

MTH 9831. LECTURE 9

ELENA KOSYGINA

ABSTRACT. After considering a few more examples related to connections with PDEs we turn to exotic options (barrier and Asian).

1. Further examples of connections with PDEs:
Pricing of a zero-coupon bond when the interest rate is random.
2. Pricing of barrier options:
 - (a) a probabilistic approach;
 - (b) a PDE approach.
3. Pricing of Asian options. New ideas: augmentation of the state space and *reduction of dimension using the change of numeraire. (* next lecture)
4. Fubini's theorem for conditional expectations.

1. FURTHER EXAMPLES OF CONNECTIONS WITH PDES

Assume that the interest rate under $\tilde{\mathbb{P}}$ satisfies

$$(1) \quad dR(t) = \beta(t, R(t))dt + \gamma(t, R(t))d\tilde{B}(t).$$

Example 1.1 (Vasiček, Hull-White, CIR models).

- (Vasiček model) It assumes that the rate is driven by a mean-reverting OU process, namely,

$$dR(t) = (a - bR(t))dt + \sigma d\tilde{B}(t),$$

where $a, b, \sigma > 0$ are constants.

- (Hull-White model) This is a generalization of Vasiček model which allows the coefficients to vary in time:

$$dR(t) = (a(t) - b(t)R(t))dt + \sigma(t)d\tilde{B}(t),$$

where $a(t), b(t), \sigma(t)$ are positive non-random functions.

- (Cox-Ingersoll-Ross model) While the previous two models allow the rate to become negative with positive probability, CIR model ensures that the rate process is always non-negative. The SDE for $R(t)$ looks as follows:

$$dR(t) = (a - bR(t))dt + \sigma\sqrt{R(t)}d\tilde{B}(t),$$

where $a, b, \sigma > 0$ are constants.

These models are examples of so-called one factor short rate models. For examples of two-factor models and more general HJM framework, see Chapter 10 of Shreve, Vol. II.

We have the discount process

$$dD(t) = -R(t)D(t)dt, \quad D(0) = 1;$$

$$D(t) = \exp\left(-\int_0^t R(s)ds\right);$$

the MMA price

$$M(t) = \frac{1}{D(t)} = \exp \left(\int_0^t R(s) ds \right),$$

$$dM(t) = R(t)M(t)dt = \frac{R(t)}{D(t)}dt, \quad M(0) = 1.$$

Recall that $B(t, T)$ is the time t price of a unit zero coupon bond maturing at T (pays \$1 at T).

$$B(t, T) = \frac{1}{D(t)} \tilde{\mathbb{E}}(D(T) \cdot 1 | \mathcal{F}(t));$$

$$B(t, T) = \tilde{\mathbb{E}}[e^{-\int_t^T R(s) ds} | \mathcal{F}(t)].$$

Yield over the time interval $[t, T]$ is

$$Y(t, T) := -\frac{1}{T-t} \ln B(t, T);$$

$$B(t, T) = e^{-Y(t, T)(T-t)},$$

where $Y(t, T)$ can be thought of as the constant rate over $[t, T]$ which is consistent with the price $B(t, T)$. Since $R(t)$ is a solution of an SDE, it is a Markov process, and we would like to say that for some non-random function $f(t, y)$

$$(2) \quad B(t, T) = f(t, R(t)).$$

Rigorously speaking, $e^{-\int_t^T R(s) ds}$ is not of the form $h(R(T))$, so our definition of a Markov process does not allow us to conclude (2). But heuristically the only way $e^{-\int_t^T R(s) ds}$ depends on the path $R(s)$ for $0 \leq s \leq t$ is only through its value at $s = t$. To find the PDE for f , we use the fact that $D(t)B(t, T) = e^{-\int_0^t R(s) ds} f(t, R(t))$ is a $\tilde{\mathbb{P}}$ -martingale.

$$d(e^{-\int_0^t R(s) ds} f(t, R(t))) = e^{-\int_0^t R(s) ds} (-R(t)f dt + f_t dt + f_y(\beta dt + \gamma d\tilde{B}(t)) + \frac{1}{2} f_{yy} \gamma^2 dt)$$

$$= e^{-\int_0^t R(s) ds} ((-Rf + f_t + \beta f_y + \frac{1}{2} \gamma^2 f_{yy}) dt + \gamma f_y d\tilde{B}(t)); f = f(t, R(t)).$$

Therefore, the corresponding PDE is

$$(3) \quad f_t(t, y) + \beta(t, y) f_y(t, y) + \frac{1}{2} \gamma^2(t, y) f_{yy}(t, y) = y f(t, y),$$

or, setting $\mathcal{A}_t := \beta(t, y) \frac{\partial}{\partial y} + \gamma(t, y) \frac{\partial^2}{\partial y^2}$ we get

$$f_t(t, y) + \mathcal{A}_t f(t, y) = y f(t, y);$$

$$f(T, y) = 1.$$

See [Example 6.5.1](#) for a derivation of an explicit formula for the Hull-White model and [Example 6.5.2](#) for CIR model.

Key idea for finding a solution: these models are affine yield models, which means that

$$f(t, y) = e^{-yC_1(t, T) - C_2(t, T)}$$

for some $C_1(t, T)$ and $C_2(t, T)$ (to be determined). The name comes from the fact that the yield $Y(t, T)$ can be assumed to be of the form $y \frac{C_1(t, T)}{T-t} + \frac{C_2(T-t)}{T-t}$, which is an affine function of y .

[Example 6.5.3](#) shows that the price $C(t, y)$ of a call option on a bond with expiration $0 \leq T_1 \leq T$ requires to solve the same PDE (3), i.e. $C_t + \beta C_y + \frac{1}{2} \gamma^2 C_{yy} = yC$ but with a different terminal condition

$$C(T_1, y) = (f(T_1, y) - K)_+.$$

2. PRICING OF BARRIER OPTIONS

Standard barrier options can be either puts or calls, and each of these comes in 4 “flavors”: up-and-out, down-and-out, up-and-in, and down-and-in.

Tools for pricing barrier options:

- joint distribution of BM with a drift and its running maximum (see lecture 6);
- stopped martingales;
- risk-neutral pricing formula;
- use of Itô’s formula to determine whether a given function of a diffusion process is a martingale.

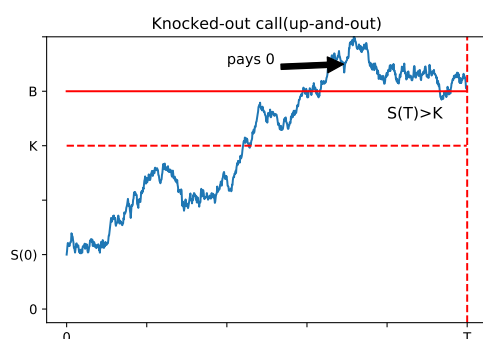


FIGURE 1. Knocked-out call

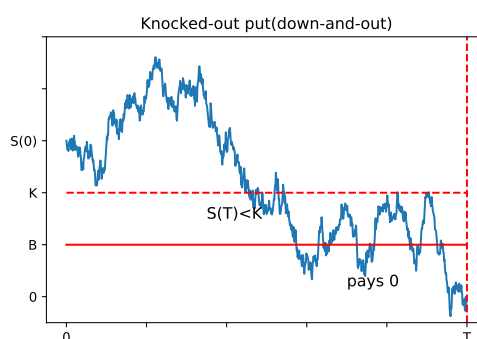


FIGURE 2. Knocked-out put

Framework: Black-Scholes-Merton model. Assume that under the risk-neutral measure the stock price satisfies

$$(4) \quad dS(t) = rS(t)dt + \sigma S(t)d\tilde{B}(t).$$

Options: we shall consider an up-and-out call with strike K , barrier $B > K$ and expiration T . The payoff of this option is the same as the payoff of a call option provided that the stock price stays below the barrier b for all times between 0 and T . If the stock price reaches the barrier at some time before T , we say that it was knocked out. In this case the option has payoff 0. We assume that $S(0) \leq B$.¹

Note that the payoffs of an up-and-in call option and an up-and-out call option with the same strike and expiration add up to the payoff of a standard call option. Therefore, the price of an up-and-in call option can be determined from the prices of the other two options.

Put options and other “flavors” are dealt with in a similar way. The most natural barrier puts are down-and-out and down-and-in put options.

Goals: develop two methods of pricing of an up-and-out call option.

- Use a probabilistic approach to set up an integral which gives the price.
- Write down the PDE and boundary conditions satisfied by the call price.

(a) We shall start with a probabilistic approach. Recall that $S(t) = S(0)e^{\sigma\tilde{B}(t) + (r - \frac{\sigma^2}{2})t} = S(0)e^{\sigma\hat{B}(t)}$, where $\hat{B}(t) = \tilde{B}(t) + \alpha t$, and $\alpha = \frac{r - \sigma^2/2}{\sigma}$.

Define $\hat{B}^*(t) = \max_{0 \leq s \leq t} \hat{B}(s)$. Since e^x is an increasing function and $\sigma > 0$, we can write

$$\max_{0 \leq t \leq T} S(t) = S(0)e^{\sigma\hat{B}^*(T)}.$$

¹Otherwise the price is 0.

The payoff of the option is

$$\begin{aligned}
 V(T) &= (S(0)e^{\sigma \hat{B}(T)} - K)_+ \mathbb{1}_{\{S(0)e^{\sigma \hat{B}^*(T)} < B\}} \\
 &= (S(0)e^{\sigma \hat{B}(T)} - K) \mathbb{1}_{\{S(0)e^{\sigma \hat{B}(T)} \geq K, S(0)e^{\sigma \hat{B}^*(T)} < B\}} \\
 &= (S(0)e^{\sigma \hat{B}(T)} - K) \mathbb{1}_{\left\{ \underbrace{\hat{B}(T) \geq \frac{1}{\sigma} \log \frac{K}{S(0)}}_{k \in \mathbb{R}}; \underbrace{\hat{B}^*(T) < \frac{1}{\sigma} \log \frac{B}{S(0)}}_{b > 0} \right\}}.
 \end{aligned}$$

Risk-neutral pricing formula tells us that the value of this option at time $t \in [0, T]$ is equal to

$$V(t) = \tilde{\mathbb{E}} \left(e^{-r(T-t)} V(T) | \mathcal{F}(t) \right), \quad 0 \leq t \leq T.$$

By the tower property, $e^{-rt}V(t)$ is a $\tilde{\mathbb{P}}$ -martingale.

To compute $V(0)$ we write

$$V(0) = e^{-rT} \tilde{\mathbb{E}} \left((S(0)e^{\sigma \hat{B}(T)} - K) \mathbb{1}_{\{\hat{B}(T) \geq k, \hat{B}^*(T) < b\}} \right).$$

The joint distribution of $(\hat{B}(T), \hat{B}^*(T))$ was computed in lecture 6. We have (under $\tilde{\mathbb{P}}$, as it is our reference measure here)

$$\hat{f}(x, a) = \begin{cases} \frac{2(2a-x)}{T\sqrt{2\pi T}} e^{\alpha x - \frac{1}{2}\alpha^2 T - \frac{1}{2T}(2a-x)^2}, & \text{if } x \leq a \text{ and } a \geq 0; \\ 0, & \text{otherwise.} \end{cases}$$

(Variable x corresponds to $\hat{B}(T)$, and a corresponds to $\hat{B}^*(T)$)

Therefore,

$$V(0) = e^{-rT} \int_k^b \left(\int_{x_+}^b (S(0)e^{\sigma x} - K) \frac{2(2a-x)}{T\sqrt{2\pi T}} e^{\alpha x - \frac{1}{2}\alpha^2 T - \frac{1}{2T}(2a-x)^2} da \right) dx.$$

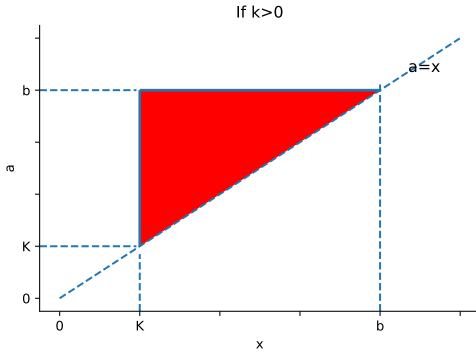


FIGURE 3. If $k > 0$

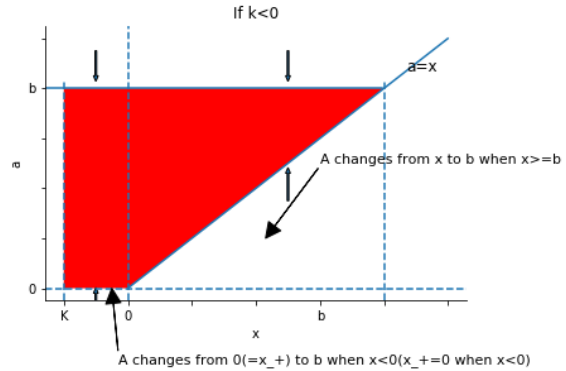


FIGURE 4. If $k < 0$

This integral can be computed in terms of $N(x)$. This gives a closed form formula. For details, see Shreve, Vol. II (pp. 304-308, Section 7.3.3).

(b) Let us now move to the PDE description.

Theorem 2.1. *Let $v(t, x)$ denote the price at time t of the up-and-out call under the assumption that the call has not knocked out prior to t and $S(t) = x$. Then $v(t, x)$ satisfies the BSM PDE*

$$v_t + rxv_x(t, x) + \frac{1}{2}\sigma^2 x^2 v_{xx}(t, x) = rv(t, x)$$

in the rectangle $\{(t, x) : 0 \leq t < T, 0 \leq x \leq B\}$ and satisfies boundary conditions

$$(5) \quad v(t, 0) = 0, \quad 0 \leq t \leq T;$$

$$(6) \quad v(t, B) = 0, \quad 0 \leq t < T;$$

$$(7) \quad v(T, x) = (x - K)_+, \quad 0 \leq x \leq B.$$

Recall that $V(t) = \tilde{\mathbb{E}}(e^{-r(T-t)}V(T)|\mathcal{F}(t))$. It is clear that $V(t)$ can not be represented as $v(t, S(t))$, since $V(t)$ "remembers" whether it has been knocked out or not, and $v(t, S(t))$ "does not remember" anything that happened prior to t .

Define $\rho = \inf\{t \geq 0 : S(t) = B\}$. Then ρ is the knock-out time (since upon reaching B , $S(t)$ will exceed B within any positive additional time increment with probability 1). Since ρ is the first passage time, it is a stopping time (namely $\{\rho \leq t\} \in \mathcal{F}(t)$ for each $t \geq 0$).

We have that $(e^{-rt}V(t))_{0 \leq t \leq T}$ is a $\tilde{\mathbb{P}}$ -martingale. We know that a martingale, stopped at a stopping time is again a martingale. Thus, $(e^{-r(t \wedge \rho)} \vee (t \wedge \rho))_{0 \leq t \leq T}$ is a $\tilde{\mathbb{P}}$ -martingale ($t \wedge \rho := \min\{t, \rho\}$).

Lemma 2.2. For $0 \leq t \leq \rho$, $V(t)$ is representable as $v(t, S(t))$. In particular, $(e^{-r(t \wedge \rho)}v(t \wedge \rho, S(t \wedge \rho)))_{0 \leq t \leq T}$ is a $\tilde{\mathbb{P}}$ -martingale.

Explanation for Lemma 2.2. On the event $\{\rho > t\}$, the option is alive, so the value of the up-and-out call is the same as for the regular call option.

Recall that for the regular call option, we had

$$V(t) = \tilde{\mathbb{E}}(e^{-r(T-t)} \underbrace{h(S(T))}_{(S(T)-K)_+} | \mathcal{F}(t)) \stackrel{\text{Markov property}}{=} v(t, S(t))$$

for some function $v(t, x)$.

Since $e^{-rt}V(t)$ is a martingale, $e^{-rt}v(t, S(t))$ is a martingale. The point here is that the same is true up to the random time ρ , or for all $t \leq \rho$, after that we stop our martingale, and a martingale stopped at ρ is still a martingale. \square

Sketch of proof of Theorem 2.1. By Lemma 2.2, $V(t) = v(t, S(t))$ for some v and $0 \leq t \leq \rho$. Moreover, $e^{-rt}v(t, S(t))$ is a martingale up to time ρ . Therefore, for $0 \leq t \leq \rho$,

$$\begin{aligned} d(e^{-rt}v(t, S(t))) &= e^{-rt}((-rv(t, S(t)) + v_t(t, S(t)) + v_x(t, S(t))S(t)r + \frac{1}{2}v_{xx}(t, S(t))\sigma^2 S^2(t))dt \\ &\quad + v_x(t, S(t))\sigma S(t)d\tilde{B}(t)), \end{aligned}$$

and setting the dt term to zero, we get for $0 \leq t \leq \rho$ that

$$-rv(t, S(t)) + v_t(t, S(t)) + v_x(t, S(t))S(t)r + \frac{1}{2}v_{xx}(t, S(t))\sigma^2 S^2(t) = 0.$$

Since $(t, S(t))$, $0 \leq t \leq \rho$, can reach any point in $(0, T) \times [0, B]$, the BSM PDE should hold for all $(t, x) \in [0, T) \times [0, B]$, i.e.

$$v_t(t, x) + v_x(t, x)xr + \frac{1}{2}v_{xx}\sigma^2 x^2 = rv(t, x), \quad (t, x) \in [0, T) \times [0, B].$$

The boundary conditions simply satisfy the conditions set forth by the option. \square

Remark 2.3. See Shreve II, p.303-4, for problems with the Δ -hedging strategy when the option is nearing the barrier close to the maturity time and what can be done about this in practice.

3. PRICING OF ASIAN OPTIONS

Let again $S(t)$ satisfy (4) under $\tilde{\mathbb{P}}$. The payoff of an Asian option with strike K and expiration T is given by

$$V(T) = \left(\frac{1}{T} \int_0^T S(u) du - K \right)_+;$$

$$V(t) = e^{-r(T-t)} \tilde{\mathbb{E}} \left(\left(\frac{1}{T} \int_0^T S(u) du - K \right)_+ \middle| \mathcal{F}(t) \right).$$

Clearly, the price depends on the whole path $S(u)$, $0 \leq u \leq t$, so we can not write $V(t)$ as a function of just t and $S(t)$, not even up to some stopping time as we did in the case of barrier options. A new idea is needed.

Idea: augmentation of the state space. To calculate $V(t)$, we definitely need to know $\int_0^t S(u) du$, since

$$\int_0^T S(u) du = \int_0^t S(u) du + \int_t^T S(u) du$$

and $\int_0^t S(u) du$ is $\mathcal{F}(t)$ -measurable, so it can be taken out of the conditional expectation. We shall introduce an auxiliary process $Y := (Y(t))_{t \in [0, T]}$ defined by

$$Y(t) = \int_0^t S(u) du,$$

or, in the differential form, $dY(t) = S(t)dt$. Y is a regular process. Consider a 2-dimensional process $(S(u), Y(u))_{t \leq u \leq T}$. This 2-dimensional process is a Markov process which we can describe by the following system of equations:

$$\begin{aligned} dS(u) &= rS(u)du + \sigma S(u)d\tilde{B}(u), \quad S(t) = x; \\ dY(u) &= S(u)du, \quad Y(t) = y. \end{aligned}$$

The generator of this process is given by

$$\mathcal{A}v(x, y) = rxv_x + xyv_y + \frac{1}{2}\sigma^2 x^2 v_{xx}.$$

The expression for the time t value is

$$V(t) = e^{-r(T-t)} \tilde{\mathbb{E}} \left(\left(\frac{1}{T} Y(T) - K \right)_+ \middle| \mathcal{F}(t) \right),$$

and $V(t)$ can be written as $v(t, S(t), Y(t))$ for some function $v(t, x, y)$.

Theorem 3.1. *Let $v(t, x, y)$ be the time t value of the Asian call option when $S(t) = x$, $Y(t) = y$. Then $v(t, x, y)$ satisfies the PDE*

$$(8) \quad v_t(t, x, y) + rxv_x(t, x, y) + xyv_y(t, x, y) + \frac{1}{2}\sigma^2 x^2 v_{xx}(t, x, y) = rv(t, x, y)$$

in $[0, T) \times [0, \infty) \times \mathbb{R}$ and the boundary conditions

$$(9) \quad v(t, 0, y) = e^{-r(T-t)} \left(\frac{y}{T} - K \right)_+, \quad 0 \leq t < T, y \in \mathbb{R};$$

$$(10) \quad \lim_{y \rightarrow -\infty} v(t, x, y) = 0, \quad 0 \leq t \leq T, x \geq 0;$$

$$(11) \quad v(T, x, y) = \left(\frac{y}{T} - K \right)_+, \quad x \geq 0, y \in \mathbb{R}.$$

Remark 3.2. There are several issues that need to be addressed.

- (1) Why are we looking also at $y < 0$? We know that $Y(t) = \int_0^t S(u)du \geq 0$, so how do we end up considering $y < 0$?
- (2) Usual existence & uniqueness questions for the above problem.

We shall start with (2): existence is not a problem but for uniqueness in an unbounded domain $(t, x, y) \in [0, T] \times [0, \infty) \times \mathbb{R}$, we need additional conditions at ∞ : what happens if $x \rightarrow \infty$ and $y \rightarrow \pm\infty$? We omit this discussion, since later we shall be able to reduce the dimension and write a simpler PDE with corresponding boundary conditions.

Turn now to (1). The function $Y(t) = \int_0^t S(u)du$ is only one specific solution of $dY(u) = S(u)du$, $0 \leq u \leq T$. Namely, the one which satisfies the condition $Y(0) = 0$. Yet we shall have to solve this equation for $u \in [t, T]$ ($t \leq T$) subject to the condition $Y(t) = y$. So we get

$$Y(u) = y + \int_t^u S(s)ds, \quad t \leq u \leq T.$$

Notice that while

$$S(t) = 0 \Rightarrow S(u) = 0 \quad \text{for all } u \in [t, T] \quad \text{and} \quad Y(t) = y \quad \text{for all } u \in [t, T],$$

we do not have the same property for the Y process: if $Y(t) = 0$, then $Y(u) = \int_t^u S(s)ds$ need not be 0, so we can not determine the value of $v(t, x, 0)$, and we can not provide a boundary condition for $y = 0$. But, at least mathematically, there is no problem considering $y \in \mathbb{R}$, and that is what we do.

Idea of a proof of Theorem 3.1. Since Y is a regular process, $d[Y]_t$ and $d[Y, S]_t$ are equal to 0 a.s.. Recall that $e^{-rt}V(t) = e^{-rt}v(t, S(t), Y(t))$ is $\tilde{\mathbb{P}}$ -martingale. Compute

$$d(e^{-rt}v(t, S(t), Y(t))) = e^{-rt}(-rv + v_t + rSv_x + Sv_y + \frac{1}{2}\sigma^2 S^2 v_{xx})dt + e^{-rt}\sigma S v_x d\tilde{B}(t).$$

Setting the dt term to zero, we get

$$\begin{aligned} v_t(t, S(t), Y(t)) + rS(t)v_x(t, S(t), Y(t)) + S(t)v_y(t, S(t), Y(t)) + \frac{1}{2}\sigma^2 S^2(t)v_{xx}(t, S(t), Y(t)) \\ = rv(t, S(t), Y(t)). \end{aligned}$$

This equation has to hold at all points (x, y) which can be possibly hit by S and Y process. Since we did not restrict ourselves to $y > 0$, the process $Y(t)$ can hit any point in \mathbb{R} , and $S(t)$ can hit any $x \geq 0$, so we arrive at PDE (8).

Now we turn to boundary conditions. If $S(t) = 0$, then $S(u) = 0$ for all $u \in [t, T]$, and $Y(t) = y$ for all $u \in [t, T]$. Thus the value of an Asian call at time t is

$$e^{-r(T-t)} \left(\frac{y}{T} - K \right)_+.$$

Thus, we get (9).

If $S(t) = x$, $Y(t) = y$ and we let $y \rightarrow -\infty$, then it is very unlikely that $(\frac{Y(T)}{T} - K)_+ > 0$. In other words,

$$\lim_{y \rightarrow -\infty} v(t, x, y) = 0.$$

This is (10).

Finally, (11) is just the payoff of the option. □

Remark 3.3. We have

$$d(e^{-rt}v(t, S(t), Y(t))) = \sigma e^{-rt}S(t)v_x(t, S(t), Y(t))d\tilde{B}(t).$$

On the other hand, the discounted value of the portfolio that has $\Delta(t)$ shares of this stock is given by (see Lecture 5)

$$d(e^{-rt}X(t)) = e^{-rt}\sigma S(t)\Delta(t)d\tilde{B}(t).$$

Thus, to hedge a short position in the Asian call, we should have $\Delta(t) = v_x(t, S(t), Y(t))$.

In the next lecture we shall perform a change of numeraire. The new numeraire will be the stock price. This will allow us to reduce our 2-dimensional problem to 1-dimensional.

4. FUBINI'S THEOREM FOR CONDITIONAL EXPECTATIONS

In conclusion, I shall state and prove one useful fact about conditional expectations. This is a corollary of Fubini's theorem.

Lemma 4.1 (Fubini's theorem for conditional expectations). *Let $X(t, \omega)$, $0 \leq t \leq T$, be a stochastic process adapted to some filtration $\mathcal{F}(t)$, $0 \leq t \leq T$, and such that either $X(t, \omega) \geq 0$ on $[0, T] \times \Omega$ or*

$$E \int_0^T |X(s)| ds < \infty.$$

Then for every $t \in [0, T]$

$$E \left[\int_t^T X(s) ds \mid \mathcal{F}(t) \right] = \int_t^T E[X(s) \mid \mathcal{F}(t)] ds \quad \text{a.s.}$$

Proof. This is a direct consequence of Fubini's theorem (Jacod and Protter, Probability Essentials, p.67) and the definition of conditional expectation. Fix $t \in [0, T]$ and set

$$Y(\omega) = E \left[\int_t^T X(s, \omega) ds \mid \mathcal{F}(t) \right], \quad Z(s, \omega) = E[X(s, \omega) \mid \mathcal{F}(t)].$$

Since t is fixed, we can drop it from the notation for simplicity. The definition of conditional expectation says that for every $A \in \mathcal{F}(t)$

$$\int_A Y(\omega) dP = \int_A \left(\int_t^T X(s, \omega) ds \right) dP.$$

By Fubini's theorem,

$$\int_A \int_t^T X(s, \omega) ds dP = \int_t^T \int_A X(s, \omega) dP ds.$$

By the definition of conditional expectation,

$$\int_A X(s, \omega) dP = \int_A Z(s, \omega) dP.$$

Again by Fubini's theorem,

$$\int_t^T \int_A Z(s, \omega) dP ds = \int_A \int_t^T Z(s, \omega) ds dP.$$

Putting all pieces together we obtain that for every $A \in \mathcal{F}(t)$

$$\int_A Y(\omega) dP = \int_A \left(\int_t^T Z(s, \omega) ds \right) dP.$$

Since $Y(\omega)$ and $\int_t^T Z(s, \omega) ds$ are $\mathcal{F}(t)$ -measurable and the above equality of integrals holds for all $A \in \mathcal{F}(t)$, we conclude that

$$Y = \int_t^T Z(s, \omega) ds \quad \text{a.s.,}$$

which is exactly what we need. □