Aalborg Universitet



Bones segmentation from lower extremity MRI for patientspecific modeling

Jung, Yunsub; Simonsen, Morten Bilde; Andersen, Michael Skipper

Published in: **Program & Abstracts**

Publication date: 2023

Link to publication from Aalborg University

Citation for published version (APA): Jung, Y., Simonsen, M. B., & Andersen, M. S. (2023). Bones segmentation from lower extremity MRI for patientspecific modeling. In Program & Abstracts: 15th Annual Meeting of the Danish Society of Biomechanics

General rights

Copyright and moral rights for the publications made accessible in the public portal are retained by the authors and/or other copyright owners and it is a condition of accessing publications that users recognise and abide by the legal requirements associated with these rights.

- Users may download and print one copy of any publication from the public portal for the purpose of private study or research.
- You may not further distribute the material or use it for any profit-making activity or commercial gain You may freely distribute the URL identifying the publication in the public portal -

Take down policy

If you believe that this document breaches copyright please contact us at vbn@aub.aau.dk providing details, and we will remove access to the work immediately and investigate your claim.

BONES SEGMENTATION FROM LOWER EXTREMITY MRI FOR PATIENT-SPECIFIC MODELING

Yunsub Jung^{1,2*}, Morten Blide Simonsen^{1,2} and Michael Skipper Andersen^{1,2}

¹Department of Materials and Production, Aalborg University, Denmark

²Center for Mathematical Modeling of Knee Osteoarthritis, Aalborg University, Denmark

*Ph.D.student, yunsubj@mp.aau.dk

INTRODUCTION

Medical imaging is widely used in the field of biomechanics to construct patient-specific musculoskeletal models [1]. In particular, magnetic resonance imaging (MRI) provides the most comprehensive evaluation of geometries, joint damage, and osteoarthritis and is a good multi-plane image for soft tissue contrast [2]. However, manual segmentation of human structures is a time-consuming and labor-intensive task and a bottleneck for future advancements in the field. This study presents a method to extract bones from lower extremity MRI images automatically.

METHODS

The total number of segmentation target bones is 10 (pelvis, femur, tibia, patella, and tarsal). Lower extremity MRI images (field strength 1.5T, sequence: T1-fl2d) stitched into three or four scans were used. A multi-atlas-based method was used to segment bones, and a total of 27 atlas data sets were created through manual segmentation. Among these, 9 atlas data were finally selected considering image quality and registration performance. All atlas data include information on manually marked bone regions and the original MRI images.



Fig. 1 Schematic diagram of the segmentation algorithm

Segmentation proceeded in the following steps: atlas selection, segmentation of body surface, localization, image registration using a pyramid model, similarity computation between target and atlas image, and weighted voting (Fig. 1). Body surface information was used to reduce the amount of computation when computing the registration and similarity. In addition, the number of iterations of the pyramid model was determined by considering the segmentation results and computation time. The 10 bones finally segmented have different label values in the image, and the segmentation results for each bone were quantitatively compared with manual segmentations. Four metrics (sensitivity, specificity, accuracy, and precision) were used for quantitative comparison.

RESULTS AND DISCUSSION

Table 1 shows the segmentation results according to the similarity criterion. The two similarity criterion methods showed similar results but different performances depending on the bone segmented. It is known that the pelvis shapes in males and females are different. Atlas-based segmentation is based on the morphological information of the target object. However, our atlas data did not distinguish between male and female pelvis. In the future, we plan to improve this through additional research.



Fig. 1 Segmentation results: original image (left), manual marking (middle), segmented result (right)

CONCLUSIONS

Atlas-based methods have shown good overall performance for bone segmentation, but expansion of the atlas set is required for the pelvis.

REFERENCES

1. Satanik M., et al. *Front in Bioe Biot* **8**: 1-20, 2020 2. Qi L., et al. *Imaging* **8**: 418-28, 201

Table 1 Exp	perimental	results	
-------------	------------	---------	--

Metric	Left				Right					
	Tibia	Femur	Pelvis	Patella	Tarsal	Tibia	Femur	Pelvis	Patella	Tarsal
Sensitivity	0.771	0.526	0.279	0.696	0.360	0.714	0.631	0.258	0.715	0.564
Specificity	0.997	0.988	0.999	0.999	0.994	0.997	0.994	0.999	0.998	0.994
Accuracy	0.991	0.966	0.986	0.991	0.964	0.990	0.977	0.985	0.991	0.974
Precision	0.866	0.703	0.771	0.925	0.757	0.849	0.853	0.782	0.896	0.837
Sensitivity	0.772	0.526	0.281	0.692	0.351	0.714	0.631	0.262	0.717	0.561
Specificity	0.997	0.988	0.999	0.999	0.994	0.997	0.994	0.999	0.998	0.994
Accuracy	0.991	0.966	0.986	0.991	0.964	0.990	0.977	0.985	0.991	0.973
Precision	0.867	0.703	0.771	0.925	0.756	0.851	0.852	0.784	0.899	0.835
	Metric Sensitivity Specificity Accuracy Precision Sensitivity Specificity Accuracy Precision	MetricTibiaSensitivity0.771Specificity0.997Accuracy0.991Precision0.866Sensitivity0.772Specificity0.997Accuracy0.991Precision0.867	Metric Tibia Femur Sensitivity 0.771 0.526 Specificity 0.997 0.988 Accuracy 0.991 0.966 Precision 0.866 0.703 Sensitivity 0.772 0.526 Specificity 0.997 0.988 Accuracy 0.997 0.988 Accuracy 0.991 0.966 Precision 0.867 0.703	Metric Left Tibia Femur Pelvis Sensitivity 0.771 0.526 0.279 Specificity 0.997 0.988 0.999 Accuracy 0.991 0.966 0.986 Precision 0.866 0.703 0.771 Specificity 0.772 0.526 0.281 Specificity 0.997 0.988 0.999 Accuracy 0.991 0.966 0.986 Precision 0.867 0.703 0.771	Metric Left Tibia Femur Pelvis Patella Sensitivity 0.771 0.526 0.279 0.696 Specificity 0.997 0.988 0.999 0.999 Accuracy 0.991 0.966 0.986 0.991 Precision 0.866 0.703 0.771 0.925 Sensitivity 0.772 0.526 0.281 0.692 Specificity 0.997 0.988 0.999 0.999 Accuracy 0.991 0.966 0.986 0.991 Precision 0.867 0.703 0.771 0.925	Metric Left Tibia Femur Pelvis Patella Tarsal Sensitivity 0.771 0.526 0.279 0.696 0.360 Specificity 0.997 0.988 0.999 0.999 0.994 Accuracy 0.991 0.966 0.986 0.991 0.964 Precision 0.866 0.703 0.771 0.925 0.757 Sensitivity 0.772 0.526 0.281 0.692 0.351 Specificity 0.997 0.988 0.999 0.994 0.964 Precision 0.867 0.703 0.771 0.925 0.756	Metric Left Tibia Femur Pelvis Patella Tarsal Tibia Sensitivity 0.771 0.526 0.279 0.696 0.360 0.714 Specificity 0.997 0.988 0.999 0.999 0.994 0.997 Accuracy 0.991 0.966 0.986 0.991 0.964 0.990 Precision 0.866 0.703 0.771 0.925 0.757 0.849 Sensitivity 0.772 0.526 0.281 0.692 0.351 0.714 Specificity 0.997 0.988 0.999 0.994 0.997 Accuracy 0.997 0.988 0.999 0.994 0.997 Accuracy 0.997 0.988 0.999 0.994 0.997 Accuracy 0.991 0.966 0.986 0.991 0.964 0.990 Precision 0.867 0.703 0.771 0.925 0.756 0.851	Metric Left Tibia Femur Pelvis Patella Tarsal Tibia Femur Sensitivity 0.771 0.526 0.279 0.696 0.360 0.714 0.631 Specificity 0.997 0.988 0.999 0.999 0.994 0.997 0.994 Accuracy 0.991 0.966 0.986 0.991 0.964 0.990 0.977 Precision 0.866 0.703 0.771 0.925 0.757 0.849 0.853 Sensitivity 0.772 0.526 0.281 0.692 0.351 0.714 0.631 Specificity 0.997 0.988 0.999 0.994 0.997 0.994 Accuracy 0.997 0.988 0.999 0.994 0.997 0.994 Accuracy 0.991 0.966 0.986 0.991 0.964 0.990 0.977 Precision 0.867 0.703 0.771 0.925 0.756 0.851	Metric Left Right Tibia Femur Pelvis Patella Tarsal Tibia Femur Pelvis Sensitivity 0.771 0.526 0.279 0.696 0.360 0.714 0.631 0.258 Specificity 0.997 0.988 0.999 0.994 0.997 0.994 0.999 Accuracy 0.991 0.966 0.986 0.991 0.964 0.990 0.977 0.985 Precision 0.866 0.703 0.771 0.925 0.757 0.849 0.853 0.782 Sensitivity 0.772 0.526 0.281 0.692 0.351 0.714 0.631 0.262 Specificity 0.997 0.988 0.999 0.994 0.997 0.994 0.999 Accuracy 0.997 0.988 0.999 0.994 0.997 0.994 0.999 Accuracy 0.991 0.966 0.986 0.991 0.964 0.990 0.977	Metric Left Right Tibia Femur Pelvis Patella Tarsal Tibia Femur Pelvis Patella Sensitivity 0.771 0.526 0.279 0.696 0.360 0.714 0.631 0.258 0.715 Specificity 0.997 0.988 0.999 0.999 0.994 0.997 0.994 0.999 0.998 Accuracy 0.991 0.966 0.986 0.991 0.964 0.990 0.977 0.985 0.991 Precision 0.866 0.703 0.771 0.925 0.757 0.849 0.853 0.782 0.896 Sensitivity 0.772 0.526 0.281 0.692 0.351 0.714 0.631 0.262 0.717 Specificity 0.997 0.988 0.999 0.994 0.997 0.994 0.999 0.998 Accuracy 0.991 0.966 0.986 0.991 0.964 0.990 0.977 0.985

Note.-NMI = normalized mutual information, NCC = normalized cross correlation *Similarity criterion