

# Multi-scale model for tissue engineered articular cartilage

Simone Cassani, Sarah D Olson

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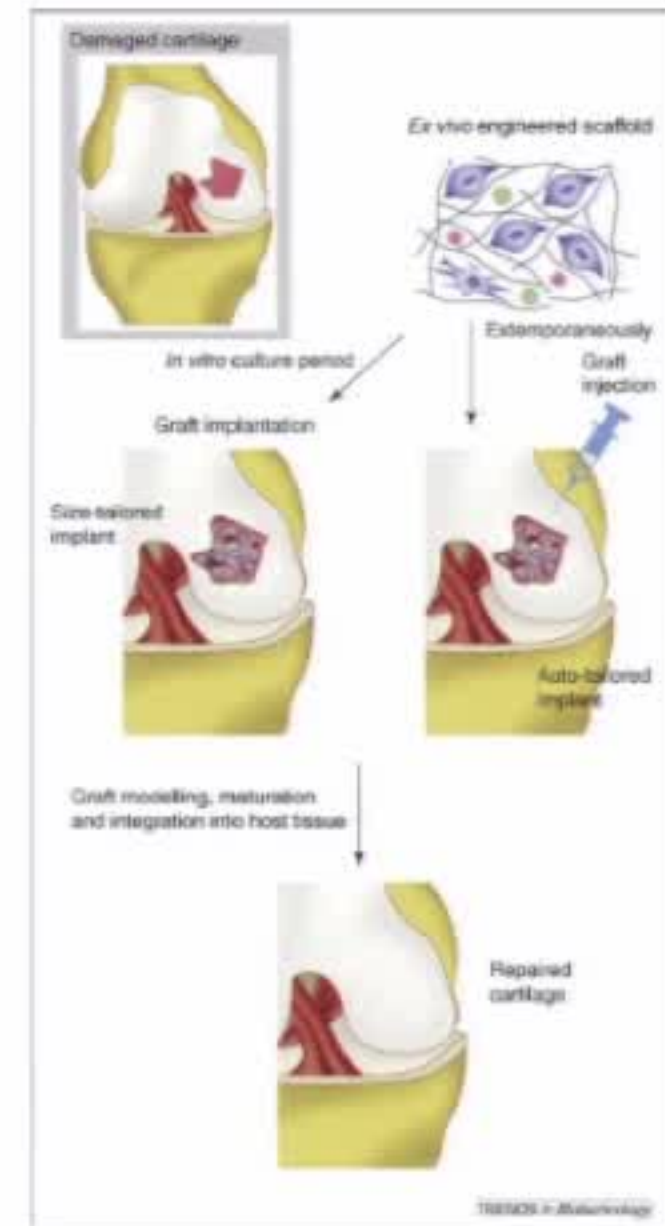
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Worcester Polytechnic Institute  
Department of Mathematical Sciences  
*scassani@wpi.edu*



# Introduction

- Articular cartilage in the knee has a complex structure composed of a dense extracellular matrix (ECM), which includes fluid, a collagen network, and other proteins (proteoglycans).
- Distributed in the matrix there are chondrocytes (cells) that synthesize the building blocks of the ECM.
- Pathologies such as osteoarthritis and injuries can cause the erosion and damage of articular cartilage
- Cartilage has a low repair capability

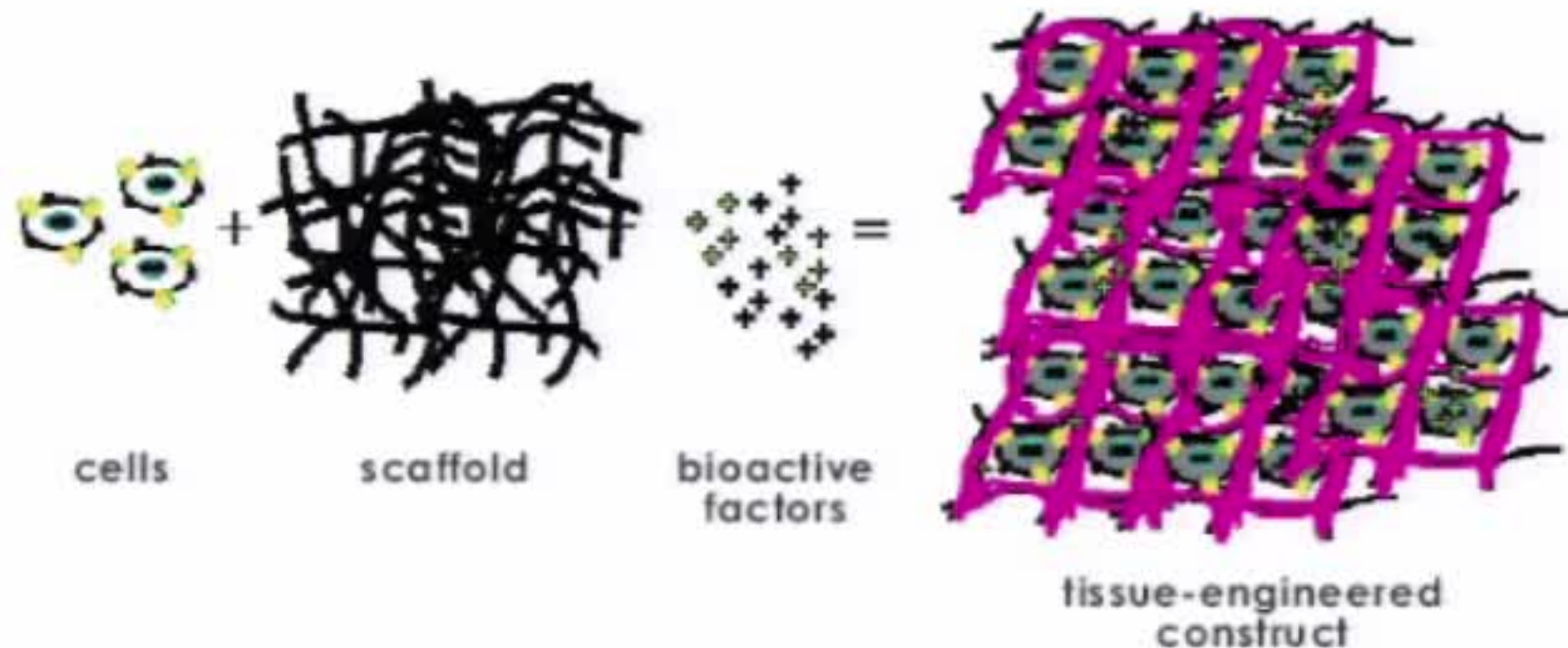


Vinatier, C. et al. Trends in Biotechnology  
Vol. 27(5), 2009.



# Introduction

Tissue engineering represents a promising path towards the treatment of damaged cartilage

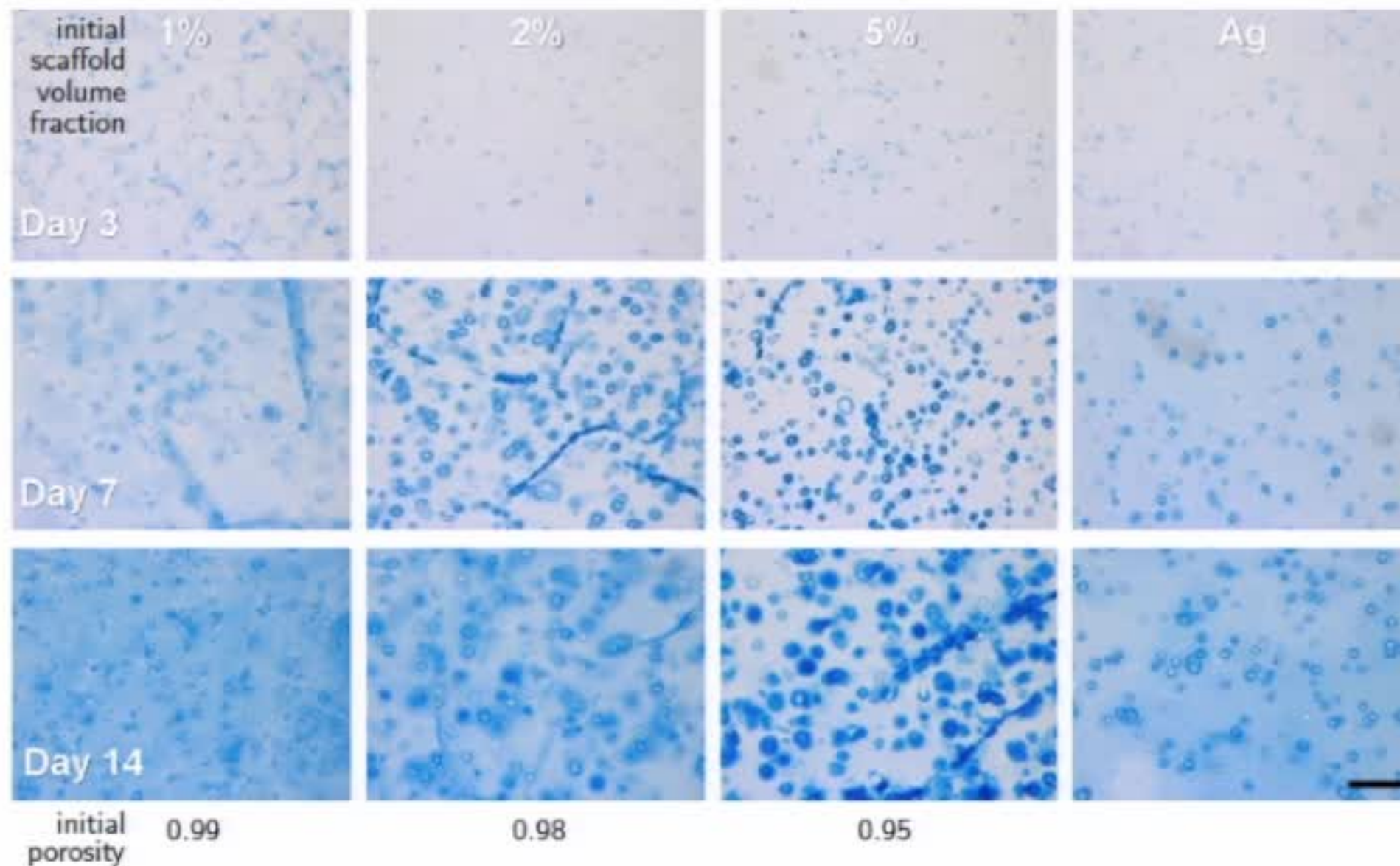


[http://www.bioeng.nus.edu.sg/\\_images/gallery/research4.gif](http://www.bioeng.nus.edu.sg/_images/gallery/research4.gif)

The outcomes of tissue engineered cartilage experiments strongly depend on different factors such as the properties of the gel or scaffold (porosity of gel, mechanical properties, diffusion of nutrients)

# Experiments: effect of porosity

$$\text{Porosity} = 1 - (\text{solid volume fraction}) = \text{fluid region}$$

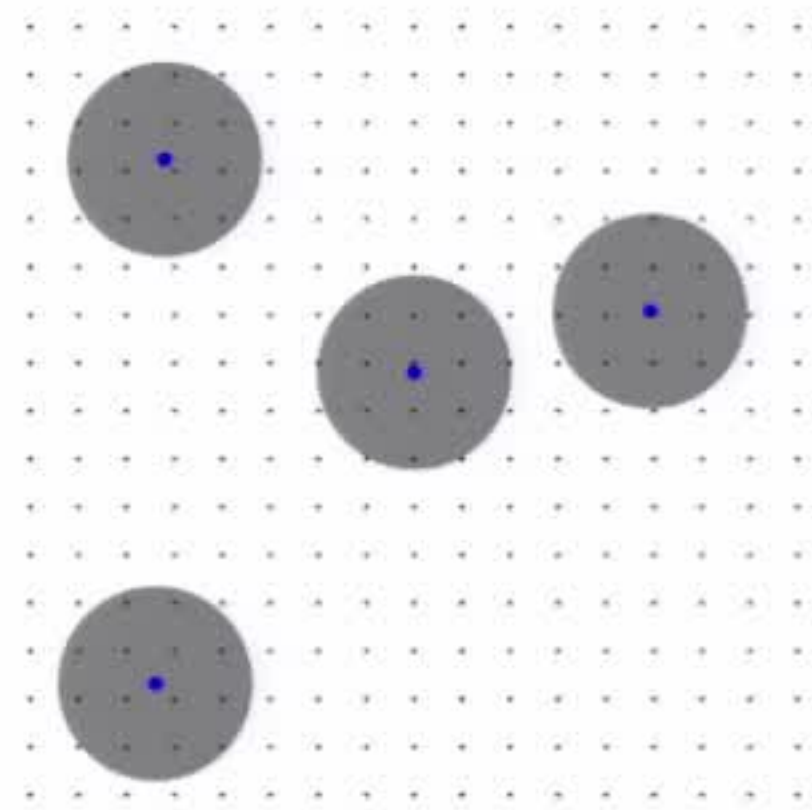
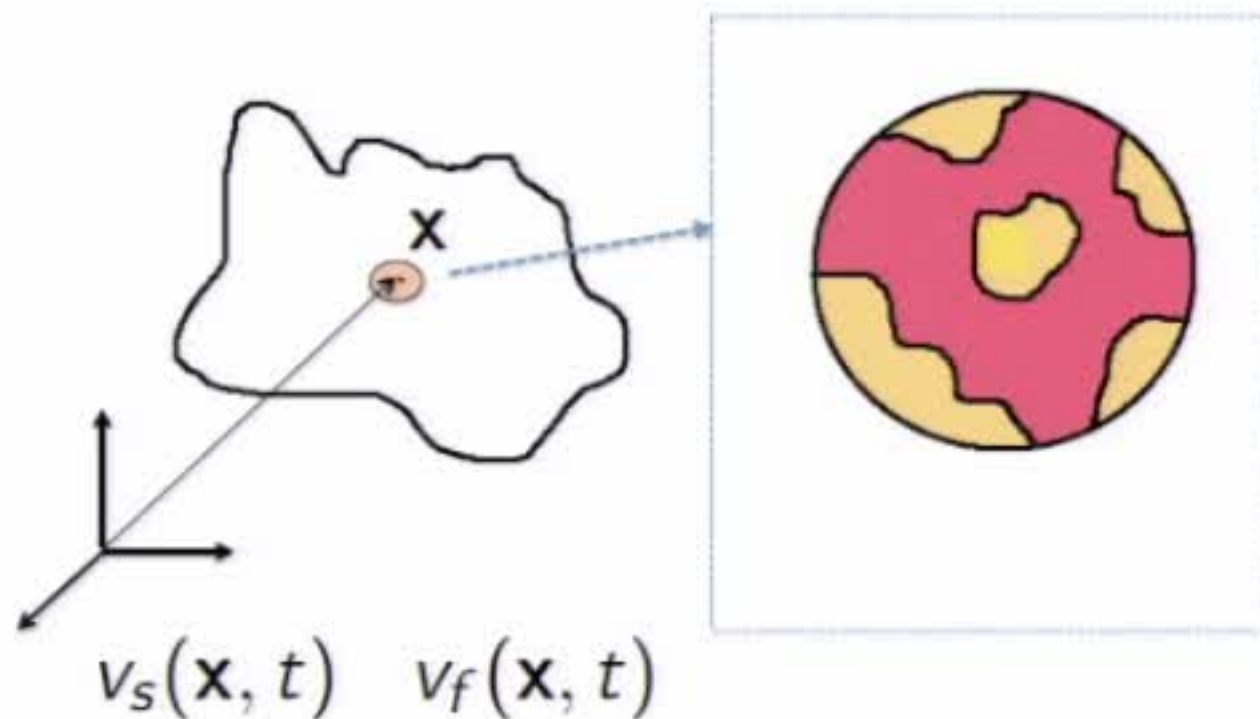


Localization of proteoglycans (blue) in gels of different porosity:  
Erickson, I.E. et al Osteoarthritis and Cartilage Vol. 17, 2009.



# Mathematical formulation

- Cell seeded gel - multiphase construct (solid + fluid)
- Porosity: multiphase quantity
- Possible approach: Mixture theory
- In this work: neglect fluid velocity



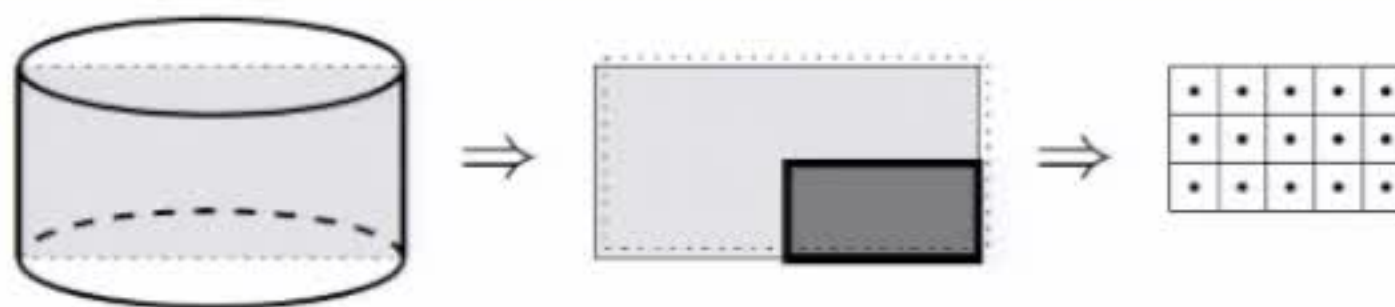
# Goal

- Develop a model that describes the process of tissue growth of engineered cartilage
  - Utilize a hybrid cellular automata framework
  - First model to account for porosity
- Investigate the mechanisms governing the growth of tissue engineered cartilage

## Main challenges

- Model calibration
- Efficiency

# Model features



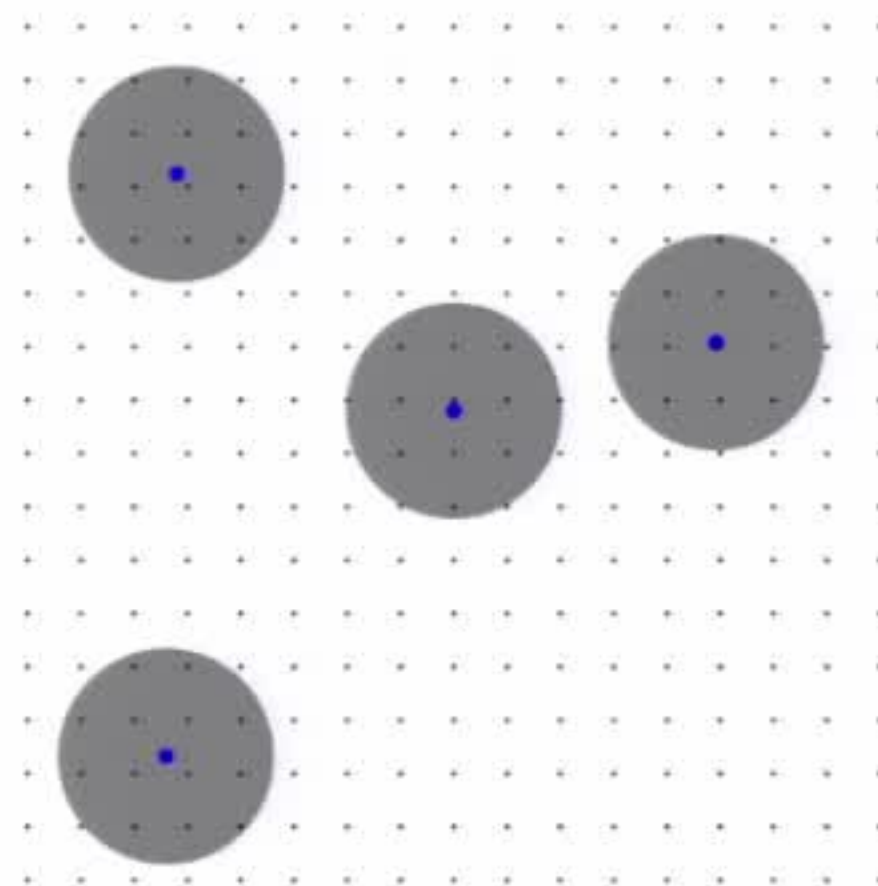
2D multiscale hybrid cellular automata model combining:

## Discrete components

- Off-grid
- Chondrocyte cells
- Rule based
- Cell center and radius

## Continuous components

- On-grid
- Nutrient concentration,  $c$
- Porosity,  $p$

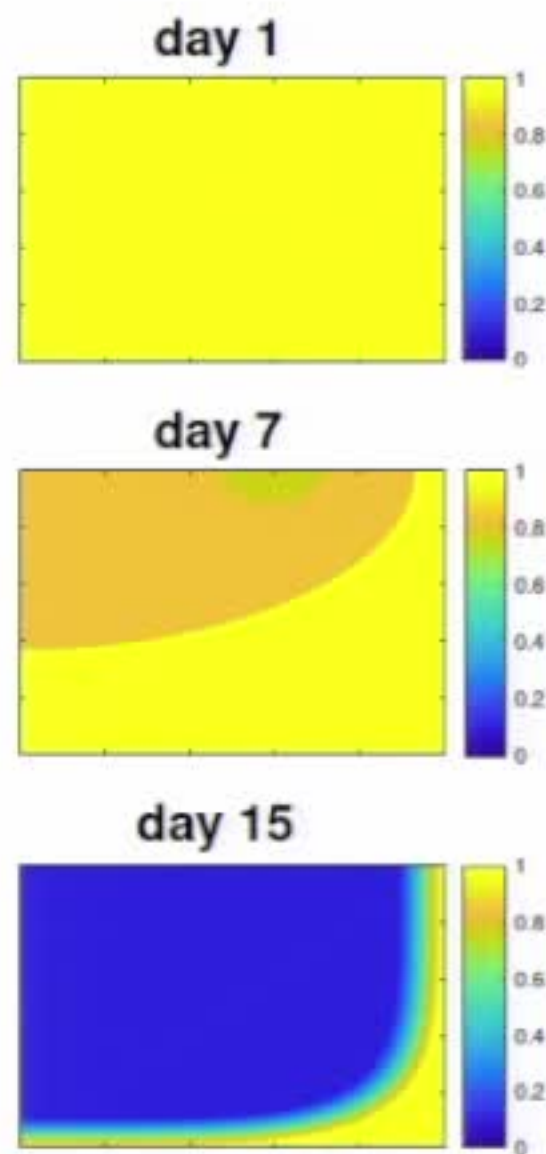




# Continuous components - nutrient concentration

Nutrient concentration,  $c$

- Based on experiments
- Qualitative approach
- Non dimensional, on  $[0,1]$  range
- Slow timescale

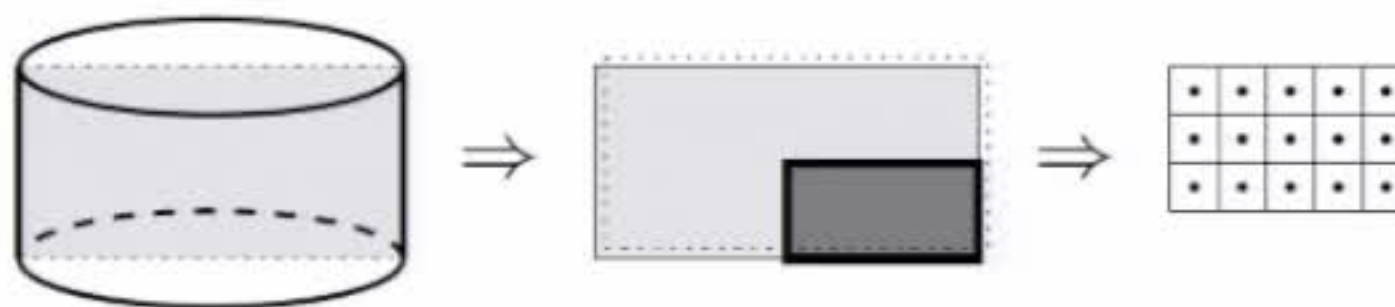


Chung et al J Theor Biol Vol 262, (2010)

Bandeiras et al Biomech Model Mechanobiol Vol 14, (2015)



# Model features



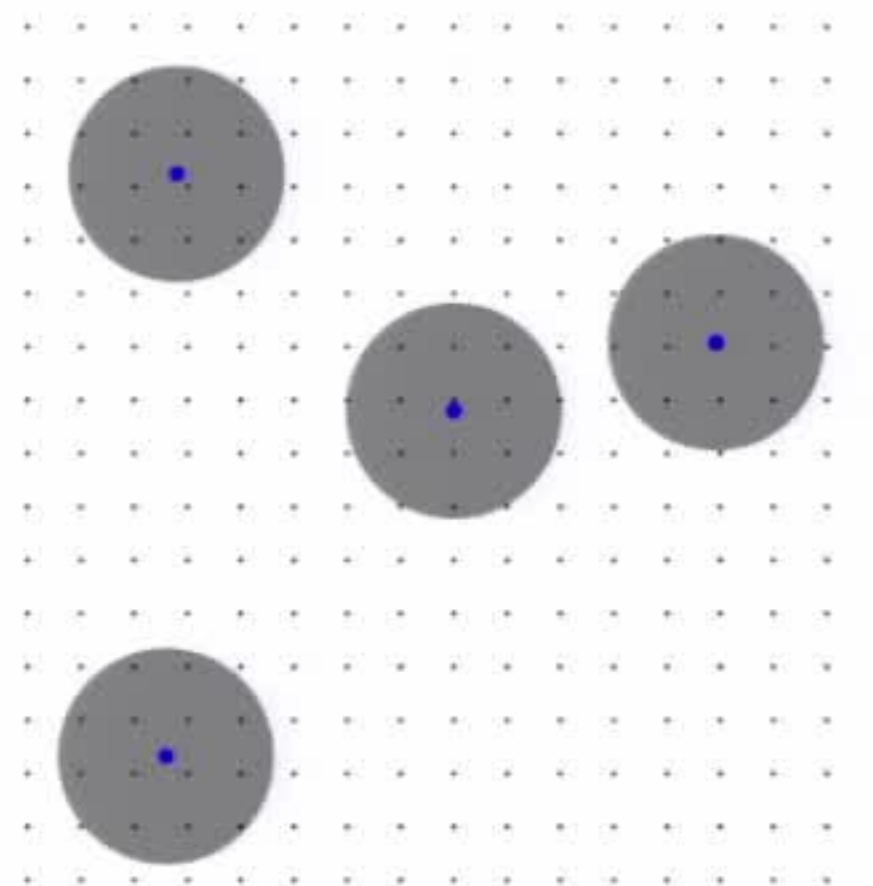
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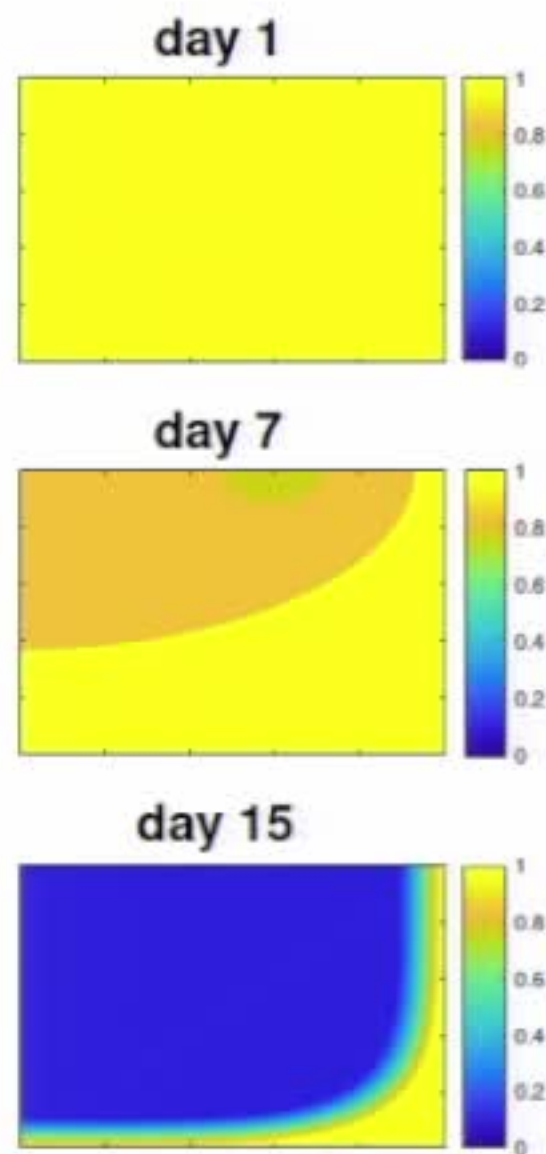
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# Continuous components - porosity

Scaffold volume fraction,  $\Phi_{SC}^{(*)}$

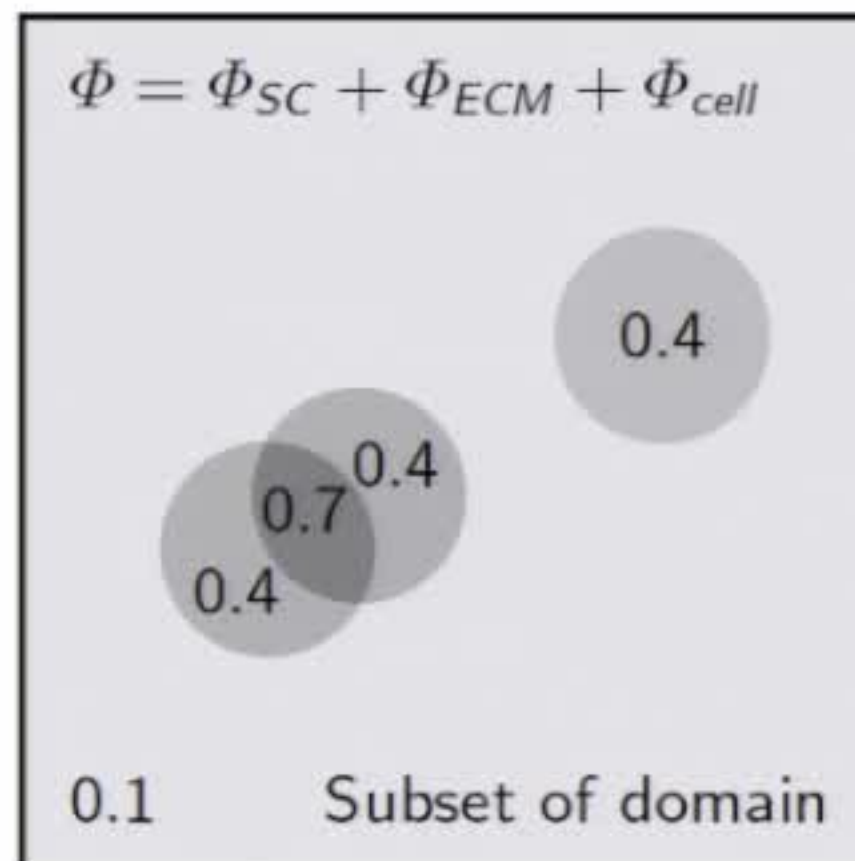
$$\Phi_{SC}(t) = \begin{cases} \Phi_{SC}^0, & t < \tau \\ \Phi_{SC}^0 e^{-k_{SC}(t-\tau)}, & t \geq \tau \end{cases}$$

ECM volume fraction,  $\Phi_{ECM}^{(*)}$

$$\Phi_{ECM}(t) = \Phi_{ECM}^{SS} \left(1 - e^{-k_{ECM} \cdot t}\right)$$

Cellular volume fraction,  $\Phi_{cell}$

- Depends on cell positions on grid
- Single cell contribution,  $\hat{\Phi}$



(\*) Wilson C.G. et al Arch Biochem Biophys Vol 408, (2002)

- Solid Volume fraction  $\Phi = \Phi_{SC} + \Phi_{ECM} + \Phi_{cell}$
- $\rho = 1 - \Phi$
- Fast timescale

# Discrete components

## Chondrocytes (cells)

- Cellular division
- Cell death
- Biased random movement
  - Chemotaxis
  - Porosoty
- Cell-to-cell contact inhibition

Numerical value	Cell status
0	moving
1	dividing
2	quiescent
3	dying

The variable `cell_status` monitor the current status of each cell.



# Discrete components

**Cell division:** to perform cellular division a cell must

- 1 Not be in quiescent state
- 2 Have reached cellular maturity
- 3 Be in a region with enough nutrient
- 4 Be near to a region with high enough porosity to host a new cell

**Cell death**

- Determined by a random rule
- Removed from computational domain

# Discrete components

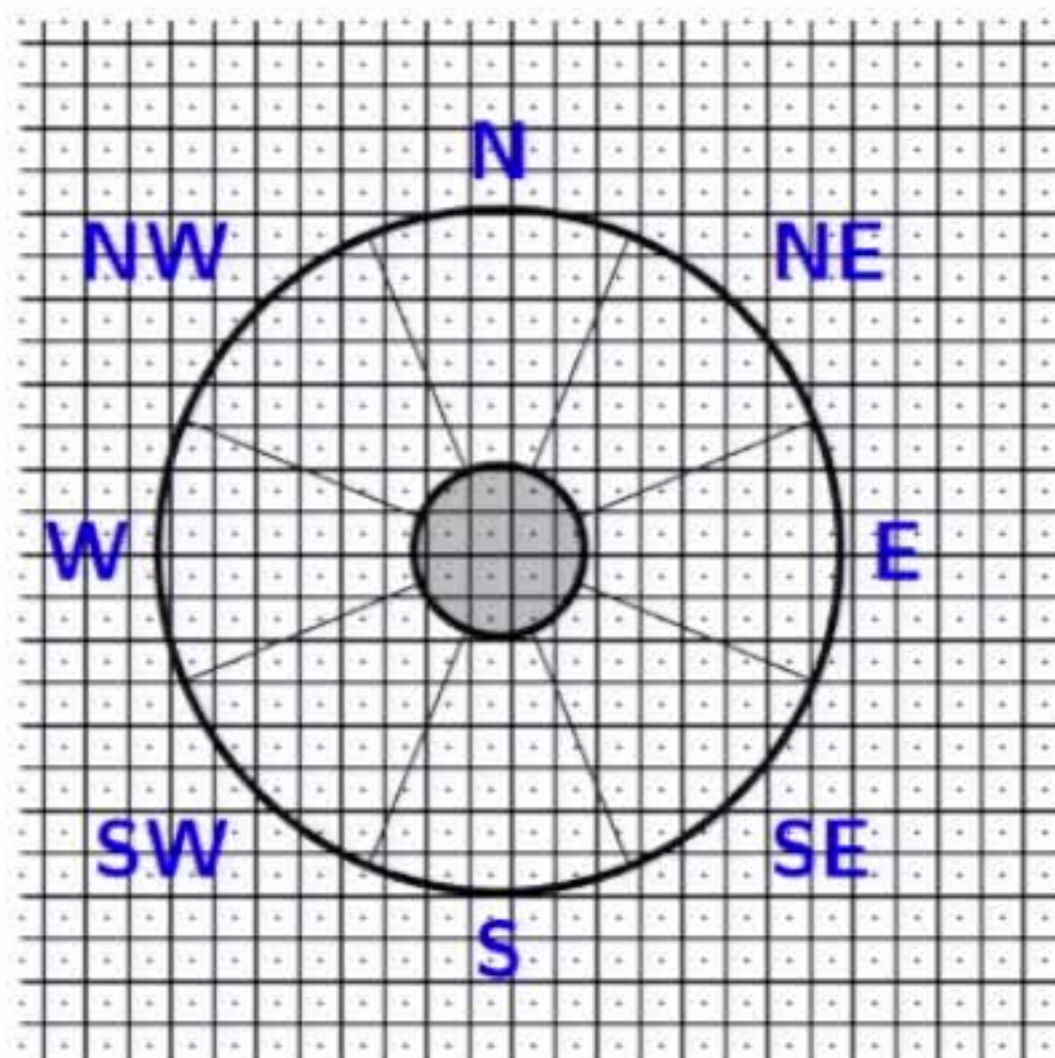
## Cell movement

- Off-grid
- 8 possible directions
- Random, biased
  - towards higher  $c$

$$\omega_{n,j} = \frac{c_{avg,j}}{\sum_{i=\emptyset}^{SE} c_{avg,i}}$$

- towards higher  $p$

$$\omega_{p,j} = \frac{p_{avg,j}}{\sum_{i=\emptyset}^{SE} p_{avg,i}}$$



$k$  direction probability vectors ( $k$  cells)

$$\omega_j = \hat{\omega}_{n,j} \cdot \hat{\omega}_{p,j}, \quad j = \emptyset, \dots, SE$$

$\emptyset$	E	NE	N	NW	W	SW	S	SE
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0

1



# Discrete components

## Cell-to-cell contact inhibition

- Collisions among cells
- Collided cells will enter the quiescent state
- Cells in quiescent state stay idle for a time  $t_{idle}$
- No movement nor cell division

## Boundary conditions

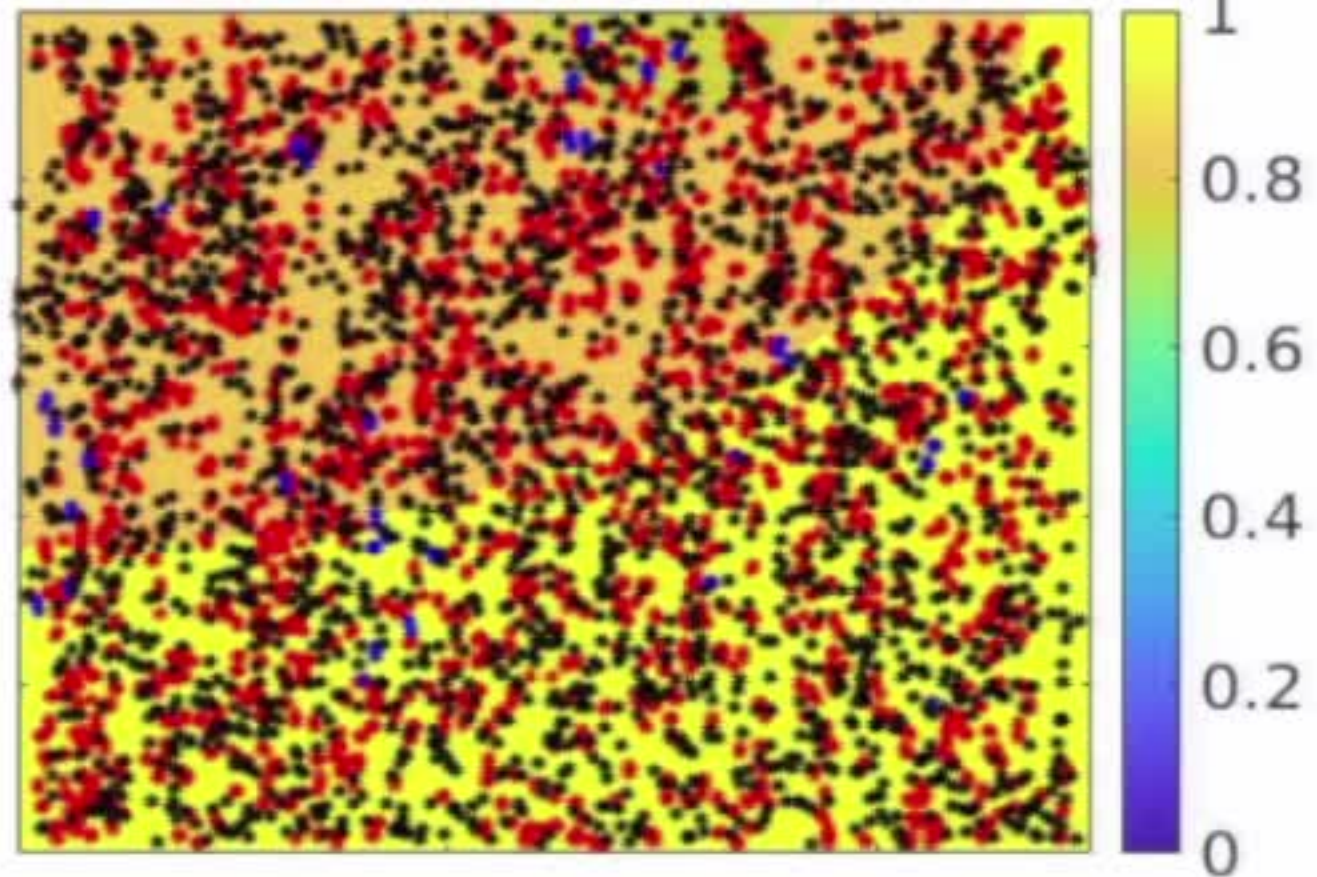
- No-flux bc
- Ghost nodes
- Trajectory corrections



# Results

**t=9.6667 days**

- color nutrient  
concentration
- \* moving state
  - \* dividing state
  - \* quiescent state

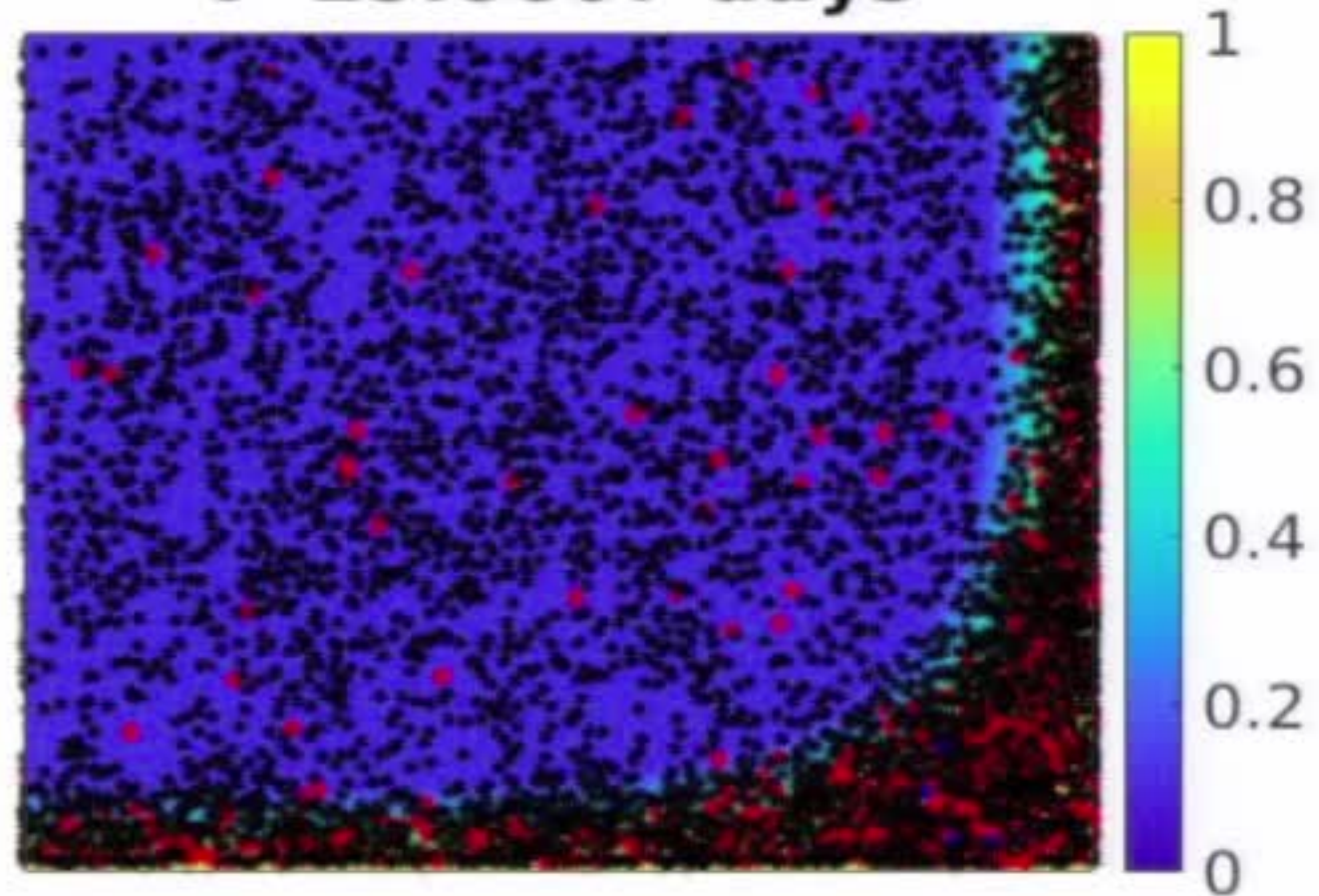




# Results

**t=29.6667 days**

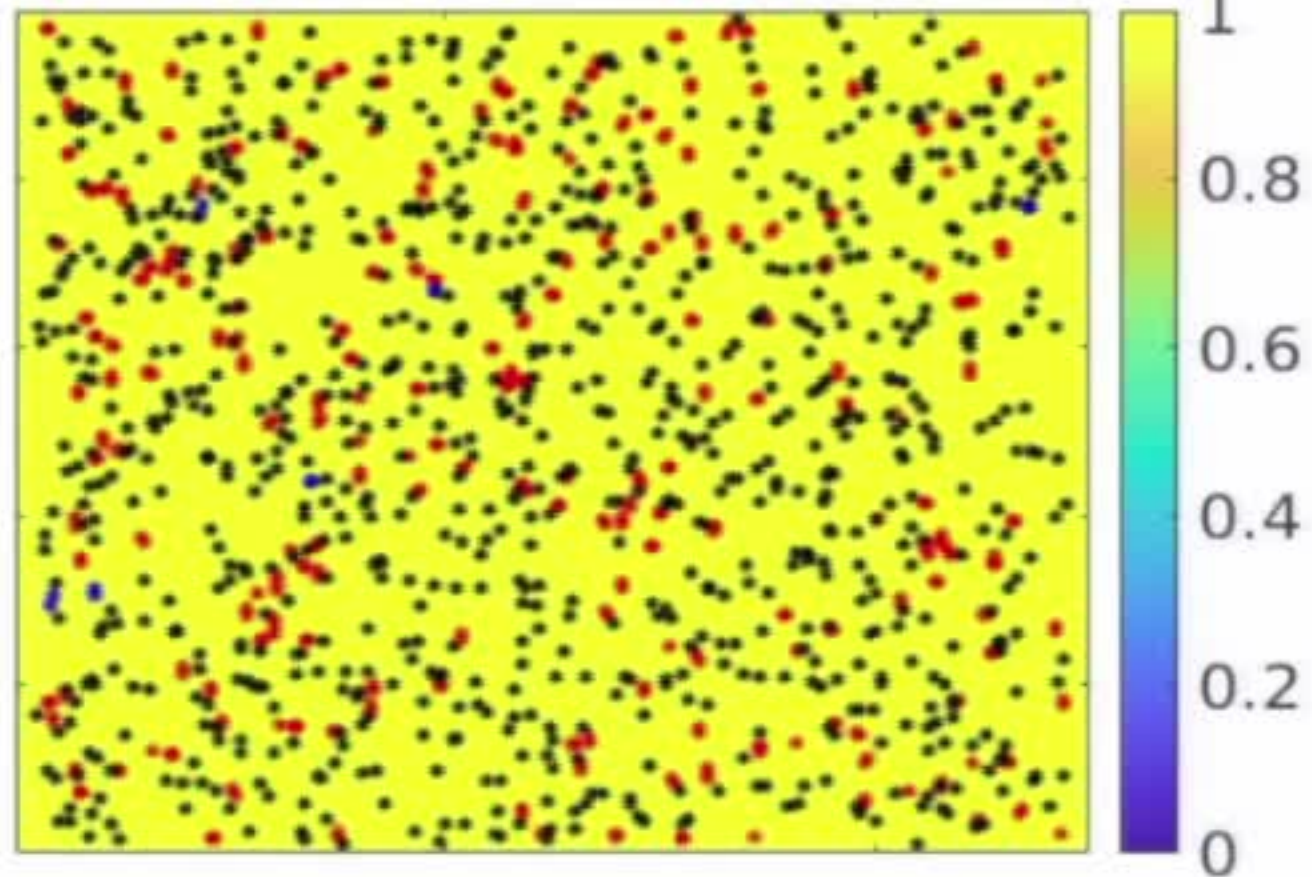
- color nutrient concentration
- \* moving state
  - \* dividing state
  - \* quiescent state



# Results

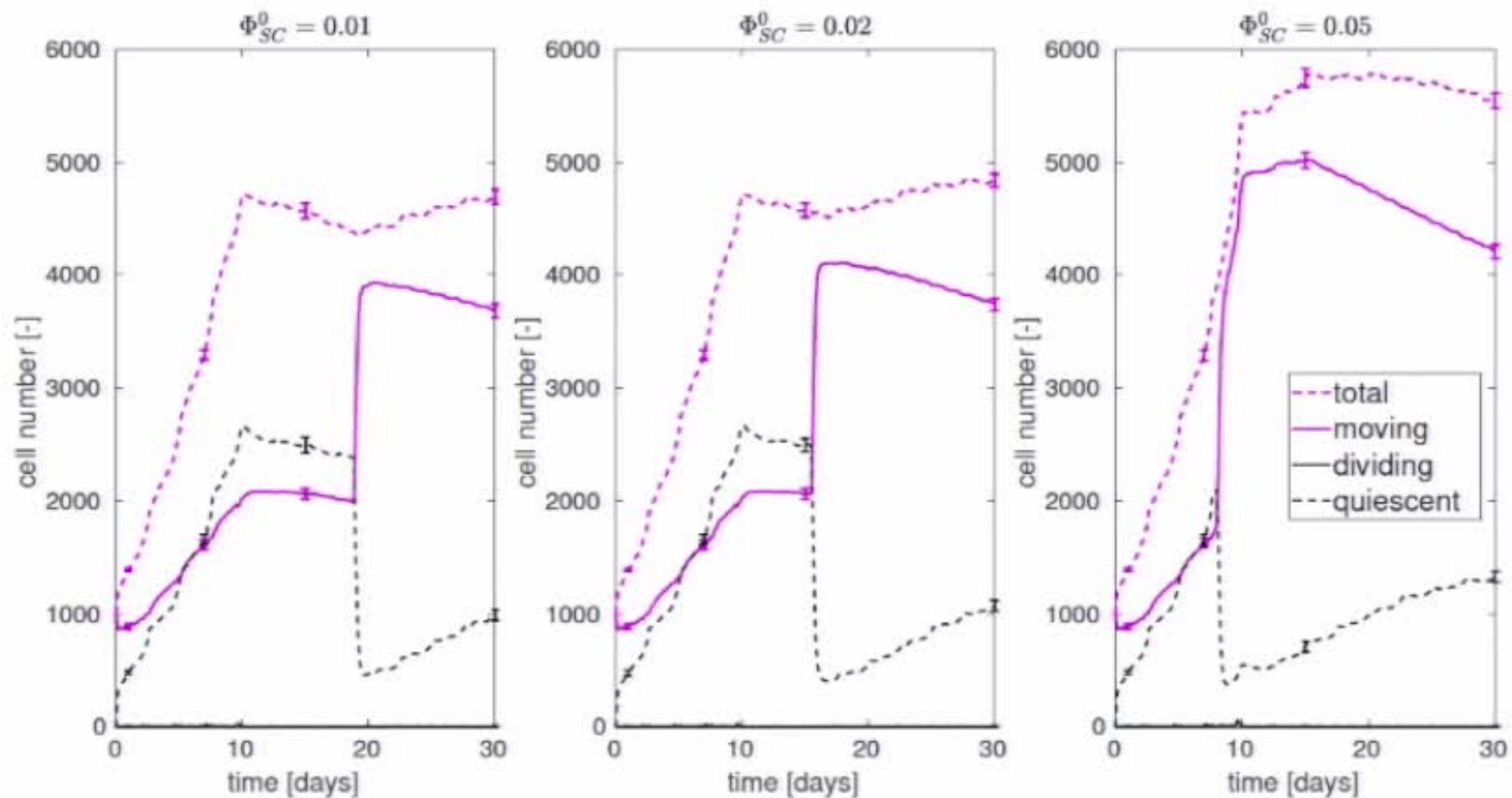
**t=0.66667 days**

- color nutrient concentration
- \* moving state
  - \* dividing state
  - \* quiescent state



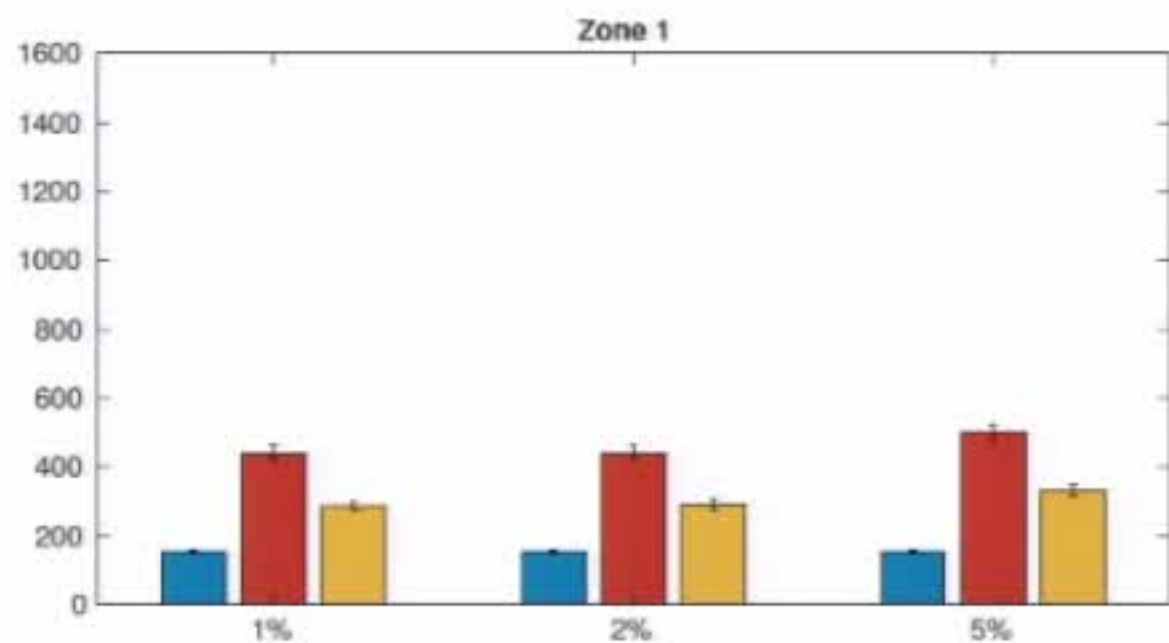


# Results - cell status and porosity

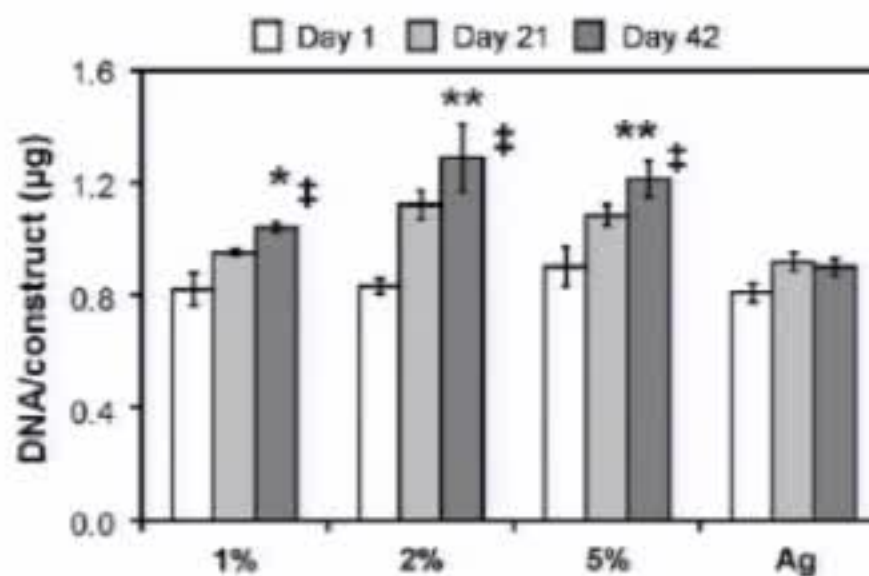
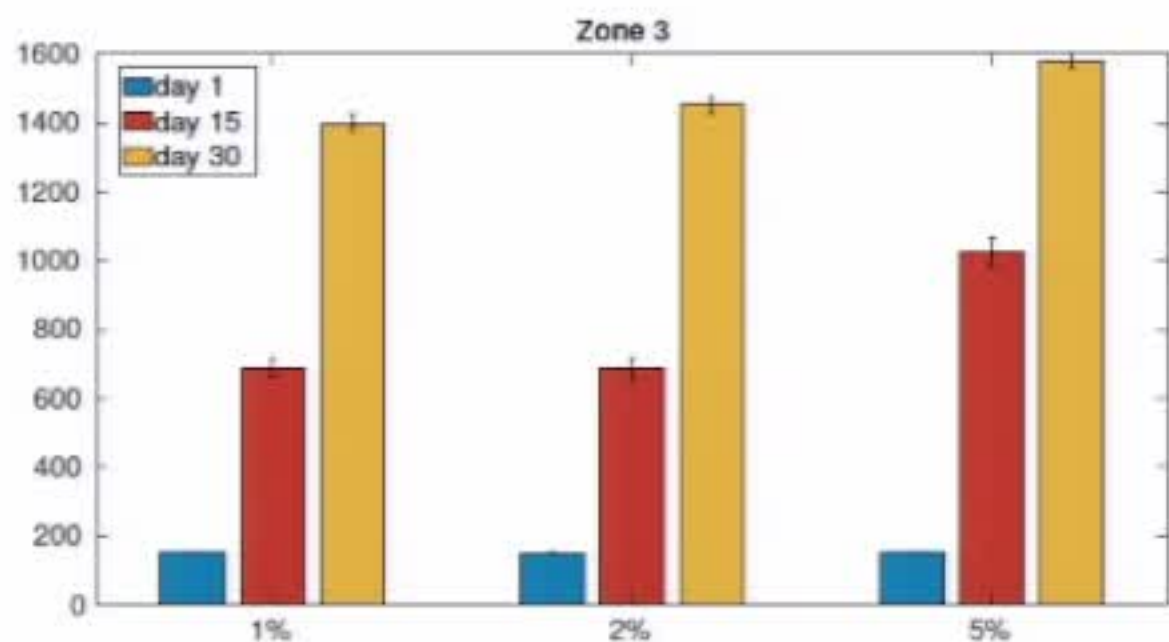




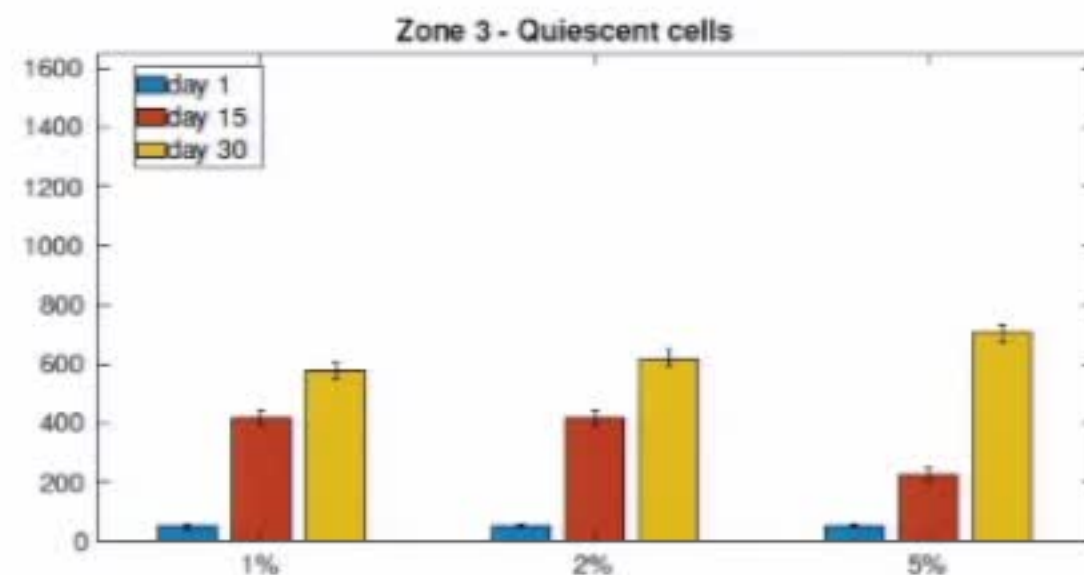
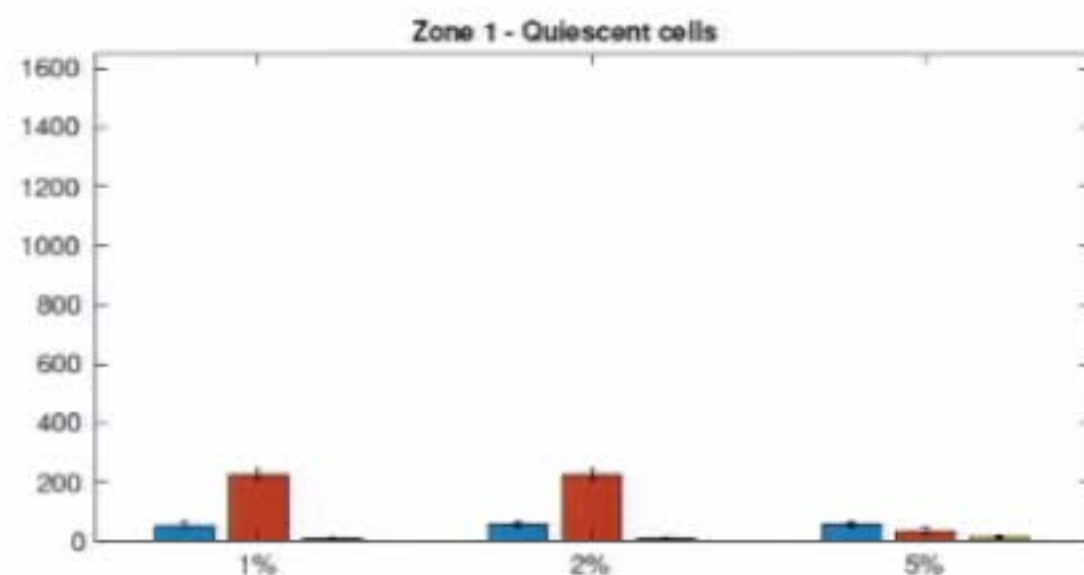
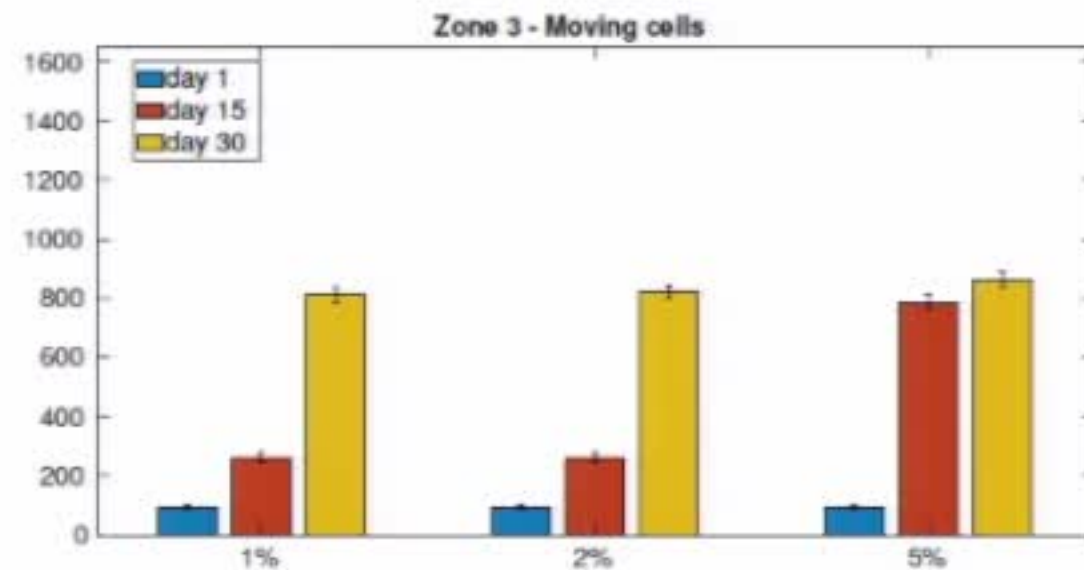
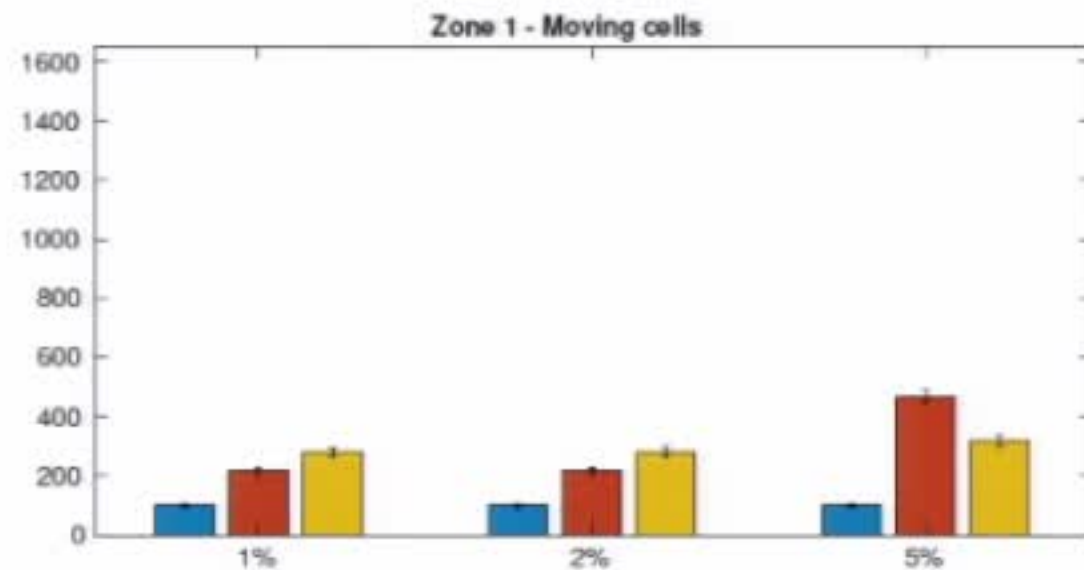
# Results - spatial variability



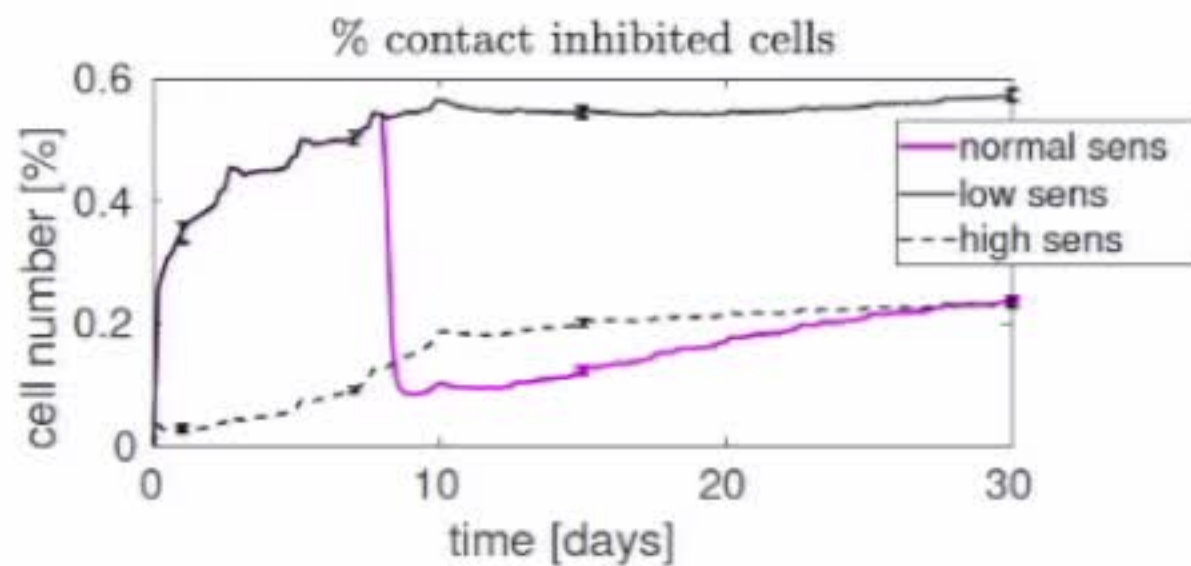
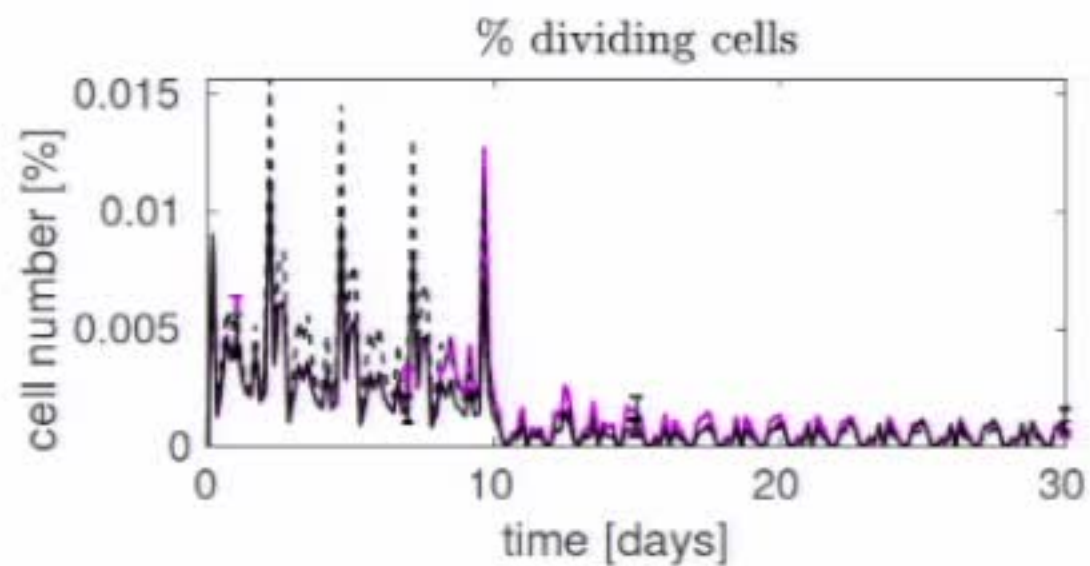
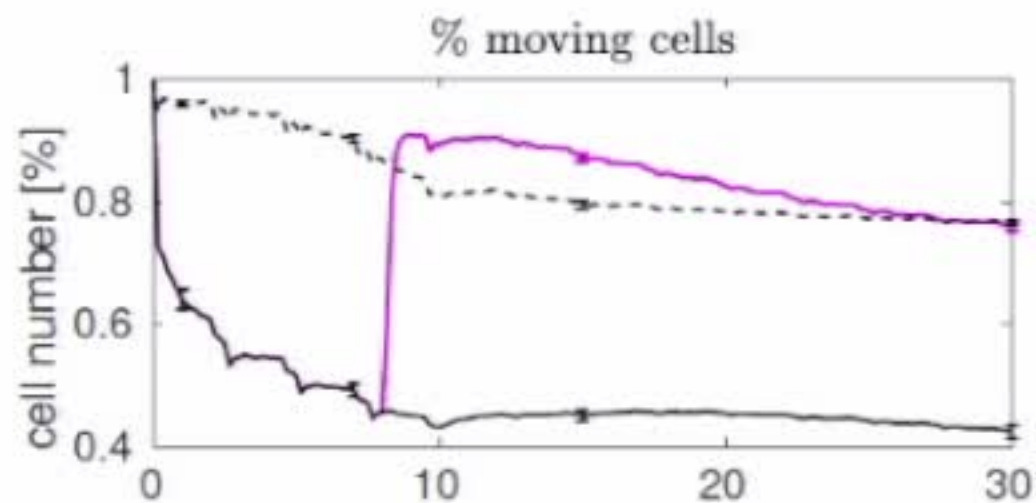
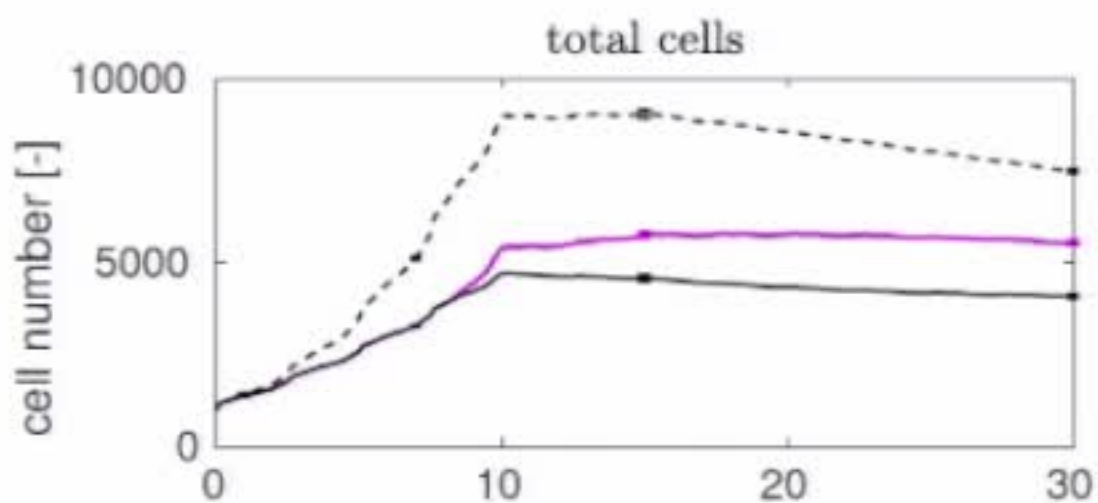
Zone 1		
		Zone 3



# Results - spatial variability



# Results - sensitivity to porosity





## Conclusions and future works

- Developed a hybrid model that describes the process of tissue growth of engineered cartilage accounting for porosity.
  - Captures different behavior for low (0.95) and high (0.99,0.98) initial value of porosity.
  - Spatial heterogeneity.
  - Non trivial dynamics due to different sensitivity to porosity.
- 
- Refine rules to capture differences between 1% and 2% cases.
  - Develop cell dependent nutrient and ECM production.
  - Expand the model to include a mixture theory formulation.

# Results - sensitivity to porosity

