

Chapter 9

Pricing in terms of Market Probabilities: The Radon-Nikodym Theorem.

9.1 Radon-Nikodym Theorem

Theorem 1.27 (Radon-Nikodym) Let P and \widetilde{P} be two probability measures on a space (Ω, \mathcal{F}) . Assume that for every $A \in \mathcal{F}$ satisfying $P(A) = 0$, we also have $\widetilde{P}(A) = 0$. Then we say that \widetilde{P} is absolutely continuous with respect to P . Under this assumption, there is a nonnegative random variable Z such that

$$\widetilde{P}(A) = \int_A Z dP, \quad \forall A \in \mathcal{F}, \quad (1.1)$$

and Z is called the Radon-Nikodym derivative of \widetilde{P} with respect to P .

Remark 9.1 Equation (1.1) implies the apparently stronger condition

$$\widetilde{E}X = E[XZ]$$

for every random variable X for which $E|XZ| < \infty$.

Remark 9.2 If \widetilde{P} is absolutely continuous with respect to P , and P is absolutely continuous with respect to \widetilde{P} , we say that P and \widetilde{P} are *equivalent*. P and \widetilde{P} are equivalent if and only if

$$P(A) = 0 \text{ exactly when } \widetilde{P}(A) = 0, \quad \forall A \in \mathcal{F}.$$

If P and \widetilde{P} are equivalent and Z is the Radon-Nikodym derivative of \widetilde{P} w.r.t. P , then $\frac{1}{Z}$ is the Radon-Nikodym derivative of P w.r.t. \widetilde{P} , i.e.,

$$\widetilde{E}X = E[XZ] \quad \forall X, \quad (1.2)$$

$$EY = \widetilde{E}\left[Y \cdot \frac{1}{Z}\right] \quad \forall Y. \quad (1.3)$$

(Let X and Y be related by the equation $Y = XZ$ to see that (1.2) and (1.3) are the same.)

Example 9.1 (Radon-Nikodym Theorem) Let $\Omega = \{HH, HT, TH, TT\}$, the set of coin toss sequences of length 2. Let P correspond to probability $\frac{1}{3}$ for H and $\frac{2}{3}$ for T , and let \tilde{P} correspond to probability $\frac{1}{2}$ for H and $\frac{1}{2}$ for T . Then $Z(\omega) = \frac{\tilde{P}(\omega)}{P(\omega)}$, so

$$Z(HH) = \frac{9}{4}, \quad Z(HT) = \frac{9}{8}, \quad Z(TH) = \frac{9}{8}, \quad Z(TT) = \frac{9}{16}.$$

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9.2 Radon-Nikodym Martingales

Let Ω be the set of all sequences of n coin tosses. Let P be the market probability measure and let \tilde{P} be the risk-neutral probability measure. Assume

$$P(\omega) > 0, \quad \tilde{P}(\omega) > 0, \quad \forall \omega \in \Omega,$$

so that P and \tilde{P} are equivalent. The Radon-Nikodym derivative of \tilde{P} with respect to P is

$$Z(\omega) = \frac{\tilde{P}(\omega)}{P(\omega)}.$$

Define the P -martingale

$$Z_k \triangleq \mathbb{E}[Z | \mathcal{F}_k], \quad k = 0, 1, \dots, n.$$

We can check that Z_k is indeed a martingale:

$$\begin{aligned} \mathbb{E}[Z_{k+1} | \mathcal{F}_k] &= \mathbb{E}[\mathbb{E}[Z | \mathcal{F}_{k+1}] | \mathcal{F}_k] \\ &= \mathbb{E}[Z | \mathcal{F}_k] \\ &= Z_k. \end{aligned}$$

Lemma 2.28 *If X is \mathcal{F}_k -measurable, then $\tilde{\mathbb{E}}X = \mathbb{E}[X Z_k]$.*

Proof:

$$\begin{aligned} \tilde{\mathbb{E}}X &= \mathbb{E}[X Z] \\ &= \mathbb{E}[\mathbb{E}[X Z | \mathcal{F}_k]] \\ &= \mathbb{E}[X \cdot \mathbb{E}[Z | \mathcal{F}_k]] \\ &= \mathbb{E}[X Z_k]. \end{aligned}$$

■

Note that Lemma 2.28 implies that if X is \mathcal{F}_k -measurable, then for any $A \in \mathcal{F}_k$,

$$\tilde{\mathbb{E}}[I_A X] = \mathbb{E}[Z_k I_A X],$$

or equivalently,

$$\int_A X d\tilde{P} = \int_A X Z_k dP.$$

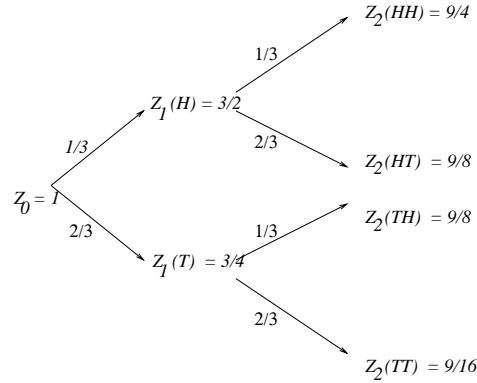


Figure 9.1: Showing the Z_k values in the 2-period binomial model example. The probabilities shown are for P , not \tilde{P} .

Lemma 2.29 If X is \mathcal{F}_k -measurable and $0 \leq j \leq k$, then

$$\tilde{\mathbb{E}}[X|\mathcal{F}_j] = \frac{1}{Z_j} \mathbb{E}[X Z_k|\mathcal{F}_j].$$

Proof: Note first that $\frac{1}{Z_j} \mathbb{E}[X Z_k|\mathcal{F}_j]$ is \mathcal{F}_j -measurable. So for any $A \in \mathcal{F}_j$, we have

$$\begin{aligned} \int_A \frac{1}{Z_j} \mathbb{E}[X Z_k|\mathcal{F}_j] d\tilde{\mathbb{P}} &= \int_A \mathbb{E}[X Z_k|\mathcal{F}_j] d\mathbb{P} \quad (\text{Lemma 2.28}) \\ &= \int_A X Z_k d\mathbb{P} \quad (\text{Partial averaging}) \\ &= \int_A X d\tilde{\mathbb{P}} \quad (\text{Lemma 2.28}) \end{aligned}$$

■

Example 9.2 (Radon-Nikodym Theorem, continued) We show in Fig. 9.1 the values of the martingale Z_k . We always have $Z_0 = 1$, since

$$Z_0 = \mathbb{E}Z = \int_{\Omega} Z d\mathbb{P} = \tilde{\mathbb{P}}(\Omega) = 1.$$

■

9.3 The State Price Density Process

In order to express the value of a derivative security in terms of the market probabilities, it will be useful to introduce the following *state price density process*:

$$\zeta_k = (1+r)^{-k} Z_k, \quad k = 0, \dots, n.$$

We then have the following pricing formulas: For a **Simple European derivative security** with payoff C_k at time k ,

$$\begin{aligned} V_0 &= \widetilde{\mathbb{E}} \left[(1+r)^{-k} C_k \right] \\ &= \mathbb{E} \left[(1+r)^{-k} Z_k C_k \right] \quad (\text{Lemma 2.28}) \\ &= \mathbb{E}[\zeta_k C_k]. \end{aligned}$$

More generally for $0 \leq j \leq k$,

$$\begin{aligned} V_j &= (1+r)^j \widetilde{\mathbb{E}} \left[(1+r)^{-k} C_k | \mathcal{F}_j \right] \\ &= \frac{(1+r)^j}{Z_j} \mathbb{E} \left[(1+r)^{-k} Z_k C_k | \mathcal{F}_j \right] \quad (\text{Lemma 2.29}) \\ &= \frac{1}{\zeta_j} \mathbb{E}[\zeta_k C_k | \mathcal{F}_j] \end{aligned}$$

Remark 9.3 $\{\zeta_j V_j\}_{j=0}^k$ is a martingale under \mathbf{P} , as we can check below:

$$\begin{aligned} \mathbb{E}[\zeta_{j+1} V_{j+1} | \mathcal{F}_j] &= \mathbb{E}[\mathbb{E}[\zeta_k C_k | \mathcal{F}_{j+1}] | \mathcal{F}_j] \\ &= \mathbb{E}[\zeta_k C_k | \mathcal{F}_j] \\ &= \zeta_j V_j. \end{aligned}$$

Now for an **American derivative security** $\{G_k\}_{k=0}^n$:

$$\begin{aligned} V_0 &= \sup_{\tau \in T_0} \widetilde{\mathbb{E}} \left[(1+r)^{-\tau} G_\tau \right] \\ &= \sup_{\tau \in T_0} \mathbb{E} \left[(1+r)^{-\tau} Z_\tau G_\tau \right] \\ &= \sup_{\tau \in T_0} \mathbb{E}[\zeta_\tau G_\tau]. \end{aligned}$$

More generally for $0 \leq j \leq n$,

$$\begin{aligned} V_j &= (1+r)^j \sup_{\tau \in T_j} \widetilde{\mathbb{E}} \left[(1+r)^{-\tau} G_\tau | \mathcal{F}_j \right] \\ &= (1+r)^j \sup_{\tau \in T_j} \frac{1}{Z_j} \mathbb{E} \left[(1+r)^{-\tau} Z_\tau G_\tau | \mathcal{F}_j \right] \\ &= \frac{1}{\zeta_j} \sup_{\tau \in T_j} \mathbb{E}[\zeta_\tau G_\tau | \mathcal{F}_j]. \end{aligned}$$

Remark 9.4 Note that

(a) $\{\zeta_j V_j\}_{j=0}^n$ is a supermartingale under \mathbf{P} ,

(b) $\zeta_j V_j \geq \zeta_j G_j \forall j$,

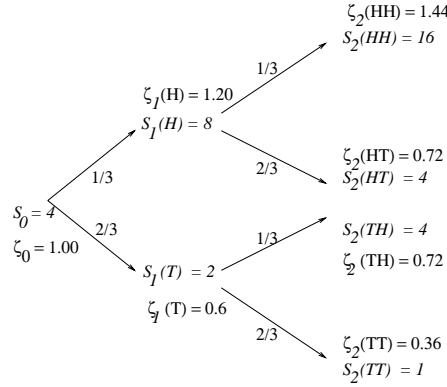


Figure 9.2: Showing the state price values ζ_k . The probabilities shown are for P , not \widetilde{IP} .

(c) $\{\zeta_j V_j\}_{j=0}^n$ is the smallest process having properties (a) and (b).

We interpret ζ_k by observing that $\zeta_k(\omega)IP(\omega)$ is the value at time zero of a contract which pays \$1 at time k if ω occurs.

Example 9.3 (Radon-Nikodym Theorem, continued) We illustrate the use of the valuation formulas for European and American derivative securities in terms of market probabilities. Recall that $p = \frac{1}{3}$, $q = \frac{2}{3}$. The state price values ζ_k are shown in Fig. 9.2.

For a **European Call** with strike price 5, expiration time 2, we have

$$V_2(HH) = 11, \quad \zeta_2(HH)V_2(HH) = 1.44 \times 11 = 15.84.$$

$$V_2(HT) = V_2(TH) = V_2(TT) = 0.$$

$$V_0 = \frac{1}{3} \times \frac{1}{3} \times 15.84 = 1.76.$$

$$\frac{\zeta_2(HH)}{\zeta_1(HH)} V_2(HH) = \frac{1.44}{1.20} \times 11 = 1.20 \times 11 = 13.20$$

$$V_1(H) = \frac{1}{3} \times 13.20 = 4.40$$

Compare with the risk-neutral pricing formulas:

$$V_1(H) = \frac{2}{5}V_1(HH) + \frac{2}{5}V_1(HT) = \frac{2}{5} \times 11 = 4.40,$$

$$V_1(T) = \frac{2}{5}V_1(TH) + \frac{2}{5}V_1(TT) = 0,$$

$$V_0 = \frac{2}{5}V_1(H) + \frac{2}{5}V_1(T) = \frac{2}{5} \times 4.40 = 1.76.$$

Now consider an **American put** with strike price 5 and expiration time 2. Fig. 9.3 shows the values of $\zeta_k(5 - S_k)^+$. We compute the value of the put under various stopping times τ :

(0) Stop immediately: value is 1.

(1) If $\tau(HH) = \tau(HT) = 2$, $\tau(TH) = \tau(TT) = 1$, the value is

$$\frac{1}{3} \times \frac{2}{3} \times 0.72 + \frac{2}{3} \times 1.80 = 1.36.$$

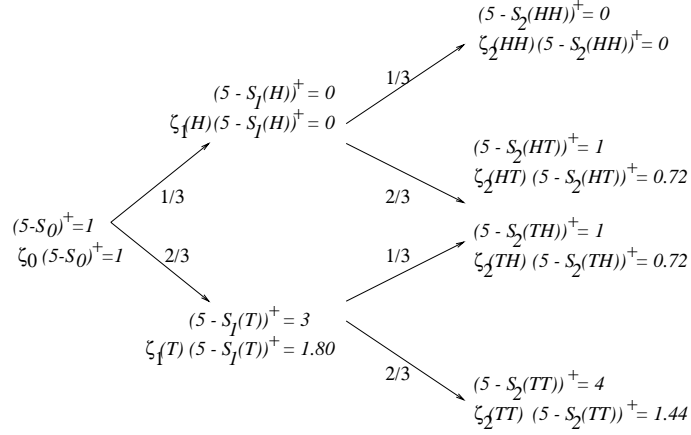


Figure 9.3: Showing the values $\zeta_k(5 - S_k)^+$ for an American put. The probabilities shown are for P , not \widetilde{P} .

(2) If we stop at time 2, the value is

$$\frac{1}{3} \times \frac{2}{3} \times 0.72 + \frac{2}{3} \times \frac{1}{3} \times 0.72 + \frac{2}{3} \times \frac{2}{3} \times 1.44 = 0.96$$

We see that (1) is optimal stopping rule. ■

9.4 Stochastic Volatility Binomial Model

Let Ω be the set of sequences of n tosses, and let $0 < d_k < 1 + r_k < u_k$, where for each k , d_k, u_k, r_k are \mathcal{F}_k -measurable. Also let

$$\tilde{p}_k = \frac{1 + r_k - d_k}{u_k - d_k}, \quad \tilde{q}_k = \frac{u_k - (1 + r_k)}{u_k - d_k}.$$

Let \widetilde{P} be the risk-neutral probability measure:

$$\widetilde{P}\{\omega_1 = H\} = \tilde{p}_0,$$

$$\widetilde{P}\{\omega_1 = T\} = \tilde{q}_0,$$

and for $2 \leq k \leq n$,

$$\widetilde{P}[\omega_{k+1} = H | \mathcal{F}_k] = \tilde{p}_k,$$

$$\widetilde{P}[\omega_{k+1} = T | \mathcal{F}_k] = \tilde{q}_k.$$

Let P be the market probability measure, and assume $P\{\omega\} > 0 \forall \omega \in \Omega$. Then P and \widetilde{P} are equivalent. Define

$$Z(\omega) = \frac{\widetilde{P}(\omega)}{P(\omega)} \quad \forall \omega \in \Omega,$$

$$Z_k = \mathbb{E}[Z | \mathcal{F}_k], \quad k = 0, 1, \dots, n.$$

We define the *money market price process* as follows:

$$M_0 = 1,$$

$$M_k = (1 + r_{k-1})M_{k-1}, \quad k = 1, \dots, n.$$

Note that M_k is \mathcal{F}_{k-1} -measurable.

We then define the *state price process* to be

$$\zeta_k = \frac{1}{M_k} Z_k, \quad k = 0, \dots, n.$$

As before the portfolio process is $\{\Delta_k\}_{k=0}^{n-1}$. The self-financing value process (wealth process) consists of X_0 , the non-random initial wealth, and

$$X_{k+1} = \Delta_k S_{k+1} + (1 + r_k)(X_k - \Delta_k S_k), \quad k = 0, \dots, n-1.$$

Then the following processes are martingales under $\widetilde{\mathbb{P}}$:

$$\left\{ \frac{1}{M_k} S_k \right\}_{k=0}^n \quad \text{and} \quad \left\{ \frac{1}{M_k} X_k \right\}_{k=0}^n,$$

and the following processes are martingales under \mathbb{P} :

$$\{\zeta_k S_k\}_{k=0}^n \quad \text{and} \quad \{\zeta_k X_k\}_{k=0}^n.$$

We thus have the following pricing formulas:

Simple European derivative security with payoff C_k at time k :

$$\begin{aligned} V_j &= M_j \widetilde{\mathbb{E}} \left[\frac{C_k}{M_k} \middle| \mathcal{F}_j \right] \\ &= \frac{1}{\zeta_j} \mathbb{E} [\zeta_k C_k | \mathcal{F}_j] \end{aligned}$$

American derivative security $\{G_k\}_{k=0}^n$:

$$\begin{aligned} V_j &= M_j \sup_{\tau \in T_j} \widetilde{\mathbb{E}} \left[\frac{G_\tau}{M_\tau} \middle| \mathcal{F}_j \right] \\ &= \frac{1}{\zeta_j} \sup_{\tau \in T_j} \mathbb{E} [\zeta_\tau G_\tau | \mathcal{F}_j]. \end{aligned}$$

The usual hedging portfolio formulas still work.

9.5 Another Application of the Radon-Nikodym Theorem

Let (Ω, \mathcal{F}, Q) be a probability space. Let \mathcal{G} be a sub- σ -algebra of \mathcal{F} , and let X be a non-negative random variable with $\int_{\Omega} X dQ = 1$. We construct the conditional expectation (under Q) of X given \mathcal{G} . On \mathcal{G} , define two probability measures

$$\begin{aligned} \mathbb{P}(A) &= Q(A) \quad \forall A \in \mathcal{G}; \\ \widetilde{\mathbb{P}}(A) &= \int_A X dQ \quad \forall A \in \mathcal{G}. \end{aligned}$$

Whenever Y is a \mathcal{G} -measurable random variable, we have

$$\int_{\Omega} Y d\mathbb{P} = \int_{\Omega} Y dQ;$$

if $Y = \mathbf{1}_A$ for some $A \in \mathcal{G}$, this is just the definition of \mathbb{P} , and the rest follows from the “standard machine”. If $A \in \mathcal{G}$ and $\mathbb{P}(A) = 0$, then $Q(A) = 0$, so $\widetilde{\mathbb{P}}(A) = 0$. In other words, the measure $\widetilde{\mathbb{P}}$ is absolutely continuous with respect to the measure \mathbb{P} . The Radon-Nikodym theorem implies that there exists a \mathcal{G} -measurable random variable Z such that

$$\widetilde{\mathbb{P}}(A) \triangleq \int_A Z d\mathbb{P} \quad \forall A \in \mathcal{G},$$

i.e.,

$$\int_A X dQ = \int_A Z d\mathbb{P} \quad \forall A \in \mathcal{G}.$$

This shows that Z has the “partial averaging” property, and since Z is \mathcal{G} -measurable, it is the conditional expectation (under the probability measure Q) of X given \mathcal{G} . The existence of conditional expectations is a consequence of the Radon-Nikodym theorem.