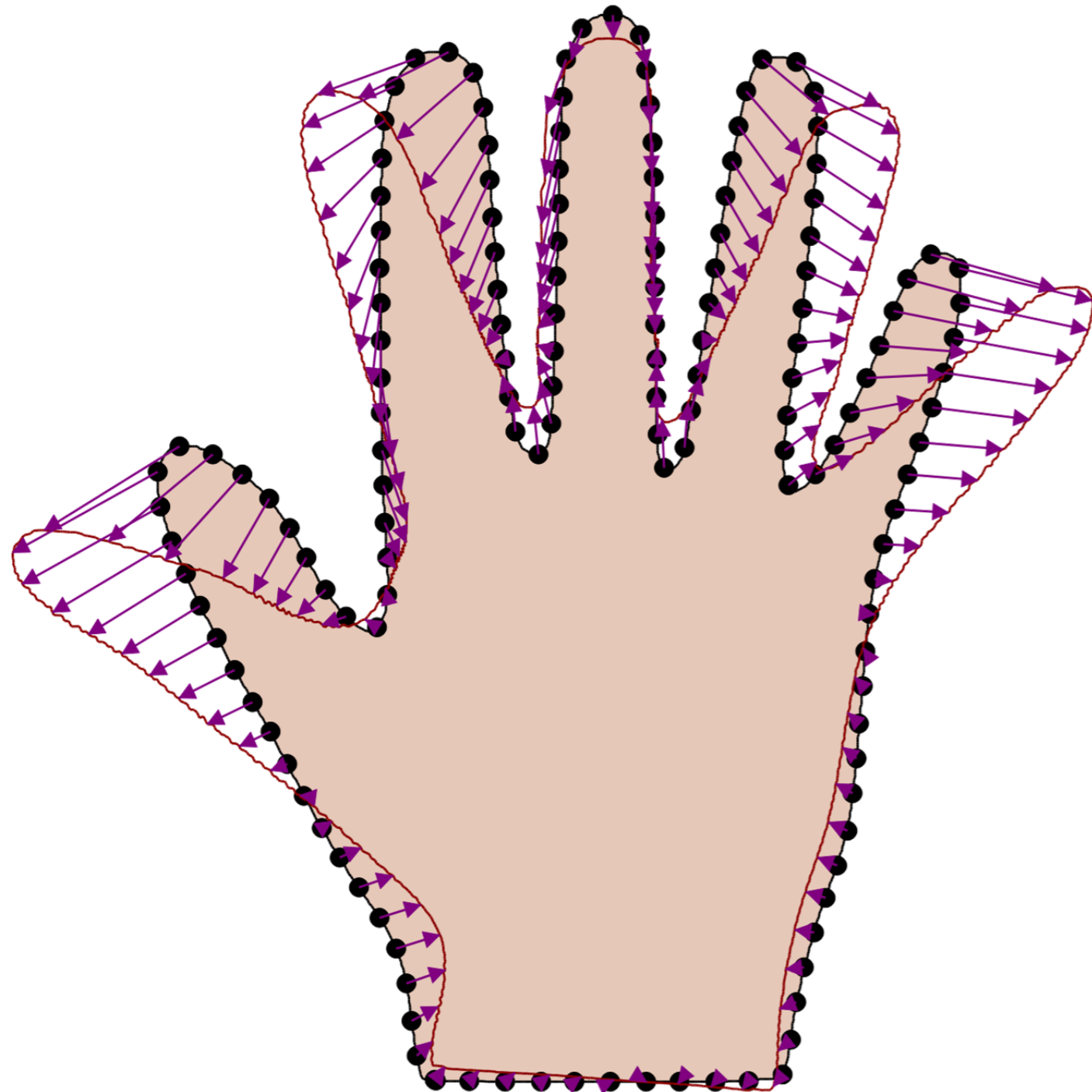


**University
of Basel**

Learning shape variations
from data

Correspondence

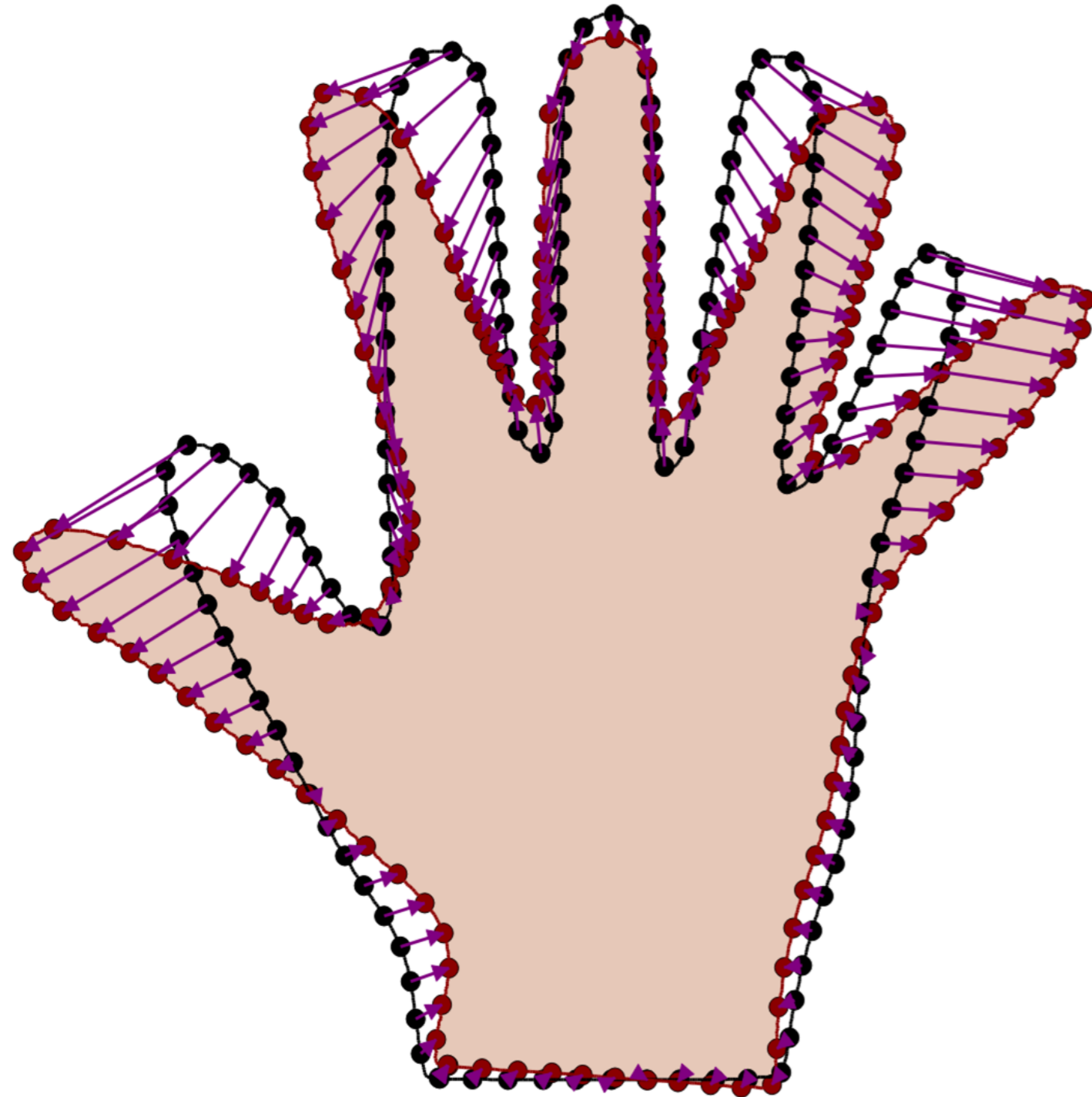


We can describe a shape as

$$\Gamma = \{x + u(x) \mid x \in \Gamma_R\}$$

for some $u : \Gamma_R \rightarrow \mathbb{R}^2$.

Correspondence



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$$\Gamma = \{x + u(x) \mid x \in \Gamma_R\}$$

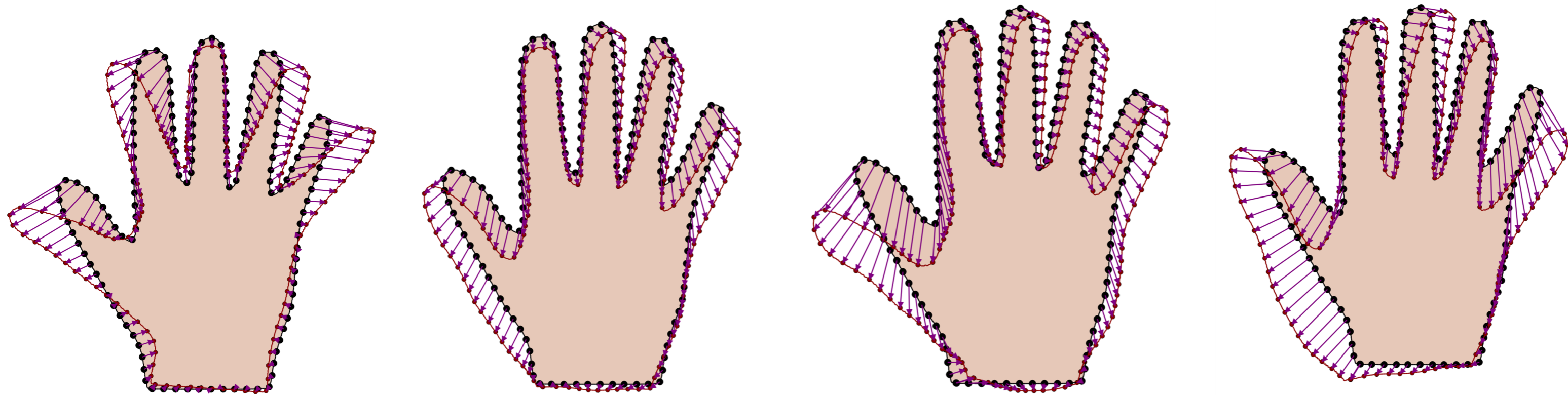
for some $u : \Gamma_R \rightarrow \mathbb{R}^2$.

u establishes **correspondence** between the points of the shapes.

Correspondence

Correspondence makes shapes comparable

- We can measure differences and compute statistics



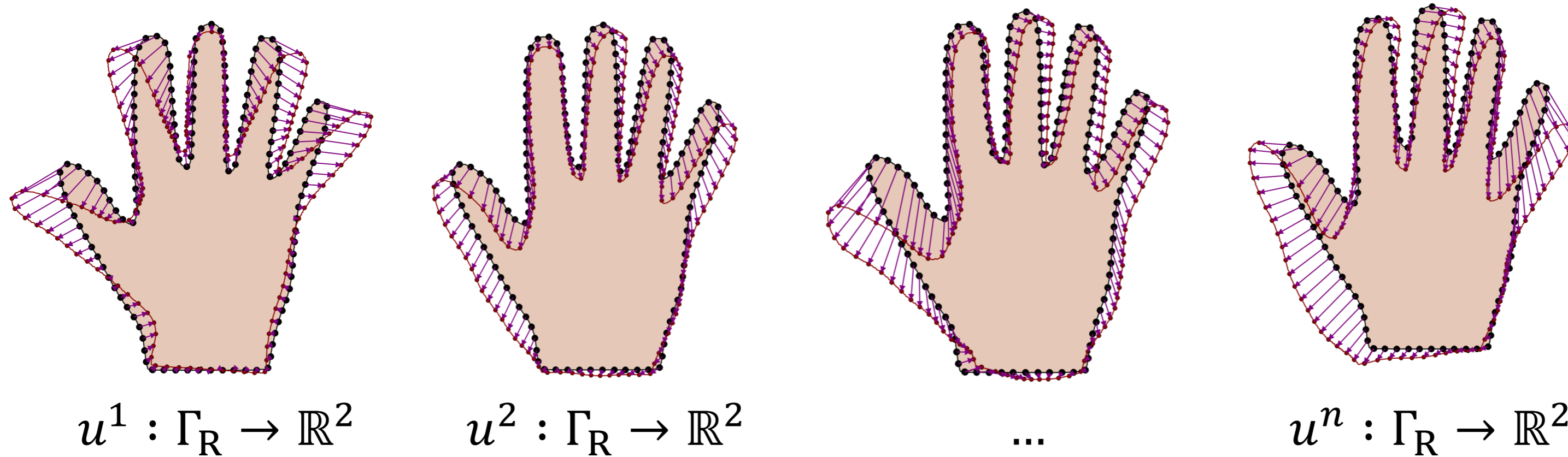
$$u^1 : \Gamma_R \rightarrow \mathbb{R}^2$$

$$u^2 : \Gamma_R \rightarrow \mathbb{R}^2$$

...

$$u^n : \Gamma_R \rightarrow \mathbb{R}^2$$

Learning shape variability

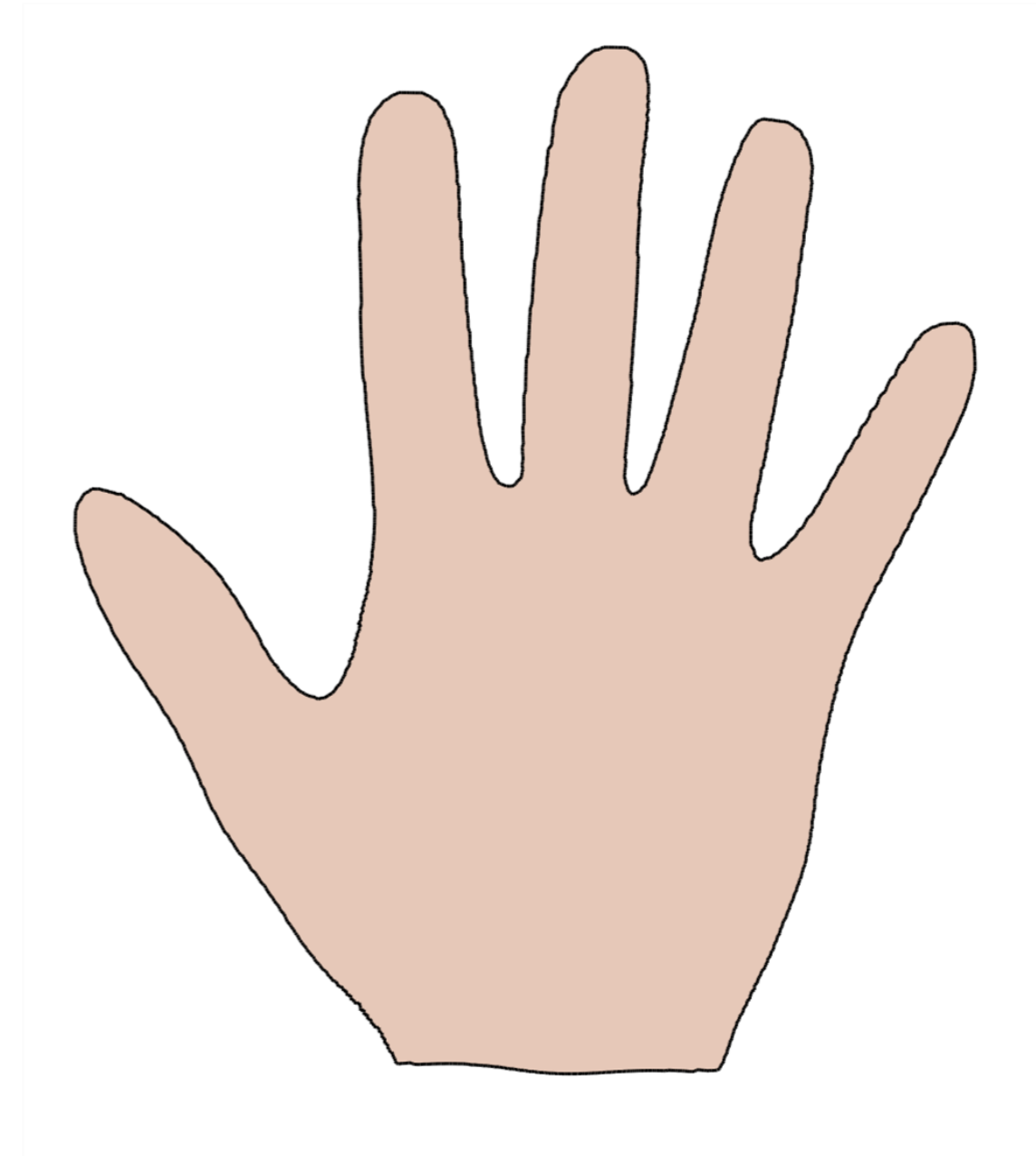


$$\mu(x) = \bar{u}(x) = \frac{1}{n} \sum_{i=1}^n u^i(x)$$
$$k(x, x') = \frac{1}{n-1} \sum_i^n (u^i(x) - \bar{u}(x))(u^i(x') - \bar{u}(x'))^T$$

Building statistical shape models

1. Define a reference shape
2. Find deformations $\{u^1, \dots, u^n\}$ from example data
3. Estimate mean μ and covariance function k from $\{u^1, \dots, u^n\}$
4. Define $u \sim GP(\mu, k)$
5. Final shape model:
$$\Gamma = \{x + u(x) \mid x \in \Gamma_R\}$$

Open problem: How do we find the deformations?



Mean hand shape