



**Department of Computer Science & Engineering**  
**IPS Academy, Institute of Engineering & Science**  
**Indore**

**Computing Midcurve with Multi-layer and  
Convolutional Neural Networks**

**Paper ID: 655**

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# PAPER TITLE

Paper ID: **655**

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# OVERVIEW

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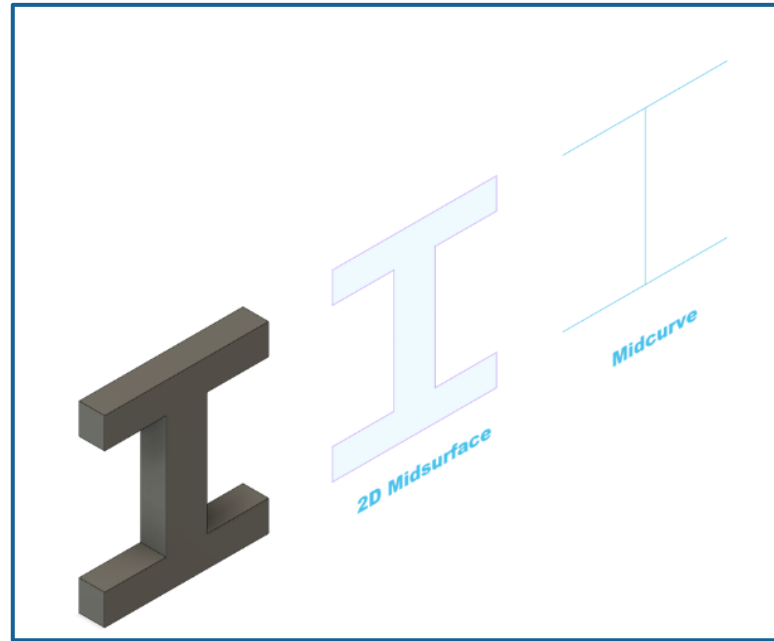
# Abstract

- Midcurve computation: dimension reduction from 2D to 1D
- Applications in CAD/CAE and robot path planning
- Two new neural architectures:
  - Multi-layer dense network
  - CNN-based architecture with skip connections
- Key achievement: Tenfold reduction in average loss



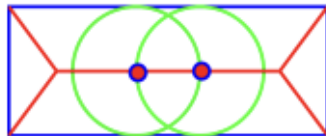
# Introduction

- What is Midcurve?
- Why is it important?
- Applications: • FEA • Robot path planning • Character animation

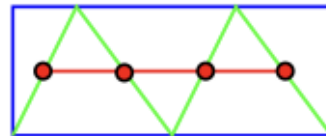


# Literature/Related Work

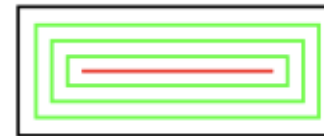
- Traditional Methods:
  - MAT (Medial Axis Transform)
  - CAT (Chordal Axis Transform)
  - Thinning-based methods
- Limitations of existing approaches



MAT



CAT



Thinning



Pairs

# Methodology

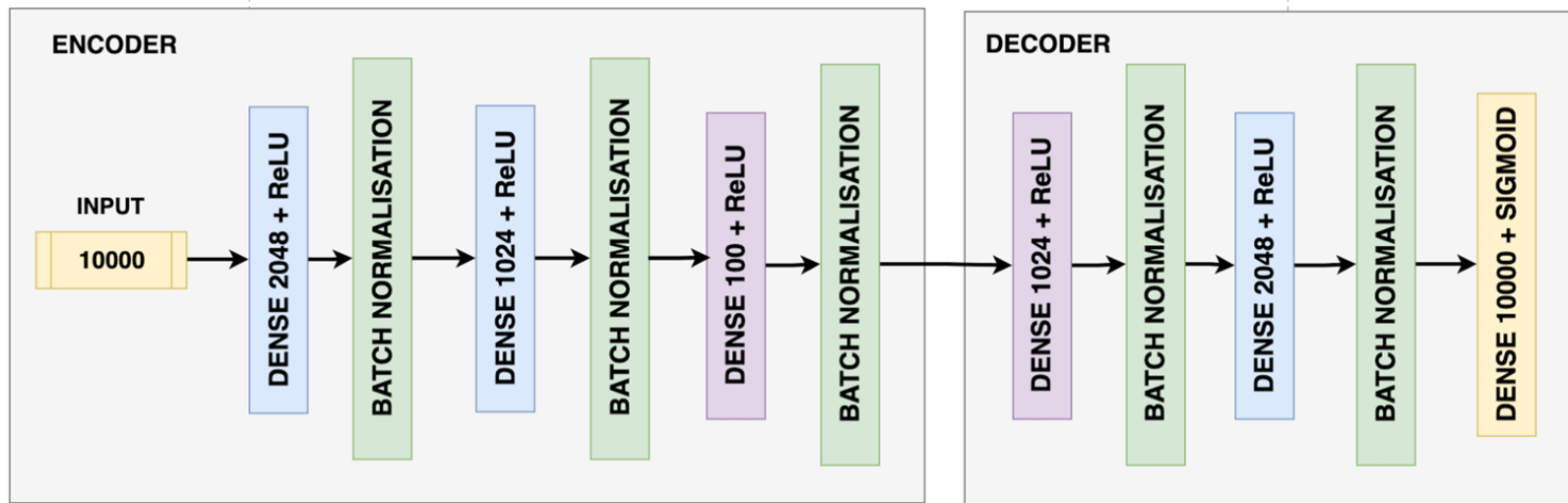
- Image-based representation
- Two proposed architectures:
  - Dense Network
  - CNN with skip connections
- Data preprocessing: 128x128 pixels





# Dense Network Architecture

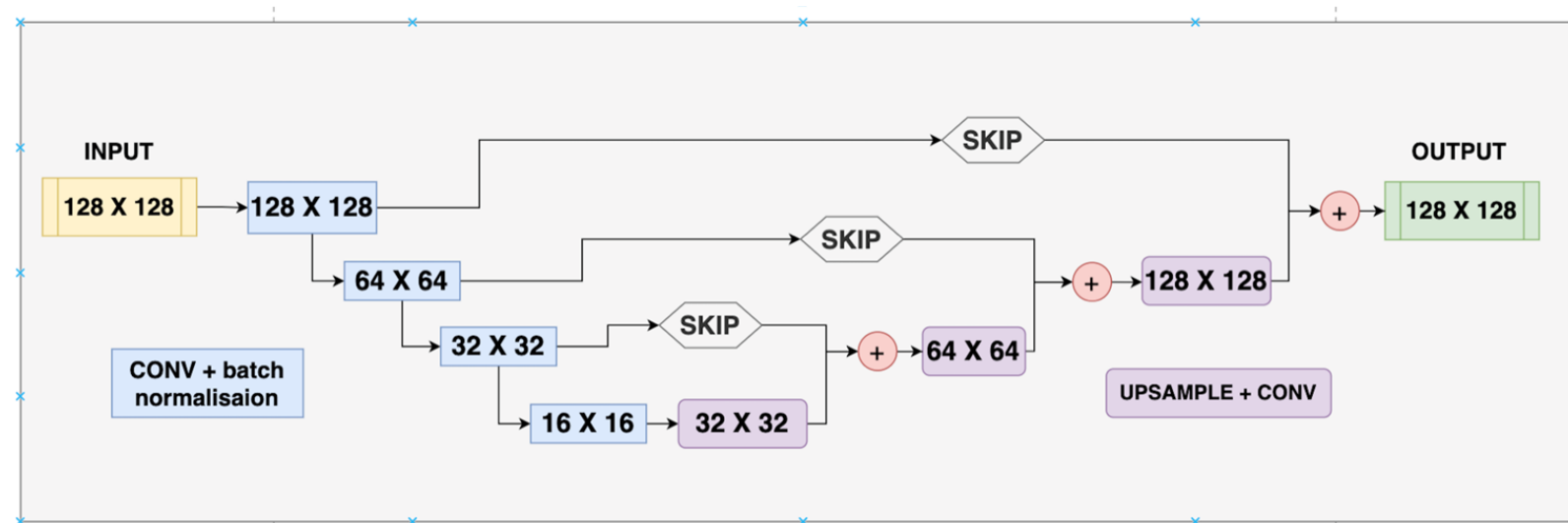
- Gradual dimension reduction
- Multiple dense layers
- ReLU activation
- Symmetric encoder-decoder





# CNN Architecture

- 4 convolutional blocks
- Skip connections
- Batch normalization
- Dynamic learning rate



# Results

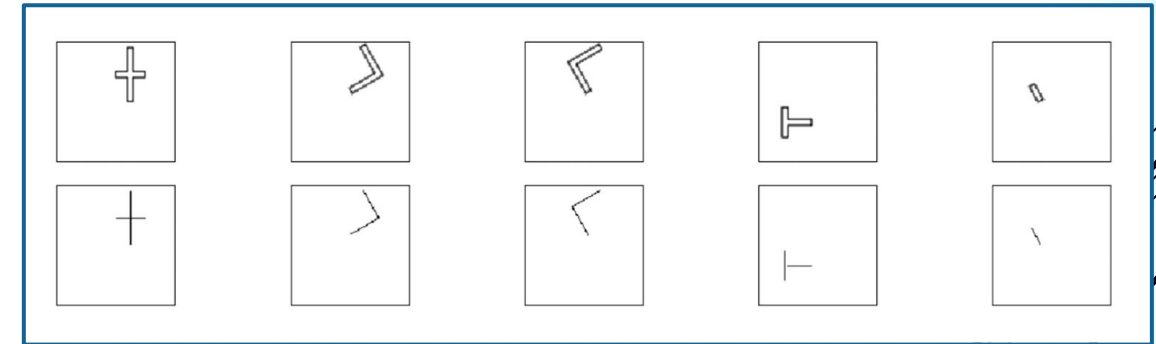
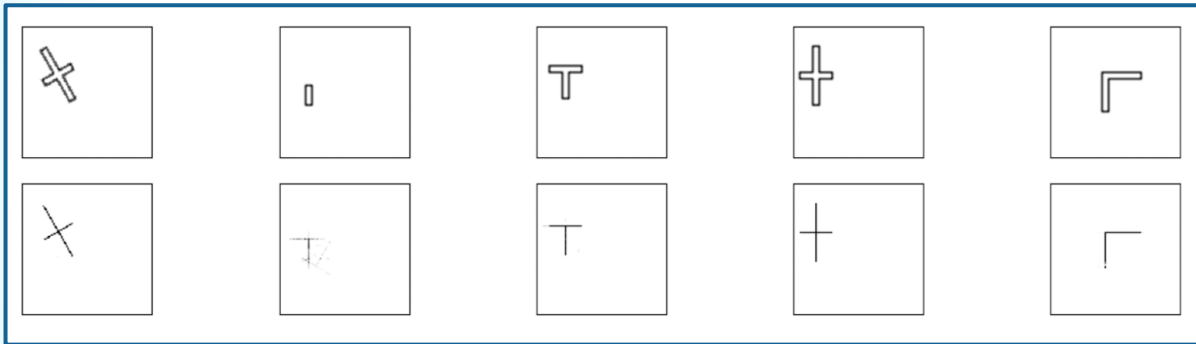
- Performance Metrics
- Comparative Analysis

Metric	Simple	Dense	CNN
Best Epoch	100	62	93
Training Loss	0.0034	0.0049	<b>0.0003</b>
Training MAE	0.0023	0.0032	<b>0.0003</b>
Validation Loss	0.0080	0.0121	<b>0.0005</b>



# Visual Results

- Input shapes
- Generated Midcurves
- Quality comparison



# Conclusion

- CNN architecture performs best
- 10x reduction in loss
- Improved geometric accuracy

