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# 40.002 OPTIMIZATION

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## Problem 1

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## Question 1

a) Let  $D$  and  $F$  denote Domestic and Foreign stocks (in millions) respectively. We formulate the Linear Program (LP):

$$\begin{aligned} \max_{D,F} \quad & 0.11D + 0.17F \\ \text{s.t.} \quad & D + F \leq 12 \\ & 0 \leq D \leq 10 \\ & 0 \leq F \leq 7 \\ & D \geq \frac{1}{2}F \\ & F \geq \frac{1}{2}D \end{aligned}$$

b) Graphically, we have Figure 1 with corner points labelled and the feasible region shaded:

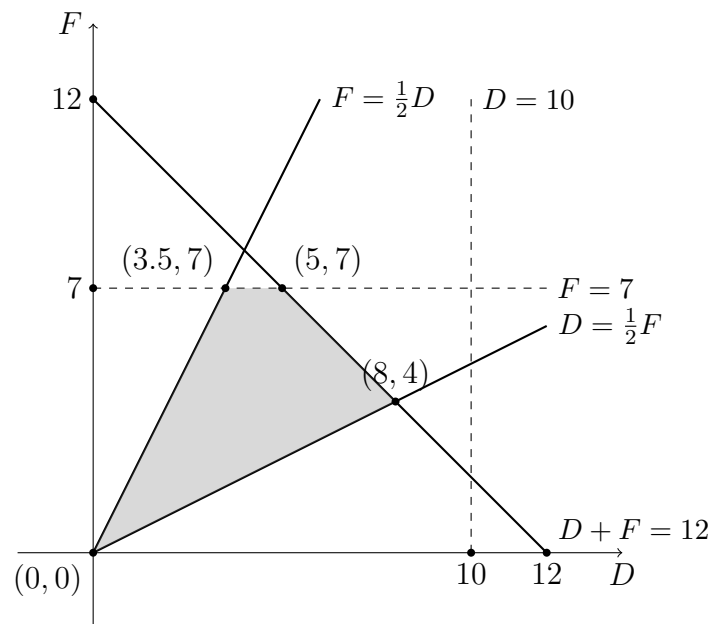


Figure 1: Feasible Region of LP

c) Solving graphically, the optimal value of  $F$  and  $D$  that maximises returns are  $D = 5$ ,  $F = 7$  from Figure 1, for a total (maximum) return of:

$$0.11(5) + 0.17(7) = \$1.74 \text{ million}$$

To prove that this is the global maximum, we brute-force attempt to find the returns of the other corner points:

For  $(3.5, 7)$ , we have:

$$0.11(3.5) + 0.17(7) = \$1.575 \text{ million} < \$1.74 \text{ million}$$

For  $(8, 4)$ , we have:

$$0.11(8) + 0.17(4) = \$1.56 \text{ million} < \$1.74 \text{ million}$$

And thus \$1.74 million is indeed the global maximum return for the fund. The mutual fund manager should divide \$5 million into Domestic stocks and \$7 million into Foreign stocks.

## Question 2

a) Given the LP constraints

$$\begin{aligned} -x_1 + x_2 &\geq 1 \\ x_1 &\geq 0 \end{aligned}$$

We have the feasible region shaded and corner points labelled:

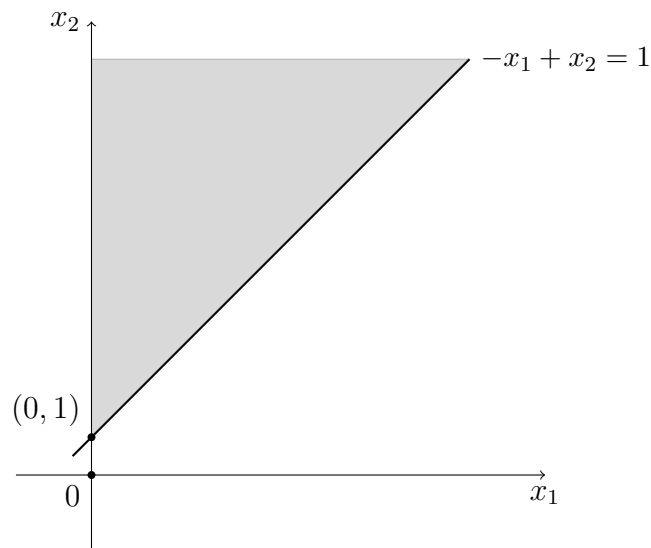


Figure 2: Feasible Region of LP

b) For a minimisation problem, we have the complete LP:

$$\begin{aligned} \min_{x_1, x_2} \quad & \alpha x_1 + x_2 \\ \text{s.t.} \quad & -x_1 + x_2 \geq 1 \\ & x_1 \geq 0 \end{aligned}$$

i) For the optimal solution to be unique, we need  $\alpha > -1$ , and the unique optimum is at the point  $(0, 1)$ . For any value of  $\alpha$  within this range, the optimum solution will always converge to this minimum point.

ii) For there to be multiple optimal solutions with finite optimal objective values, we need  $\alpha = -1$ , and this results in slope = 1. The line now lies on  $-x_1 + x_2 = 1$  on Figure 2, with multiple optimal solutions and optimal objective value of 1.

iii) For there to be an unbounded optimal objective value, we need  $\alpha < -1$ , resulting in a slope = 1.

### Question 3

a) Since  $Z$  is a discrete random variable that takes values in the set  $\{1, 2, \dots, 10\}$ , and  $p_i = \text{Prob}(Z = i) \forall i \in \{1, 2, \dots, 10\}$ , with sum and non-negative constraints, we can formulate the LP:

$$\begin{aligned} \max_{p_i} \quad & \sum_{i=5}^{10} p_i \\ \text{s.t.} \quad & \sum_{i=1}^{10} p_i = 1 \\ & \sum_{i=1}^{10} i \cdot p_i = 4 \\ & p_i \geq 0 \end{aligned}$$

We want to optimise the probability distribution of  $Z$  such that the likelihood of  $Z$  being 5, 6, 7, 8, 9, or 10 is maximised. The first and last constraints come from the fact that the probabilities must sum to one and be non-negative. The second constraint arises from the average value (and hence the expectation)  $\mathbb{E}(X) = \sum_{i=1}^{10} i \cdot p_i = 4$ .

Using the following *JuMP* code and GLPK solver, we have the final probability distribution of  $Z$  being:

$$p = \{0.25, 0, 0, 0, 0.75, 0, 0, 0, 0, 0\}$$

## Question 4

For the minimisation problem:

$$\begin{aligned} \min \quad & x_1 - x_2 \\ \text{s.t.} \quad & 2x_1 + 3x_2 - x_3 + x_4 \leq 0 \end{aligned} \tag{1}$$

$$3x_1 + x_2 + 4x_3 - 2x_4 \geq 3 \tag{2}$$

$$-x_1 - x_2 + 2x_3 + x_4 = 6 \tag{3}$$

$$x_1 \leq 0 \tag{4}$$

$$x_2, x_3 \geq 0 \tag{5}$$

We can rearrange the LP into the form:

$$\begin{aligned} \max \quad & \mathbf{c}^\top \mathbf{x} \\ \text{s.t.} \quad & \mathbf{Ax} \leq \mathbf{b} \end{aligned}$$

Note that to transform the LP from a minimisation to maximisation problem, we multiply the objective function by  $-1$ :

$$\max_{\mathbf{x} \in \chi} f(x_1, \dots, x_n) = -\min_{\mathbf{x} \in \chi} -f(x_1, \dots, x_n) \implies \min -x_1 + x_2$$

For constraint (3), we separate into two cases:

$$\begin{aligned} -x_1 - x_2 + 2x_3 + x_4 &\geq 6 \\ x_1 + x_2 - 2x_3 - x_4 &\leq -6 \end{aligned}$$

where:

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}, \quad \mathbf{c} = \begin{bmatrix} -1 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \quad \mathbf{A} = \begin{bmatrix} 2 & 3 & -1 & 1 \\ -3 & -1 & -4 & 2 \\ -1 & -1 & 2 & 1 \\ 1 & 1 & -2 & -1 \\ 1 & 0 & 0 & 0 \\ 0 & -1 & 0 & 0 \\ 0 & 0 & -1 & 0 \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} 0 \\ -3 \\ 6 \\ -6 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

After transformation to the maximisation problem, in both LPs, the optimum solution remains the exact same. However, since we multiply the objective value by  $-1$ , turning it into a maximisation problem means that the optimal objective value now differs from the minimisation problem by  $-1$ , i.e. its negative value.

## Question 5

The given optimisation problem is non-linear in nature, due to the absolute value function  $f(z)$  in the objective function. We linearise the problem by rewriting  $f(x)$  as a piecewise linear function:

$$\begin{aligned} \min \quad & \sum_{i=1}^m \max(0, \mathbf{a}_i^\top \mathbf{x} - b_i - 5, -\mathbf{a}_i^\top \mathbf{x} - b_i - 5) \\ \text{s.t.} \quad & \mathbf{x} \in \mathbb{R}^n \end{aligned}$$

We let  $y_i$  be auxiliary variables for each  $i$ , and  $z_i$  be the argument of the function. Enforcing additional constraints to the problem, we now formulate an appropriate LP:

$$\begin{aligned} \min \quad & \sum_{i=1}^m y_i \\ \text{s.t.} \quad & z_i = \mathbf{a}_i^\top \mathbf{x} - b_i, \quad i = 1, \dots, m \\ & y_i \geq z_i - 5, \quad i = 1, \dots, m \\ & y_i \geq -z_i - 5, \quad i = 1, \dots, m \\ & y_i \geq 0, \quad i = 1, \dots, m \\ & \mathbf{x} \in \mathbb{R}^n \end{aligned}$$

Note that  $y_i \in \mathbb{R}^n$  as well. The constraints ensure that we are obtaining the minimum value of all  $i$ 's as the optimal solution. Graphically, refer to Figure 3 below for  $f(z) = \max(0, z - 5, -z - 5)$ :

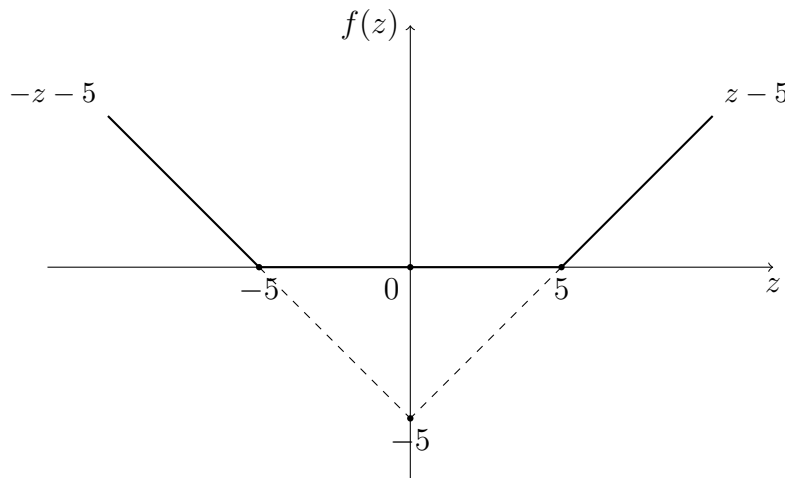


Figure 3: Graph of Piecewise Function

## Question 6

We are given that  $f(\mathbf{x})$  is convex, and we need to show that  $\forall t \in \mathbb{R}$ , the set of  $\mathbf{x}$  (denoted by  $\chi_t$ ) that satisfies  $f(\mathbf{x}) \leq t$  is a convex set. Geometrically, for any 2 points  $\mathbf{x}_1$  and  $\mathbf{x}_2$  in  $\chi_t$ , the line segment containing them must also be in  $\chi_t$ . i.e., if  $\mathbf{x}_1, \mathbf{x}_2 \in \chi_t$ , then  $\lambda \mathbf{x}_1 + (1 - \lambda) \mathbf{x}_2 \in \chi_t$ ,  $\forall \lambda \in [0, 1]$ .

*Proof.*

$\because \mathbf{x}_1, \mathbf{x}_2 \in \chi_t$ , then  $f(\mathbf{x}_1) \leq t$  and  $f(\mathbf{x}_2) \leq t$ . Since  $f(\mathbf{x})$  is convex, we use the property:

$$\begin{aligned} f(\lambda \mathbf{x}_1 + (1 - \lambda) \mathbf{x}_2) &\leq \lambda f(\mathbf{x}_1) + (1 - \lambda) f(\mathbf{x}_2) \\ &\leq \lambda t + (1 - \lambda) t \\ &= \lambda t + t - \lambda t \\ &= t \end{aligned}$$

$\therefore \forall \lambda \in [0, 1]$ ,  $f(\lambda \mathbf{x}_1 + (1 - \lambda) \mathbf{x}_2) \leq t$ , and thus  $\lambda \mathbf{x}_1 + (1 - \lambda) \mathbf{x}_2 \in \chi_t$ , which is convex.  $\square$

An appropriate illustration of this is Figure 4 below.

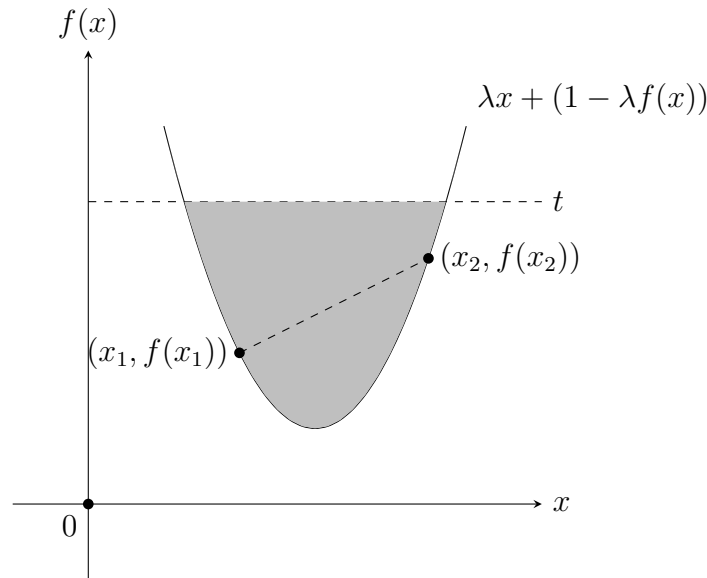


Figure 4: Illustration of Convexity



## Question 7

(1)

$$\begin{aligned}
 \max \quad & 5x_1x_2 + 12x_3 \\
 \text{s.t.} \quad & x_3 \leq 2 \\
 & x_2 \leq 5 \\
 & 3.2x_2 + 5.1x_1 = 5.1 \\
 & x_1 + \log x_2 - x_3 = 0 \\
 & x_1, x_2, x_3 \in \{0, 1\}
 \end{aligned}$$

For this LP, only Certainty holds, while Proportionality, Additivity, and Divisibility assumptions do not. The **logarithmic term in the constraint** does not yield a proportional contribution, so Proportionality does not hold. Similarly, the **total value of the objective function is not the sum of the contribution of each of the terms**, due to the  $x_1x_2$  term, so Additivity does not hold. Lastly, the final constraint suggests that  $x_1, x_2, x_3$  are **constrained to values either 0 or 1 in the set, which are only binary whole numbers, and not fractional values**. Thus Divisibility does not hold.

(2)

$$\begin{aligned}
 \max \quad & 5x_1 - x_2 - 2x_3 \\
 \text{s.t.} \quad & x_3 \leq 2.9 \\
 & x_2 \geq \theta \\
 & x_3 + x_2 - x_1^2 = 0 \\
 & 0.4 \frac{x_1}{x_2} + x_3 \leq 0 \\
 & x_1, x_2, x_3 \geq 0 \\
 & \theta \in U(0, 1)
 \end{aligned}$$

For this LP, only Divisibility holds, while Proportionality, Additivity, and Certainty do not. Rearranging the terms in constraint (4), we have  $0.4x_1 \leq -x_2x_3$ . Additivity does not hold in this case due to the  $-x_2x_3$  term. Proportionality does not hold due to the  $-x_1^2$  term in constraint (3), as it does not yield a proportional change in the contribution to the function. For Certainty, since random variables are stochastic in nature, there is no way to know for sure the value of  $x_2$ .

**(3)**

$$\begin{array}{ll}\max & x_1 - 15x_2 - 10x_3 \\ \text{s.t.} & \frac{x_3}{x_2} \leq 20 \\ & x_2 - 4x_1 \leq 0 \\ & x_3 + x_2 - x_1 = 0 \\ & x_1, x_2, x_3 \geq 0\end{array}$$

For this LP, all 4 assumptions of Proportionality, Additivity, Divisibility, and Certainty hold.