

Abstract

Development of AI-based segmentation and anatomical reconstruction for orbital floor implants using medical image data

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Orbital floor fractures are common in craniomaxillofacial trauma, and patient specific implants (PSI) are required to restore the anatomical structure. The current workflow to design the PSI involves manual segmentation of the skull, manual mirroring the healthy orbit to the defected one and manual adapting the implant design to the patient's anatomy. [1] The aim of the work presented here is to automate the process by using AI. The methods used are based on research in the field of cranial implants [2]. The first step involves AI-based preprocessing of the medical image data. For this purpose, a Convolutional Neural Network, the Dense U-Net [3], was trained and used to segment the skull. The next step is the virtual reconstruction of the orbital anatomy. Instead of mirroring, a Statistical Shape Model (SSM) was created. The SSM was based on healthy skulls and its shape can be fitted onto a defected skull. [4] Both process steps were implemented in Python and training runs were conducted. Due to data protection, less than 100 datasets were available for training. In addition, for orbital defects, no image data is available prior to the occurrence of the defect. For this reason, defects were virtually inserted into healthy skull models to test the created SSM. Therefore, more datasets are needed to validate the application in practice.

AUTHOR'S STATEMENT

Conflict of interest: Authors state no conflict of interest. Animal models: Animal models have not been used in the present research. Informed consent: Informed consent has been obtained from all individuals included in this study. Ethical approval: The research related to human use complies with all the relevant national regulations, institutional policies and was performed in accordance with the tenets of the Helsinki Declaration, and has been approved by the authors' institutional review board or equivalent committee.

Acknowledgments: The work presented is part of the project DigiMed and has received funding by the European Regional Development Fund in the framework of the REACT-EU program.

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