

Problem Set #5

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Problem 7.1

Let two elements of S be $\lambda_1 x_1 + \lambda_2 x_2 + \dots + \lambda_n x_n$ and $\alpha_1 y_1 + \alpha_2 y_2 + \dots + \alpha_m y_m$. We then choose a $\beta \in (0, 1)$ such that:

$$\frac{1-\beta}{2} \left[\frac{\lambda_1}{1-\beta} x_1 + \dots + \frac{\lambda_n}{1-\beta} x_n \right] + \frac{\beta}{2} \left[\frac{\alpha_1}{\beta} x_1 + \dots + \frac{\alpha_n}{\beta} x_n \right]$$

Since both elements are convex combinations, combination of those two will be a convex combination as well. We divide both sets by 2 so that the coefficients will all sum up to 1 which is a requirement for the elements of S .

Problem 7.2

- (i) Let $p_1, p_2 \in P$. We want to show that the combination of those two points $\alpha p_1 + (1-\alpha)p_2$ is in P as well. By rules of inner product spaces,

$$\begin{aligned} \langle a, \alpha p_1 + (1-\alpha)p_2 \rangle &= \alpha \langle a, p_1 \rangle + (1-\alpha) \langle a, p_2 \rangle \\ &= \alpha b + (1-\alpha)b \\ &= b \end{aligned}$$

Which implies that the point is in the arbitrary hyperplane. Thus, the hyperplane is convex.

- (ii) Let $h_1, h_2 \in H$. We want to show that the combination of those two points $\alpha h_1 + (1-\alpha)h_2$ is in H as well. We know that $\langle a, h_1 \rangle \leq b$ and $\langle a, h_2 \rangle \leq b$.

$$\begin{aligned} \langle a, \alpha h_1 + (1-\alpha)h_2 \rangle &= \alpha \langle a, h_1 \rangle + (1-\alpha) \langle a, h_2 \rangle \\ &\leq \alpha b + (1-\alpha)b \\ &= b \end{aligned}$$

Thus, the point is in the half space H .

Problem 7.4

- (i)

$$\begin{aligned} \|x - y\|^2 &= \|x - p + p - y\|^2 \\ &= \|x - p\|^2 + \|p - y\|^2 + 2\langle x - p, p - y \rangle \end{aligned}$$

(ii)

$$\begin{aligned} \|x - y\|^2 &= \|x - p\|^2 + \|p - y\|^2 + 2\langle x - p, p - y \rangle \\ &= \langle x - p, p - y \rangle^2 \\ &> \langle x - p, x - p \rangle^2 \\ &= \|x - p\|^2 \end{aligned}$$

Thus, $\|x - y\| > \|x - p\|$

(iii)

$$\begin{aligned} \|x - z\|^2 &= \|x - \lambda y - (1 - \lambda)p\|^2 \\ &= \|(x - p) + \lambda(p - y)\|^2 \\ &= \|x - p\|^2 + \lambda^2\|y - p\|^2 + 2\lambda\langle x - p, p - y \rangle^2 \end{aligned}$$

(iv)

$$\|x - z\|^2 - \|x - p\|^2 = 2\lambda\langle x - p, p - y \rangle + \lambda^2\|y - p\|^2 \geq 0$$

since $\|x - z\|^2$ is larger than $\|x - p\|^2$ as the latter is the projection of x on C .

Part (ii) proves the theorem in the (\Leftarrow) direction and part (iv) proves it in the (\Rightarrow) direction.

Problem 7.6

Let $x_1, x_2 \in S = \{x \in \mathbb{R}^n | f(x) \leq c\}$. We want to show that $\alpha x_1 + (1 - \alpha)x_2 \in S$, which is to say that $f(\alpha x_1 + (1 - \alpha)x_2) \leq c$. Let $\alpha = \lambda$

$$\begin{aligned} f(\lambda x_1 + (1 - \lambda)x_2) &\leq \lambda f(x_1) + (1 - \lambda)f(x_2) \\ &\leq \lambda c + (1 - \lambda)c \\ &= c \end{aligned}$$

Problem 7.7

We want to show that $f(\alpha x_1 + (1 - \alpha)x_2) \leq \alpha f(x_1) + (1 - \alpha)f(x_2)$.

$$\begin{aligned} f(\alpha x_1 + (1 - \alpha)x_2) &= \sum_{i=1}^k \lambda_i f_i(\alpha x_1 + (1 - \alpha)x_2) \\ &\leq \sum_{i=1}^k \lambda_i \left[\alpha f_i(x_1) + (1 - \alpha)f_i(x_2) \right] \\ &= \alpha f(x_1) + (1 - \alpha)f(x_2) \end{aligned}$$

Problem 7.13

Let the minimum point be x_1 . Choose an arbitrary point x_2 . WLOG, assume that $x_2 \geq x_1$. We construct a line segment from $(x_2, f(x_2))$ to $(x_1, f(x_1))$, the minimum point. The segment of function $f(x), x \in [x_1, x_2]$ should lie between line $y = \frac{f(x_2)-f(x_1)}{x_2-x_1}(x_2 - x_1) + f(x_1)$ and the line $y = f(x_1)$.

Therefore, $\frac{f(x_2)-f(x_1)}{x_2-x_1}(x_2 - x_1) + f(x_1) \geq f(x) - f(x_1) \geq 0$ for any x and x_2 as long as $x \in [x_1, x_2]$. For an arbitrary x , the inequality holds as $x_2 \rightarrow \infty$, but we know that $\lim_{x_2 \rightarrow \infty} \frac{f(x_2)-f(x_1)}{x_2-x_1}(x - x_1) = 0$. Recall that $f(x_2), f(x_1)$ are bounded.

Thus, $0 \geq f(x) - f(x_1) \geq 0$ which implies that $f(x) = f(x_1)$.

Problem 7.20

$$\begin{aligned} f(\lambda x_1 + (1 - \lambda)x_2) &\leq \lambda f(x_1) + (1 - \lambda)f(x_2) \\ -f(\lambda x_1 + (1 - \lambda)x_2) &\leq -\lambda f(x_1) - (1 - \lambda)f(x_2) \\ \Rightarrow f(\lambda x_1 + (1 - \lambda)x_2) &= \lambda f(x_1) + (1 - \lambda)f(x_2) \end{aligned}$$

This proves linearity. We set $c = 0$ and thus f is affine.

Problem 7.21

(\Rightarrow) If x^* is a local minimizer for the optimization problem, then $\phi \circ f(x^*) \leq \phi \circ f(x)$ for all $x \in \mathbb{R}^n$. Since ϕ is strictly increasing, $f(x^*) \leq f(x)$. (\Leftarrow) We know that $f(x^*) \leq f(x)$. Since ϕ is strictly increasing, we know that $\phi \circ f(x^*) \leq \phi \circ f(x)$ as well.