



# Enhancing Computer-Aided Design with Deep Learning Frameworks: A Literature Review

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## Keywords:

Machine Learning, Deep Learning, Generative Design, Deep CAD, CAD Modelling, Product Engineering, Early Design Phase, Design Generation, Design Reconstruction, Design Retrieval, Engineering Design

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The following literature list forms the foundation for the analysis of deep learning frameworks in CAD across three key design areas: Design Generation, Design Reconstruction, and Design Retrieval.

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