

# From Discrete Time to Continuous Time Modeling

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## Arrow-Debreu Securities

- Consider a simple one-period economy containing a single stock and a single savings account.
- The economy has two possible states in one time step from now.
- Is the model arbitrage free?



## Arrow-Debreu Securities

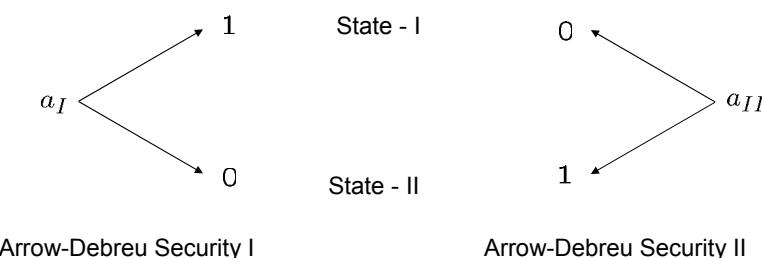
- An arbitrage portfolio is a portfolio which costs zero at  $t=0$ , but may have a positive pay-off at  $t=1$
- The model is arbitrage free if and only if

$$S_d < (1 + r) S_0 < S_u$$

- Suppose that  $S_u > S_d \geq (1+r) S_0$ , then it is always better to invest in the asset than the money-market account
- An example of an arbitrage portfolio is
  - Long 1 unit of Asset
  - Short  $1/S_0$  units of the Bank Account

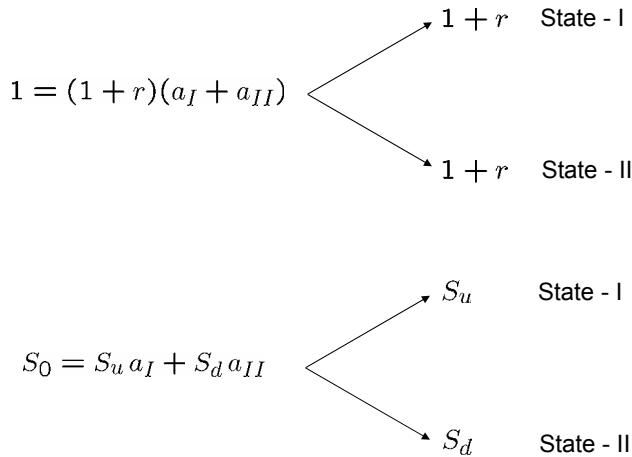
## Arrow-Debreu Securities

- Consider two fictitious assets which pay exactly 1 in one of the two states of the world and zero in the other.
- What is a rational price for these assets?



## Arrow-Debreu Securities

- Make a portfolio of AD securities which generate the pay-offs of the existing claims:



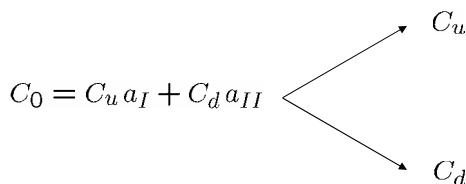
## Arrow-Debreu Securities

- Solving the linear system gives the prices of the AD securities:

$$a_I = \frac{1}{1 + r} \frac{(1 + r) S_0 - S_d}{S_u - S_d}$$

$$a_{II} = \frac{1}{1 + r} \frac{S_u - (1 + r) S_0}{S_u - S_d}$$

- Given these prices, the price of a contingent claim paying  $C_u$  and  $C_d$  in the two states of the world must be (otherwise an arbitrage exists):



## Arrow-Debreu Securities

- The price of the claim can be rewritten as follows:

$$\begin{aligned} C_0 &= C_u a_I + C_d a_{II} \\ &= \frac{1}{1+r} \{q C_u + (1-q) C_d\} \end{aligned}$$

where,

$$q = \frac{(1+r)S_0 - S_d}{S_u - S_d}$$

- Notice that the state probabilities do not appear in price!
- Is  $q$  a probability?
- YES - since no arbitrage requires  $S_d < (1+r)S_0 < S_u$

## Arrow-Debreu Securities

- Introduce the relative price of an asset/claim as,  $\tilde{C}_t \equiv \frac{C_t}{M_t}$
- Where  $M_t$  denotes the value of the money-market account (bank account) at time  $t$  and is equal to  $(1+r)^t$
- Then,  

$$\tilde{C}_0 = \mathbf{E}^Q[\tilde{C}_1] = q\tilde{C}_u + (1-q)\tilde{C}_d$$
- This implies that the relative prices of assets have zero expected change under the probability measure  $Q$ .
- Random variables of this type are called **Martingales**

## Hedge-based Pricing

- A dual, yet equivalent, method for determining prices is by utilizing a hedging strategy
- Consider a portfolio set up at  $t = 0$  consisting of:
  - $x$  units of stock
  - $y$  units of the bank account
- At time  $t = 1$  this portfolio will be worth
  - $x S_u + y (1+r)$  in state-I
  - $x S_d + y (1+r)$  in state-II
- It is possible to choose  $x$  and  $y$  such that the pay-off of the contingent is matched exactly regardless of which state prevails at time  $t=1$

## Hedge-based Pricing

- This leads to the linear system

$$\begin{aligned} C_u &= x S_u + y (1 + r) & x &= \frac{C_u - C_d}{S_u - S_d} \\ C_d &= x S_d + y (1 + r) & \Rightarrow & \\ & & y &= \frac{1}{1 + r} \frac{C_d S_u - C_u S_d}{S_u - S_d} \end{aligned}$$

- Since the pay-off of the claim and the portfolio are identical, they must have the same price today,

$$C_0 = x S_0 + y = \frac{1}{1 + r} \{q C_u + (1 - q) C_d\}$$

## Multi-Period Binomial Model

- This model extends naturally to multiple periods
- Let  $x_0, x_1, x_2, \dots$  represent a set of identical independent random Bernoulli variables with,

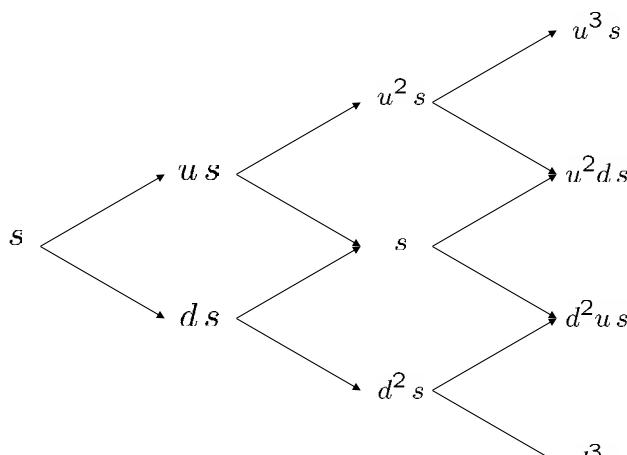
$$IP(x_t = +1) = p \quad \text{and} \quad IP(x_t = -1) = 1 - p \quad \forall t = 0, 1, \dots$$

- Then assume that the asset price dynamics satisfies,

$$S_t = S_{t-1} \exp\{\alpha x_{t-1}\}$$

- That is, the asset has a (continuously compounded) return of  $\pm \alpha$  each period.
- Such dynamics can be represented by a recombining tree as shown on the next slide

## Multi-Period Binomial Model



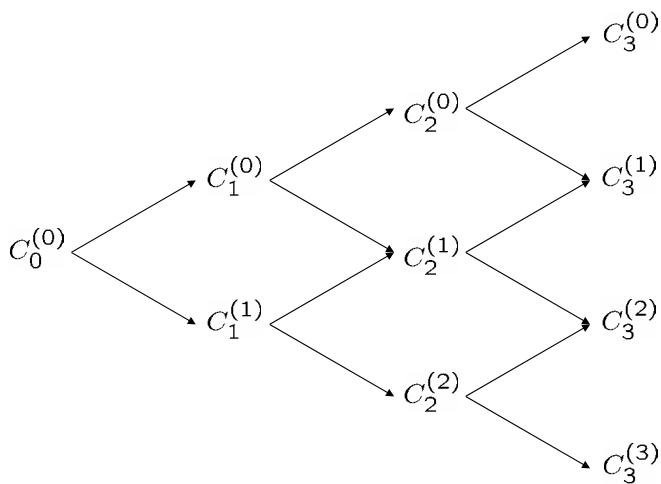
with  $u = \frac{1}{d} = \exp\{\alpha\}$

## Multi-Period Binomial Model

- Prices of European contingent claims can be obtained through backwards recursion
- The claim will define the pay-off at maturity
- Use the discounted expectation in the risk-neutral measure (the probability  $p$  is replaced by  $q$ ) to compute the prices on the nodes one-time prior
- Repeat the process until time  $t = 0$  is reached.
- Denoting the price of the claim at time  $t$  with asset level  $S_t = S_0 u^{t-2}$  by  $C_t^{(j)}$  the recursion formula can be compactly written as follows:

$$C_t^{(j)} = \frac{1}{1+r} (q C_{t+1}^{(j)} + (1-q) C_{t+1}^{(j+1)})$$

## Multi-Period Binomial Model



## Multi-Period Binomial Model

- The hedging strategy at each node can also be obtained in a similar manner

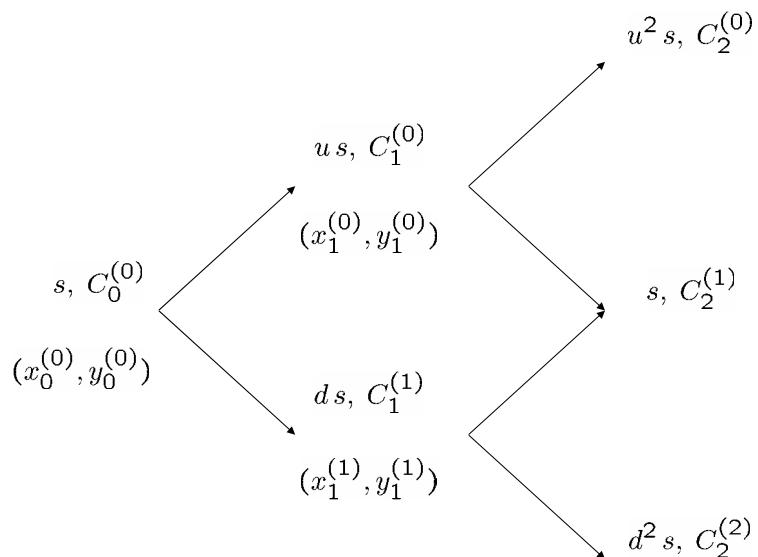
$$x_t^{(j)} = \frac{1}{S_0} \frac{C_{t+1}^{(j)} - C_{t+1}^{(j+1)}}{u^{t-2j+1} - u^{t-2j-1}}$$

$$y_t^{(j)} = \frac{1}{1+r} \frac{C_{t+1}^{(j+1)} u^{t-2j+1} - C_{t+1}^{(j)} u^{t-2j-1}}{u^{t-2j+1} - u^{t-2j-1}}$$

- Of course the relationship between the hedging parameters and the price still holds

$$C_t^{(j)} = x_t^{(j)} S_0 u^{t-2j} + y_t^{(j)}$$

## Multi-Period Binomial Model



## Multi-Period Binomial Model

- It is possible to price the claim without resorting to computing the value at every single node..
- Recall that the asset price dynamics is given by

$$S_t = S_{t-1} \exp\{\alpha x_{t-1}\}$$

- Where (in the risk neutral measure)

$$P(x_t = +1) = q \quad \text{and} \quad P(x_t = -1) = 1 - q \quad \forall t = 0, 1, \dots$$

- For a European claim the pay-off function depends only on the terminal value of the asset,  $S_T$ , but,

$$\begin{aligned} S_T &= S_0 \exp\{\alpha (x_0 + x_1 + \dots + x_{T-1})\} \\ &= S_0 \exp\{\alpha X_T\} \end{aligned}$$

- Where  $X_T = x_0 + x_1 + \dots + x_{T-1}$  and is binomial random variable of degree  $T$  and success probability  $q$ .

## Multi-Period Binomial Model

- Consequently, the price of a European contingent claim, with payoff function  $\phi(S_T)$ , in the binomial model is,

$$\begin{aligned} C_0 &= \frac{1}{(1+r)^T} \mathbf{E}^Q [\phi(S_T) | \mathcal{F}_0] \\ &= \frac{1}{(1+r)^T} \mathbf{E}^Q [\phi(S_0 e^{\alpha X_T}) | \mathcal{F}_0] \\ &= \frac{1}{(1+r)^T} \sum_{n=0}^T \binom{T}{n} q^n (1-q)^{T-n} \phi(S_0 e^{n\alpha}) \end{aligned}$$

- This is the Cox-Ross-Rubinstein (CRR) representation for the price of a European option

## Volatility matching

- The parameter  $\alpha$  can be specified through the volatility of the asset dynamics.
- In particular, the asset will be forced to have a variance of  $\sigma^2 \Delta t$  (where  $\Delta t$  is size of the time step in the tree) when  $\Delta t \ll 1$
- This leads to system of two equations (one for risk-neutrality and one for variance matching)

$$\mathbf{E}^Q[S_{t+\Delta t} | \mathcal{F}_t] = (1 + r\Delta t) S_t$$

$$\text{Var}^Q \left[ \ln \left( \frac{S_{t+\Delta t}}{S_t} \right) \middle| \mathcal{F}_t \right] = \sigma^2 \Delta t$$

- The solution when  $\Delta t \ll 1$  can be expressed as,

$$q = \frac{1}{2} \left( 1 + \frac{r - \frac{1}{2}\sigma^2}{\sigma} \sqrt{\Delta t} \right) \quad \text{and} \quad \alpha = \sigma \sqrt{\Delta t}$$

## Continuous Time Limit

- Recall that the asset dynamics was,

$$S_t = S_{t-1} \exp\{\alpha x_{t-1}\}$$

$$\Rightarrow S_t = S_0 \exp\{\alpha (x_0 + x_1 + \dots + x_{t-1})\}$$

- Letting  $\mathbf{X}_t = \mathbf{x}_0 + \mathbf{x}_1 + \dots + \mathbf{x}_{t-1}$ . Then, since  $\mathbf{x}_0, \mathbf{x}_1, \dots, \mathbf{x}_{t-1}$  are all i.i.d. Bernoulli random variables,

$$\mathbf{E}^Q[X_t | \mathcal{F}_0] = n\alpha \mathbf{E}^Q[x_0] = \left( r - \frac{1}{2}\sigma^2 \right) n \Delta t$$

$$\text{Var}^Q[X_t | \mathcal{F}_0] = n\alpha^2 \text{Var}^Q[x_0] = \sigma^2 n \Delta t$$

- Furthermore, the central limit theorem says that  $\mathbf{X}_t$  is a normal r.v.

- Finally,  $\mathbf{X}_t - \mathbf{X}_{t+s}$  (for  $s > 0$ ) has a distribution that is independent of  $t$  (it depends only on  $s$ .)

## Continuous Time Limit

- We can therefore summarize the properties of the asset dynamics as follows:
 
$$S_t = \exp\{X_t\}$$
- Where,  $X_t$  has the following characteristics:
  - Starts at 0
    - $X_0 = 0$
  - Has independent increments
    - $X_t - X_s$  is independent of  $X_v - X_u$  whenever  $(t,s) \cap (u,v) = \emptyset$
  - Has stationary increments
    - $X_t - X_{t+s} \sim N(r - \frac{1}{2}\sigma^2 s; \sigma^2 s)$
- The above properties describe a stochastic process known as **Brownian motion** (or a Weiner process.)

## Continuous Time Limit

- The price of a European contingent claim using the Brownian motion representation of the asset dynamics in the continuous time is,

$$\begin{aligned}
 C_t &= e^{-r(T-t)} \mathbf{E}^Q [\phi(S_T) | \mathcal{F}_t] \\
 &= e^{-r(T-t)} \mathbf{E}^Q [\phi(S_t e^{X_T - X_t}) | \mathcal{F}_t] \\
 &= e^{-r(T-t)} \int_{-\infty}^{\infty} \phi(S_t e^x) f(x; r, \sigma, T-t) dx
 \end{aligned}$$

- Where  $f(x; r, \sigma, T-t)$  represents the normal distribution with appropriate mean and variance,

$$f(x; r, \sigma, T-t) = \frac{\exp\left\{-\frac{1}{2} \frac{(x - (r - \frac{1}{2}\sigma^2)(T-t))^2}{\sigma^2(T-t)}\right\}}{\sqrt{2\pi\sigma^2(T-t)}}$$

## The Black-Scholes Pricing Formula

- When the pay-off function is that of a call option,  $\max(S_T - K, 0)$ , the integral can be carried out explicitly,

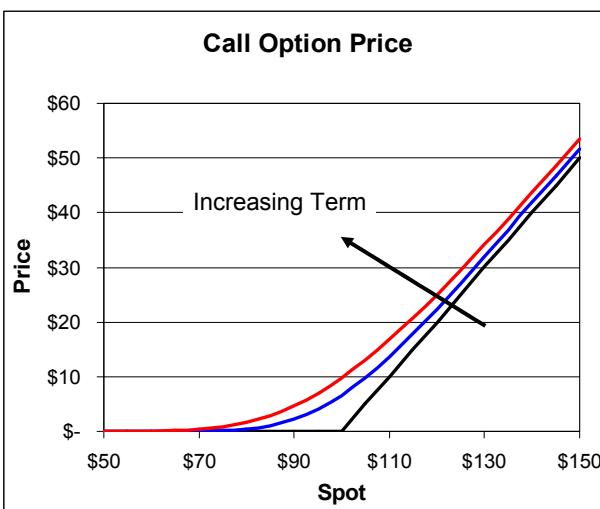
$$\begin{aligned}
 C_t(S_t; K, T) &= e^{-r(T-t)} \int_{-\infty}^{\infty} \max(S_t e^x - K, 0) f(x; r, \sigma, T-t) dx \\
 &= e^{-r(T-t)} \int_{\ln \frac{K}{S_t}}^{\infty} (S_t e^x - K) f(x; r, \sigma, T-t) dx \\
 &= S_t \Phi(d_+) - e^{-r(T-t)} K \Phi(d_-)
 \end{aligned}$$

- Where,

$$d_{\pm} = \frac{\ln \frac{S_t}{K} + (r \pm \frac{1}{2}\sigma^2)(T-t)}{\sigma\sqrt{T-t}}$$

- and  $\Phi(x)$  is the cumulative density function of a standard normal r.v.

## The Black-Scholes Pricing Formula



## The Black-Scholes Pricing Formula

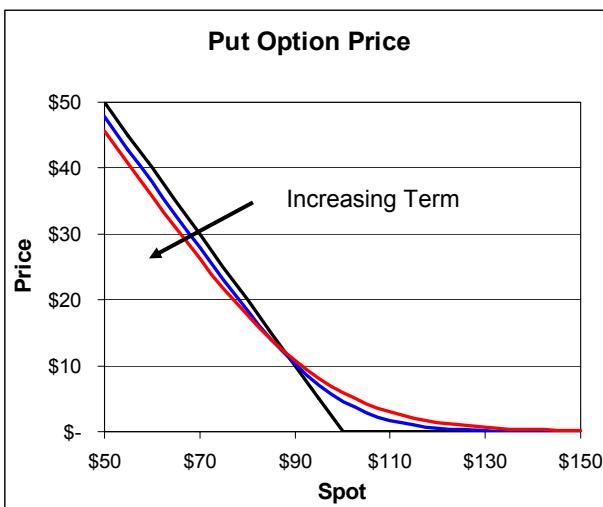
- The price of a put option can be obtained in a similar manner, or through put-call parity,

$$C_t(S_t; K, T) - P_t(S_t; K, T) = S_t - e^{-r(T-t)} K$$

- The result is,

$$P_t(S_t; K, T) = e^{-r(T-t)} K \Phi(-d_-) - S_t \Phi(-d_+)$$

## The Black-Scholes Pricing Formula



## Stochastic Integrals

- We would like to define integrals w.r.t. the stochastic variable  $\mathbf{W}(t)$  where  $\mathbf{W}(t)$  is a standard Brownian process
- Consider a simple stochastic process (piecewise constant)  $\mathbf{g}(s)$  with jump points at  $a < t_0 < t_1 < \dots < t_n < b$
- The integral of such a process w.r.t to  $\mathbf{W}(t)$  can be represented as a finite sum:

$$\int_a^b g(s) dW(s) = \sum_{k=0}^{n-1} g(t_k) [W(t_{k+1}) - W(t_k)]$$

- For a general non-simple process  $\mathbf{h}(s)$  take an approximating sequence of simple processes  $\mathbf{h}_1(s), \mathbf{h}_2(s), \dots$  s.t.,

$$\int_a^b \mathbf{E} [(h_n(s) - h(s))^2] ds \rightarrow 0$$

## Stochastic Integrals

- It is possible to prove that

$$\int_a^b h_n(s) dW(s) \rightarrow Z \quad \text{in } \mathcal{L}^2$$

- Where  $Z$  is some r.v. Then define the stochastic integral as follows:

$$\int_a^b h(s) dW(s) = \lim_{n \rightarrow \infty} \int_a^b h_n(s) dW(s)$$

- Stochastic integrals are often written in “differential form”

$$X(b) - X(a) = \int_a^b h(s) dW(s) \quad \leftrightarrow \quad dX(s) = h(s) dW(s)$$

## Stochastic Integrals

- General diffusion processes are often defined through stochastic differential equations:

$$\begin{aligned} dX(s) &= \mu(s, X(s))ds + \sigma(s, X(s))dW(s) \\ X(0) &= x \end{aligned}$$

- The integral representation is:

$$X(t) = x + \int_0^t \mu(s, X(s))ds + \int_0^t \sigma(s, X(s))dW(s)$$

- Ito's lemma tells one how the SDE changes under a transformation:

$$\begin{aligned} df(x_t, t) &= \left[ \left( \frac{\partial}{\partial t} + \mu(x_t, t) \frac{\partial}{\partial x_t} + \frac{1}{2} \sigma^2(x_t, t) \frac{\partial^2}{\partial x_t^2} \right) f(x_t, t) \right] dt \\ &\quad + \left[ \sigma(x_t, t) \frac{\partial}{\partial x_t} f(x_t, t) \right] dW_t \end{aligned}$$

## Black-Scholes Differential Equation

- Suppose that an investor follows a self-financing strategy which consists of
  - $\Delta(S_t, t)$  units of the asset
  - $\theta_t(S_t, t)$  units in the money-market account;
  - 1 units in the option in the interval  $(t, t+dt]$ .
- Let  $V_t(S_t)$  denote the value-process for such a strategy. Then,

$$V(S_t, t) = \Delta(S_t, t)S_t + \theta(S_t, t)M_t - P(S_t, t)$$

- Choose  $\Delta(S_0, 0)$  and  $\theta(S_0, 0)$  such that  $V(S_0, 0) = 0$

$$\begin{aligned} dV(S_t, t) &= d(\Delta(S_t, t)S_t + \theta(S_t, t)M_t) - dP(S_t, t) \\ &= \Delta(S_t, t)dS_t + \theta(S_t, t)dM_t - dP(S_t, t) \end{aligned}$$

## Black-Scholes Differential Equation

- Using Ito's Lemma one finds,

$$dV(S_t, t) = \left[ \mu S_t \Delta + r M_t \theta - \left( \frac{\partial}{\partial t} + \mu S_t \frac{\partial}{\partial S_t} + \frac{1}{2} \sigma^2 S_t^2 \frac{\partial^2}{\partial S_t^2} \right) P \right] dt + \left[ \sigma S_t \Delta - \sigma S_t \frac{\partial}{\partial S_t} P \right] dW_t$$

- By choosing  $\Delta(S_t, t)$  appropriately, the stochastic term can be removed,

$$\Delta(S_t, t) = \frac{\partial}{\partial S_t} P(S_t, t)$$

- Since the return is non-stochastic, to avoid arbitrage it must grow at the risk-free rate, which leads to the Black-Scholes PDE:

$$\left( \frac{\partial}{\partial t} + r S_t \frac{\partial}{\partial S_t} + \frac{\sigma^2}{2} S_t^2 \frac{\partial^2}{\partial S_t^2} \right) P(S_t, t) = r P(S_t, t)$$