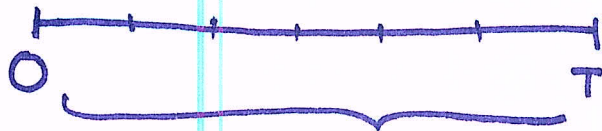


Realized Returns.

W: 02/08/2019

Recall:



$n \dots \# \text{ of periods}$

\Rightarrow length of every period is $h_n = \frac{T}{n}$

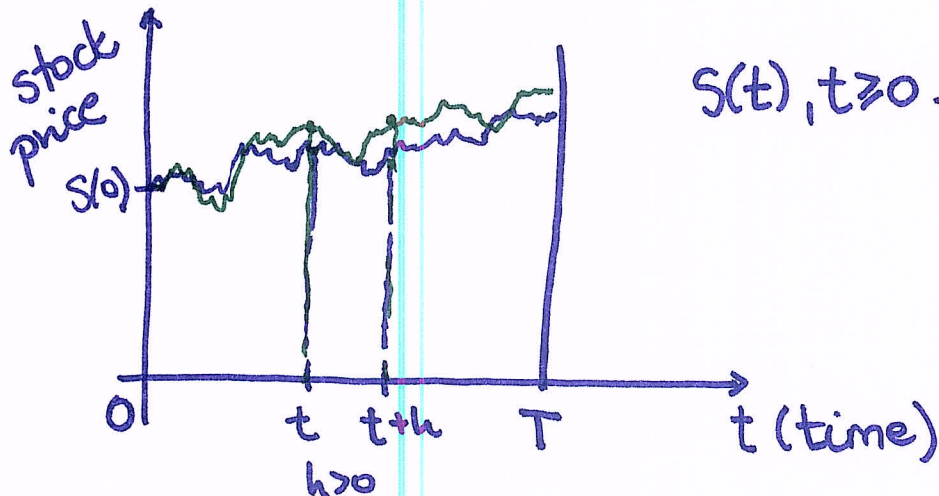
$\Rightarrow u_n, d_n, p_n^*$ dependent on (n)

Facts about our model:

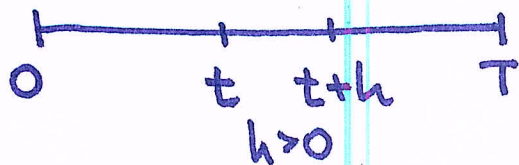
- \rightarrow realized returns are independent between periods;
- \rightarrow realized returns are identically distributed over different periods of the same length.

We want to inherit these properties in a continuous time model.

In continuous time:



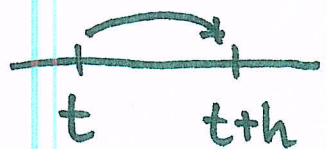
$S(t), t \geq 0 \dots$ time t stock price



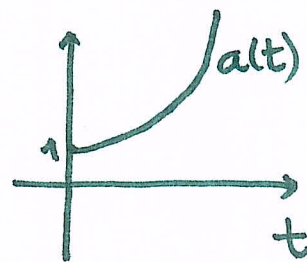
For every t, h , we define the **realized return** as

$$\underbrace{R(t, t+h)}_{\substack{\uparrow \\ \text{random variable}}} := \ln \left(\frac{S(t+h)}{S(t)} \right)$$

Think back to the accumulation function and the c.c.r.f.i.r. (r) :



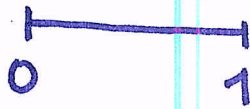
$$\begin{aligned} & \underbrace{a(t)} \quad a(t+h) \\ & a(t+h) = a(t) \cdot e^{r \cdot h} \\ & \underline{r \cdot h} = \ln \left(\frac{a(t+h)}{a(t)} \right) \end{aligned}$$



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

Recall: σ ... volatility parameter;
the standard deviation of the
realized return over a time period of
length one year

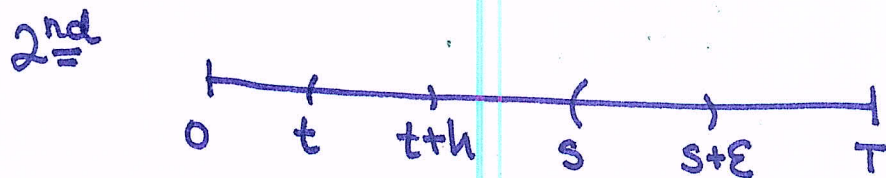
We should have



$$\text{Var}[R(0,1)] = \sigma^2, \text{ i.e., } \text{SD}[R(0,1)] = \sigma$$



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

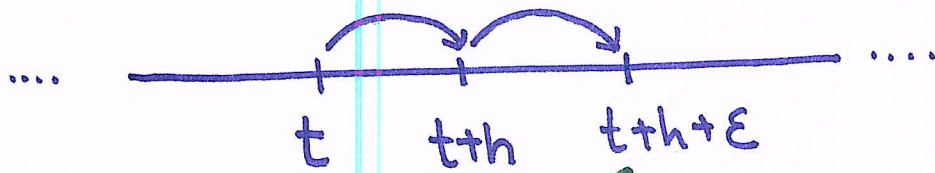


it's allowed to have $t+h = s$
(the intervals can touch @ the boundary)

$R(t, t+h)$ and $R(s, s+E)$ are
required to be independent.

These are the requirements we place on our
model which are "inherited" from the discrete
model

3rd Just from the def'n of realized returns, we have:



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

by def'n

$$= \ln \left(\frac{S(t+h+\epsilon)}{S(t+h)} \cdot \frac{S(t+h)}{S(t)} \right)$$

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

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

Realized returns are additive. ■

We decide to model

$R(t, t+h)$ as normally distributed
for every choice of t, h .

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

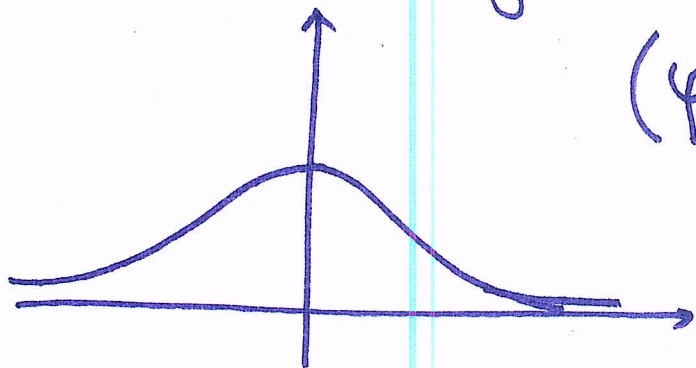
Normal Distribution.

* Standard Normal random variable:

$$Z \sim N(\text{mean} = 0, \text{sd} = 1)$$

Its dist'n is given by its density (pdf):

$$\begin{aligned} (\varphi(z)) &= \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}, \quad z \in \mathbb{R} \\ &=: f_Z(z) \end{aligned}$$



\Rightarrow Its cumulative dist'n f'n (cdf):

$$N(a) = \mathbb{P}_a[Z \leq a]$$

$$= \int_{-\infty}^a f_Z(z) dz = \dots \text{we find in the std normal tables or using an online calculator}$$

$$f_Z(z) = N'(z)$$

* Normal random variable:

$$\boxed{X \sim N(\text{mean} = m, \text{sd} = \tau)}$$

is given through its relationship w/ the $\boxed{Z \sim N(0,1)}$

Since any normal random variable is a linear transform of the std normal, we get

$$X^{(d)} = m + \tau \cdot Z \quad \Leftrightarrow \quad \frac{X - m}{\tau} \stackrel{(d)}{=} Z$$

Moment Generating Function

For any random variable Y , its moment generating f'tion is:

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

wherever it's
well-defined