

Stochastic Calculus in Finance
MATH 6910 - Salisbury

Girsanov Transformations - A short introduction

- (1) Two probability measures P and \tilde{P} are equivalent on a sigma field \mathcal{G} if

$$P(A) = 0 \iff \tilde{P}(A) = 0$$

for every $A \in \mathcal{G}$. The Radon-Nikodym theorem (from measure theory) states that this is equivalent to the existence of a density $M > 0$ for \tilde{P} with respect to P . That is, that there exists an \mathcal{G} -measurable random variable $M > 0$ such that

$$\tilde{E}[X] = E[MX]$$

for every suitably integrable \mathcal{G} -measurable X . In particular,

$$\tilde{P}(A) = E[M1_A] \tag{A}$$

for every $A \in \mathcal{G}$. One thinks of $M(\omega)$ as the factor by which one reweights the probability $P(d\omega)$ to get $\tilde{P}(d\omega)$; $\tilde{P}(d\omega) = M(\omega)P(d\omega)$. In the discrete case this is exactly what happens (see (6) below).

Note also that it must be the case that $E[M] = 1$. Turning things around, let M be any \mathcal{G} -measurable random variable with $M > 0$ and $E[M] = 1$. Then the formula (A) can be taken to *define* a probability measure \tilde{P} , which is automatically equivalent to P .

- (2) If P and \tilde{P} are equivalent on \mathcal{F}_t , for every $t \geq 0$, then we get a family of densities M_t , one for each t . In other words, M_t is \mathcal{F}_t -measurable, and satisfies

$$\tilde{P}(A) = E[M_t 1_A] \tag{B}$$

for every $A \in \mathcal{F}_t$. In this case, we have that

$$\boxed{M_t \text{ is a martingale under } P.}$$

PROOF: for $s < t$ we need that $E[M_t | \mathcal{F}_s] = M_s$. According to the definition, this means that

$$E[M_s 1_A] = E[M_t 1_A]$$

for every $A \in \mathcal{F}_s$. But the LHS equals $\tilde{P}(A)$ by definition, and so does the RHS, because A also belongs to \mathcal{F}_t .

Conversely, let M_t be any P -martingale with $M_t > 0$ P -a.s. and $E[M_t] = 1$, for every t . Then (B) defines a measure \tilde{P} on \mathcal{F}_t . And the argument just given can be reversed to show

that these measures are consistent (that is, if $A \in \mathcal{F}_{t_1}$ and if we compute $\tilde{P}(A)$ from (B) with $t = t_1$, then we get the same answer as if we compute it from (B) using some other $t = t_2 > t_1$). In other words, there is a precise correspondence between strictly positive P -martingales on the one hand, and probabilities equivalent to P on the other.

It is worth noting that none of this says that P and \tilde{P} are equivalent on the whole σ -field $\mathcal{F} = \mathcal{F}_\infty$. In our financial models they typically aren't; we can't tell μ from r with certainty based on the information in \mathcal{F}_t , that is, up to any finite time t . But we can tell them apart if we let $t = \infty$.

- (3) Let B_t be a Brownian motion under P , and let \mathcal{F}_t be its filtration. Then for any predictable process γ_t for which the stochastic integral $\int_0^t \gamma_s dB_s$ makes sense, we know by Itô's lemma that

$$M_t = e^{-\int_0^t \gamma_s dB_s - \frac{1}{2} \int_0^t \gamma_s^2 ds} \quad (\text{C})$$

defines a local martingale. If a suitable integrability condition is satisfied then M_t is also a martingale. For example, this is true if γ_t is bounded, or more generally, if the *Novikov condition* $E[e^{\frac{1}{2} \int_0^t \gamma_s^2 ds}] < \infty \forall t$ is satisfied. In this case, the passage from P to \tilde{P} is known as a *Girsanov transformation*, and we have the following result:

$$\boxed{\tilde{B}_t = B_t + \int_0^t \gamma_s ds \text{ is a Brownian motion under } \tilde{P}.} \quad (\text{D})$$

- (4) Take $\gamma_t = \gamma = (\mu - r)/\sigma$. For our GBM model of stock prices, γ is the *market price of risk*, or *Sharpe ratio*, or *equity risk premium*. It is the risk premium per unit volatility that investors will demand before they buy the stock. The fact that it is strictly positive is a consequence of the capital asset pricing model (CAPM), and is based on the sub-linear utility of wealth.

Under P we have the equation

$$dS_t = \mu S_t dt + \sigma S_t dB_t.$$

substituting $d\tilde{B}_t = dB_t + \gamma dt$ gives that

$$dS_t = r S_t dt + \sigma S_t d\tilde{B}_t.$$

Our pricing and hedging arguments are all based on the hypothesis that we can find an equivalent probability measure \tilde{P} under which \tilde{B}_t is a Brownian motion. The obvious candidate

is the Girsanov transformation by (C), and the result of (D) tells us that this works.

- (5) It remains to prove the result (D) above. First we prove the following:

$$\boxed{N_t \text{ is a } \tilde{P}\text{-martingale} \iff N_t M_t \text{ is a } P\text{-martingale.}}$$

PROOF: Let N_t be a \tilde{P} -martingale. Then $E[N_t | \mathcal{F}_s] = N_s$ for $s < t$. This means that

$$\tilde{E}[N_t 1_A] = \tilde{E}[N_s 1_A]$$

for every $A \in \mathcal{F}_s$. Turning the \tilde{E} 's into E 's, we get that $E[M_t N_t 1_A] = E[M_s N_s 1_A]$ for every $A \in \mathcal{F}_s$, which means exactly that $E[M_t N_t | \mathcal{F}_s] = M_s N_s$. In other words, we do get that $M_t N_t$ is a P -martingale. The converse is obtained by simply reversing the argument.

PROOF OF (D): We use Itô's lemma on $\tilde{B}_t M_t$, to check that it is a P -martingale. First note that $dM_t = -\gamma_t M_t dB_t$, so $d[\tilde{B}, M]_t = d[B, M]_t = -\gamma_t M_t dt$. Thus

$$\begin{aligned} d\tilde{B}_t M_t &= M_t d\tilde{B}_t + \tilde{B}_t dM_t + d[\tilde{B}, M]_t \\ &= M_t (dB_t + \gamma_t dt) + \tilde{B}_t dM_t - \gamma_t M_t dt \\ &= M_t dB_t + \tilde{B}_t dM_t. \end{aligned}$$

Since both B_t and M_t are martingales under P , we see that $\tilde{B}_t M_t$ is one too (actually, this only shows that it's a local martingale, but it isn't hard to verify the appropriate integrability condition). So by the previous argument, \tilde{B}_t is a martingale under \tilde{P} .

To finish the proof, we use Lévy's theorem (see Sharpe, p. 168), that a continuous martingale is a Brownian motion if and only if it has quadratic variation t . In our case, \tilde{B}_t is clearly continuous, and we have just verified that it is a \tilde{P} -martingale. Since it only differs from B_t by a term of bounded variation, also $[\tilde{B}, \tilde{B}]_t = [B, B]_t = t$, and we're done.