

Homework #2

CSE 446/546: Machine Learning
Profs. Jamie Morgenstern and Ludwig Schmidt
Ray Chen

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B1:

• **Answer :**

$$\|x\|_2 \leq \|x\|_1: \quad \|x\|_2^2 = \sum_{i=1}^n |x_i|^2 \leq \left(\sum_{i=1}^n |x_i|^2 + 2 \sum_{i \neq j} |x_i| |x_j| \right) = \|x\|_1^2,$$

$$\|x\|_\infty \leq \|x\|_2: \quad \|x\|_\infty^2 = (\max_{i=1, \dots, n} |x_i|)^2 = (\max_{i=1, \dots, n} |x_i|^2) \leq \sum_{i=1}^n |x_i|^2 = \|x\|_2^2.$$

we have: $\|x\|_\infty \leq \|x\|_2 \leq \|x\|_1$.

B2:

- **Part a:**

Firstly, we need to prove $f(x) + g(x)$ is also a convex function. Because f and g are convex, we have:

$$\begin{aligned}f(\lambda x + (1 - \lambda)y) + g(\lambda x + (1 - \lambda)y) &\leq \lambda f(x) + (1 - \lambda)f(y) + \lambda g(x) + (1 - \lambda)g(y) \\f(\lambda x + (1 - \lambda)y) + g(\lambda x + (1 - \lambda)y) &\leq \lambda(f(x) + g(x)) + (1 - \lambda)(f(y) + g(y)) \\f(\lambda x + (1 - \lambda)y) + g(\lambda x + (1 - \lambda)y) &\leq \lambda f(x) + (1 - \lambda)f(y) \\f(\lambda x + (1 - \lambda)y) &\leq \lambda f(x) + (1 - \lambda)f(y)\end{aligned}$$

So, $f(x) + g(x)$ is a convex function

For $\sum_{i=1}^n l_i(w) + \lambda||w||$, if it is convex, it should be Triangle Inequality, Absolute Scalability and Non-Negativity.

$l_i(w)$ is convex for any $x, y \in \mathbb{R}$, so $\sum_{i=1}^n l_i(w)$ is convex.

Because $||\cdot||$ is convex by given, $\lambda||w||$ is convex.

As prove before if $f(x)$ and $g(x)$ are convex, then $f(x) + g(x)$ is convex.

So, $\sum_{i=1}^n l_i(w) + \lambda||w||$ is convex.

- **Part b:**

We prefer to use loss functions and regularized loss functions because this can ensure every local minima is the global minima for all the training data.

B3:

- Part :

- Part :