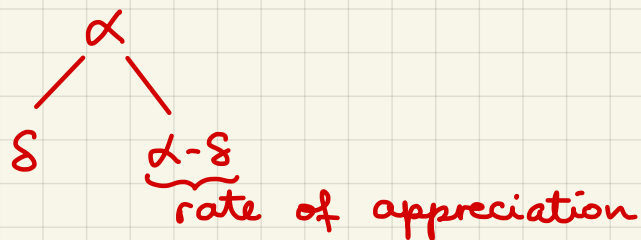


M339W: February 10<sup>th</sup>, 2021.

Review:

Parameters . •  $\alpha$  ... mean rate of return (per annum)  
satisfies  $E[S(\tau)] = S(0)e^{(\alpha - \delta) \cdot \tau}$   
w/  $\delta$  ... dividend yield

Note:



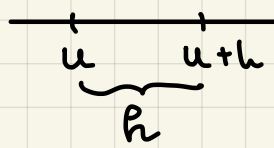
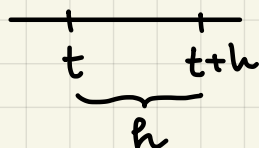
•  $\sigma$  ... volatility,  
i.e., the standard deviation of realized  
returns on an annual basis

Realized returns. For any  $t, h > 0$ , we set

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

Modeling Assumptions. In continuous time:

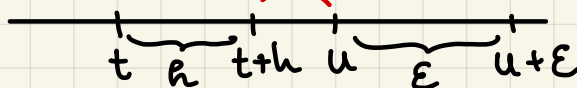
i.



We require that  $R(t, t+h)$  and  $R(u, u+h)$  be  
identically distributed.

ii.

These can be equal!



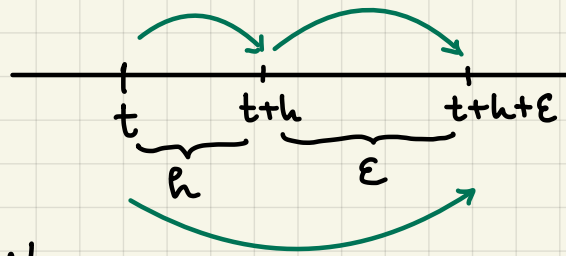
$t, h, u, \epsilon > 0$

We require that  $R(t, t+h)$  and  $R(u, u+\epsilon)$  be  
independent.

We want  
them  
to be  
inherited  
from the  
binomial  
tree.

Property. iii.

$t, h, \epsilon > 0$



By def'n:

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

$$R(t, t+h+\epsilon) = R(t, t+h) + R(t+h, t+h+\epsilon)$$

Realized Returns are ADDITIVE.

Q: Which probabilistic model would you propose for our realized returns?

→: Think back to our discussion of what happens in the limit w/ a binomial tree 😊

⇒ We decide to model our realized returns

$R(t, t+h)$  as normally distributed,

i.e.,

$$R(t, t+h) \sim \text{Normal}(\text{mean} = \underline{m}, \text{variance} = \underline{\sigma^2})$$

Note: By def'n:

$$\underline{S(t+h) = S(t)e^{R(t, t+h)}}$$

## Moment Generating Functions.

For any random variable  $Y$ ,  
and for independent arguments denoted by  $t$ ,  
we define the **moment generating f'tion** of  $Y$  as the  
following function of  $t$ :

$$M_Y(t) := \mathbb{E}[e^{Y \cdot t}]$$

for all  $t$  such that the  
expectation exists, i.e., when it  
is finite

Note: •  $M_Y(0) = 1$

$\Rightarrow$  @ least  $t=0$  is in the domain of  $M_Y$

Goal: Understanding  $e^X$  where  $X \sim \text{Normal}(\text{mean}=m, \text{var}=\sigma^2)$ .

$\rightarrow$ : Recall the standard normal:  $Z \sim N(0,1)$   
Then, our  $X$  can be expressed as:

$$X = m + \sigma \cdot Z$$

In general: Take constants  $a$  and  $b$ ;  
define  $\tilde{Y} = a \cdot Y + b$  w/  $Y$  any r.v.

By def'n:  $M_{\tilde{Y}}(t) = \mathbb{E}[e^{t \cdot \tilde{Y}}]$

$$\begin{aligned} &= \mathbb{E}[e^{t(a \cdot Y + b)}] \\ &= \mathbb{E}[e^{t \cdot a \cdot Y} \cdot e^{t \cdot b}] \\ &= e^{t \cdot b} \cdot \mathbb{E}[e^{t \cdot a \cdot Y}] \\ &= e^{t \cdot b} \cdot M_Y(a \cdot t) \end{aligned}$$

In particular: Let  $X \sim \text{Normal}(\text{mean}=m, \text{var}=\sigma^2)$ .

$$\Rightarrow M_X(t) = e^{m \cdot t} \cdot M_Z(\sigma \cdot t) = e^{m \cdot t} \cdot e^{\frac{\sigma^2 \cdot t^2}{2}}$$

Recall:  $M_Z(s) = e^{\frac{s^2}{2}}$

$$\Rightarrow M_X(t) = e^{m \cdot t + \frac{\sigma^2 \cdot t^2}{2}}$$