

# COT 5615 Math for Intelligent Systems, Fall 2021

## Midterm 1

A2.1 (14 pts)

$$f(x) = (\mathbf{e}_m - \frac{1}{n}\mathbf{1}_n)^T x.$$

3.25 (22 pts) We went over this question during the lecture on Sept. 7.

(a) (10 pts) The mean return of the portfolio is the average of the vector  $p$ :

$$\begin{aligned}\mathbf{avg}(p) &= \mathbf{avg}(\theta r + (1 - \theta)\mu^{\text{rf}}\mathbf{1}) \\ &= \theta\mathbf{avg}(r) + (1 - \theta)\mu^{\text{rf}}\mathbf{avg}(\mathbf{1}) \\ &= \theta\mu + (1 - \theta)\mu^{\text{rf}}.\end{aligned}$$

On the last line we use  $\mathbf{avg}(r) = \mu$ , and  $\mathbf{avg}(\mathbf{1}) = 1$ .

The risk is the standard deviation of the vector  $p$ :

$$\begin{aligned}\mathbf{std}(p) &= \|p - \mathbf{avg}(p)\mathbf{1}\|/\sqrt{T} \\ &= \|\theta r + (1 - \theta)\mu^{\text{rf}}\mathbf{1} - (\theta\mu + (1 - \theta)\mu^{\text{rf}})\mathbf{1}\|/\sqrt{T} \\ &= \|\theta(r - \mu\mathbf{1})\|/\sqrt{T} \\ &= |\theta|\|(r - \mu\mathbf{1})\|/\sqrt{T} \\ &= |\theta|\mathbf{std}(r) \\ &= |\theta|\sigma.\end{aligned}$$

On line 2 we use the expression for  $\mathbf{avg}(p)$  that we derived in part (a). The last step is the definition of  $\sigma = \mathbf{std}(r)$ .

(b) (6 pts) To achieve the target risk  $\sigma^{\text{tar}}$ , we need  $|\theta| = \sigma^{\text{tar}}/\sigma$ , so there are two choices:

$$\theta = \sigma^{\text{tar}}/\sigma, \quad \theta = -\sigma^{\text{tar}}/\sigma.$$

To choose the sign of  $\theta$  we consider the portfolio return

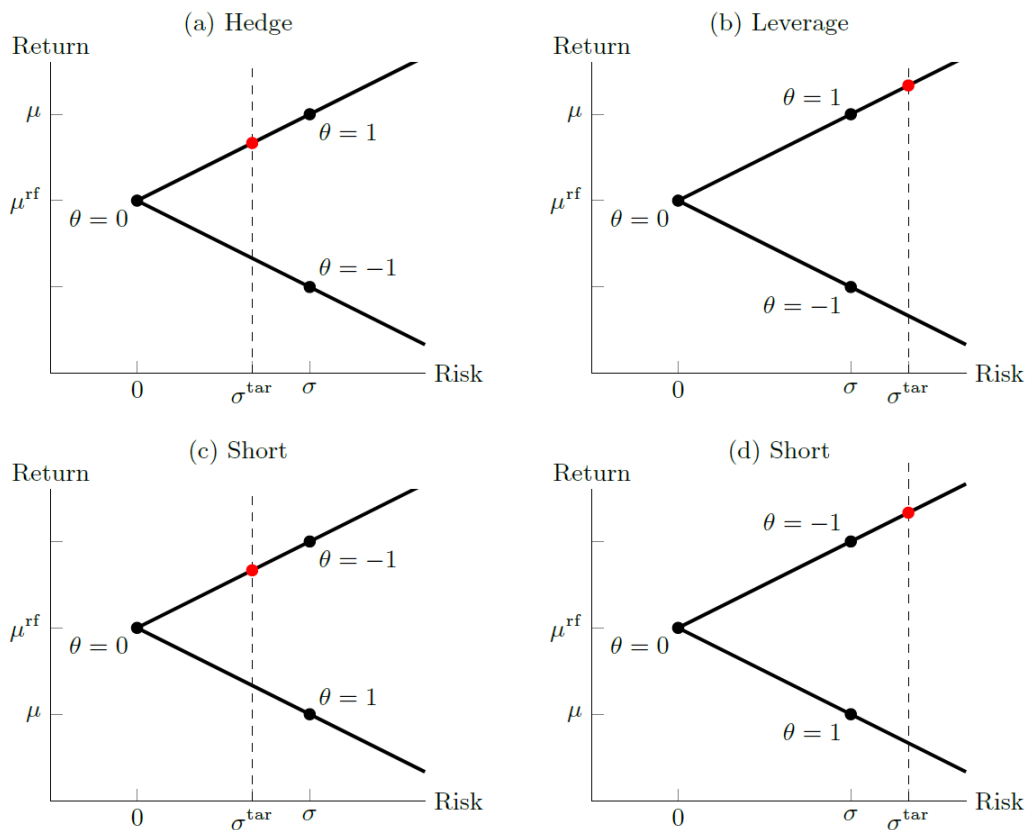
$$\theta\mu + (1 - \theta)\mu^{\text{rf}} = \mu^{\text{rf}} + \theta(\mu - \mu^{\text{rf}}).$$

To maximize this, for given  $|\theta|$ , we choose  $\theta$  positive if  $\mu > \mu^{\text{rf}}$  and  $\theta$  negative if  $\mu < \mu^{\text{rf}}$ . This means we short the asset when its return is less than the risk-free return.

(c) (6 pts) We can distinguish the cases shown in the figure below. The solid lines show

$$(\text{risk}, \text{return}) = (\mathbf{std}(p), \mathbf{avg}(p)) = (|\theta|\sigma, \mu^{\text{rf}} + \theta(\mu - \mu^{\text{rf}}))$$

for all values of  $\theta$ . The red dot shows the portfolio with the highest return for the given target value of risk.



In case (a), the asset return is more than the risk-free return ( $\mu > \mu^{\text{rf}}$ ) and its risk is higher than the target risk ( $\sigma > \sigma^{\text{tar}}$ ). In this case we hedge ( $0 < \theta < 1$ ). In case (b), the asset return is more than the risk-free return ( $\mu > \mu^{\text{rf}}$ ) and its risk is less than the target risk ( $\sigma < \sigma^{\text{tar}}$ ). In this case we leverage ( $\theta > 1$ ). In cases (c) and (d), the asset return is less than the risk-free return ( $\mu < \mu^{\text{rf}}$ ). In this case we short ( $\theta < 0$ ).

To summarize, we short the asset when its return is less than the risk-free return. We hedge when the asset return is more than the risk-free return and the asset risk is higher than the target risk. We leverage when the asset return is more than the risk-free return and the asset risk is less than the target risk.

A8.1 (15 pts)

$$p(0) = p(1) \Rightarrow c_1 = c_1 + c_2 + c_3 + c_4 + c_5.$$

Besides,

$$p'(x) = c_2 + 2c_3x + 3c_4x^2 + 4c_5x^3,$$

so

$$p'(0) = p'(1) \Rightarrow c_2 = c_2 + 2c_3 + 3c_4 + 4c_5.$$

The set of linear equations

$$\begin{aligned}c_2 + c_3 + c_4 + c_5 &= 0 \\ 2c_3 + 3c_4 + 4c_5 &= 0\end{aligned}$$

can be written as  $Ac = b$  with

$$A_{2 \times 5} = \begin{bmatrix} 0 & 1 & 1 & 1 & 1 \\ 0 & 0 & 2 & 3 & 4 \end{bmatrix}, \quad b_{2 \times 1} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

8.8 (15 pts) The interpolation conditions

$$\frac{c_1 + c_2 t_i + c_3 t_i^2}{1 + d_1 t_i + d_2 t_i^2} = y_i, \quad i = 1, \dots, K,$$

can be written as linear equations

$$c_1 + c_2 t_i + c_3 t_i^2 - y_i d_1 t_i - y_i d_2 t_i^2 = y_i, \quad i = 1, \dots, K.$$

If we define  $\theta = (c_1, c_2, c_3, d_1, d_2)$  then this can be written as  $A\theta = b$  with

$$A = \begin{bmatrix} 1 & t_1 & t_1^2 & -y_1 t_1 & -y_1 t_1^2 \\ 1 & t_2 & t_2^2 & -y_2 t_2 & -y_2 t_2^2 \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 1 & t_K & t_K^2 & -y_K t_K & -y_K t_K^2 \end{bmatrix}, \quad b = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_K \end{bmatrix}.$$

9.4 (14 pts)

$$z_{t+1} = \begin{bmatrix} A_1 & A_2 \\ I & 0 \end{bmatrix} z_t.$$

A10.5 (20 pts)

- (a) Makes sense. Dimension is  $10 \times 10$ .
- (b) Makes sense. Dimension is  $20 \times 10$ .
- (c) Doesn't make sense.
- (d) Makes sense. Dimension is  $10 \times 20$ .
- (e) Makes sense. Dimension is  $20 \times 10$ .