### Stochastic Calculus II

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#### Kolmogorov Backward Equation

• Consider a stochastic process  $X_t$ .  $X_t$  is a martingale if for s > t,

$$E_t[X_s] = X_t$$

In other words, conditional expectation of future value is simply the current value (example: fair gamble). The notion of a martingale makes sense for both discrete and continuous time processes.

• Now let's look at this concept when  $X_t$  is an Ito process. Suppose

$$dX_t = \mu(t, X_t) dt + \sigma(t, X_t) dZ_t$$

• If  $X_t$  were a martingale, then over a small time interval dt,

$$E_t [X_{t+dt}] = X_t$$
$$E_t [X_{t+dt} - X_t] = 0$$
$$E_t [dX_t] = 0$$

This is another way of thinking about martingales: expected changes are zero.

• But now recall the dynamics of  $dX_t$ :

$$E_t [dX_t] = E_t [\mu (t, X_t) dt + \sigma (t, X_t) dZ_t]$$
$$= \mu (t, X_t) dt$$

- So X<sub>t</sub> is a martingale if and only if μ(t, X<sub>t</sub>) = 0. This is intuitive: the drift term is responsible for expected change whereas the diffusion term is responsible for variance of change.
- The Kolmogorov Backward Equation simply formalizes this idea.

### Continued

• Suppose we have an underlying process  $X_t$  where

$$dX_t = \mu(t, X_t) dt + \sigma(t, X_t) dZ_t$$

- Consider Y<sub>t</sub> = f (t, X<sub>t</sub>). Suppose we know for some reason that Y<sub>t</sub> is a martingale. We have seen two such examples: 1) conditional expectation, 2) security price discounted at risk-free rate under the risk-neutral measure. Then by the above argument, we know that the drift term of dY<sub>t</sub> = d (f (t, X<sub>t</sub>)) should be zero.
- By Ito's Lemma,

$$df(t, X_t) = \left(\frac{\partial f(t, X_t)}{\partial t} + \frac{\partial f(t, X_t)}{\partial X_t} \mu(t, X_t) + \frac{1}{2} \frac{\partial^2 f(t, X_t)}{\partial X_t^2} (\sigma(t, X_t))^2 \right) dt + \frac{\partial f(t, X_t)}{\partial X_t} \sigma(t, X_t) dZ_t$$

• So if  $Y_t = f(t, X_t)$  is a martingale, then

$$\frac{\partial f(t,X_t)}{\partial t} + \frac{\partial f(t,X_t)}{\partial X_t} \mu(t,X_t) + \frac{1}{2} \frac{\partial^2 f(t,X_t)}{\partial X_t^2} (\sigma(t,X_t))^2 = 0$$

### **Option Replication**

- How do you dynamically replicate an European option in the Black-Scholes setting? Answer: Match the "delta".
- Suppose you have the option price f (t, S<sub>t</sub>) on the one hand and the value of a replicating portfolio consisting of the stock and the bond, W<sub>t</sub>. At time t, this replicating portfolio holds θ<sub>t</sub> number of stocks and the rest is invested in the risk-free bond. This means

$$dW_t = \theta_t dS_t + r \left( W_t - \theta_t S_t \right) dt$$

• To have perfect replication, we need

$$d(f(t,S_t))=dW_t$$

In particular we need the coefficient in front of  $dS_t$  to match:

$$\frac{\partial f(t,S_t)}{\partial S_t} = \theta_t$$

• This argument is exactly as in yesterday's lecture where we tried to replicate an option in an environment with stochastic volatility using a stock and another option.

### State Price Density and Risk-Neutral Measure

- We can always move from one to the other very easily in both discrete time and continuous time settings.
- In discrete time, if the state price density is given by

$$\frac{\pi_t}{\pi_{t-1}} = \exp\left(-r_{t-1} - \eta_{t-1}\varepsilon_t - \frac{1}{2}\eta_{t-1}^2\right)$$

then the change of measure from  $\mathsf{P}$  to  $\mathsf{Q}$  is given by

$$\left(\frac{dQ}{dP}\right)_{t} = \exp\left(-\eta_{t-1}\varepsilon_{t} - \frac{1}{2}\eta_{t-1}^{2}\right)$$

• In continuous time, if the state price density is given by

$$\pi_t = \exp\left(-\int_0^t r_s ds - \int_0^t \eta_s dZs - \frac{1}{2}\int_0^t \eta_s^2 ds\right)$$

then the change of measure from  $\mathsf{P}$  to  $\mathsf{Q}$  is given by

$$\left(\frac{dQ}{dP}\right)_t = \exp\left(-\int_0^t \eta_s dZs - \frac{1}{2}\int_0^t \eta_s^2 ds\right)$$

• We can always decompose the state price density into time discount and change of measure from P to Q.

## Change of Measure in Continuous Time

• Similar as in the discrete time case. If

$$\left(\frac{dQ}{dP}\right)_{t} = \exp\left(-\int_{0}^{t}\eta_{s}dZs - \frac{1}{2}\int_{0}^{t}\eta_{s}^{2}ds\right)$$

then the Brownian motion under P,  $Z_t^P$  acquires a drift but its diffusion coefficient does not change. Therefore

$$dZ_t^P = dZ_t^Q - \eta_t dt$$

- Suppose W<sup>P</sup><sub>t</sub> is another Brownian motion under P and it has correlation ρ with Z<sup>P</sup><sub>t</sub>. How does W<sup>P</sup><sub>t</sub> look under the Q-measure?
- Do the decomposition

$$W_t^P = \rho Z_t^P + \sqrt{1 - \rho^2} V_t^P$$

where  $V_t^P$  is another Brownian motion under P that's independent of  $Z_t^P$ . Conclude

$$dW_t^P = dW_t^Q - \rho \eta_t dt$$

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