

Exercise 4

- ① If the subdifferential of $f: \mathbb{R}^d \rightarrow \mathbb{R}$ is nonempty at every point, then f is a convex function.

The subdifferential of f , $\sigma f(x)$, is the set of all subgradients of f at x .

$$\sigma f(x) := \{ g \text{ s.t. } f(y) \geq f(x) + g^T(y-x) \quad \forall y \in X \}$$

We know that $\sigma f(x)$ is nonempty at every point

$$\Rightarrow \forall x, \exists g \text{ s.t. } f(y) \geq f(x) + g^T(y-x)$$

Let $z = \lambda x + (1-\lambda)y$, where $\lambda \in [0,1]$, $g \in \sigma f(x)$.

$$\begin{aligned} f(x) &\geq f(z) + g^T(x-z) = \\ &= f(z) + g^T(x - \lambda x - (1-\lambda)y) = \\ &= f(z) + g^T(\lambda(x-y)) = \\ &= f(z) + \lambda g^T(x-y) = \\ &= f(z) + (1-\lambda)g^T(x-y) \end{aligned}$$

$$\begin{aligned} f(y) &\geq f(z) + g^T(y-z) = \\ &= f(z) + g^T(y - \lambda x - (1-\lambda)y) = \\ &= f(z) + g^T(-\lambda(x-y)) = \\ &= f(z) - \lambda g^T(x-y) \end{aligned}$$

$$\begin{aligned} \lambda f(x) + (1-\lambda)f(y) &\geq \lambda f(z) + \lambda(1-\lambda)g^T(x-y) + (1-\lambda)f(z) - (1-\lambda)\lambda g^T(x-y) = \\ &= \lambda f(z) + (1-\lambda)f(z) = \\ &= f(z) \end{aligned}$$

We have that:

$$\begin{aligned} \lambda f(x) + (1-\lambda)f(y) &\geq f(z) = \\ &= f(\lambda x + (1-\lambda)y) \end{aligned}$$

$\Rightarrow f$ is convex

□