# A metric paradigm for geometrical shape comparison

Patrizio Frosini<sup>1,2</sup>

<sup>1</sup>Department of Mathematics, University of Bologna, Italy <sup>2</sup>ARCES - Vision Mathematics Group, University of Bologna, Italy patrizio.frosini@unibo.it

Fifth Workshop on Non-Rigid Shape Analysis and Deformable Image Alignment (NORDIA'12) Florence, 7 October 2012

# Outline



Our setting and some examples





# Our setting and some examples



Every comparison of properties involves the presence of

- an observer perceiving the properties
- a methodology to compare the properties

It follows that shape comparison is affected by subjectivity.

Let us give some examples illustrating this fact.

The perception properties depend on the subjective interpretation of an observer:



# Magnifying glass or cup of coffee?

The perception properties depend on the subjective interpretation of an observer:



#### Duck or rabbit?

Patrizio Frosini (University of Bologna)

The perception properties depend on the subjective interpretation of an observer:



The black shapes are NOT the camels, the narrow stripes below the shapes are. The black shapes are the shadows of the camels, as this photo was taken from overhead.

Patrizio Frosini (University of Bologna)

The perception properties depend on the subjective interpretation of an observer:



#### How many rabbits?

Patrizio Frosini (University of Bologna)

The perception properties depend on the subjective interpretation of an observer:



# "Glued in Florence" by Christiaan Triebert

Patrizio Frosini (University of Bologna)

The perception properties depend on the subjective interpretation of an observer:



# Our working hypothesis

- In shape comparison objects are not accessible directly, but only via measurements made by an observer.
- The comparison of two shapes is usually based on a family *F* of "measuring functions", which are defined on a set *M* and take values in a set *V*. Each function in *F* represents a measurement obtained via a measuring instrument.
- In most cases, the family *F* of measuring functions is invariant with respect to a given group G of transformations, that depends on the type of measurement we are considering.
- A *G*-invariant pseudo-metric *d<sub>F</sub>* is available for the set *F*, so that we can quantify the difference between the measuring functions in *F*. (pseudo-metric = metric without the property

 $d(x,y)=0\implies x=y)$ 

### Example 1



- $M = S^1$ ,  $V = \{$ the set of pictures $\}$
- Every *f* ∈ *F* is a function associating each point of S<sup>1</sup> with a picture.
- *G* is the set of rotations around the *z*-axis, and  $F \circ G = F$
- We can set  $d_F(f_1, f_2) = \inf_{g \in G} \sup_{\mu \in M} \|f_1(\mu) f_2(g(\mu))\|$

### Example 2



- $M = a \text{ decagon}, V = \mathbb{R}^4$ .
- Every *f* ∈ *F* is a differentiable function taking each point *p* of the sea star (whose body is diffeomorphic to *M*) to a quadruple (*R*(*p*), *r*(*p*), *g*(*p*), *b*(*p*)). *R*(*p*) is the distance of *p* from the center of mass of the sea star, while (*r*(*p*), *g*(*p*), *b*(*p*)) describes its color.
- *G* is the group of the diffeomorphisms from *M* to *M* ( $F \circ G = F$ ).

• We can set 
$$d_F(f_1, f_2) = \inf_{g \in G} \sup_{\mu \in M} \|f_1(\mu) - f_2(g(\mu))\|$$
.

# Our shape pseudo-distance $d_F$ (formal definition)

Assume that the following objects are given:

- A set *M*. Each point  $\mu \in M$  represents a measurement.
- A set V. Each point v ∈ V represents the value taken by a measurement.
- A set *F* of functions from *M* to *V*. Each function *f* ∈ *F* describes a possible set of results for all measurements in *M*.
- A group *G* acting on *M*, such that *F* is invariant with respect to *G* (i.e., for every *f* ∈ *F* and every *g* ∈ *G* we have that *f* ∘ *g* ∈ *F*).
- A pseudo-metric *d<sub>F</sub>* defined on the set *F*, that is invariant under the action of the group *G* (in other words, if *f*<sub>1</sub>, *f*<sub>2</sub> ∈ *F* and *g* ∈ *G* then *f*<sub>2</sub> ∘ *g* ∈ *F* and *d<sub>F</sub>*(*f*<sub>1</sub>, *f*<sub>2</sub>) = *d<sub>F</sub>*(*f*<sub>1</sub>, *f*<sub>2</sub> ∘ *g*)).

We call each pair  $(F, d_F)$  a (pseudo-)metric shape space.

# An interesting case

It often happens that M is a topological space and V is a metric space, endowed with a metric  $d_V$ . In this case the functions in F are assumed to be continuous, and the group G is assumed to be a subgroup of the group of all bicontinuous maps from M to M. As an example, let us think of a CT scanning.



Patrizio Frosini (University of Bologna)

## An interesting case

In this example

- $M = S^1$  represents the topological space of all directions that are orthogonal to a given axis;
- V = ℝ represents the metric space of all possible quantities of matter encountered by the X-ray beam in the considered direction.
- Every *f* ∈ *F* is a function taking each direction in S<sup>1</sup> to the quantity of matter encountered by the X-ray beam along that direction.
- *G* is the group of the rotations of  $S^1$  ( $F \circ G = F$ ).

We can set

$$d_{F}(f_{1}, f_{2}) = \inf_{g \in G} \sup_{\mu \in M} |f_{1}(\mu) - f_{2}(g(\mu))|$$

for  $f_1, f_2 \in F$ .

#### Other interesting cases

If  $V = \mathbb{R}^k$  we can use the shape pseudo-distance

$$d_F(f_1, f_2) = \inf_{g \in G} \sup_{\mu \in M} \|f_1(\mu) - f_2(g(\mu))\|_{\infty}$$

for  $f_1, f_2 \in F$ . The functional  $\sup_{\mu \in M} \|f_1(\mu) - f_2(g(\mu))\|_{\infty}$  quantifies the change in the measurement induced by the transformation *g*.

The pseudo-metric  $d_F$  is produced by the attempt of minimizing this functional, varying the transformation g in the group G, and is called natural pseudo-distance.

More generally, if V is a metric space endowed with the metric  $d_V$ , we can set

$$d_{\mathsf{F}}(f_1, f_2) = \inf_{g \in \mathcal{G}} \sup_{\mu \in \mathcal{M}} d_V\left(f_1(\mu), f_2(g(\mu))\right)$$

for  $f_1, f_2 \in F$ .

#### Other interesting cases

If  $V = \mathbb{R}$  and *M* is a compact subset of  $\mathbb{R}^m$ , we can set

$$d_{\mathsf{F}}(f_1, f_2) = \inf_{g \in G} \left( \int_{\mu \in M} |f_1(\mu) - f_2(g(\mu))|^p \ d\mu \right)^{\frac{1}{p}}$$

after fixing  $p \ge 1$ .

The functional  $\left(\int_{\mu \in M} |f_1(\mu) - f_2(g(\mu))|^p d\mu\right)^{\frac{1}{p}}$  quantifies the change in the measurement induced by the transformation *g*.

If G is the group of all isometries of M,  $d_F$  is invariant under the action of G.

#### An important remark

# Every pseudo-distance that we have shown until now is a particular case of the shape pseudo-metric $d_F$ .





In the rest of this talk we shall illustrate the theoretical advantages of the framework that we have described.

We will assume that M is a topological space and V is a metric space.

These assumptions allow us to require that if two measurements are close to each other (in some reasonable sense), then the values obtained by these measurements are close to each other, too.

The functions in *F* will be assumed to be continuous. The group *G* will be assumed to be a subgroup of the group Hom(M) of all bicontinuous maps from *M* to *M*.

### **Usual assumptions**

Why do we just consider bicontinuous maps from *M* to *M*?

Why couldn't we use, e.g., relations on M?



# A result that suggests not to use relations in our setting

The following result highlights a problem about doing that:

# **Non-existence Theorem**

Let *M* be a Riemannian manifold. Call Hom(M) the set of all bicontinuous maps from *M* to *M*. Let us endow Hom(M) with the uniform convergence metric  $d_{UC}$ :  $d_{UC}(h, h') = \max_{x \in M} d_M(h(x), h'(x))$  for every  $h, h' \in Hom(M)$ , where  $d_M$  is the geodesic distance on *M*. Then  $(Hom(M), d_{UC})$  cannot be embedded in any compact metric space  $(K, d_K)$  endowed with an internal binary operation  $\bullet$  that extends the usual composition  $\circ$  between bicontinuous maps in Hom(M) and commutes with the passage to the limit in *K*. In particular, Hom(M)cannot be embedded into the set of binary relations on *M*.

P. Frosini, C. Landi, *No embedding of the automorphisms of a topological space into a compact metric space endows them with a composition that passes to the limit*, Applied Mathematics Letters, 24 (2011), n. 10, 1654–1657.

# A result that suggests not to use relations in our setting

In plain words, the previous theorem shows that, if our space of measurements M is a Riemannian manifold, no reasonable embedding of the set of all bicontinuous maps from M to M into another compact metric space (K,  $d_K$ ) exists. In particular, there does not exist any reasonable embedding into the space of binary relations on M.

This is due to the fact that any such embedding couldn't preserve the usual composition between bicontinuous maps in Hom(M) and commute with the passage to the limit in K.

### Remark

The previous theorem can be extended to topological spaces that are far more general than manifolds. It is sufficient that they contain a subset U that is homeomorphic to an n-dimensional open ball for some  $n \ge 1$ .

# Some theoretical results about the natural pseudo-distance

Until now, most of the results about the natural pseudo-distance  $d_F$  have been proven for the case when M is a closed manifold, G equals the set Hom(X) of all bicontinuous maps from M to M and the measuring functions take their values in  $V = \mathbb{R}$ . Here we recall three of these results:

# Theorem (for manifolds)

Assume that M is a closed manifold of class  $C^1$  and that  $f_1, f_2 : M \to \mathbb{R}$  are two functions of class  $C^1$ . Then, if d denotes the natural pseudo-distance between  $f_1$  and  $f_2$ , at least one of the following properties holds:

- a positive odd integer m exists, such that m · d equals the distance between a critical value of f<sub>1</sub> and a critical value of f<sub>2</sub>;
- a positive even integer m exists, such that m · d equals the distance between either two critical values of f<sub>1</sub> or two critical values of f<sub>2</sub>.

### Some theoretical results about the natural pseudo-distance

# Theorem (for surfaces)

Assume that M is a closed surface of class  $C^1$  and that  $f_1, f_2 : M \to \mathbb{R}$  are two functions of class  $C^1$ . Then, if d denotes the natural pseudo-distance between  $f_1$  and  $f_2$ , at least one of the following properties holds:

- d equals the distance between a critical value of f<sub>1</sub> and a critical value of f<sub>2</sub>;
- d equals half the distance between two critical values of f1;
- *d* equals half the distance between two critical values of *f*<sub>2</sub>;
- d equals one third of the distance between a critical value of f<sub>1</sub> and a critical value of f<sub>2</sub>.

In plain words, referring to the theorem for manifolds, we have that m = 1, or m = 2, or m = 3.

### Some theoretical results about the natural pseudo-distance

# Theorem (for curves)

Assume that M is a closed curve of class  $C^1$  and that  $f_1, f_2 : M \to \mathbb{R}$  are two functions of class  $C^1$ . Then, if d denotes the natural pseudo-distance between  $f_1$  and  $f_2$ , at least one of the following properties holds:

- d equals the distance between a critical value of f<sub>1</sub> and a critical value of f<sub>2</sub>;
- *d* equals half the distance between two critical values of f<sub>1</sub>;
- *d* equals half the distance between two critical values of *f*<sub>2</sub>.

In plain words, referring to the theorem for manifolds, we have that m = 1, or m = 2.

# Some references about the natural pseudo-distance for topological spaces

- P. Frosini, M. Mulazzani, Size homotopy groups for computation of natural size distances, Bulletin of the Belgian Mathematical Society - Simon Stevin, 6 (1999), 455-464.
- P. Donatini, P. Frosini, Natural pseudodistances between closed topological spaces, Forum Mathematicum, 16 (2004), n. 5, 695-715.
- P. Donatini, P. Frosini, *Natural pseudodistances between closed surfaces*, Journal of the European Mathematical Society, 9 (2007), 331-353.
- P. Donatini, P. Frosini, *Natural pseudodistances between closed curves*, Forum Mathematicum, 21 (2009), n. 6, 981-999.

### Natural pseudo-distance and size functions

- The natural pseudo-distance is usually difficult to compute.
- Lower bounds for the natural pseudo-distance can be obtained by computing the size functions.



#### **Definition of size function**

# Given a topological space *M* and a continuous function $f = (f_1, \ldots, f_k) : M \to \mathbb{R}^k$ ,

#### Lower level sets

For every 
$$u \in \mathbb{R}^k$$
,  $M \langle f \preceq u \rangle = \{x \in M : f(x) \preceq u\}$ .  
 $(u = (u_1, \dots, u_k) \preceq v = (v_1, \dots, v_k) \text{ means } u_j \leq v_j \text{ for every index } j.)$ 

# **Definition (Frosini 1991)**

The Size Function of (M, f) is the function  $\ell$  that takes each pair (u, v) with  $u \prec v$  to the number  $\ell(u, v)$  of connected components of the set  $M\langle f \preceq v \rangle$  that contain at least one point of the set  $M\langle f \preceq u \rangle$ .

# Example of a size function, in the case that the measuring function has only one component

Here the measuring function *f* equals the distance from *C*.



We observe that each size function can be described by giving a set of points (vertices of triangles in figure).

Patrizio Frosini (University of Bologna)

# Example of a size function, in the case that the measuring function has two components

Here the measuring function  $f = (f_1, f_2)$  is given by the coordinates of each point of the curve *M*.



### Some references about size functions

- A. Verri, C. Uras, P. Frosini, M. Ferri, *On the use of size functions for shape analysis*, Biological Cybernetics, 70 (1993), 99–107.
- A. Verri, C. Uras, *Metric-topological approach to shape* representation and recognition, Image and Vision Computing, 14, n. 3 (1996), 189–207
- C. Uras, A. Verri, *Computing size functions from edge maps*, International Journal of Computer Vision, 23, n. 2 (1997), 169–183.
- S. Biasotti, L. De Floriani, B. Falcidieno, P. Frosini, D. Giorgi, C. Landi, L. Papaleo, M. Spagnuolo, *Describing shapes by geometrical-topological properties of real functions*, ACM Computing Surveys, vol. 40 (2008), n. 4, 12:1–12:87.

#### Persistent homology groups and size homotopy groups

Size functions have been generalized by Edelsbrunner and al. to homology in higher degree (i.e., counting the number of holes instead of the number of connected components). This theory is called Persistent Homology:

H. Edelsbrunner, D. Letscher, A. Zomorodian, *Topological persistence and simplification*, Discrete & Computational Geometry, vol. 28, no. 4, 511–533 (2002).

Size functions have been also generalized to size homotopy groups:

P. Frosini, M. Mulazzani, *Size homotopy groups for computation of natural size distances*, Bulletin of the Belgian Mathematical Society, vol. 6, no. 3, 455–464 (1999).

# A metric to compare size functions

Several metrics to compare size functions exist.

One of them is the multidimensional matching distance (MMD)  $D_{match}$ . It is based on a foliation of the space where the parameters u, v of size functions vary. The foliation is made by half-planes. On each leaf of this foliation the multidimensional size function reduces to a 1-dimensional size function. The MMD is defined as the sup of the ordinary matching distance, varying the half-plane in the foliation, after a suitable normalization. Here we skip its formal definition.

Biasotti, S., Cerri, A., Frosini, P., Giorgi, D., Landi, C., *Multidimensional size functions for shape comparison*, Journal of Mathematical Imaging and Vision, vol. 32 (2008), n. 2, 161-179.

An analogous distance exists for persistent homology groups: A. Cerri, B. Di Fabio, M. Ferri, P. Frosini, C. Landi, *Betti numbers in multidimensional persistent homology are stable functions*, Mathematical Methods in the Applied Sciences (accepted for publication), available at http://amsacta.unibo.it/2923/.

# Stability of the multidimensional matching distance between size functions

An important property of the multidimensional matching distance is that size functions (and the ranks of persistent homology groups) are stable with respect to it. This stability can be expressed by means of the inequality

# $D_{match} \leq d_F$

where  $d_F$  denotes the natural pseudo-distance.

Therefore, the multidimensional matching distance between size functions gives a lower bound for the natural pseudo-distance.

In plain words, the stability of  $D_{match}$  means that a small change of the measuring function induces a small change of the size function.

#### **Comparing shapes**

#### Stability allows us to use the matching distance to compare shapes:



	T	贫	Ŕ	A	P	A.	Ý	P
8.	0.0000	0.0181	0.1411	0.1470	0.1325	0.1287	0.1171	0.1187
2	0.0000	0.0003	0.0025	0.0026	0.0023	0.0022	0.0020	0.0021
Å	0.0181	0.0000	0.1419	0.1478	0.1304	0.1265	0.1171	0.1187
	0.0003	0.0000	0.0026	0.0026	0.0023	0.0022	0.0020	0.0021
Å	0.1411	0.1419	0.0000	0.0137	0.1583	0.1370	0.1127	0.1017
	0.0025	0.0025	0.0000	0.0002	0.0028	0.0024	0.0020	0.0018
\$	0.1470	0.1478	0.0137	0.0000	0.1533	0.1381	0.1137	0.1021
T	0.0026	0.0026	0.0002	0.0000	0.0027	0.0024	0.0020	0.0018
- Re	0.1325	0.1304	0.1583	0.1533	0.0000	0.0921	0.0588	0.1000
7	0.0023	0.0023	0.0028	0.0027	0.0000	0.0014	0.0016	0.0017
ngh.	0.1287	0.1265	0.1370	0.1381	0.0921	0.0000	0.1069	0.1048
1	0.0022	0.0022	0.0024	0.0024	0.0014	0.0000	0.0019	0.0018
3.	0.1171	0.1171	0.1127	0.1137	0.0588	0.1069	0.0000	0.0350
8	0.0020	0.0020	0.0020	0.0020	0.0016	0.0019	0.0000	0.0006
2	0.1187	0.1187	0.1017	0.1021	0.1000	0.1048	0.0350	0.0000
F	0.0021	0.0021	0.0018	0.0018	0.0017	0.0018	0.0006	0.0000

Upper lines refer to the 2-dimensional matching distance associated with a suitable function  $f = (f_1, f_2)$  taking values in  $\mathbb{R}^2$ . Lower lines refer to the maximum between the 1-dimensional matching distances associated with  $f_1$  and  $f_2$ , respectively.

Patrizio Frosini (University of Bologna)

### The idea of G-invariant size function

Classical size functions and persistent homology are not tailored on the group G. In some sense, they are tailored on the group Hom(M) of all bicontinuous maps from M to M.

In order to obtain better lower bounds for the pseudo-distance  $d_F$  we need to adapt persistent homology, and to consider *G*-invariant persistent homology.

Roughly speaking, the main idea consists in defining size functions and persistent homology by means of a set of chains that is invariant under the action of G.

We skip the details of this procedure, that produces better lower bounds for the shape pseudo-distance  $d_F$ .

# Uniqueness of models with respect to size functions: the case of curves

We have seen that size functions allow to obtain lower bounds for the natural pseudo-distance. What about upper bounds? This problem is far more difficult.

However, the following statement holds:

# Theorem

Let  $f = (f_1, f_2), f' = (f'_1, f'_2) : S^1 \to \mathbb{R}^2$  be "generic" functions from  $S^1$  to  $\mathbb{R}^2$ . If the size functions of the four pairs of measuring functions  $(\pm f_1, \pm f_2), (\pm f'_1, \pm f'_2)$  (with corresponding signs) coincide, then there exists a  $C^1$ -diffeomorphism  $h : S^1 \to S^1$  such that  $f' \circ h = f$ . Moreover, it is unique.

P. Frosini, C. Landi, *Uniqueness of models in persistent homology: the case of curves*, Inverse Problems, 27 (2011), 124005.

Patrizio Frosini (University of Bologna)

#### Conclusions

In this talk we have illustrated a general metric scheme for geometrical shape comparison. The main idea of this model is that in shape comparison objects are not accessible directly, but only via measurements made by an observer. It follows that the comparison of two shapes is usually based on a family F of functions, which are defined on a topological space M and take values in a metric space (V, d). Each function in F represents a measurement obtained via a measuring instrument and, for this reason, it is called a "measuring function". In most cases, the set F of measuring functions is invariant with respect to a given group G of transformations, that depends on the type of measurement we are considering.

After endowing F with a natural pseudo-metric, we have presented some examples and results, motivating the use of this theoretical framework.

