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Deep generative models for engineering design

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Deep Generative Models for Engineering Design

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“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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