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Symmetry enables data-efficient velocity estimation Julian Suk, Christoph Brune, Jelmer M. Wolterink

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Introduction	Example results
 Hemodynamic velocity fields could be useful in diagnosis, prognosis and treatment planning Medical imaging, segmentation and computational fluid dynamics (CFD) to estimate velocity 	

- CFD is accurate but costly
- Data-driven surrogate methods that learn to infer velocity from artery shape
- Offline training on CFD-generated dataset, fast online inference for new artery
- [Problem] clincal datasets are usually small
- Make use of roto-translational (SE(3)) symmetry as inductive bias in graph neural network (GNN)
- GNN can focus on shape-intrinsic properties instead of orientation





Neural Network Architecture 100 copy & cat action O(3) tensor prod O(3) tensor prod pro SEGNN layer SEGNN layer SEGNN layer SEGNN layer SEGNN layer $\mathcal{V}^{1}\mathcal{V}^{2}$ SEGNN laye SEGNN laye SEGNN laye pood extend extend 30 tensor [%] mean eature ω 0(3) 14 $\gamma^2 \gamma^1$ $\gamma^0 \gamma^1$ $\mathcal{V}^1 \mathcal{V}^0$ 9 7 \mathcal{V}^0 \mathcal{V}^2



Data efficiency

Conclusion

- Steerable E(3)-equivariant graph neural network (SEGNN) [1]
- Three-scale pooling scheme [2]
- Predicts a discrete velocity field mapped to the vertices of the volumetric input mesh
- End-to-end equivariant to SE(3) transforms (roto-translations)

- We demonstrate how SEGNN can accurately estimate blood velocity in unseen arteries
- We achieve speed-up from 15 min (CFD) to 24.5 s (SEGNN)
- SEGNN is able to learn from a small dataset and makes data augmentation obsolete
- Physics-informed extensions are natural
- We are currently working on pulsatile flow and changing boundary conditions

Contact

[1] Brandstetter, J., Hesselink, R., van der Pol, E., Bekkers, E.J., Welling, M.: Geo- metric and physical quantities improve E(3) equivariant message passing. In: Pro- ceedings of the 10th International Conference on Learning Representations (2022)
[2] Suk, J., de Haan, P., Lippe, P., Brune, C., Wolterink, J.M.: Mesh convolutional neural networks for wall shear stress estimation in 3D artery models. In: MICCAI Workshop on Statistical Atlases and Computational Models of the Heart (2022)

References



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