The Application of Neural Networks to Predict Skin Evolution After Burn Trauma

Literature Presentation



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Biological Background

Skin Consists of Three Layers



- Wounds to the epidermis are superficial and heal quickly, without scarring
- Wounds that affect the dermis are more complex, as the damaged dermis needs to be regenerated
- The focus is on the latter

Gholizadeh, Z., Yousefi, R., & Asadi, S. (2020). The Role of Mutations on Genes KRT5, POFUT1 & POGLUT1, in Dowling-Degos Syndrome. International Journal of Innovative Research in Medical Science (IJIRMS), 5(08).

The Dermis Consists of

- Water
- Cells: **fibroblasts** being the most common ones
- ECM: 3D network of molecules that surrounds and supports the cells. Consist mostly out of **collagen** proteins



Extracellular Matrix (ECM)



- Acts as a scaffold for cells, such that they can migrate and perform their functioning
- Contains signalling molecules

Pfisterer, K., Shaw, L. E., Symmank, D., & Weninger, W. (2021). The extracellular matrix in skin inflammation and infection. Frontiers in cell and developmental biology, 9, 682414.

Wound Healing (ECM is damaged)



Ribatti, D., Benagiano, V., & Guidolin, D. (2023). Overlapping between Wound Healing Occurring in Tumor Growth and in Central Nervous System Neurodegenerative Diseases. Brain Sciences, 13(3), 398.

- Fibrin cloth is made, acting as temporary ECM
- Pathogens are removed
- Fibroblasts come in and deposit new ECM
- Signalling molecules:

fibroblasts -> myofibroblasts

- Myofibroblasts pull on the ECM
- New ECM undergoes reorganisation

Contraction



Nascimento-Filho, C. H., Silveira, E. J., Goloni-Bertollo, E. M., de Souza, L. B., Squarize, C. H., & Castilho, R. M. (2020). Skin wound healing triggers epigenetic modifications of histone H4. Journal of Translational Medicine, 18(1), 1-11.

Egberts, G., Vermolen, F., & van Zuijlen, P. (2023). Stability of a two-dimensional biomorphoelastic model for post-burn contraction. Journal of Mathematical Biology, 86(4), 59.



Mechanical Background

Description of Motion

Lagrangian / Material Description:

$$\mathbf{x} = \mathbf{f}(\mathbf{X}, t)$$

Eulerian / Spatial Description:

$$\mathbf{X} = \mathbf{f}^{-1}(\mathbf{x}, t)$$



Deformation Gradient Tensor



- Fundamental measure of deformation
- Maps line elements in reference configuration to line elements in current configuration

 $d\mathbf{x} = \mathbf{F} \, d\mathbf{X}$

Strain tensor

- Strain is a measure of deformation of a body with respect to a reference configuration
- Presence of a stress will generally lead to strain

$$oldsymbol{arepsilon} oldsymbol{arepsilon} = rac{1}{2} egin{bmatrix}
abla \mathbf{u} + (
abla \mathbf{u})^T \end{bmatrix} egin{array}{c} oldsymbol{arepsilon} = egin{pmatrix} arepsilon_{11} & arepsilon_{12} & arepsilon_{13} \ arepsilon_{21} & arepsilon_{22} & arepsilon_{23} \ arepsilon_{31} & arepsilon_{32} & arepsilon_{33} \end{pmatrix} \end{array}$$

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Stress tensor



File:Components stress tensor cartesian.svg. (2022, March 16). Wikimedia Commons. Retrieved 21:10, October 24, 2023 from https://commons.wikimedia.org/w/index.php?title=File:Components_stress_tensor_cartesian.svg&oldid=639858293.

- Stress expresses the internal forces that neighbouring material particles exert on each other
- This can be as a reaction to external forces on the surface

$$oldsymbol{\sigma} oldsymbol{\sigma} = egin{pmatrix} \sigma_{11} & \sigma_{12} & \sigma_{13} \ \sigma_{21} & \sigma_{22} & \sigma_{23} \ \sigma_{31} & \sigma_{32} & \sigma_{33} \end{pmatrix}$$



Morphoelastic Model for Burn Injuries

Morphoelasticity







Initial Configuration

Zero Stress State

F





Decomposition:

Т

$$egin{aligned} \mathbf{F} &= rac{\partial \mathbf{x}}{\partial \mathbf{X}} \ &= rac{\partial \mathbf{x}}{\partial oldsymbol{\chi}} rac{\partial oldsymbol{\chi}}{\partial oldsymbol{\chi}} = \mathbf{F}_e \ \mathbf{F}_g \end{aligned}$$

Current Configuration

Hall, C. L. (2008). Modelling of some biological materials using continuum mechanics (Doctoral dissertation, Queensland University of Technology).

Mathematical Model

$$egin{aligned} rac{\mathrm{D} z_i}{\mathrm{D} t} + z_i (
abla \cdot \mathbf{v}) &= -
abla \cdot \mathbf{J}_i + R_i, \
ho_t \left(rac{\mathrm{D} \mathbf{v}}{\mathrm{D} t} + \mathbf{v} (
abla \cdot \mathbf{v})
ight) &=
abla \cdot oldsymbol{\sigma} + \mathbf{f}, \end{aligned}$$

$$\frac{\mathrm{D}\boldsymbol{\varepsilon}}{\mathrm{D}t} + \boldsymbol{\varepsilon}\operatorname{skw}(\nabla\mathbf{v}) - \operatorname{skw}(\nabla\mathbf{v})\boldsymbol{\varepsilon} + (\operatorname{tr}(\boldsymbol{\varepsilon}) - 1)\operatorname{sym}(\nabla\mathbf{v}) = -\mathbf{G}.$$

conservation of cell density / concentration

conservation of linear momentum

evolution equation of strain, morphoelasticity of dermal layer

Biological Constituents				Mechanical Components			
N	fibroblasts	С	signalling molecule	u	displacement	${oldsymbol {\mathcal E}}$	strain
M	myo- fibroblasts	ho	collagen	\mathbf{V}	velocity		



Machine Learning Background

Neural Networks (NNs)



Hidden Layers

Nuzzi, R., Boscia, G., Marolo, P., & Ricardi, F. (2021). The impact of artificial intelligence and deep learning in eye diseases: a review. Frontiers in Medicine, 8, 710329.

- Layers of interconnected nodes (neurons)
- Each neuron receives input. Inputs are weighted, summed and passed through an activation function to produce the neuron's output
- NN minimises a loss function by adjusting weights and biases during training

Physics Informed Neural Networks (PINNs)



Guo, Y., Cao, X., Liu, B., & Gao, M. (2020). Solving partial differential equations using deep learning and physical constraints. Applied Sciences, 10(17), 5917.

Use the physical prior knowledge (PDEs) to constrain the NN's space of admissible solutions

$$u_t + \mathcal{N}[u;\lambda] = 0, \; x \in \Omega, t \in [0,T]$$

$$f:=u_t+\mathcal{N}[u;\lambda]$$

 $MSE = MSE_u + MSE_f$

$$egin{aligned} MSE_u &= rac{1}{N_u} \sum_{i=1}^{N_u} |u(t^i_u, x^i_u) - u^i|^2, \ MSE_f &= rac{1}{N_f} \sum_{i=1}^{N_f} |f(t^i_f, x^i_f)|^2. \end{aligned}$$

Deep Operator Networks (DeepONets)

- Learns maps between function spaces: solution operator of PDE
- There is some input parameter u the solution depends on. Could be forcing term / source term / IC / coefficient / domain geometry
- Learn the map between *u* and the whole spatio-temporal solution



Lu, L., Jin, P., & Karniadakis, G. E. (2019). Deeponet: Learning nonlinear operators for identifying differential equations based on the universal approximation theorem of operators. arXiv preprint arXiv:1910.03193.



What Has Been Done?

ML for Wound Healing

- 1D model: simple feedforward NN with 25 parameters as inputs and RSAW, displacement vector and strain energy as output
- Same for 2D model
- Fixed wound shape (rotated square in 2D) and data generated using the FEM model
- Hybrid model: NN as surrogate for computationally expensive step in FEM for solving 1D problem



Direction of Research

Research Question

Can a neural network be trained to predict the entire time evolution of skin contraction, given only the initial geometry of the wound as input?

For this we would incorporate:

- DeepONet
- PINNs setup in the loss function
- Data using the finite element simulation

Physics Informed DeepONets



Wang, S., Wang, H., & Perdikaris, P. (2021). Learning the solution operator of parametric partial differential equations with physics-informed DeepONets. Science advances, 7(40), eabi8605.

Approach

- Start simple
- Given one material parameter, we want to predict the displacement of the wound at fixed time T
- If this is feasible using the proposed set-up, use two material parameters
- Consider the input to the branch net to be a continuous function
- Consider the input to the branch net to be discrete (image of the wound)



Thank you for your attention!

Complete Model

$$\begin{split} \frac{\mathrm{D}N}{\mathrm{D}t} + N(\nabla \cdot \mathbf{v}) &= -\nabla \cdot \left(-D_F F \nabla N + \chi_F N \nabla c\right) + r_F \left(1 + \frac{r_F^{\max} c}{a_e^I + c}\right) (1 - \kappa_F F) N^{1+q} - k_F c N - \delta_N N, \\ \frac{\mathrm{D}M}{\mathrm{D}t} + M(\nabla \cdot \mathbf{v}) &= -\nabla \cdot \left(-D_F F \nabla M + \chi_F M \nabla c\right) + r_F \left(\frac{(1 + r_F^{\max})c}{a_e^I + c}\right) (1 - \kappa_F F) M^{1+q} - k_F c M - \delta_M M, \\ \frac{\mathrm{D}c}{\mathrm{D}t} + c(\nabla \cdot \mathbf{v}) &= -\nabla \cdot \left(-D_c \nabla c\right) + \frac{k_c (N + \eta^I M)c}{a_e^{II} + c} - \delta_c \frac{(N + \eta^{II} M)\rho}{1 + a_e^{III} c} c, \\ \frac{\mathrm{D}\rho}{\mathrm{D}t} + \rho(\nabla \cdot \mathbf{v}) &= k_\rho \left(1 + \frac{k_\rho^{\max} c}{a_e^{IV} + c}\right) (N + \eta M) - \delta_\rho \frac{(N + \eta^{II} M)\rho}{1 + a_e^{III} c} \rho, \\ \rho_t \left(\frac{\mathrm{D}\mathbf{v}}{\mathrm{D}t} + \mathbf{v}(\nabla \cdot \mathbf{v})\right) &= \nabla \cdot \left(\mu_1 \mathrm{sym}(\nabla \mathbf{v}) + \mu_2 [\mathrm{tr}(\mathrm{sym}(\nabla \mathbf{v}))\mathbf{I}] + \frac{E\sqrt{\rho}}{1 + \nu} \left(\varepsilon + \mathrm{tr}(\varepsilon) \frac{\nu}{1 - 2\nu}\mathbf{I}\right)\right) + \nabla \cdot \left(\xi M \left(\frac{\rho}{R^2 + \rho^2}\right)\mathbf{I}\right), \\ \frac{\mathrm{D}\varepsilon}{\mathrm{D}t} + \varepsilon \operatorname{skw}(\nabla \mathbf{v}) - \operatorname{skw}(\nabla \mathbf{v})\varepsilon + (\operatorname{tr}(\varepsilon) - 1)\operatorname{sym}(\nabla \mathbf{v}) = -\xi \left(\frac{(N + \eta^{II} M)c}{1 + a_e^{III} c}\right)\varepsilon. \end{split}$$