

## COURSEWORK 1

IMPERIAL COLLEGE LONDON

DEPARTMENT OF COMPUTING

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# Computational Finance

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## 1 Exercise 1. Dividend Discount Model

a) The price of stock is given by:

$$\begin{aligned} PV &= \sum_{t=1}^{\infty} \frac{D_t}{(1+r)^{t-1}} \\ &= \sum_{t=1}^{\infty} \frac{D_1(1+g)^{t-1}}{(1+r)^{t-1}} \\ &= \sum_{t=0}^{\infty} \frac{D_1(1+g)^t}{(1+r)^t} \\ &= D_1 \left( \frac{1}{1 - \frac{(1+g)}{(1+r)}} \right) \\ &= D_1 \left( \frac{(1+r)}{(1+r) - (1+g)} \right) \\ &= D_1 \left( \frac{1+r}{r-g} \right) \end{aligned} \tag{1}$$

b) The price of stock is given by:

$$\begin{aligned} PV &= \sum_{t=1}^{\infty} \frac{D_t}{(1+r)^{t-1}} \\ &= \sum_{t=1}^{\infty} \frac{D_1 + (t-1)I}{(1+r)^{t-1}} \\ &= \sum_{t=0}^{\infty} \frac{D_1 + tI}{(1+r)^t} \\ &= \sum_{t=0}^{\infty} \frac{D_1}{(1+r)^t} + \sum_{t=0}^{\infty} \frac{tI}{(1+r)^t} \\ &= \sum_{t=0}^{\infty} \frac{D_1}{(1+r)^t} + \sum_{t=0}^{\infty} tI(1+r)^{-t} \\ &= \sum_{t=0}^{\infty} \frac{D_1}{(1+r)^t} - \sum_{t=0}^{\infty} I(1+r) \frac{d}{d(1+r)} (1+r)^{-t} \\ &= \sum_{t=0}^{\infty} \frac{D_1}{(1+r)^t} - I(1+r) \frac{d}{d(1+r)} \left( \sum_{t=0}^{\infty} (1+r)^{-t} \right) \\ &= \frac{D_1(r+1)}{r} - I(1+r) \frac{d}{d(1+r)} \left( \frac{r+1}{r} \right) \\ &= \frac{D_1(r+1)}{r} + \frac{I(r+1)}{r^2} \end{aligned} \tag{2}$$

## 2 Exercise 2. Fixed-Income Securities

a) The price of this bond is given by its present value:

$$P = \frac{F}{[1 + s_n]^n} + \sum_{k=1}^n \frac{C}{[1 + s_k]^k} \quad (3)$$

For  $F = 100$ ,  $C = 10$ ,  $n = 2$ , and the given spot rates, the price,  $P$  of the bond is:

$$\begin{aligned} P &= \frac{100}{[1 + s_2]^2} + \frac{10}{[1 + s_1]} + \frac{10}{[1 + s_2]^2} \\ P &= 107.5 \end{aligned} \quad (4)$$

b) The yield of the bond is such that:

$$P = \frac{F}{[1 + (\frac{\lambda}{m})]^n} + \sum_{k=1}^n \frac{\frac{C}{m}}{[1 + (\frac{\lambda}{m})]^k} \quad (5)$$

Specifically for this bond:

$$\begin{aligned} P &= \frac{F}{(1 + \lambda)^2} + \frac{C}{(1 + \lambda)^2} + \frac{C}{(1 + \lambda)} \\ P(1 + \lambda)^2 &= F + C + C + C\lambda \\ P\lambda^2 + (2P - C)\lambda + (P - F - 2C) &= 0 \\ \lambda &= \frac{-(2P - C) \pm \sqrt{(2P - C)^2 - 4P(P - F - 2C)}}{2P} \end{aligned} \quad (6)$$

Which when evaluated is:

$$\begin{aligned} \lambda &= 0.059 \\ &= 5.9\% \end{aligned} \quad (7)$$

c) The bond's Macauley duration is given by:

$$D = \frac{1 + \frac{\lambda}{m}}{\lambda} - \frac{1 + \frac{\lambda}{m} + n(\frac{c}{m} - \frac{\lambda}{m})}{c[(1 + \frac{\lambda}{m})^n - 1] + \lambda} \quad (8)$$

$$D = \frac{1 + \lambda}{\lambda} - \frac{1 + \lambda + n(c - \lambda)}{c[(1 + \lambda)^n - 1] + \lambda} \quad (9)$$

$$D = \frac{1 + \lambda}{\lambda} - \frac{1 + \lambda + n(c - \lambda)}{c[(1 + \lambda)^n - 1] + \lambda} \quad (10)$$

Using the result for  $\lambda$ , and  $c = 0.1$ :

$$D = 1.91 \quad (11)$$

- d) The price of the bond next year can be calculated by first calculating the forward rates from year 1:

$$f_{i,j} = \left[ \frac{(1 + s_j)^j}{(1 + s_i)^i} \right]^{\frac{1}{j-i}} - 1 \quad (12)$$

$$\begin{aligned} f_{1,2} &= \left[ \frac{(1 + s_2)^2}{(1 + s_1)} \right] - 1 \\ &= 0.08 \\ &= 8\% \end{aligned} \quad (13)$$

$$\begin{aligned} f_{1,3} &= \left[ \frac{(1 + s_3)^3}{(1 + s_1)} \right]^{\frac{1}{2}} - 1 \\ &= 0.10 \\ &= 1\% \end{aligned} \quad (14)$$

$$P = \frac{100}{[1 + f_{1,3}]^2} + \frac{10}{[1 + f_{1,2}]} + \frac{10}{[1 + f_{1,3}]^2} \quad (15)$$

$$P = 100.1 \quad (16)$$

### 3 Exercise 3. Portfolio Optimisation

- a) The expected return,  $\bar{r}_P$ , and the variance,  $\bar{\sigma}_P^2$  is given as follows:

$$\begin{aligned} \bar{r}_P &= \mathbb{E}[\mathbf{r}_P] \\ &= \mathbb{E} \left[ w_0 r_f + \sum_{i=1}^n w_i r_i \right] \\ &= w_0 r_f + \sum_{i=1}^n w_i \bar{r}_i \\ &= w_0 r_f + \mathbf{w}^T \bar{\mathbf{r}} \end{aligned} \quad (17)$$

$$\begin{aligned} \bar{\sigma}_P^2 &= \text{var}(r_P) \\ &= \mathbb{E} \left[ \left( \sum_{i=1}^n w_i r_i - \sum_{i=1}^n w_i \bar{r}_i \right)^2 \right] \\ &= \mathbb{E} \left[ \left( \sum_{i=1}^n w_i (r_i - \bar{r}_i) \right) \left( \sum_{j=1}^n w_j (r_j - \bar{r}_j) \right) \right] \\ &= \mathbb{E} \left[ \left( \sum_{i=1}^n w_i (r_i - \bar{r}_i) \right) \left( \sum_{j=1}^n w_j (r_j - \bar{r}_j) \right) \right] \end{aligned}$$

$$\begin{aligned}
 &= \mathbb{E} \left[ \sum_{i,j=1}^n w_i w_j (r_i - \bar{r}_i)(r_j - \bar{r}_j) \right] \\
 &= \sum_{i,j=1}^n w_i \sigma_{ij} w_j \\
 &= \mathbf{w}^T \Sigma \mathbf{w}
 \end{aligned} \tag{18}$$

b) To maximise  $\bar{r}_P - \frac{a}{2} \sigma_P^2$ , the problem can be formulated into a quadratic program:

$$\begin{aligned}
 &\text{Maximise : } \bar{r}_P - \frac{a}{2} \sigma_P^2 \\
 &\text{Where : } w_0 r_f + \mathbf{w}^T \bar{\mathbf{r}} = \bar{r}_P \\
 &\quad w_0 + \mathbf{w}^T \mathbf{e} = 1
 \end{aligned}$$

$$\begin{aligned}
 &\text{Maximise : } f(\mathbf{w}) = -\frac{a}{2} \mathbf{w}^T \Sigma \mathbf{w} + \mathbf{w}^T \bar{\mathbf{r}} + (1 - \mathbf{w}^T \mathbf{e}) r_f \\
 &\quad f(\mathbf{w}) = -\frac{a}{2} \mathbf{w}^T \Sigma \mathbf{w} + \mathbf{w}^T (\bar{\mathbf{r}} - r_f \mathbf{e}) + r_f
 \end{aligned} \tag{19}$$

Set derivative = 0, and solve for  $\mathbf{w}$ :

$$f'(\mathbf{w}) = -2\mathbf{w}^T \Sigma + \bar{\mathbf{r}}^T - r_f \mathbf{e}^T = 0 \tag{20}$$

c) The problem can also be solved by deriving the optimality conditions. The Lagrangian function,  $L$  can be obtained by substituting the constraints into the objective function:

$$L = -\frac{a}{2} \mathbf{w}^T \Sigma \mathbf{w} + \mathbf{w}^T (\bar{\mathbf{r}} - r_f \mathbf{e}) + r_f \tag{21}$$

The optimality condition can be obtained by differentiating the Lagrangian w.r.t.  $\mathbf{w}$  set derivative to 0:

$$\begin{aligned}
 \frac{dL}{d\mathbf{w}} &= -a \Sigma \mathbf{w} - r_f \mathbf{e} + \bar{\mathbf{r}} = 0 \\
 &\Rightarrow a \Sigma \mathbf{w} + r_f \mathbf{e} - \bar{\mathbf{r}} = 0 \\
 \mathbf{w} &= \frac{1}{a} \Sigma^{-1} (\bar{\mathbf{r}} - r_f \mathbf{e})
 \end{aligned} \tag{22}$$

Maximum value found by substituting into the quadratic problem:

$$\begin{aligned}
 &-\frac{a}{2} \left( \frac{1}{a} \Sigma^{-1} (\bar{\mathbf{r}} - r_f \mathbf{e}) \right)^T \Sigma \left( \frac{1}{a} \Sigma^{-1} (\bar{\mathbf{r}} - r_f \mathbf{e}) \right) + \left( \frac{1}{a} \Sigma^{-1} (\bar{\mathbf{r}} - r_f \mathbf{e}) \right)^T (\bar{\mathbf{r}} - r_f \mathbf{e}) + r_f \\
 &= -\frac{1}{2a} (\bar{\mathbf{r}} - r_f \mathbf{e})^T \Sigma^{-1} (\bar{\mathbf{r}} - r_f \mathbf{e}) + \frac{1}{a} (\bar{\mathbf{r}} - r_f \mathbf{e})^T \Sigma^{-1} (\bar{\mathbf{r}} - r_f \mathbf{e}) + r_f \\
 &= \frac{1}{2a} (\bar{\mathbf{r}} - r_f \mathbf{e})^T \Sigma^{-1} (\bar{\mathbf{r}} - r_f \mathbf{e}) + r_f
 \end{aligned} \tag{23}$$

d) As  $a$  tends to  $+\infty$ :

$$\lim_{a \rightarrow \infty} \frac{1}{2a} (\bar{\mathbf{r}} - r_f \mathbf{e})^T \Sigma^{-1} (\bar{\mathbf{r}} - r_f \mathbf{e}) + r_f = r_f$$

e) Let  $f$  denote the optimal value function (expression 23):

$$\begin{aligned} f(\bar{\mathbf{r}}) &= \frac{1}{2a} (\bar{\mathbf{r}} - r_f \mathbf{e})^T \Sigma^{-1} (\bar{\mathbf{r}} - r_f \mathbf{e}) + r_f \\ \Rightarrow f(\mathbb{E}[\hat{\mathbf{r}}]) &= \frac{1}{2a} (\mathbb{E}[\hat{\mathbf{r}}] - r_f \mathbf{e})^T \Sigma^{-1} (\mathbb{E}[\hat{\mathbf{r}}] - r_f \mathbf{e}) + r_f \end{aligned} \quad (24)$$

The average estimation of the optimal value of the investor is:

$$\mathbb{E}[f(\hat{\mathbf{r}})] = \mathbb{E}\left[\frac{1}{2a} (\hat{\mathbf{r}} - r_f \mathbf{e})^T \Sigma^{-1} (\hat{\mathbf{r}} - r_f \mathbf{e}) + r_f\right] \quad (25)$$

Jenson's inequality states that:

$$\mathbb{E}[f(x)] \geq f(\mathbb{E}[x]) \quad (26)$$

Therefore, it can be shown that the investor will on average overestimate the optimal value of the portfolio:

$$\mathbb{E}[f(\hat{\mathbf{r}})] \geq f(\mathbb{E}[\hat{\mathbf{r}}]) \quad (27)$$