

Machine Visual Screening

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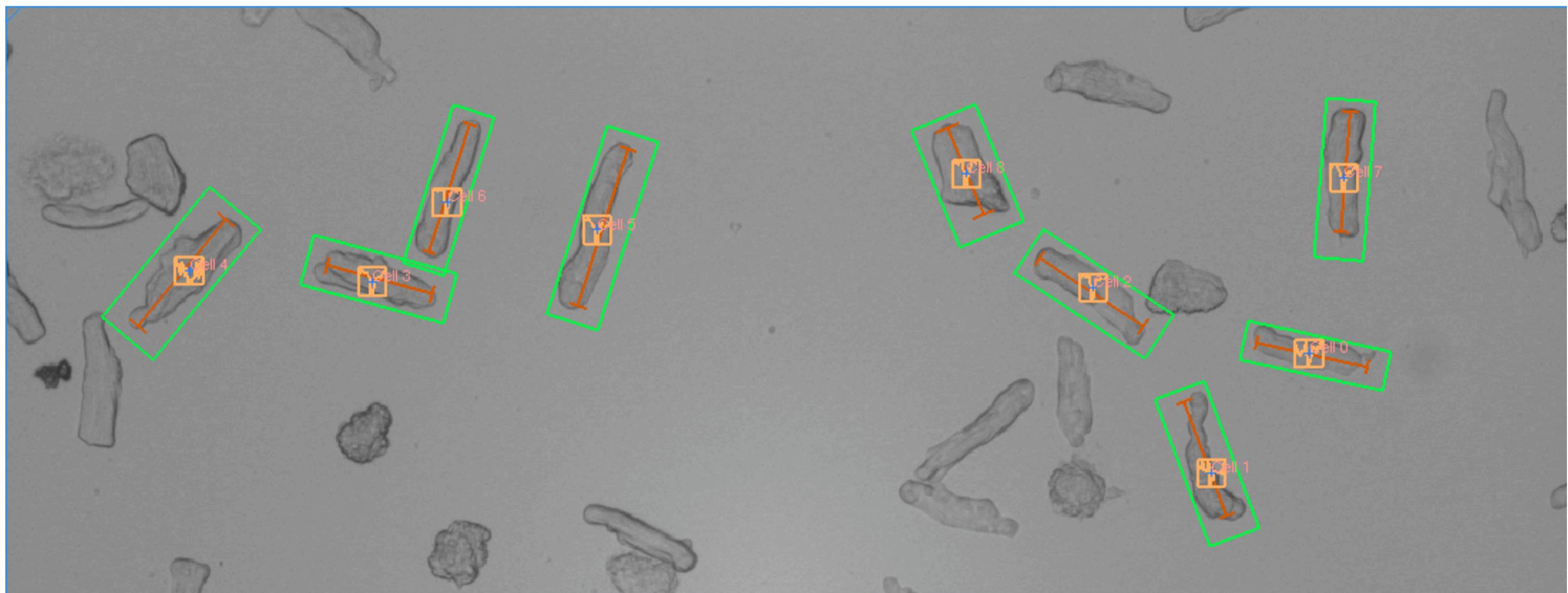
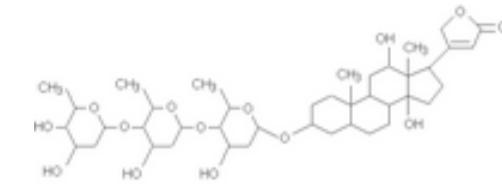
Outline of talk

- Motivation
- Why is it hard ?
- Why is interesting ?

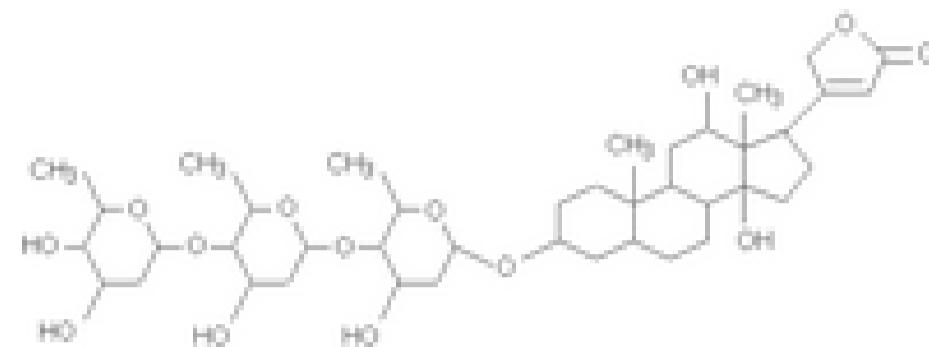
From bedside to cells



One of the few treatments for CHF Digitoxin

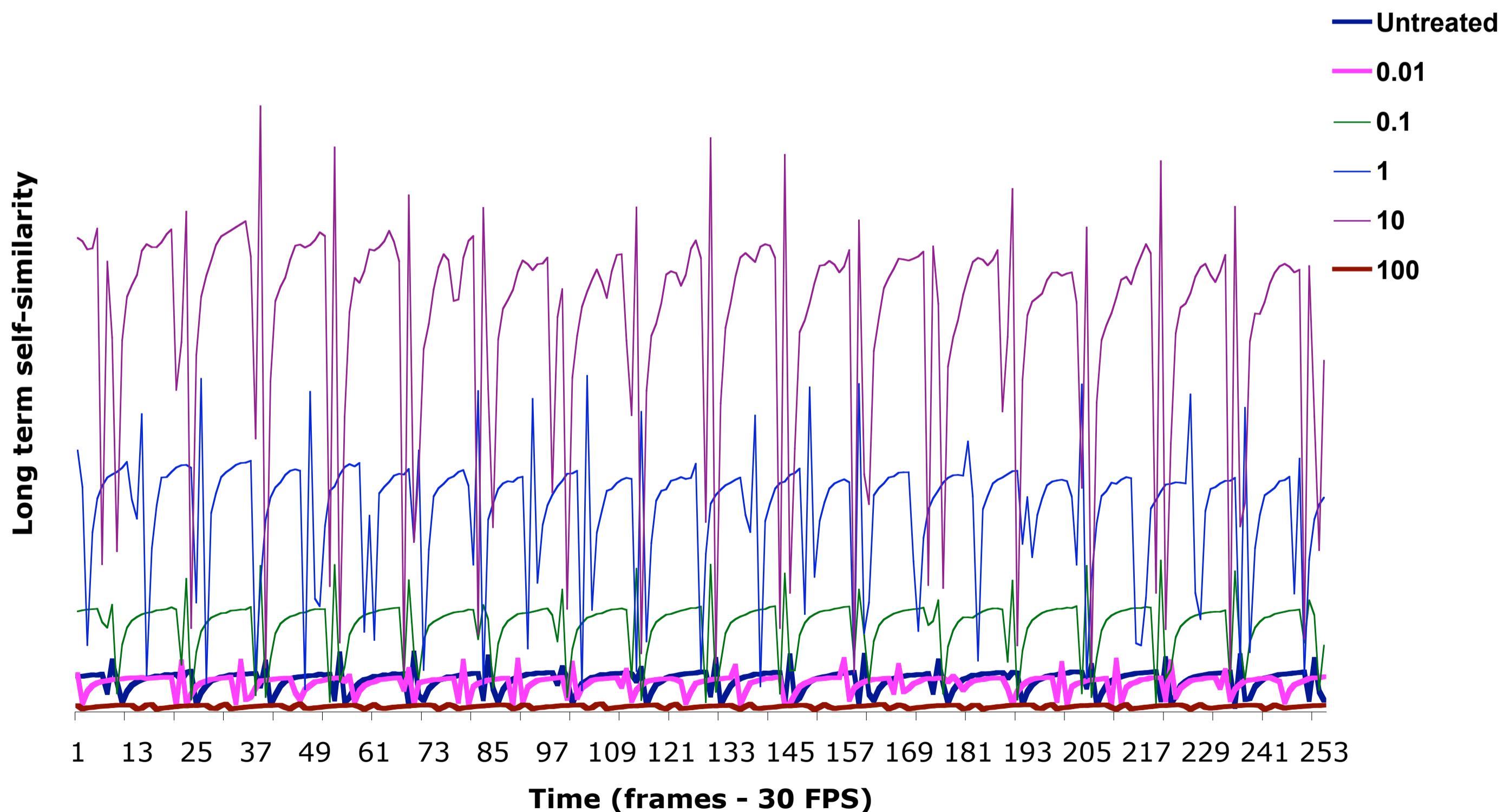


Digitoxin

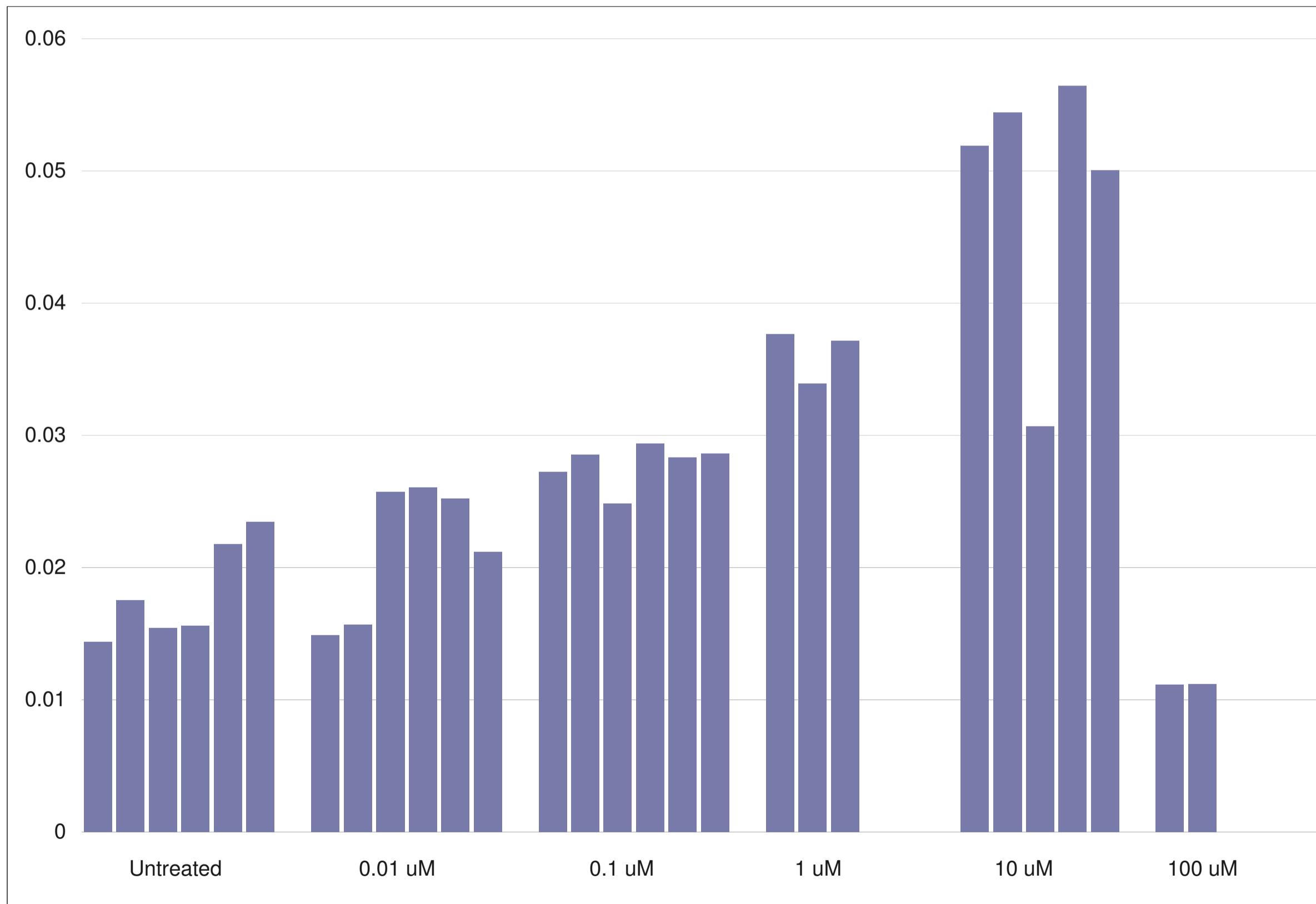


Digitoxin Paced Cardiomiocyte Dose Experiments

Whole Plate Readout (uM)

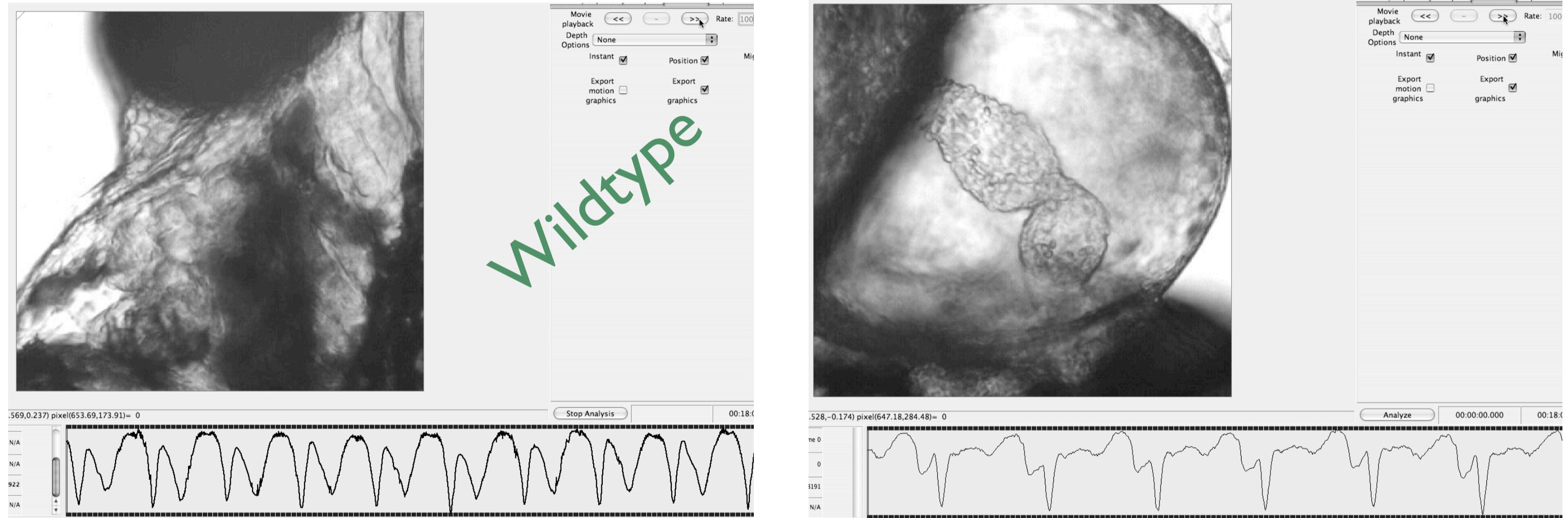


Geometric vs Morphometric

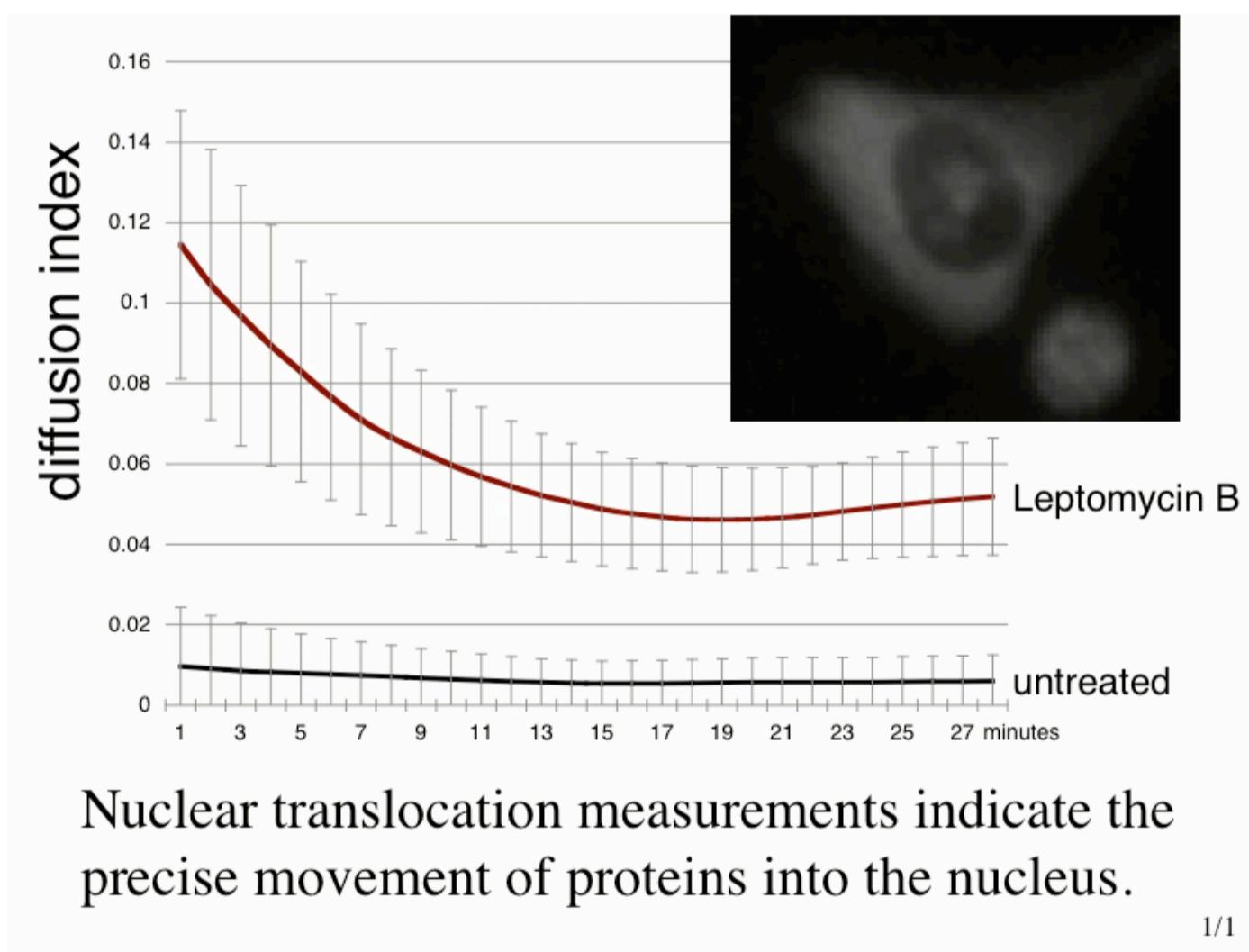


Same Organ Different Species

Same Algorithm Same Parameters

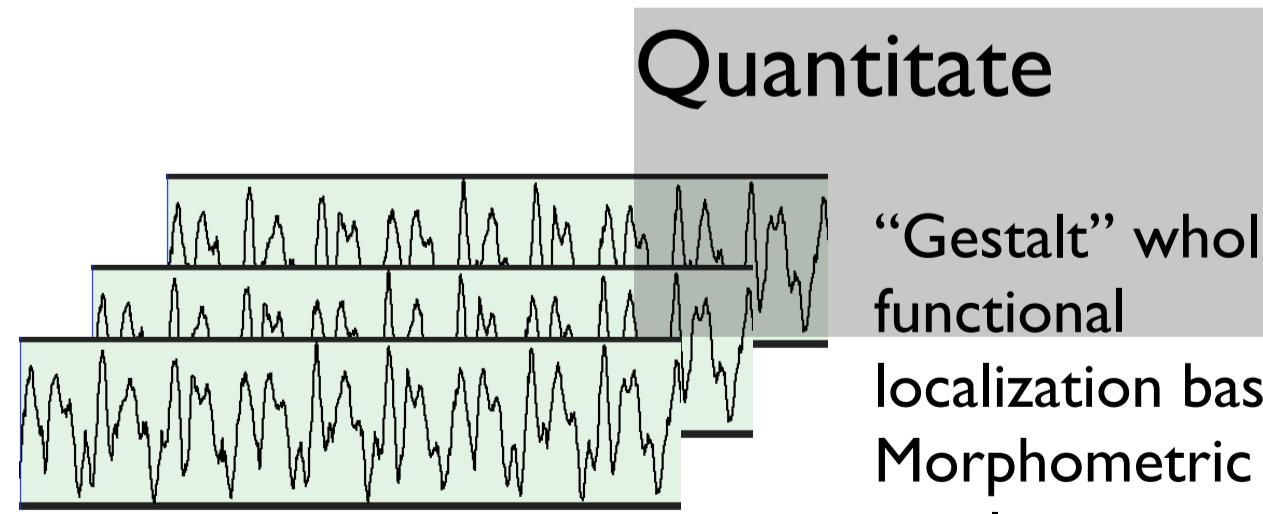
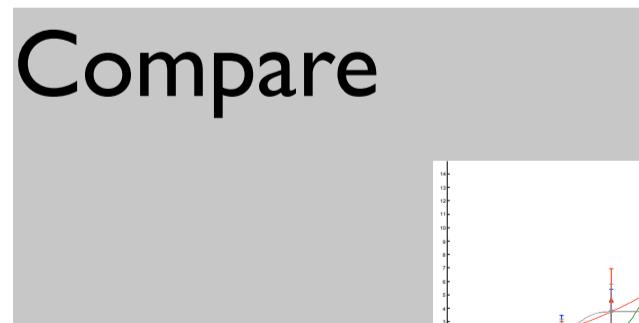
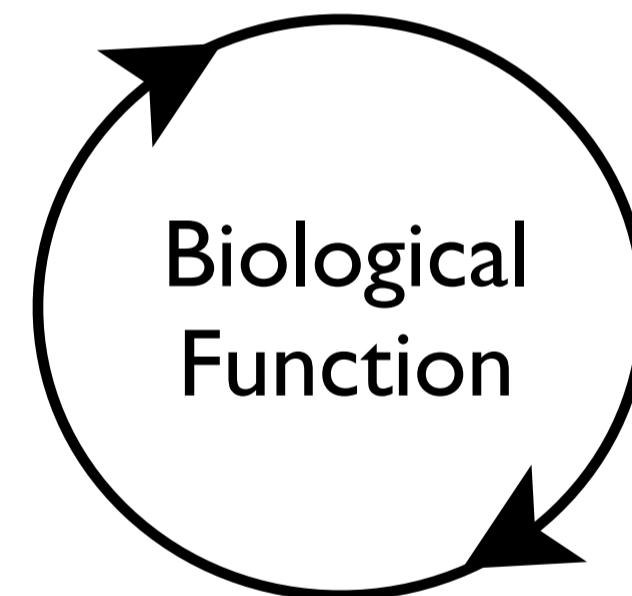
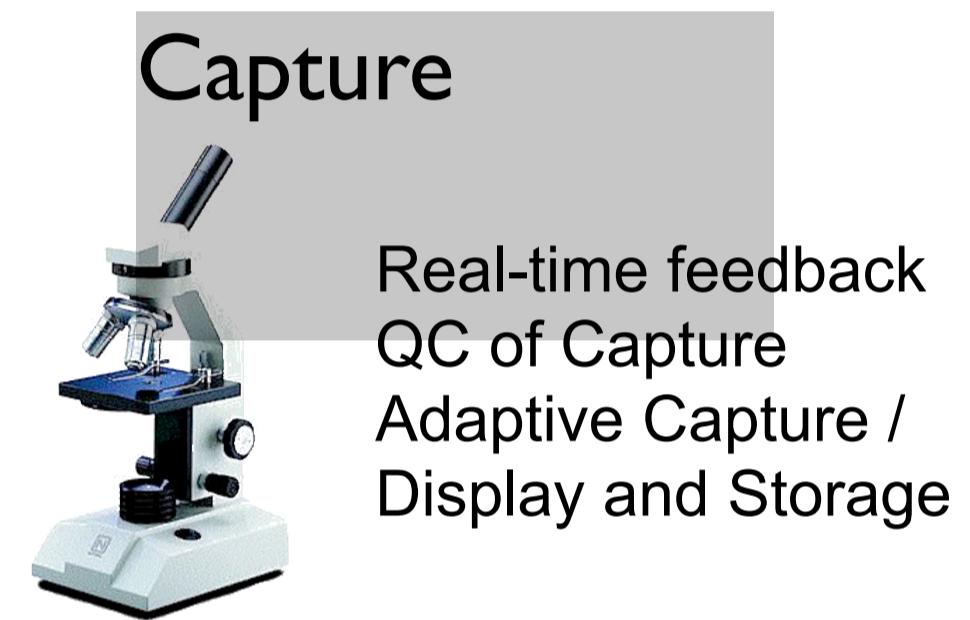
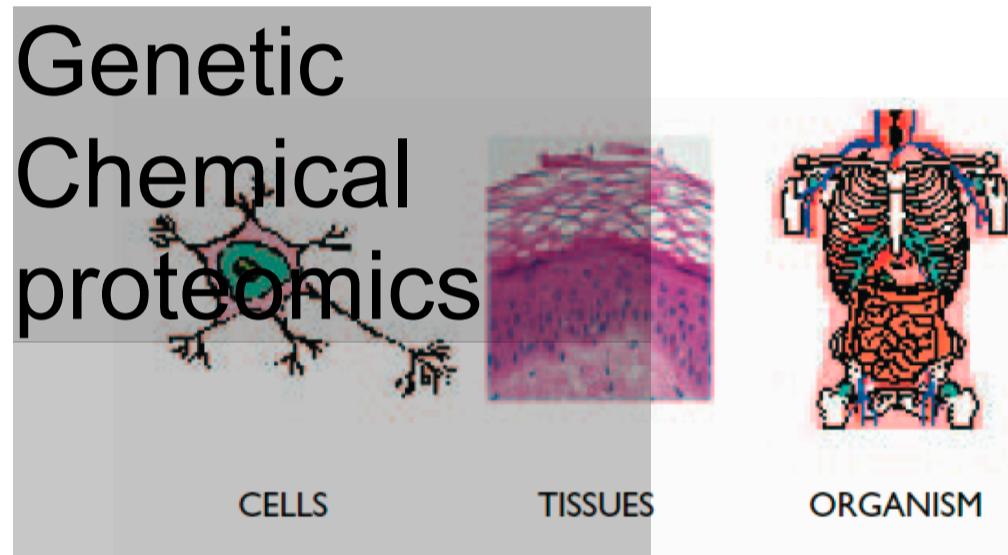


Movement is the phenotype



Machine Visual Screening

Better screening...faster



History

Industry

Thresholding/Binary

Scale Space

Assisted Tracking

Differences (DIAS)

Average Template

Academic Research

cell migration: Lauffenberger

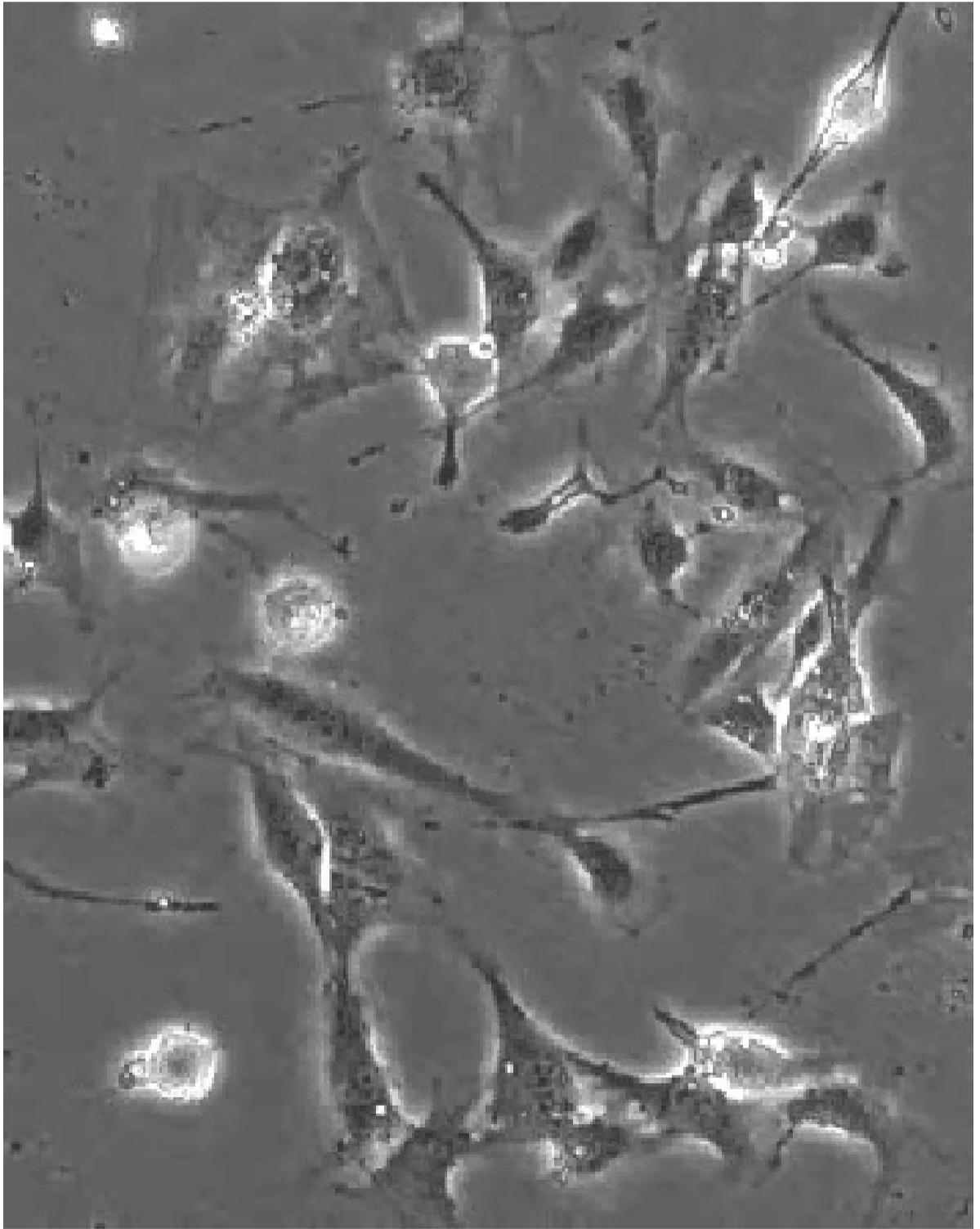
Mathematical Modeling: Blum, Zucker,

Random walks: Brown, Turing

Population Alignment:Lutscher

Phenomological .vs.biophysical: Dunn / brown

Why is it hard ?



What
where &
when

Sampling

To optimize the balance between sampling rate and data volume, *attentive capture* continuously modifies temporal sampling parameters (i.e. frame rate, shutter speed, exposure, etc...) based on events of interest in the scene.

Attribute of interest

Under sampled

Over sampled

Attentive capture

Morphological Change Over Time

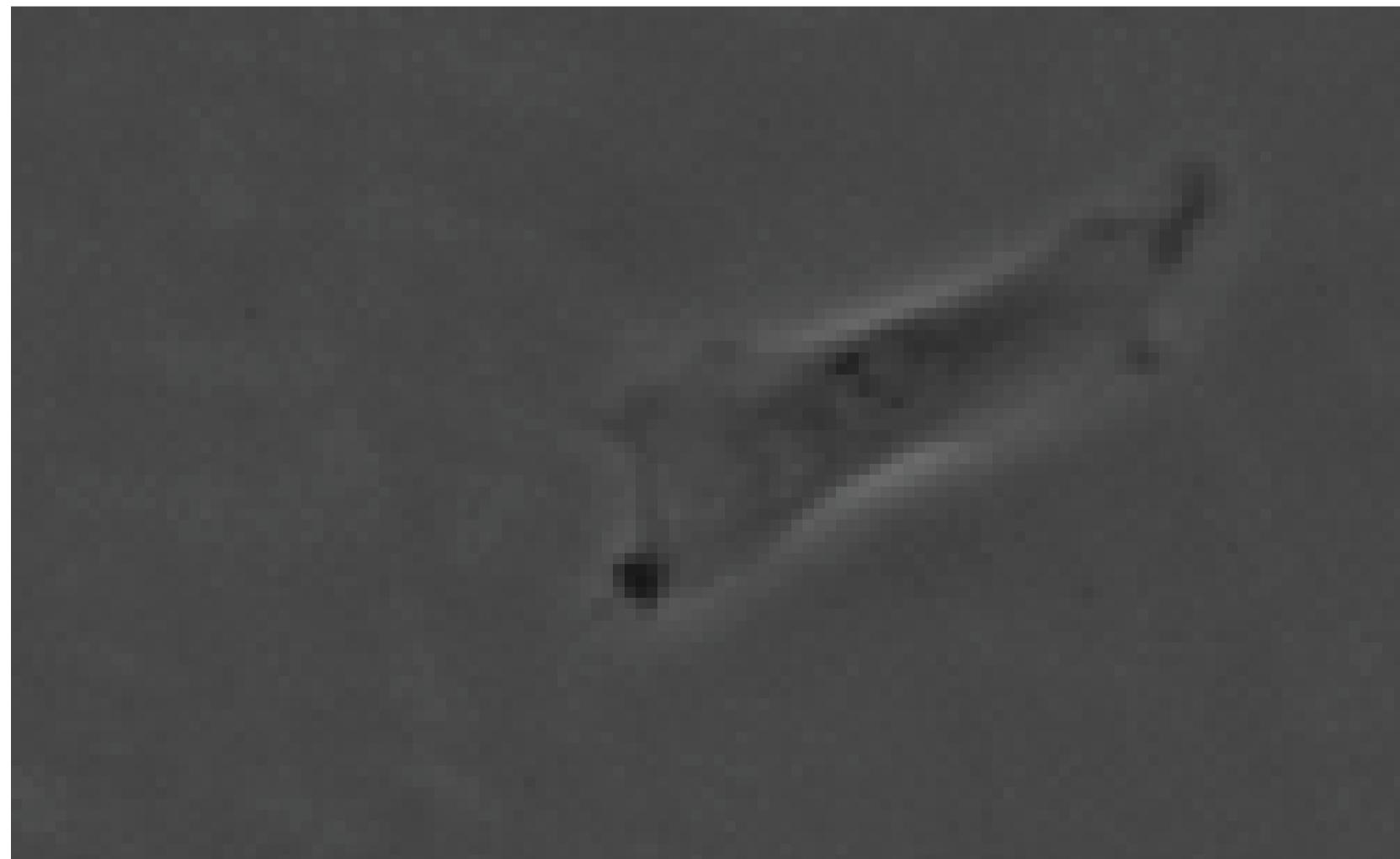
- ▶ Imaging relationship:
 - ▶ $I(P(x)) = M(P^{-1}(x)) + \text{noise}$
 - ▶ I image pixel data
 - ▶ P pose of an object
 - ▶ x position vector
 - ▶ M function image formation function: lighting + shape + etc
 - ▶ $I(P(x, t)) = M(P^{-1}(x, t)) + \text{noise}$
 - ▶ t time

Register: Fine but not finder

$$I(P(x, t)) = M(P^{-1}(x, t)) + \text{noise} ?$$

discrete pixel data

↑
Mapping Function
surface orientation, reflection,



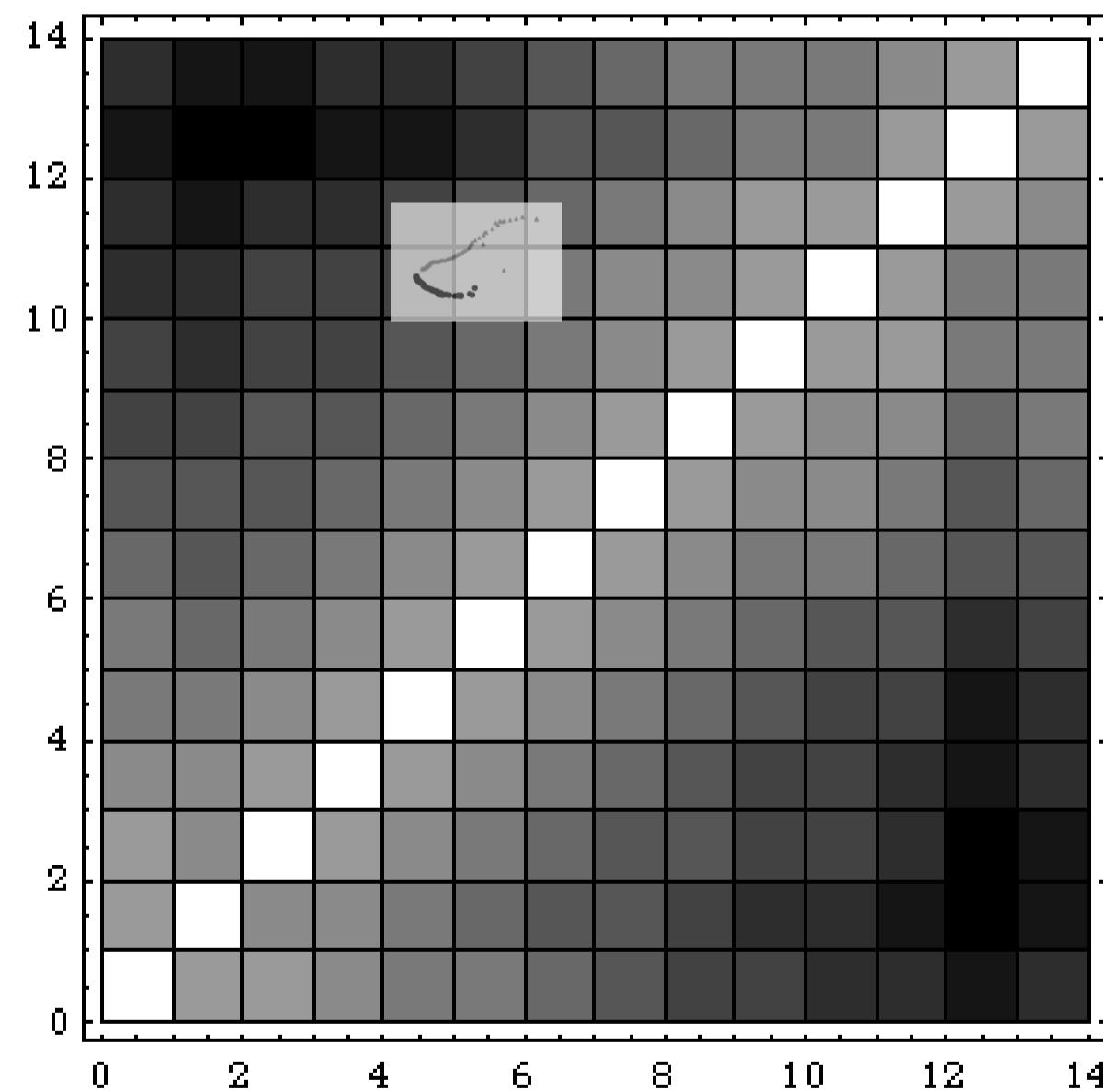
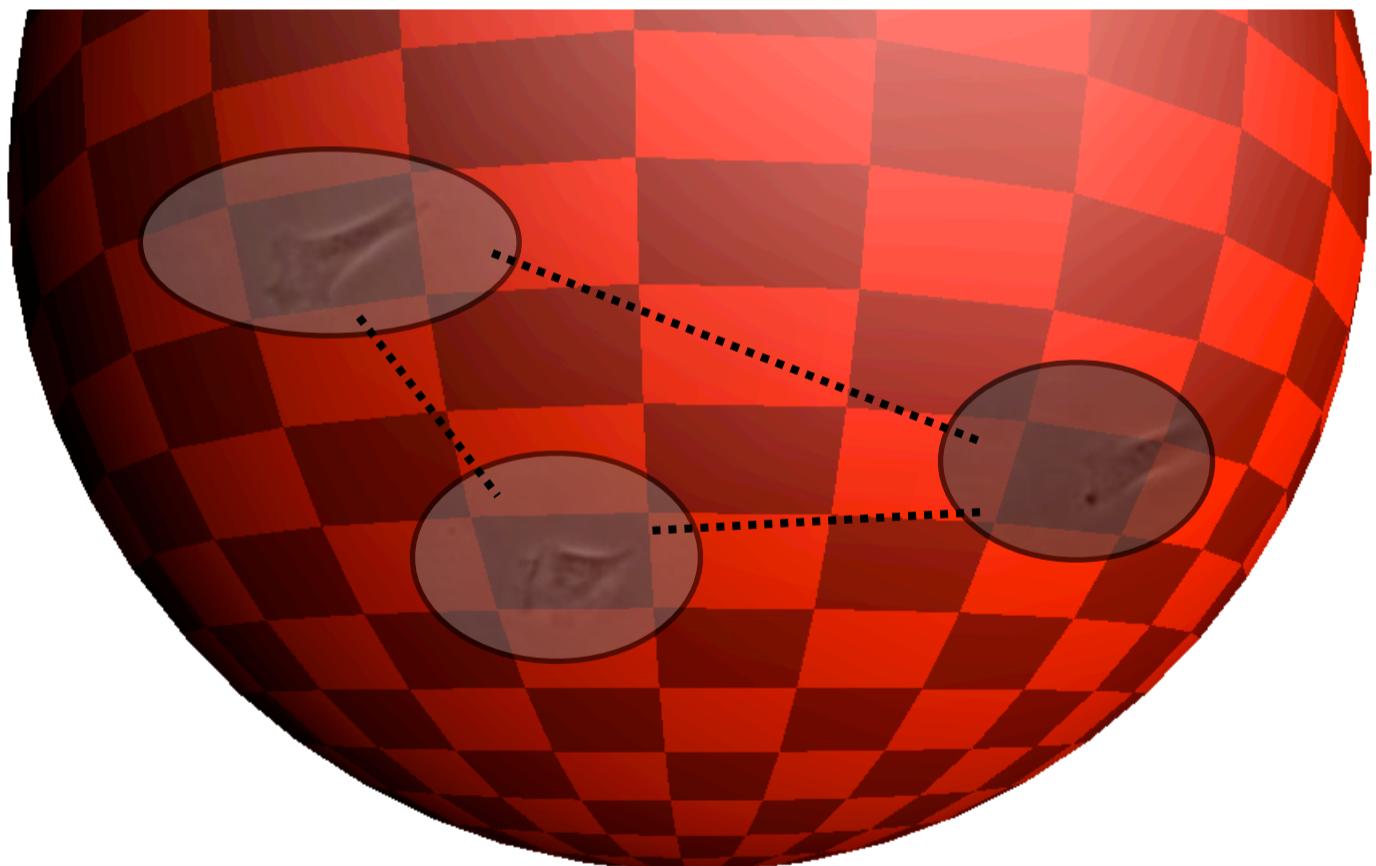
Morphological Change Over time

$$I(P(x, t)) = M(P^{-1}(x, t)) + \text{noise}$$

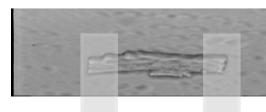


Diffusions
Oscillations

Morphological Change Over time

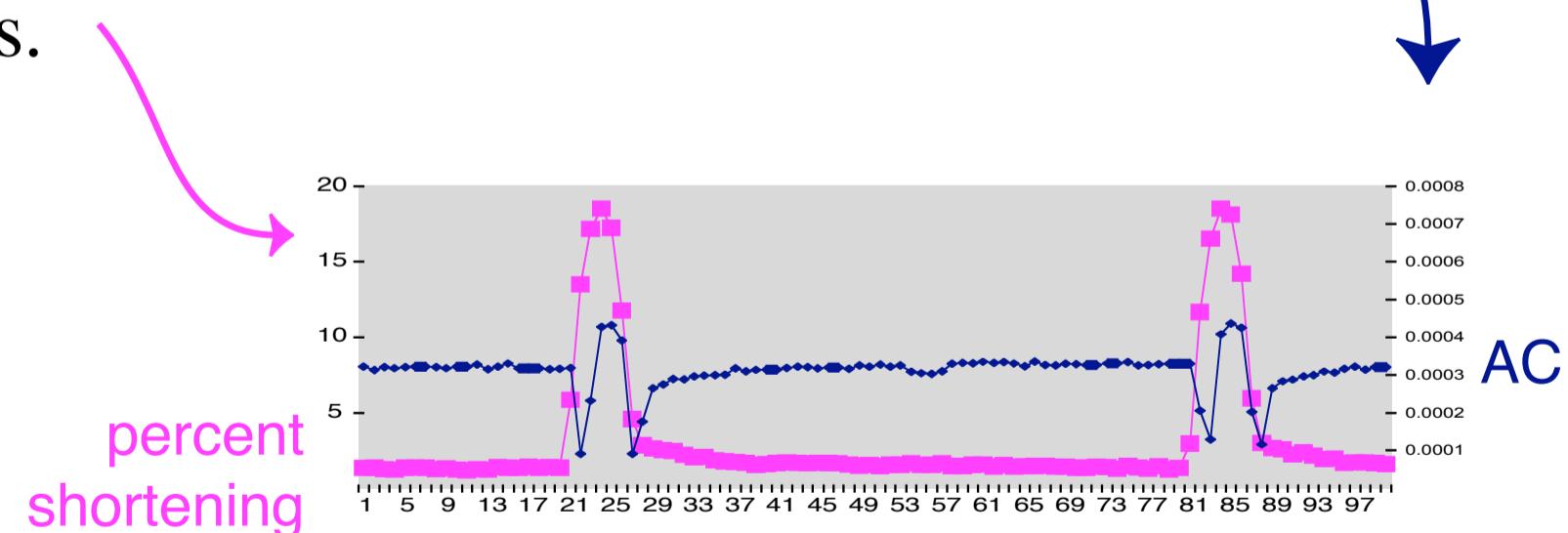


Muscle length vs. Aggregate Change Index



Use information from all the frames for the unsupervised ACI measurement.

Measure the length directly using a supervised approach that involves identifying the cell boundaries.

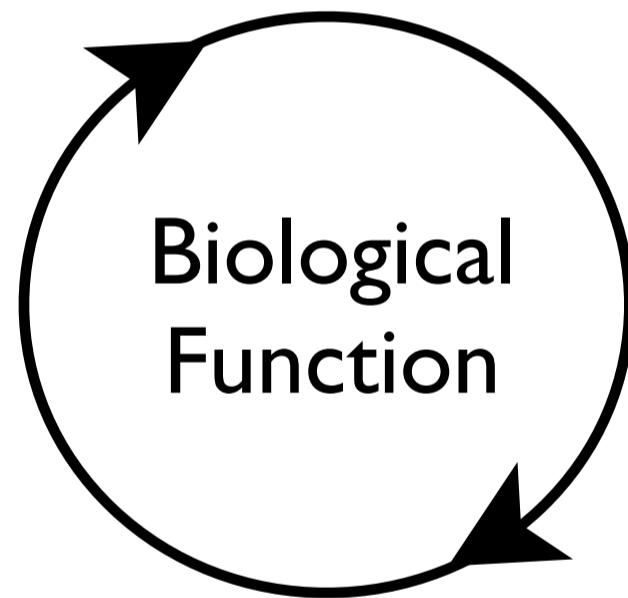
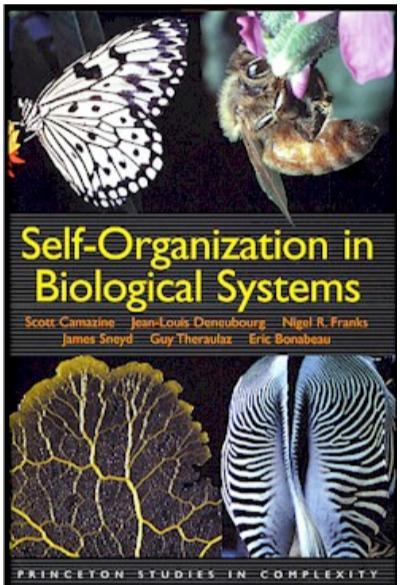


Self-Similarity

“A time varying phenomena is called self-similar if the spatial distribution of its properties at various moments can be obtained from one another by a similarity transformation” (Barenblatt in “Scaling, self-similarities, and intermediate asymptotics”)

“Equality Reynolds Numbers imply geometric similarities in flow patterns “ (Steven Vogel in “Life in Moving Fluid”)

Why is it interesting ?



Capture to maximize
mutual information
or joint statistics
over time

Self-similarity
Locally and
globally

Thank you for listening!

- Collaborators
- Everyone at Reify
- Bill Silver and Marilyn Matz at Cognex