



3D CT Proximal Femora Segmentation **Using Optimized nnU-Net**

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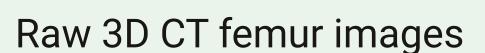
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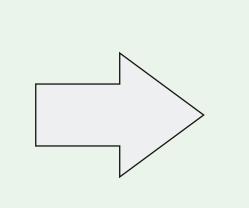
Institute of Visual Computing and Human-Centered Technology Rsearch Unit: Computer Visison

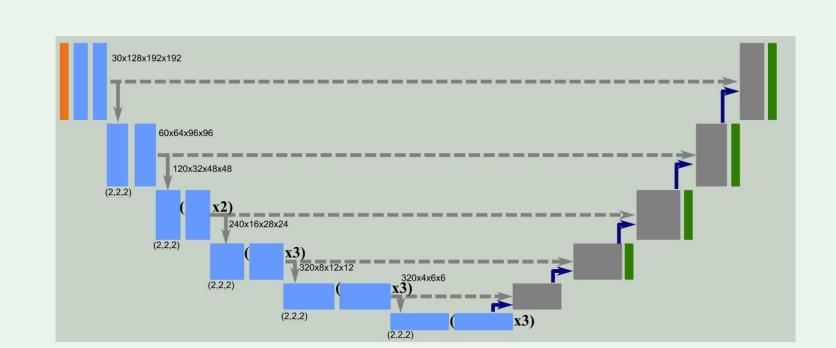
Co-Supervisor: Univ.Lektor Dr.techn. Alexander Pacha MScn

Main Objective

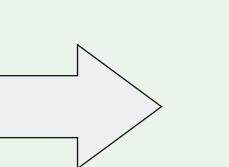


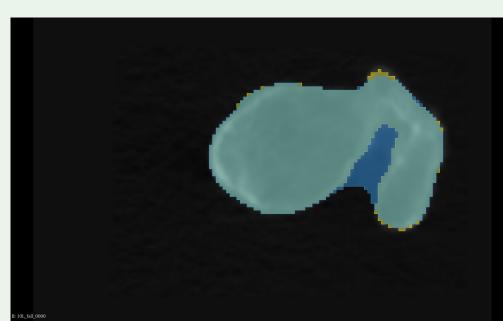


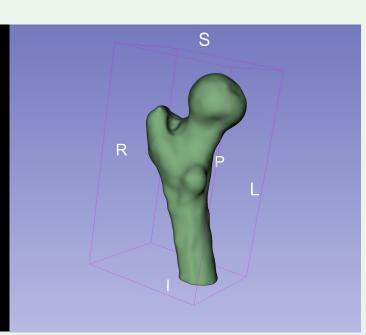




Train U-Net based segmentation model on ground truth labels







Automatically generate accurate segmentations, needed for simulation in biomechanics

Problem Definition

Accurate segmentation of bones from CT images is crucial for biomechanical modeling and medical planning. Manual segmentation is time-consuming and introduces human error, underlining the need for automated deep learning models. Classical algorithms using morphological operations often struggle with complex geometries, while deep learning models promise to have better local context. Despite the relatively recent introduction of larger transformer models, older U-Net-based models like the selfadapting nnU-Net continue to compete effectively. However, the optimal U-Net configurations, especially when comparing 2D slice-wise approaches to full 3D approaches, remain unclear. This project focuses on implementing and optimizing both 2D and 3D nnU-Net models for femur 3D CT images and analyzing their performance.

Results

Model	Dice	Sensitivity	Specificity	ASD
3D corrected	0.9895	0.9914	0.9981	0.1246
2D corrected	0.9879	0.9895	0.9979	0.1424
3D original	0.9923	0.9909	0.9990	0.0922
2D original	0.9892	0.9876	0.9985	0.1308

Model	Training Time [s]	Training VRAM [MiB]	Inference Time [s]	Inference VRAM [MiB]
3D	7343.8	4830	1.68	2000
2D	14017.2	9456	6.00	408

The parameter search only yielded slightly better performance for the 2D network. High performance scores were obtained on the test set for both the 2D and 3D models. In all cases the 3D model outperformed the 2D model in all performance metrics. The results for the models trained and tested on the original, not corrected labels were slightly better. The inference and training times for the 3D model was faster, while the VRAM usages were higher than the 2D model.

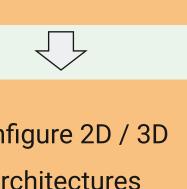
Methodology

Parameter Search

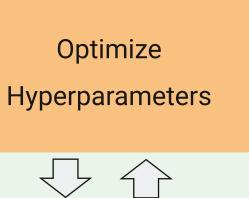
Various nnU-Net configurations were explored by tuning key hyperparameters like batch size, use of residual encoders, and data augmentation (DA5). Grid search was conducted on the chosen parameters on a separate validation split.

Preprocess Data / Manual Corrections in 3D Slicer

The best performing 2D and 3D configurations were selected for further training and evaluation on the test set.



Configure 2D / 3D architectures



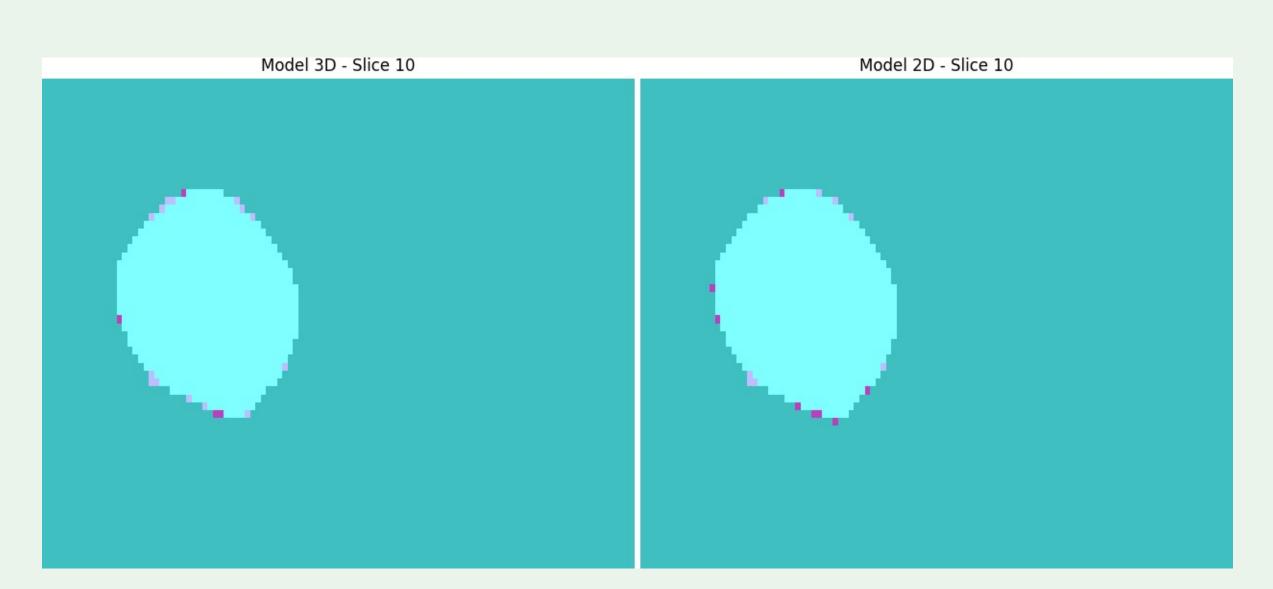
Evaluation in Valudation Split

Publish on Github to ensure reproducibility

> Analysis of model performance

Prediction on Set aside **Test Split**

Visualizations



Interactive visualizations, including an error map were implemented to compare model predictions against ground truth labels. These allow one to explore segmentation differences across CT slices and between models, providing a more detailed look into prediction differences.

Conclusion

This project successfully implemented both 2D and 3D nnU-Net models for femur segmentation in 3D CT images. The 3D model outperformed the 2D model across all performance metrics, including sensitivity, specificity, and average surface distance (ASD), while also providing faster inference times although higher resource requirements. The parameter search revealed that the default nnU-Net settings already performed well. Additionally, as the model trained and tested on the original labels performed better, it is indicated that more consistent while realistic labeling could further improve model performance.

nnU-Net Code Repository