

Problem 1: let $z_1, z_2 \in \mathbb{Z}$. Then $\exists x_1 \in X_1, x_2 \in X_2$ such that

$$z_1 = a_1 x_1 + a_2 x_2$$

Similarly, $\exists y_1 \in X_1, y_2 \in X_2$ s.t.

$$z_2 = a_1 y_1 + a_2 y_2$$

Now,

$$\begin{aligned} \lambda z_1 + (1-\lambda) z_2 &= \lambda (a_1 x_1 + a_2 x_2) + (1-\lambda) (a_1 y_1 + a_2 y_2) \\ &= a_1 (\underbrace{\lambda x_1 + (1-\lambda) y_1}_{\in X_1}) + a_2 (\underbrace{\lambda x_2 + (1-\lambda) y_2}_{\in X_2}) \end{aligned}$$

Hence $\lambda z_1 + (1-\lambda) z_2 \in \mathbb{Z} \quad \forall \lambda \in [0, 1], z_1, z_2 \in \mathbb{Z}$.

Problem 2: S is epigraph of the function $f(x, y) = x^2 + y^2$, where f is a convex function of $\begin{pmatrix} x \\ y \end{pmatrix}$.

Therefore, S is a convex set.

Problem 3: $f(\lambda x_1 + (1-\lambda)x_2)$

$$= h(g(\lambda x_1 + (1-\lambda)x_2))$$

$$\leq h(\lambda g(x_1) + (1-\lambda)g(x_2)) \quad \begin{array}{l} \text{since } g \text{ is concave and} \\ h \text{ is non-increasing} \end{array}$$

$$\leq \lambda h(g(x_1)) + (1-\lambda)h(g(x_2)) \quad \text{since } h \text{ is convex}$$

$$= \lambda f(x_1) + (1-\lambda)f(x_2).$$

The above holds for all $x_1, x_2 \in \text{dom } f$ and $\lambda \in [0, 1]$.

Hence, f is a convex function

Problem 4: The function is convex.

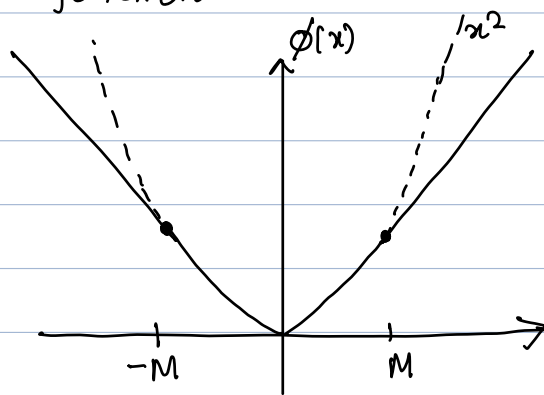
need to show that

$$f(y) \geq f(x) + \nabla f(x)^T (y-x)$$

$\forall x, y$.

If both $|x|, |y| \leq M$

or $x, y > M$ or $x, y < -M$, then easy.



Need to show the above holds when $|x| \leq M$ and $|y| > M$.

Use the fact that $f(x)$ is monotonically increasing in x .

Problem 5: any $x \in [a, b]$ can be written as $\lambda a + (1-\lambda)b$.

$$\begin{aligned} f(x) = f(\lambda a + (1-\lambda)b) &\leq \lambda f(a) + (1-\lambda)f(b) \\ &\leq \lambda \max(f(a), f(b)) + (1-\lambda) \max(f(a), f(b)) \\ &= \max(f(a), f(b)). \end{aligned}$$

Problem 6: The given problem is equivalent to $\min_x C^T x$
s.t. $Ax = b$
 $x \geq 0$

with $C^T = [0, 1, -1]$

$$A = \begin{bmatrix} 0 & -2 & 0.5 \\ -1 & -1 & 1 \end{bmatrix}, \quad b = \begin{bmatrix} 1 \\ 1 \end{bmatrix}.$$

Dual:

$$\max_{y \in \mathbb{R}^2} y_1 + y_2$$

$$\text{s.t.}$$

$$A^T y \leq C$$

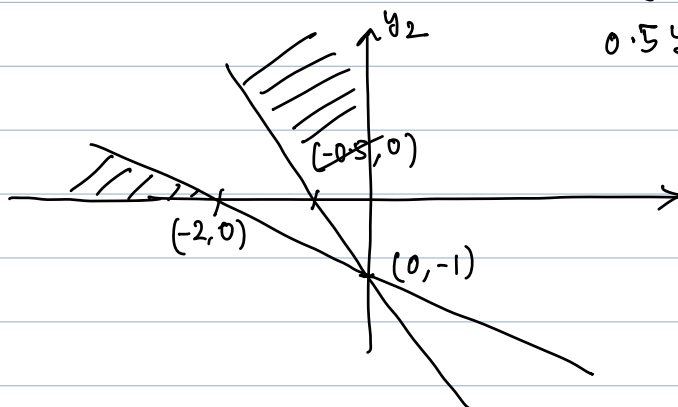
$\Leftrightarrow$

$$-y_2 \leq 0$$

$$-2y_1 - y_2 \leq 1$$

$$0.5y_1 + y_2 \leq -1$$

} infeasible



- Dual is infeasible.

- Since primal is feasible, it is unbounded.

$x = \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix}$ is feasible to primal.

Problem 7: The given problem is equivalent to

$$\begin{aligned} \min_{x \in \mathbb{R}} \quad & 1+x^2 \\ \text{s.t.} \quad & x^2 \geq 1 \end{aligned}$$

$\Rightarrow$ optimal solⁿ is

$$x^* = 1, \text{ opt. value} = 2$$

$$\begin{aligned} L(x, \lambda) &= 1+x^2 + \lambda(1-x^2) \\ &= 1+\lambda + x^2(1-\lambda) \end{aligned}$$

$$\inf_x L(x, \lambda) = \begin{cases} -\infty & \text{if } 1-\lambda < 0 \\ 1+\lambda & \text{if } 1-\lambda \geq 0 \end{cases}$$

$$\text{dual: } \begin{array}{ll} \max_{\lambda} & 1+\lambda \\ \text{s.t.} & 1-\lambda \geq 0 \\ & \lambda \geq 0 \end{array} \quad \} \quad 0 \leq \lambda \leq 1$$

optimal dual solution : $\lambda^* = 1$

optimal dual value = 2

Strong duality holds.
