

M339 W: September 13th, 2021: Part II

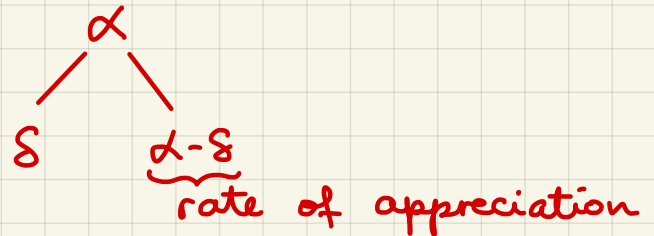
Review:

Parameters . • α ... mean rate of return (per annum)

satisfies $E[S(\tau)] = S(0) e^{(\alpha - \delta) \cdot \tau}$

w/ δ ... dividend yield

Note:



• σ ... volatility,

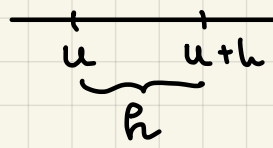
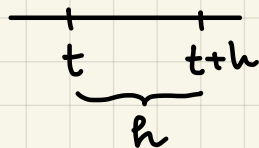
i.e., the standard deviation of realized returns on an annual basis

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:

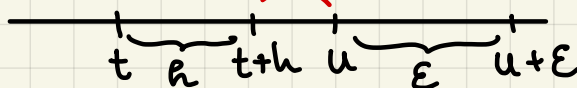
i.



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

ii.

These can be equal!



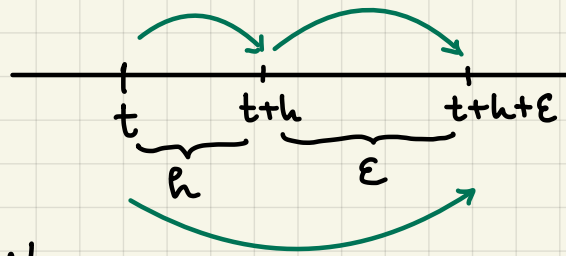
$$t, h, u, \epsilon > 0$$

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

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