Technical Appendix to Risk, Return, and Dividends

Andrew Ang* Columbia University and NBER

> Jun Liu[†] UC San Diego

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*Columbia Business School, 3022 Broadway 805 Uris, New York NY 10027, ph: (212) 854-9154; fax: (212) 662-8474; email: aa610@columbia.edu; WWW: http://www.columbia.edu/~aa610.

[†]Rady School of Management, Pepper Canyon Hall, Room 320, 9500 Gilman Dr, MC 0093, La Jolla, CA 92093-0094; ph: (858) 534-2022; fax: (858) 534-0745; email: junliu@ucsd.edu; WWW: http://rady.ucsd.edu/cms/showcontent.aspx?ContentID=183

A Relation of Proposition 2.1 to Pricing Kernel Formulations

By definition, given the dividend process D_t , the price of the stock is given by:

$$P_t = \mathcal{E}_t \left[\int_t^\infty \Lambda_s D_s \, ds \right],\tag{A-1}$$

under the pricing kernel process Λ_t , together with a transversality assumption. We assume that the pricing kernel follows:

$$\frac{d\Lambda_t}{\Lambda_t} = -r_f(x_t)dt - \xi_x(x_t)dB_t^x - \xi_d(x_t)dB_t^d,$$
(A-2)

where $r_f(\cdot)$ is the risk-free rate process, and ξ_x and ξ_d are prices of risk corresponding to shocks to the state variable x_t and dividend growth, respectively. Using equation (A-1), we can express the price-dividend ratio as:

$$\begin{aligned} \frac{P_t}{D_t} &= \mathcal{E}_t \left[\int_t^\infty \exp\left(-\int_t^s (r_f + \frac{1}{2}(\xi_x^2 + \xi_d^2)) \, du + \xi_x \, dB_u^x + \xi_d \, dB_u^d \right) \\ & \times \exp\left(\int_t^s \mu_d du + \sigma_d dB_u^d \right) ds \right], \end{aligned}$$

assuming that $\sigma_{dx} = 0$ for simplicity. This can be equivalently written as:

$$\frac{P_t}{D_t} = \mathcal{E}_t^Q \left[\int_t^\infty \exp\left(-\int_t^s (r_f - \mu_d - \frac{1}{2}(\sigma_d - \xi_d)^2) \, du\right) ds \right],\tag{A-3}$$

where the Radon-Nikodym derivative defining the risk-neutral measure Q is given by:

$$\frac{dQ}{dP} = \exp\left(-\int_{t}^{s} \frac{1}{2}(\xi_{x}^{2} + (\sigma_{d} - \xi_{d})^{2}) \, du - \xi_{x} \, dB_{u}^{x} - (\sigma_{d} - \xi_{d}) dB_{u}^{d}\right). \tag{A-4}$$

Note that equation (A-3) is a function $f(\cdot)$ of x.

We describe how a particular choice of a return process dR_t , together with assumptions on dividends, places restrictions on the underlying pricing kernel process $d\Lambda_t$ through the following proposition:

Proposition A.1 Suppose the state of the economy is described by x_t , which follows equation (1), and a stock is a claim to the dividends D_t that are described by equation (2) with $\sigma_{dx} = 0$. If the stock return follows equation (8) and the pricing kernel process follows equation (A-2), then the price-dividend ratio $P_t/D_t = f(x_t)$ satisfies the following relation:

$$(\mu_x - \xi_x \sigma_x)f' + \frac{1}{2}\sigma_x^2 f'' - (r_f - \mu_d - \frac{1}{2}\sigma_d^2 + \xi_d \sigma_d)f = -1,$$
(A-5)

which determines the price-dividend ratio f. This implies that the expected return $\mu_r(\cdot)$ and volatility $\sigma_{rx}(\cdot)$ of the return are given by:

$$\mu_r = r_f + \xi_x \sigma_x (\ln f)' + \xi_d \sigma_d,$$

$$\sigma_{rx} = \sigma_x (\ln f)'$$
(A-6)

Proof: Equation (A-5) is the standard Feynman-Kac pricing equation. Once the price-dividend ratio f is obtained from solving equation (A-5), we can derive equation (A-6) by equating terms from the drift term of dR_t and the diffusion term on dB_t^x in equation (8).

Proposition A.1 states that, given the dividend stream, the pricing kernel completely determines the price-dividend ratio f, the expected return of the stock μ_r , and the volatility of the stock σ_{rx} . However, if we specify the price of the stock, the expected return, or the volatility of the stock (each one being sufficient to determine the other two from Proposition 2.1), the short rate r^f , the prices of risk ξ_x and ξ_d , or the pricing kernel Λ_t are not uniquely determined. For example, suppose we specify μ_r . There are potentially infinitely many pairs of r^f and $\xi = (\xi_x, \xi_d)$ that can produce the same μ_r . For example, one (trivial) choice of ξ is $\xi = (0, 0)$ corresponding to risk neutrality, and the stock return is the same as the risk-free rate. Whereas Proposition 2.1 shows that specifying μ_r , σ_{rx} , or f completely determines the return process, the result from Proposition A.1 implies that a single choice of μ_r , σ_{rx} , or f does not necessarily determine the pricing kernel.

B Multivariate State Variables

Suppose that there are K state variables, so that $x = (x_1, \ldots, x_K)^{\top}$ represents a $K \times 1$ vector of diffusion processes. We let x follow the diffusion process in equation (1), where $\mu_x(\cdot)$ is a vector function of x and $\sigma_x(\cdot)$ is a matrix function of x. Similarly, dividend growth satisfies equation (2) where the scalars $\mu_d(\cdot)$, $\sigma_d(\cdot)$ are potentially functions of x. For expositional simplicity, we assume that $\sigma_{dx} = 0$ and denote the scalar price-dividend ratio by P/D = f(x).

Suppose that the return R_t satisfies the following diffusion equation:

$$dR_t = \mu_r(x_t)dt + \sigma_{rx}(x_t)dB_t^x + \sigma_{rd}(x_t)dB_t^d = \mu_r(x_t)dt + \sum_{i=1}^K \sigma_{rx_i}(x_t)dB_t^{x_i} + \sigma_{rd}(x_t)dB_t^d,$$
(B-1)

where $\mu_r(\cdot)$ is a scalar function of x, $\sigma_{rx}(\cdot)$ is a matrix function of x, σ_{rx_i} represents the *i*th row of σ_{rx} , and the vector of Brownian motions dB_t^x is partitioned as $dB_t^x = (dB_t^{x_1} \dots dB_t^{x_K})^\top$. From the definition of the return $dR_t = df_t/f_t + dD_t/D_t + 1/f_t dt$, the diffusion term of the return is given by:

$$\left(\frac{\partial \ln f}{\partial x}\right)^{\top} \sigma_{rx} dB_t^x + \sigma_d dB_t^d.$$
(B-2)

Thus, in order for σ_{rx} to represent the diffusion coefficients of a return, we must have:

$$\sigma_{rx} = \left(\frac{\partial \ln f}{\partial x}\right)^{\top} \sigma_x, \tag{B-3}$$

or, equivalently:

$$\sigma_{rx_i} = \sum_{j=1}^{K} \frac{\partial \ln f}{\partial x_j} (\sigma_x)_{ji},$$

where $(\sigma_x)_{ji}$ is the element of σ_x in the *j*th row and *i*th column. From equation (B-3), there must be a function *f* such that:

$$\frac{\partial \ln f}{\partial x} = (\sigma_{rx} \sigma_x^{-1})^\top.$$
(B-4)

The necessary and sufficient condition for this is:

Assumption B.1 The diffusion coefficients σ_x and σ_{rx} satisfy the integrability condition:

$$\frac{\partial}{\partial x_j} (\sigma_{rx} \sigma_x^{-1})_i = \frac{\partial}{\partial x_i} (\sigma_{rx} \sigma_x^{-1})_j.$$
(B-5)

Note that unlike the univariate case, we cannot arbitrarily specify the diffusion coefficients σ_{rx} of the return. If σ_{rx} does not satisfy the integrability condition, then equation (B-1) cannot represent a return implied from a pricing function. The multivariate version of equations (8) and (11) in Proposition 2.1 are:

Proposition B.1 Suppose that the return R_t follows the diffusion equation (B-1) and that diffusion ratio $\sigma_{rx}\sigma_x^{-1}$ satisfies the integrability condition Assumption B.1. Then, the price-dividend ratio and the expected return are determined up to integration constant.

Proof: From the integrability condition (B-5), it follows from elementary calculus that there exists a function f that satisfies:

$$d\ln f = \sigma_{rx}\sigma_x^{-1}dx. \tag{B-6}$$

Equation (B-6) is the multivariate version of equation (11). The function f is the dividend yield, and it is unique up to a multiplicative constant (since $\ln f$ is determined up to an additive constant). The expected return is then determined by:

$$\mu_r(x) = \frac{\mu_x^\top f' + \frac{1}{2} f''^\top \sigma_x \sigma_x^\top f'' + 1}{f} + \mu_d + \frac{1}{2} \sigma_d^2.$$
(B-7)

Equation (B-7) is the multivariate version of equation (8). \blacksquare

The integrability condition in Assumption B.1 imposes strong restrictions on multivariate stochastic volatility processes. For example, the diffusion process

$$\sqrt{v_1}dB_t^1 + v_1\sqrt{v_2}dB_t^2,$$

where v_1 and v_2 are stochastic processes, cannot represent the return diffusion process of a valid pricing function because it violates the condition (B-5).