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## Deep learning for tomographic reconstruction with limited data

Hendriksen, A.A.

### Citation

Hendriksen, A. A. (2022, March 3). *Deep learning for tomographic reconstruction with limited data*. Retrieved from <https://hdl.handle.net/1887/3277969>

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# **Deep learning for tomographic reconstruction with limited data**

Proefschrift

ter verkrijging van  
de graad van doctor aan de Universiteit Leiden,  
op gezag van rector magnificus prof. dr. ir. H. Bijl,  
volgens besluit van het college voor promoties  
te verdedigen op donderdag 3 maart 2022  
klokke 15:00 uur

door

Allard Adriaan Hendriksen  
geboren te Muscat, Oman  
in 1991

**Promotor:**

Prof. dr. K. J. Batenburg

**Co-promotor:**

Dr. D. M. Pelt

**Promotiecommissie:**

Prof. dr. F.A.J. de Haas

Prof. dr. R.M. van Luijk

Prof. dr. P.D. Grünwald

Prof. dr. ir. B.P.F. Lelieveldt

Dr. F. Marone Welford (Paul Scherrer Institute)

Prof. dr. O. Öktem (KTH Royal Institute of Technology)



The research presented in this dissertation was carried out at the Centrum Wiskunde & Informatica (CWI) in Amsterdam.

Financial support was provided by the Netherlands Organisation for Scientific Research (NWO), programme 639.073.506.

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