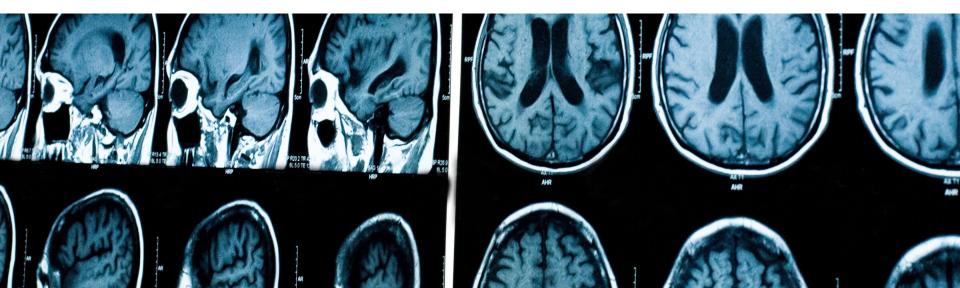


Welcome to ECE 594n

Geometric Machine Learning for Biomedical Imaging & Shape Analysis

Nina Miolane, Assistant Professor @ BioShape Lab

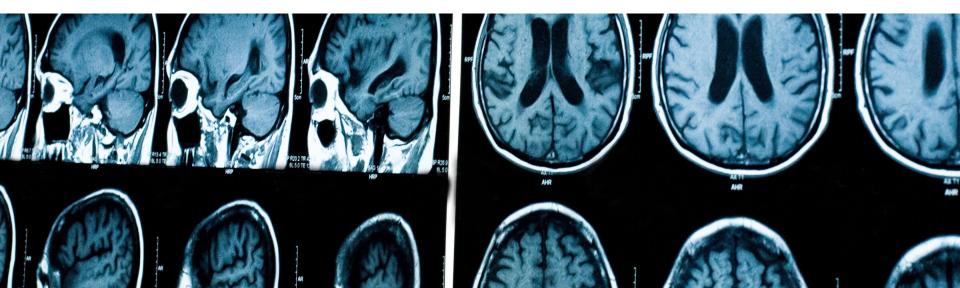




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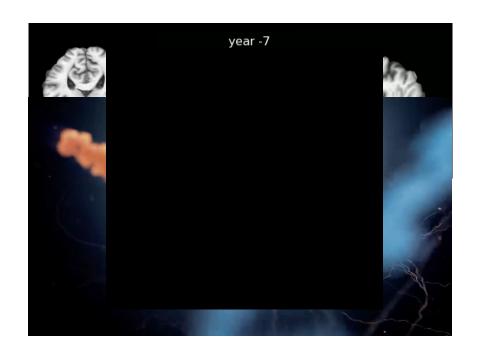
The Many Shapes of Alzheimer's Disease

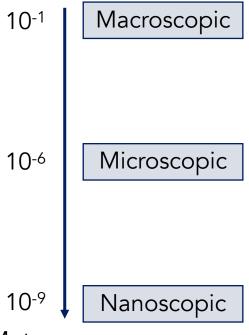
The Many Shapes of Alzheimer's Disease

Macroscopic

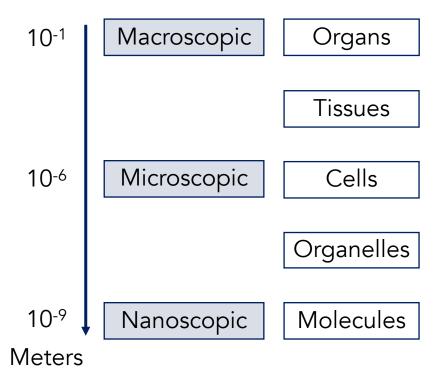
Microscopic

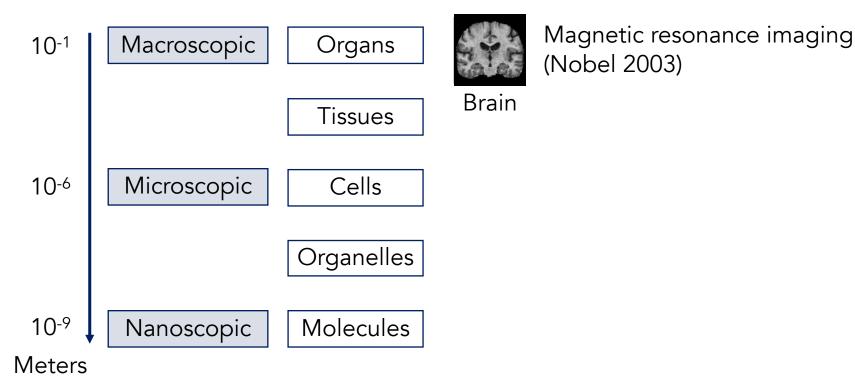
Nanoscopic

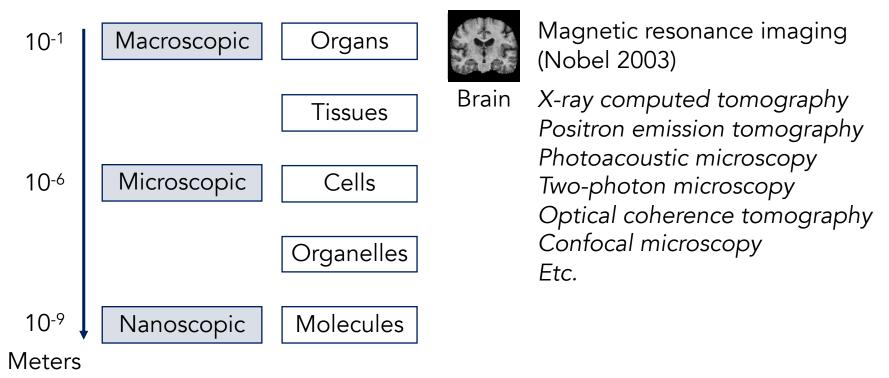


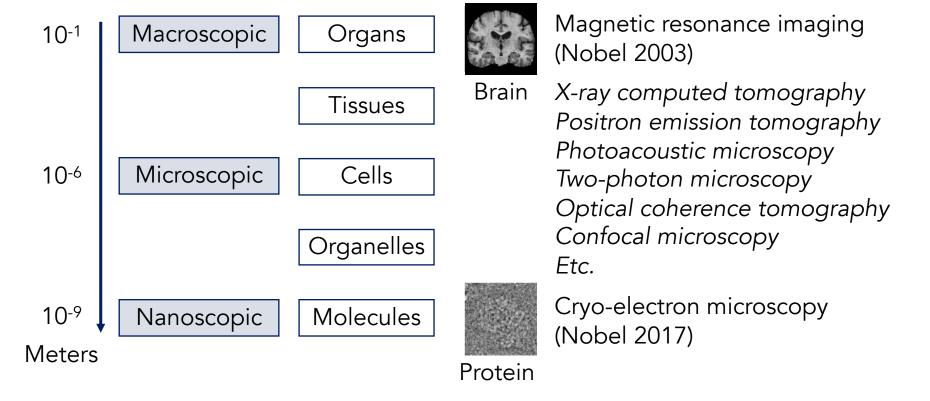


Meters









From BioShapes to Biological Insights

From BioShapes to Biological Insights

Biophysics

Healthy/pathological state Function

→ BioShapes

From BioShapes to Biological Insights

Biophysics

Healthy/pathological state Function

→ BioShapes

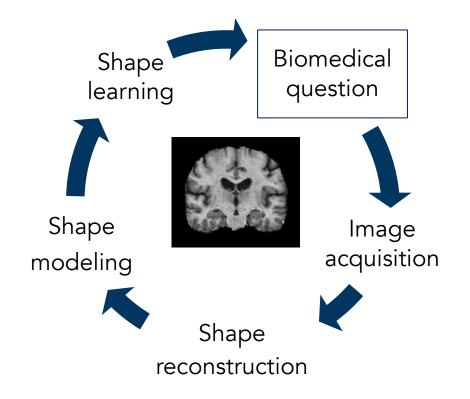
Biomedical insights

← BioShapes

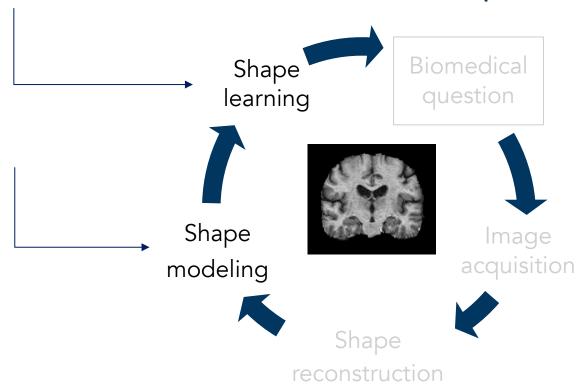
Geometric

Machine Learning

Geometric Machine Learning for Biomedical Imaging & Shape Analysis

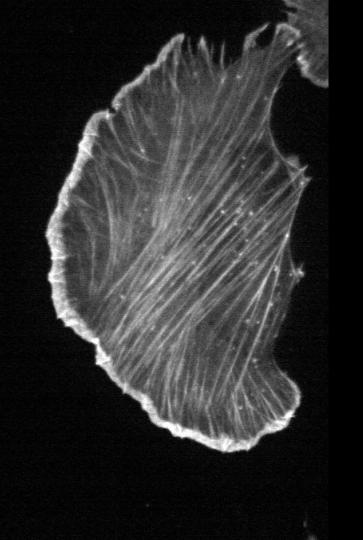


Geometric Machine Learning for Biomedical Imaging & Shape Analysis

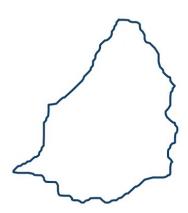


- Mathematical...
- Computational...
- Statistical...

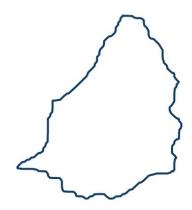
...shape models



Translation



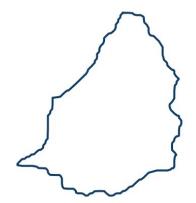
Translation



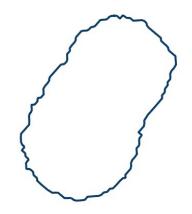
Shapes Equivalence classes

= Elements of "Quotient space" Q

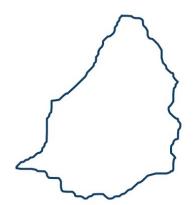
Translation



Smooth deformation

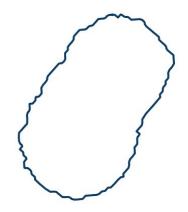


Translation



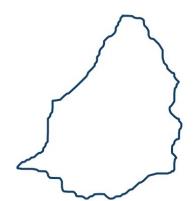
Shapes Equivalence classes = Elements of "Quotient space" Q

Smooth deformation



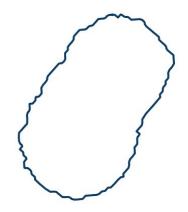
Shapes \bigcirc Deformations = Elements of "Lie group" G

Translation



Shapes Equivalence classes
= Elements of "Quotient space" Q

Smooth deformation



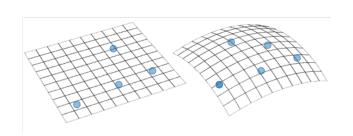
Shapes → Deformations = Elements of "Lie group" *G*

= "Manifolds"

Computations on Manifolds

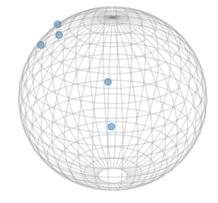
Computations on Manifolds

Computing with data on curved spaces



Data on a vector space

Data on a manifold



Example:
Data on the sphere

Statistics and Machine Learning on Manifolds

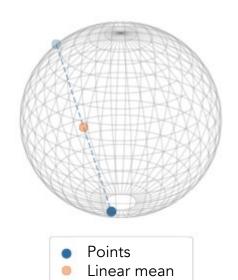
Statistics and Machine Learning on Manifolds

• Traditional statistics and machine learning (ML) fail

```
from geomstats.geometry.hypersphere
  import Hypersphere

sphere = Hypersphere(dim=2)
points = sphere.random_uniform(
    n_samples=2)

linear_mean = gs.sum(
    points, axis=0) / n_samples
```



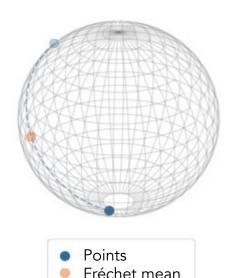
Statistics and Machine Learning on Manifolds

Need geometric statistics and machine learning (ML)

```
from geomstats.learning.frechet_mean import \
    FrechetMean

estimator = FrechetMean(metric=sphere.metric)
estimator.fit(points)

frechet_mean = estimator.estimate_
```



Outline of ECE 594 N

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- 1. (Geometry): Differential Geometry for Engineers
- 2. (Shapes): Computational Representations of Shapes
- 3. (Machine Learning): Geometric Machine Learning

With applications from cutting-edge research in biomedicine.

Software: Geomstats

Software: Geomstats

• Computations, statistics and machine learning on manifolds

1. Instantiate manifold of interest

```
sphere = Hypersphere(dim=2)
```

2. Apply machine learning method

```
estimator = FrechetMean(metric=sphere.metric)
estimator.fit(points)
```

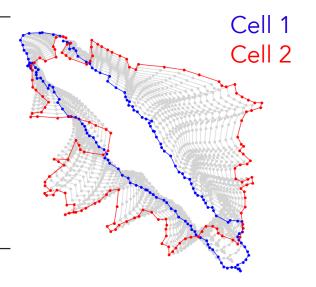
Run Operations on 20+ Manifolds

Run Operations on 20+ Manifolds

```
from geomstats.geometry.discrete_curves \
    import R2, DiscreteCurves

curves = DiscreteCurves(R2)
metric = curves.square_root_velocity_metric

geodesic = metric.geodesic(
    initial_curve=cells_shape[i],
    end_curve=cells_shape[j])
```

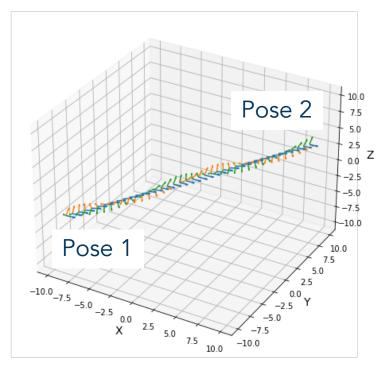


Run Operations on 20+ Manifolds

```
from geomstats.geometry.special_euclidean \
    import SpecialEuclidean

se3 = SpecialEuclidean(n=3, point_type='vector')
metric = se3.left_canonical_metric

initial_point = se3.identity
initial_tangent_vec = gs.array(
    [1.8, 0.2, 0.3, 3., 3., 1.])
geodesic = metric.geodesic(
    initial_point=initial_point,
    initial_tangent_vec=initial_tangent_vec)
```



...Statistics and Machine Learning

	Statistics	Machine Learning	•••
Riemannian		(2019)	
Affine			
Stratified spaces	(2017-18)	(2020)	
Lie groups	(2015)		
Quotient spaces	(2017-21)		
Subriemannian	(2015)		

Miolane, Pennec: Computing bi-invariant pseudo-metrics on Lie groups for consistent statistics (2015).

Miolane, Pennec: A survey of mathematical structures for extending 2D neurogeometry to 3D image processing (2015).

Miolane, Holmes, Pennec: Template shape estimation: correcting an asymptotic bias (2017).

Miolane, Holmes, Pennec: Topologically constrained template (2018).

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Questions?

