# Geometric Deep Learning Beyond Euclidean Domains

### 1 Geometric Priors

## **Definition 1.1.** Our compact euclidean domain $\Omega$

 $\Omega := \textstyle \prod_{i \in I} [0,1].$ 

#### **Definition 1.2.** Classification

Let  $x \in L^2 := L^2(\Omega)$  then  $f:L^2 \to \mathscr{C}$  surjective is said to be a classification of  $L^2$  on the set  $\mathscr{C}$ .

#### **Definition 1.3.** Training set

Let f be a classification of  $L^2$  on  $\mathscr{C}$  and  $\{x_i\}_{i\in I}\subset L^2$  then the set  $\{(x_i,f(x_i))\}_{i\in I}$  is called a training set for f.

#### **Proposition 1.1.** The classification f is not injective

Let f be a classification of  $L^2$  on  $\mathscr C$  then, given the inevitable noise acting on data, there exists a real positive  $\varepsilon$  such that  $\forall (x, x_{\varepsilon}) \in L^2 \times L^2 : \int\limits_{\Omega} |x - x_{\varepsilon}|^2 < \varepsilon$  we have that  $f(x) = f(x_{\varepsilon})$ .

Given ideal data classification we can define two functions f-equivalent if and only if their images via the classification f are equal according to an equivalence on  $\mathscr C$  which so far can be any set.

#### **Proposition 1.2.** The relation $\simeq$ is an equivalence relation

Let  $x, y, z \in L^2$  we define  $x \simeq y \iff f(x) = f(y)$  where f is a classification of  $L^2$  on  $\mathscr{C}$ , then:

(i)  $x \simeq x$ 

(ii)  $x \simeq y \iff y \simeq x$ 

(iii)  $x \simeq y, y \simeq z \implies x \simeq z$ 

*Proof.* (i),(ii) and (iii) follow from the equivalence on  $\mathscr{C}$  by which they are defined.

#### **Definition 1.4.** Translation operator

Let  $x \in L^2$  and  $v \in \Omega$  then  $T_v : L^2 \to L^2$  such that  $x(\xi) \mapsto x(\xi - v)$  is said to be a translation operator.

#### **Definition 1.5.** Local deformation operator

Let  $x \in L^2$  and  $\tau \in C^{\infty}(\Omega,\Omega)$  then  $L_{\tau}: L^2 \to L^2$  such that  $x(\xi) \mapsto x(\xi - \tau(\xi))$  is said to be a local deformation operator according to the smooth vector field  $\tau$ .

#### **Definition 1.6.** Invariance

A classification f of  $L^2$  on  $\mathscr C$  is said to be A-invariant, where  $A:L^2\to L^2$ , if and only if f(A(x))=f(x)  $\forall x\in L^2$ .

#### **Definition 1.7.** Equivariance

A classification f of  $L^2$  on  $\mathscr C$  is said to be A-equivariant, where  $A:L^2\to L^2$ , if and only if  $f(A(x))=A(f(x))\ \forall x\in L^2$ . This is well defined only if A is defined to act on  $\mathscr C$ 

#### **Proposition 1.3.** If f is translation invariant then it is stable under local deformations

Let f be a translation invariant classification of  $L^2$  on  $\mathscr C$  then  $|f(L_\tau(x)) - f(x)| \approx |J_\tau|$  where  $(J_\tau)_{ij} = (\frac{\partial \tau_i}{\partial \xi_j})$  under some misterious norm.

*Proof.* To be found...  $\Box$