

Lecture Note of EE: 541

A Computational Introduction to Deep Learning

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Part I

Basic theory

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what is

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History

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Python

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Decision and Logistic Regression

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Convolutional Neural Networks(CNNs)

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Deep Learning workflow

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Dimensionality Reduction: PCA and LDA

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