

ℓ_p Norms — Product Metrics and Induced Distances

Definition

Let (U_i, D_i) be metric spaces for $i = 1, \dots, d$.
Their **product space** is

$$\hat{U} = U_1 \times U_2 \times \cdots \times U_d$$

For some p with $1 \leq p \leq \infty$, define the **product metric** as

$$\hat{D}_p(x, y) = \left(\sum_{i=1}^d D_i(x_i, y_i)^p \right)^{1/p}$$

If all $U_i = \mathbb{R}$ and $D_i(a, b) = |a - b|$, then $\hat{U} = \mathbb{R}^d$ and

$$D_p(x, y) = \|x - y\|_p$$

is the standard ℓ_p **distance**.

Common ℓ_p Norms

Name	Symbol	Definition	Unit Ball Shape
Manhattan (L_1)	$\ x\ _1$	$\sum_{i=1}^d x_i $	Diamond
Euclidean (L_2)	$\ x\ _2$	$\left(\sum_{i=1}^d x_i^2 \right)^{1/2}$	Circle / Sphere
Chebyshev (L_∞)	$\ x\ _\infty$	$\max_i x_i $	Square / Cube

Induced Distance Functions

Metric	Formula	Interpretation
ℓ_1 Distance (Manhattan)	$D_1(x, y) = \sum_i x_i - y_i $	Grid-based “taxicab” distance
ℓ_2 Distance (Euclidean)	$D_2(x, y) = \sqrt{\sum_i (x_i - y_i)^2}$	Straight-line distance
ℓ_∞ Distance (Chebyshev)	$D_\infty(x, y) = \max_i x_i - y_i $	Largest coordinate difference

Properties

1. Each $\|\cdot\|_p$ defines a **valid norm** on \mathbb{R}^d .
2. The induced distance $D_p(x, y) = \|x - y\|_p$ is a **metric** satisfying all metric axioms.
3. Changing p changes the geometry of the space:
 - Small $p \rightarrow$ sharper shapes (L_1)
 - Large $p \rightarrow$ rounder shapes (L_2)
 - $p \rightarrow \infty \rightarrow$ box-shaped geometry (L_∞)
4. All ℓ_p norms are **equivalent** in finite-dimensional spaces:

$$c_1 \|x\|_p \leq \|x\|_q \leq c_2 \|x\|_p$$

for constants $c_1, c_2 > 0$ depending on dimension d .

Geometric Intuition

p	Equation of Unit Ball in \mathbb{R}^2	Shape
1	$\{ x_1 + x_2 = 1\}$	Diamond
2	$\{x_1^2 + x_2^2 = 1\}$	Circle
∞	$\{\max(x_1 , x_2) = 1\}$	Square

As p increases, the ℓ_p ball transitions from a diamond ($p = 1$) to a circle ($p = 2$) to a square ($p = \infty$). This affects neighborhood shape, volume growth, and nearest-neighbor relationships.

Construction Context

Clarkson (2006) defines these as **product metrics**, meaning ℓ_p distances arise naturally by combining d metric coordinates (U_i, D_i) into one higher-dimensional space. Thus, (\mathbb{R}^d, D_p) is a **metric space** for all $1 \leq p \leq \infty$.