

Convexity II

Topics we'll cover

- ① Second derivative test for convexity
- ② Convexity examples

Second-derivative test for convexity

A function of several variables, $F(z)$, is convex if its second-derivative matrix $H(z)$ is positive semidefinite for all z .

More formally:

Suppose that for $f : \mathbb{R}^d \rightarrow \mathbb{R}$, the second partial derivatives exist everywhere and are continuous functions of z . Then:

- ① $H(z)$ is a symmetric matrix
- ② f is convex $\Leftrightarrow H(z)$ is positive semidefinite for all $z \in \mathbb{R}^d$

Example

Is $f(x) = \|x\|^2$ convex?

Example

Fix any vector $u \in \mathbb{R}^d$. Is this function $f : \mathbb{R}^d \rightarrow \mathbb{R}$ convex?

$$f(z) = (u \cdot z)^2$$

Least-squares regression

Recall loss function: for data points $(x^{(i)}, y^{(i)}) \in \mathbb{R}^d \times \mathbb{R}$,

$$L(w) = \sum_{i=1}^n (y^{(i)} - (w \cdot x^{(i)}))^2$$