

$Y(t): t \geq 0$  BMD  $\mu, \sigma^2$

show:

$$E[X(t) | X(u); 0 \leq u \leq s] = X(s) e^{(t-s)(\mu + \sigma^2/2)}$$

Since  $X(t)$  is a Geometric Brownian Motion process, which is memoryless, the conditional expectation of all subsequent values depends only on the most recent value  $[X(s)]$ , and is not affected by all prior values  $[X(u); 0 \leq u < s]$ .

$$\therefore X(t) | X(u); 0 \leq u \leq s$$

||

$$X(t) | X(s)$$

|| implies

$$X(t) | X(s) = X(s) e^{Y(t-s)}$$

$X(t)$  has a lognormal distribution with parameters

$(\ln(X(s)) + \mu(t-s))$  and  $\sigma\sqrt{(t-s)}$

$$E[X(t) | X(s)] = E[X(s) e^{Y(t-s)}]$$

$$= X(s) E[e^{Y(t-s)}], \text{ where } Y(t-s) \text{ has}$$

using  $E[e^{aY}] = e^{aE[Y] + a^2 \text{Var}(Y)/2}$  param:  $\mu, \sigma^2$

$$E[e^{Y(t-s)}] = e^{(t-s)(\mu + \sigma^2/2)}$$

$$E[X(t) | X(s)] = X(s) e^{(t-s)(\mu + \sigma^2/2)}$$