T} K_{s}^{2} \mathrm{~d} \widehat{B}_{s}^{2}
$$

But the second integral can not be written as an integral w.r.t. $\mathrm{d} \tilde{S}_{s}$.

## Incomplete market models

Under a stochastic volatility model, the market is incomplete.

- No unique price.
- More random sources than traded assets.
- It is not always possible to hedge a generic contingent claim.
- Captures more empirical characteristics.


## Limitations

- Analytically less tractable.

■ No closed form solutions for option prices. Option prices can only be calculated by simulation

- The practical applications of stochastic volatility models are limited.


## Trinomial model

An attempt to improve the Binomial Model (CRR)...
We add a third possible state at which the stock price will not change.


## Trinomial Model

Absence of arbitrage $\Rightarrow \quad a<R<b$.
Indeed, absence of arbitrage implies the existence of a probability $\mathbb{P}^{*}$ such that discounted prices are $\mathbb{P}^{*}$-martingales.

Let $\mathbb{P}^{*}\left(S_{1}=1+a\right)=p_{1}$ and $\mathbb{P}^{*}\left(S_{1}=1+b\right)=p_{2}$.
Then,
$S_{0}=\mathbb{E}^{*}\left[\frac{S_{1}}{1+R}\right]=\frac{S_{0}(1+a) p_{1}+S_{0}(1+b) p_{2}+S_{0}\left(1-p_{1}-p_{2}\right)}{1+R}$,
or equivalently

$$
1+R=(1+a) p_{1}+(1+b) p_{2}+\left(1-p_{1}-p_{2}\right)
$$

Hence, necessarily $a<R<b$.

## Pricing in the Trinomial Model

For the one-step trinomial model, the discounted price is a $\mathbb{P}^{*}$-martingale if and only if

$$
1+R=(1+a) p_{1}+(1+b) p_{2}+\left(1-p_{1}-p_{2}\right)
$$

where

$$
\mathbb{P}^{*}\left(S_{1}=1+a\right)=p_{1} \quad \text { and } \quad \mathbb{P}^{*}\left(S_{1}=1+b\right)=p_{2}
$$

We have to solve one equation with two unknown quantities.
No unique risk-neutral price!

## Hedging in the Trinomial Model

Consider a financial derivative on the asset $S$ with value

$$
X_{t}=f\left(S_{t}\right)
$$

At time 0 , we want to construct a hedging strategy for $X_{1}$

$$
H_{1}^{0} S_{1}^{0}+H_{1} S_{1}=f\left(S_{1}\right)
$$

Hence, $\left(H_{1}^{0}, H_{1}\right)$ must satisfy

$$
\left\{\begin{align*}
H_{1}^{0} S_{1}^{0}+H_{1} S_{0}(1+a) & =f\left(S_{0}(1+a)\right)  \tag{1}\\
H_{1}^{0} S_{1}^{0}+H_{1} S_{0} & =f\left(S_{0}\right) \\
H_{1}^{0} S_{1}^{0}+H_{1} S_{0}(1+b) & =f\left(S_{0}(1+b)\right)
\end{align*}\right.
$$

We have to solve a system of three equations with two unknown quantities.

We are unable to replicate the portfolio!

