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Deep generative models for engineering design

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Citation

Fan, J. (2026, March 24). *Deep generative models for engineering design*.
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Deep Generative Models for Engineering Design

Proefschrift

ter verkrijging van
de graad van doctor aan de Universiteit Leiden,
op gezag van rector magnificus prof.dr. S. de Rijcke,
volgens besluit van het college voor promoties
te verdedigen op dinsdag 24 maart 2026
klokke 16:00 uur

door

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geboren te Wenzhou, China
in 1996

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This dissertation was made possible through the support of BMW AG which provided me with a full-time PhD position, thereby facilitating my research and studies as a guest PhD candidate at the Leiden Institute Of Advanced Computer Science, Leiden University.

“Nothing whets the intelligence more than a passionate suspicion, nothing develops all the faculties of an immature mind more than a trail running away into the dark.”

Stefan Zweig

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