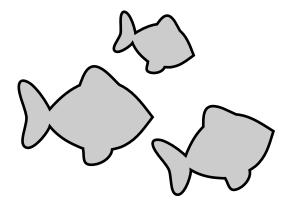
#### Introduction to Shape Manifolds

Geometry of Data

September 27, 2022

#### What is Shape?

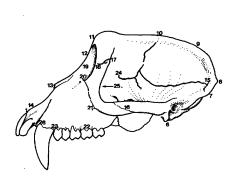


Shape is the geometry of an object modulo position, orientation, and size.

#### Geometry Representations

- Landmarks (key identifiable points)
- Boundary models (points, curves, surfaces, level sets)
- Interior models (medial, solid mesh)
- Transformation models (splines, diffeomorphisms)

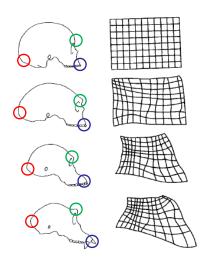
#### Landmarks



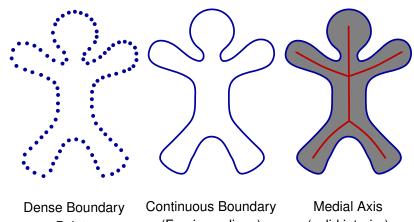
From Dryden & Mardia, 1998

- A landmark is an identifiable point on an object that corresponds to matching points on similar objects.
- This may be chosen based on the application (e.g., by anatomy) or mathematically (e.g., by curvature).

# Landmark Correspondence



# More Geometry Representations

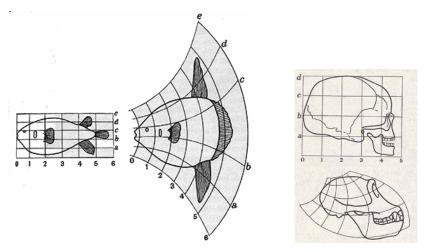


**Points** 

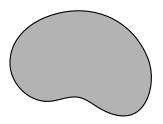
(Fourier, splines)

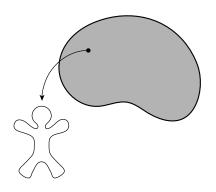
(solid interior)

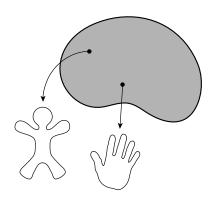
#### **Transformation Models**

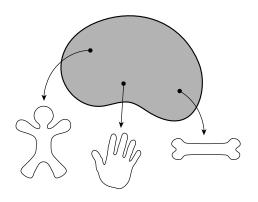


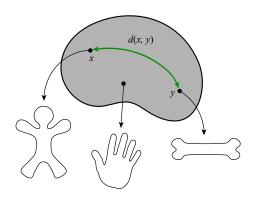
From D'Arcy Thompson, On Growth and Form, 1917.







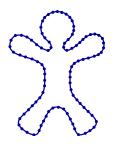




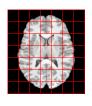
A metric space structure provides a comparison between two shapes.

# Examples: Shape Spaces

Kendall's Shape Space

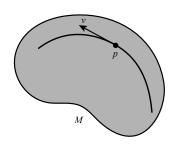


# Space of Diffeomorphisms

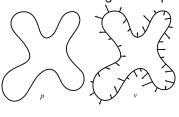




# **Tangent Spaces**



Infinitesimal change in shape:



A **tangent vector** is the velocity of a curve on M.

#### Shape Equivalences

Two geometry representations,  $x_1, x_2$ , are **equivalent** if they are just a translation, rotation, scaling of each other:

$$x_2 = \lambda R \cdot x_1 + v,$$

where  $\lambda$  is a scaling, R is a rotation, and v is a translation.

In notation:  $x_1 \sim x_2$ 

#### **Equivalence Classes**

The relationship  $x_1 \sim x_2$  is an **equivalence** relationship:

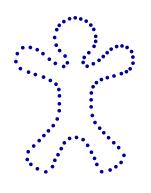
- ▶ Reflexive:  $x_1 \sim x_1$
- Symmetric:  $x_1 \sim x_2$  implies  $x_2 \sim x_1$
- ► Transitive:  $x_1 \sim x_2$  and  $x_2 \sim x_3$  imply  $x_1 \sim x_3$

We call the set of all equivalent geometries to x the equivalence class of x:

$$[x] = \{y : y \sim x\}$$

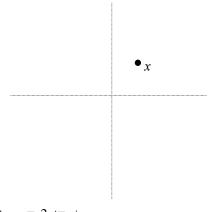
he set of all equivalence classes is our shape space.

#### Kendall's Shape Space

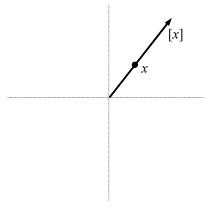


- Define object with k points.
- ▶ Represent as a vector in  $\mathbb{R}^{2k}$ .
- Remove translation, rotation, and scale.
- End up with complex projective space,  $\mathbb{CP}^{k-2}$ .

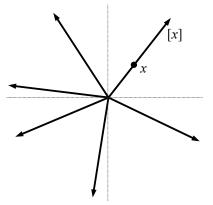
What do we get when we "remove" scaling from  $\mathbb{R}^2$ ?



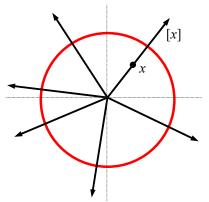
What do we get when we "remove" scaling from  $\mathbb{R}^2$ ?



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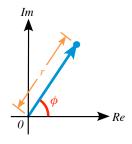
What do we get when we "remove" scaling from  $\mathbb{R}^2$ ?



#### Constructing Kendall's Shape Space

- Consider planar landmarks to be points in the complex plane.
- ▶ An object is then a point  $(z_1, z_2, ..., z_k) \in \mathbb{C}^k$ .
- ▶ Removing **translation** leaves us with  $\mathbb{C}^{k-1}$ .
- How to remove scaling and rotation?

#### Scaling and Rotation in the Complex Plane



Recall a complex number can be written as  $z=re^{i\phi}$ , with modulus r and argument  $\phi$ .

Complex Multiplication:

$$se^{i\theta} * re^{i\phi} = (sr)e^{i(\theta+\phi)}$$

Multiplication by a complex number  $se^{i\theta}$  is equivalent to scaling by s and rotation by  $\theta$ .

#### Removing Scale and Rotation

Multiplying a centered point set,  $\mathbf{z} = (z_1, z_2, \dots, z_{k-1})$ , by a constant  $w \in \mathbb{C}$ , just rotates and scales it.

Thus the shape of z is an equivalence class:

$$[\mathbf{z}] = \{(wz_1, wz_2, \dots, wz_{k-1}) : \forall w \in \mathbb{C}\}$$

This gives complex projective space  $\mathbb{CP}^{k-2}$  – much like the sphere comes from equivalence classes of scalar multiplication in  $\mathbb{R}^n$ .

#### Alternative: Shape Matrices

Represent an object as a real  $d \times k$  matrix.

#### Preshape process:

- Remove translation: subtract the row means from each row (i.e., translate shape centroid to 0).
- Remove scale: divide by the Frobenius norm.

# Orthogonal Procrustes Analysis

#### **Problem:**

Find the rotation  $R^*$  that minimizes distance between two  $d \times k$  matrices A, B:

$$R^* = \arg\min_{R \in SO(d)} ||RA - B||^2$$

#### Solution:

Let  $U\Sigma V^T$  be the SVD of  $BA^T$ , then

$$R^* = UV^T$$

#### Geodesics in 2D Kendall Shape Space

#### Let A and B be $2 \times k$ shape matrices

- 1. Remove centroids from *A* and *B*
- 2. Project onto sphere:  $A \leftarrow A/\|A\|$ ,  $B \leftarrow B/\|B\|$
- 3. Align rotation of *B* to *A* with OPA
- 4. Now a geodesic is simply that of the sphere,  $S^{2k-1}$

#### Where to Learn More

#### **Books**

- Dryden and Mardia, Statistical Shape Analysis, Wiley, 1998.
- Small, The Statistical Theory of Shape, Springer-Verlag, 1996.
- Kendall, Barden and Carne, Shape and Shape Theory, Wiley, 1999.
- Krim and Yezzi, Statistics and Analysis of Shapes, Birkhauser, 2006.