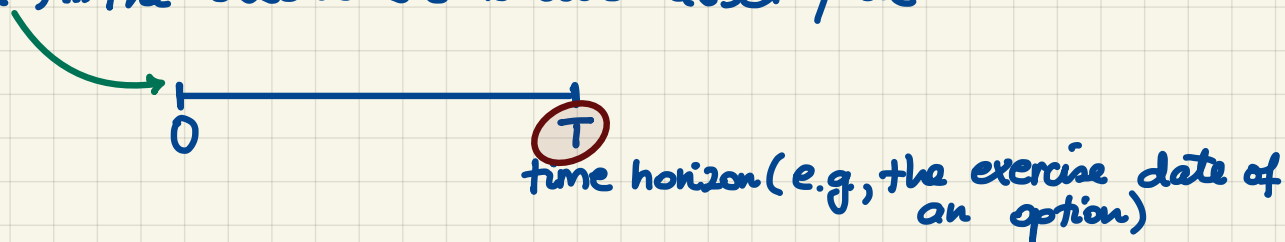


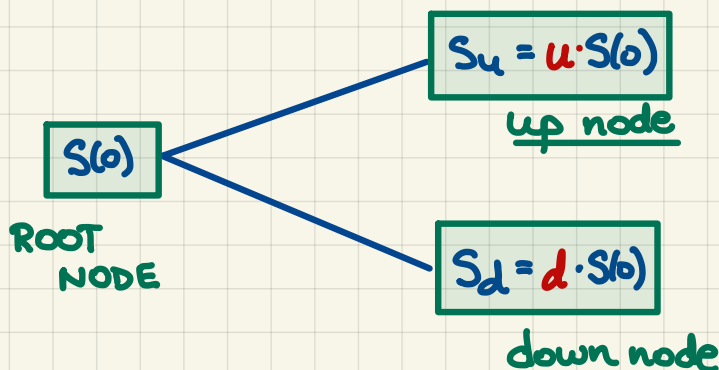
M339D: March 3rd, 2025.

The Binomial Asset Pricing Model.

$S(0)$... the observable initial asset price



time horizon (e.g., the exercise date of an option)



u ... up factor
 d ... down factor

By convention:

$$u > d$$

h ... length of a single period

one period:

$S(T) = S(h)$... a random variable denoting the time T stock price w/ two possible values S_u and S_d

As a random variable: SIMPLE RETURN

$$\frac{S(T) - S(0)}{S(0)} \quad \leftarrow$$

up node: $\frac{S_u - S(0)}{S(0)} = \frac{S_u}{S(0)} - 1 = u - 1$

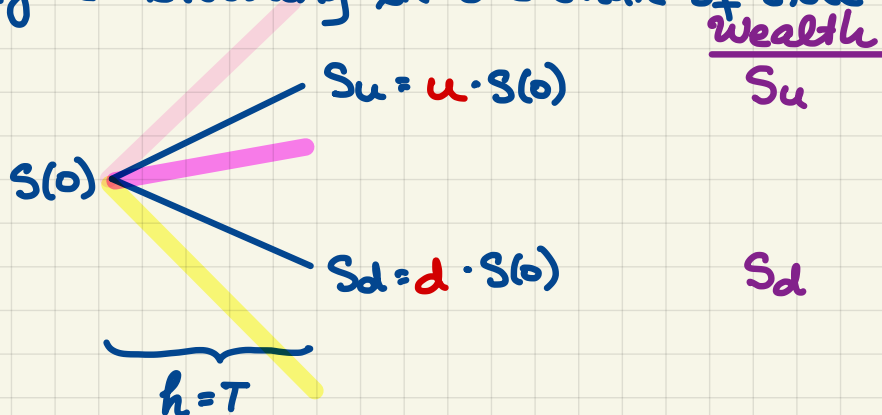
down node: $\frac{S_d - S(0)}{S(0)} = \frac{S_d}{S(0)} - 1 = d - 1$

No-Arbitrage Condition.

Market Model.

- riskless asset: @ the ccrfir (r)
- risky asset: non-dividend-paying stock

Imagine investing in one share of stock @ time 0:



At the risk-free rate, the amount $S(0)$ accumulates to $S(0)e^{rh}$ in the same time period.

The No-Arbitrage Condition.

$$S_d < S(0)e^{rh} < S_u$$
$$\cancel{d \cdot S(0)} < \cancel{S(0)e^{rh}} < \cancel{u \cdot S(0)}$$
$$\boxed{d < e^{rh} < u}$$

Half-a-Proof.

Say, to the contrary, $\boxed{e^{rh} \leq d < u}$

Propose. Long one share of stock.

Verify. Profit = Payoff - $FV_{0,T}$ (Initial Cost)
 $= S(h) - S(0)e^{rh}$

down node: $S_d - S(0)e^{rh} = d \cdot S(0) - S(0)e^{rh} = S(0)(d - e^{rh}) \geq 0$

up node: $S_u - S(0)e^{rh} = u \cdot S(0) - S(0)e^{rh} = S(0)(u - e^{rh}) > 0$

Indeed, this is an arbitrage portfolio.

Forward Binomial Tree.

"Def'n". The **volatility** σ is the standard deviation of realized returns on a continuously compounded scale and annualized.

Realized Return.

$$\begin{array}{c} | \qquad \qquad | \\ t \qquad \qquad t+s \end{array}$$

$$R(t, t+s) := \ln \left(\frac{S(t+s)}{S(t)} \right)$$

It satisfies:

$$S(t+s) = S(t)e^{R(t, t+s)}$$

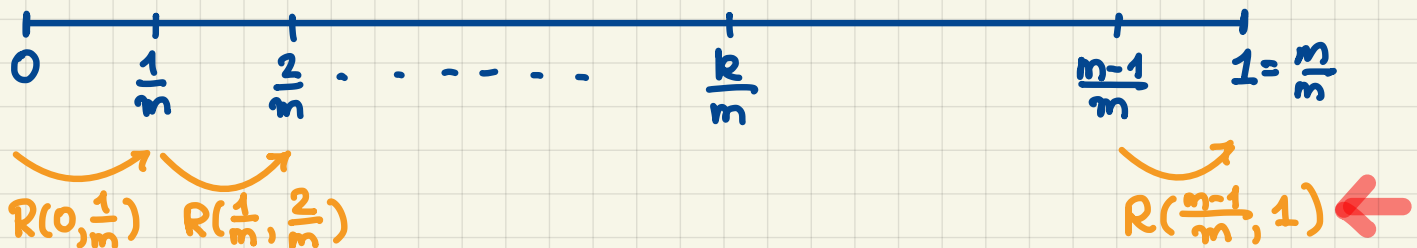
Heuristics.

$$T = 1$$

$$h_m = \frac{1}{m} \text{ (of a year)}$$

Q: What is the volatility for the time period of length h_m ?

$$\sigma_{h_m} = ?$$

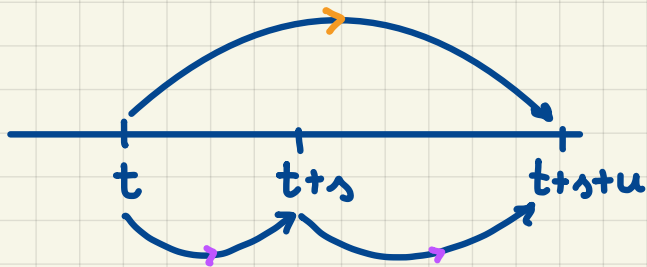


Note:

$R(\frac{k-1}{m}, \frac{k}{m})$ for $k=1 \dots m$ are all RANDOM VARIABLES.

We make the following assumptions:

- the returns over disjoint intervals are **independent**,
- all the returns over the same length intervals are **identically dist'd**.



$$\begin{aligned}
 R(t, t+s+u) &= \ln \left(\frac{S(t+s+u)}{S(t)} \right) \\
 &= \ln \left(\frac{S(t+s)}{S(t)} \cdot \frac{S(t+s+u)}{S(t+s)} \right) \\
 &= \boxed{\ln \left(\frac{S(t+s)}{S(t)} \right)} + \boxed{\ln \left(\frac{S(t+s+u)}{S(t+s)} \right)} \\
 &= R(t, t+s) + R(t+s, t+s+u)
 \end{aligned}$$

Hence, realized returns are additive.

$$\rightarrow R(0, \frac{1}{m}) + R(\frac{1}{m}, \frac{2}{m}) + \dots + R(\frac{m-1}{m}, 1) = R(0, 1)$$

Q: $\text{Var}[R(0, 1)] = \underline{\sigma^2}$

$$\begin{aligned}
 \Rightarrow \underline{\sigma^2} &= \text{Var} \left[R(0, \frac{1}{m}) + \dots + R(\frac{m-1}{m}, 1) \right] = \text{(independence)} \\
 &= \text{Var} \left[R(0, \frac{1}{m}) \right] + \dots + \text{Var} \left[R(\frac{m-1}{m}, 1) \right] = \text{(identically dist'd)} \\
 &= m \cdot \text{Var} \left[R(0, \frac{1}{m}) \right] = m \cdot \underline{\sigma_{h_m}^2}
 \end{aligned}$$

$$\sigma_{h_m}^2 = \frac{1}{m} \sigma^2$$

$$\sigma_{h_m} = \sigma \sqrt{\frac{1}{m}} = \sigma \sqrt{h_m}$$

We generalize this identity to arbitrary lengths h :

$$\sigma_h = \sigma \sqrt{h}$$