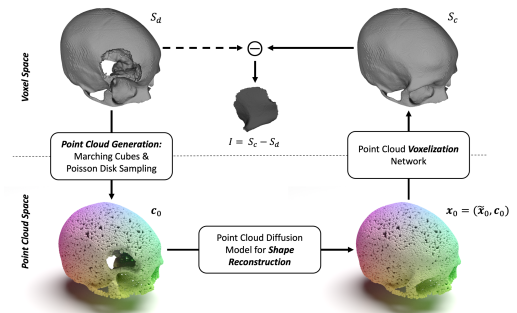


Master of Science – Biomedical Engineering, Thesis Proposal

## Improved Point Cloud Diffusion Models for Automatic Implant Generation

Advances in 3D printing of biocompatible materials make patient-specific implants increasingly popular. The design of these implants is, however, still a largely manual and time-consuming process, which is usually carried out by trained experts. To automate and speed up this design process, we developed an automatic deep learning solution presented in [1]. The method consists of a point cloud diffusion model for shape reconstruction and a voxelization network. For further information on the developed method check out the project page: <https://pfriedri.github.io/pcdiff-implant-io/>.



Due to the iterative sampling process of the used Denoising Diffusion Probabilistic Model (DDPM,  $T=1000$ ) and computationally expensive postprocessing steps, the time required for generating an implant is around 25-30 minutes, which we aim to speed up significantly. This goal could be achieved by applying different sampling strategies [2, 3] or distillation methods [4], adapting the used conditioning strategy of the method (namely by using less points for conditioning), by applying a different network architecture or by using a more efficient implementation of the postprocessing steps (mainly the undercut removal). The student should try to reduce the runtime of the overall method, while keeping the good implant quality and the methods low memory footprint. The goal of this thesis would be to implement and compare different “speed-up methods”, to adapt the existing code base accordingly, and to conduct extensive experiments to verify the success of the applied changes.

### Nature of the Thesis

Programming: 80%, Documentation: 20%

### Specific Requirements

Experience in deep learning & good programming skills (*Python/ PyTorch*)

### Supervisors

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### References

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- [2] Song, J., Meng, C., & Ermon, S. (2020). Denoising Diffusion Implicit Models. In International Conference on Learning Representations – ICLR 2020.
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