

$P_t$  price of an asset at time  $t$

$R_t$  net returns

$$R_t = \frac{P_t}{P_{t-1}} - 1 = \frac{P_t - P_{t-1}}{P_{t-1}}$$

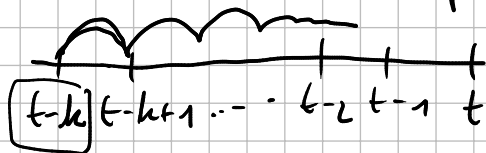
$P$  uppercase

$1+R_t$  gross return

$$1+R_t = \frac{P_t}{P_{t-1}}$$

$1+R_t(k)$  gross return over the most recent  $k$  periods

↑ ending period  
↑ holding period



$$1+R_t(k) = (1+R_t) \cdot (1+R_{t-1}) \cdot \dots \cdot (1+R_{t-k+1})$$

$r_t$  log returns

$$r_t = \log(1+R_t)$$

log of gross returns

$$\log(1+x) \approx x \quad x \text{ is small enough}$$

$$\log(1+R_t) \approx R_t \quad \text{if } R_t \text{ is small enough}$$

$$r_t \leq R_t$$

$$r_t = \log(1+R_t) = \log\left(\frac{P_t}{P_{t-1}}\right) = \log P_t - \log P_{t-1}$$

$P_t$   $P_{t-1}$  not used often in the book

$$r_t(k) = r_t + r_{t-1} + \dots + r_{t-k+1}$$

log returns over  $k$  periods = sum of log returns each period.

- No statistical assumptions made.
- No dividends. (nonstatistical assumption)

### Random walk model

$r_t$ 's are independent random variables  
log returns

The r.v.'s  $X$  &  $Y$  are independent if and only if

$$f_{X,Y}(x,y) = f_X(x) f_Y(y) \quad (\text{continuous case})$$

$$P(X=x \& Y=y) = P(X=x) P(Y=y) \quad (\text{discrete case})$$

joint distribution.

↑ involves

r.v.  $X$  only

↑ involves r.v.  $Y$  only

(marginal dist of  $Y$ )

(marginal dist of  $X$ )

another way of seeing this

discrete case:

$$P(Y=y | X=x) = P(Y=y)$$

$$P(X=x | Y=y) = P(X=x)$$

↑ conditional distribution

$$P(Y=y | X=x) > P(Y=y)$$

some dependence between  $X$  &  $Y$ .

\*  $r_t(k) = r_t + r_{t-1} + \dots + r_{t-k+1} \sim \text{dist? } N(\underline{\quad}, \underline{\quad})$

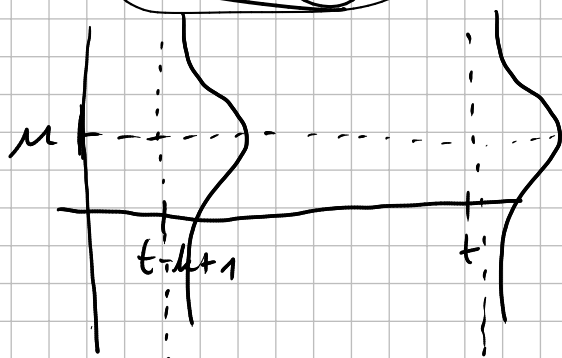
$r_t$ 's independent of each other  
share a common dist

$$N(\mu, \sigma^2)$$

$$E(r_j) = \mu$$

$j = t-k+1, \dots, t$

$$\text{Var}(r_j) = \sigma^2$$



WARNING If  $X, Y$  have the same dist, it does not automatically mean that  $X+Y$  will have the same shape of the dist of  $X$  &  $Y$ .

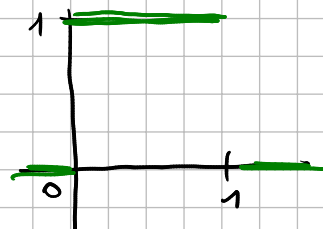
$$X \sim U(0,1)$$

$$Y \sim U(0,1)$$

$X+Y \sim ?$

pdf (prob. density function)

$$f(x) = \begin{cases} 1 & \text{if } 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$



$$r_t(k) = r_t + \dots + r_{t-k+1} \sim N\left(\frac{k\mu}{\uparrow}, \frac{k\sigma^2}{\times}\right)$$

$\mathbb{E}(r_t(k))$        $\text{Var}(r_t(k))$

$$\mathbb{E}(r_t(k)) = \mathbb{E}(r_t + r_{t-1} + \dots + r_{t-k+1})$$

$r_j \sim N(\mu, \sigma^2)$   
 $j = t-k+1, \dots, t$

expected value of a sum is the sum of the expected values.

$$\begin{aligned} &= \mathbb{E}(r_t) + \mathbb{E}(r_{t-1}) + \dots + \mathbb{E}(r_{t-k+1}) \\ &= \underbrace{\mu + \mu + \dots + \mu}_{k \text{ of them.}} \\ &= k\mu \end{aligned}$$

$$\text{Var}(r_t(k)) = \text{Var}(r_t + r_{t-1} + \dots + r_{t-k+1})$$

variance of a sum is the sum of variances

(if) the r.v.'s are independent of each other

$$\begin{aligned} &= \text{Var}(r_t) + \text{Var}(r_{t-1}) + \dots + \text{Var}(r_{t-k+1}) \\ &\quad \text{share a common dist} \\ &\stackrel{\downarrow}{=} \underbrace{\sigma^2 + \sigma^2 + \dots + \sigma^2}_{k \text{ of them}} \\ &= k\sigma^2 \end{aligned}$$

CONCLUSION :

$$r_t(k) = r_t + r_{t-1} + \dots + r_{t-k+1} \sim N(k\mu, k\sigma^2)$$

under the random walk hypothesis  
+ common  $N(\mu, \sigma^2)$  return distribution

NOTE  $X, Y$  r.v.'s (not necessarily independent)

$$\text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y) + 2\text{Cov}(X, Y)$$

$t=0$      $S_0$

$Z_1, Z_2, \dots \sim \text{i.i.d. } (N(\mu, \sigma^2))$

$t=1$      $S_1 = S_0 + Z_1$

(noise / shock)

$t=2$      $S_2 = S_1 + Z_2 = S_0 + Z_1 + Z_2$

$\vdots$

$S_t = S_0 + Z_1 + \dots + Z_t$

conditional expectation / CEF

$$\mathbb{E}(S_t | S_0) = S_0 + \mu t$$

cond. variance

$$\text{Var}(S_t | S_0) = \sigma^2 t$$

↓

$$\text{Var}(S_1 | S_0) = \sigma^2$$

$$\text{Var}(S_2 | S_0) = 2\sigma^2$$

$$S_t = S_0 + Z_1 + Z_2 + \dots + Z_t$$

$$\mathbb{E}(S_t | S_0)$$

↑ treated as constant

$$= \mathbb{E}(S_0 + Z_1 + \dots + Z_t | S_0)$$

$$= \mathbb{E}(S_0 | S_0) + \mathbb{E}(Z_1 + \dots + Z_t | S_0)$$

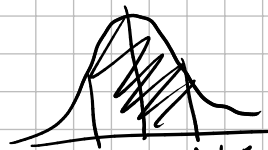
steps  
one increment  
of the starting  
point

$$= S_0 + \mathbb{E}(Z_1 + \dots + Z_t)$$

$$= S_0 + \underbrace{\mu + \mu + \dots + \mu}_t \text{ of them}$$

$$= S_0 + \mu t$$

Empirical rule for normal dist.



$$X \sim N(\mu, \sigma^2)$$

$$X \sim N(\mu, \sigma^2) \Rightarrow P(\mu - \sigma \leq X \leq \mu + \sigma) \approx 68\%$$

$$P(\mu - 2\sigma \leq X \leq \mu + 2\sigma) \approx 95\%$$

$$P(\mu - 3\sigma \leq X \leq \mu + 3\sigma) \approx 99\%$$

normal random walk

$$S_t | S_0 \sim N(S_0 + \mu t, \sigma^2 t) \Rightarrow P(S_0 + \mu t - \sigma\sqrt{t} \leq S_t \leq S_0 + \mu t + \sigma\sqrt{t})$$

$$\approx 68\%$$

⋮

(2.7)

$$P_t = P_0 \exp(r_t + r_{t-1} + \dots + r_1)$$

$$\underbrace{\log P_t}_{S_t} = \underbrace{\log P_0}_{S_0} + \underbrace{r_t + r_{t-1} + \dots + r_1}_{Z_t + Z_{t-1} + \dots + Z_1} \quad (2.5)$$

Ex 4, 5, 6 only use Section 2.1

Exercise 1

\$1000  
wealth at end of 1 trading day

$$1000(1 + R_t)$$

gross return

$$P(1000(1 + R_t) < 990)$$

$$1000(1+R_t) < 990$$

$$\Rightarrow 1+R_t < 0.99$$

$$\Rightarrow \underbrace{\log(1+R_t)}_{r_t} < \log(0.99)$$

$$P(r_t < \log(0.99))$$