

AI Experiment 10 – C3-1 – C165

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Aim: Case Study of AI Applications

Topic: Foundational Research Papers in Neural Networks and Deep Learning

Introduction:

The field of deep learning has been shaped by several groundbreaking papers, each introducing or enhancing neural network architectures that have since become foundational. Below is a summary of five key research papers that introduced Recurrent Neural Networks (RNN), Long Short-Term Memory Networks (LSTM), Convolutional Neural Networks (CNN), and Generative Pretrained Transformers (GPT), along with an additional paper to highlight another key advancement in neural networks.

Theory:

1. Learning Representations by Back-Propagating Errors (1986)

Written by Rumelhart, Hinton, and Williams in 1986, this paper introduced the concept of backpropagation for training neural networks. The method solved the problem of adjusting the weights of a neural network in a systematic way, leading to the development of recurrent neural networks (RNNs). Backpropagation allowed for gradient-based learning in neural networks, becoming a cornerstone technique for training deep learning models.

2. Long Short-Term Memory (1997)

Hochreiter and Schmidhuber proposed the Long Short-Term Memory (LSTM) network in 1997 to solve the vanishing gradient problem associated with traditional RNNs. LSTM introduced a memory cell that could maintain information over long periods, allowing it to capture long-range dependencies in sequence data. LSTMs became widely used for tasks such as speech recognition, machine translation, and time series forecasting.

3. Gradient-Based Learning Applied to Document Recognition (1998)

Yann LeCun and colleagues introduced Convolutional Neural Networks (CNNs) in this 1998 paper. The model, LeNet-5, demonstrated the effectiveness of CNNs for image classification tasks. CNNs leverage convolutional layers to automatically detect spatial hierarchies in data, making them highly effective for computer vision tasks like object detection, image classification, and video processing.

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4. Improving Language Understanding by Generative Pre-training (2018)

The first GPT model, introduced by Radford et al. in 2018, pioneered the use of a Transformer model pretrained on a large corpus of text. The idea was to pretrain a model on unsupervised tasks and fine-tune it on specific tasks such as question-answering or translation. GPT became a groundbreaking architecture in NLP, influencing models such as BERT and GPT-3.

5. ImageNet Classification with Deep Convolutional Neural Networks (2012)

Alex Krizhevsky, Ilya Sutskever, and Geoffrey Hinton's 2012 paper introduced AlexNet, a deep convolutional neural network that achieved breakthrough performance on the ImageNet dataset. AlexNet's success marked the beginning of the deep learning revolution in computer vision, thanks to its innovative use of ReLU activations, dropout regularization, and GPU acceleration.

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Summary Table:

Sr. No	Paper Title	Publication Name, Year	Algorithm / Technique Used	Dataset (if any)	Results Mentioned in the Paper	Your Observations
1	Learning Representations by Back-Propagating Errors	Rumelhart, Hinton, and Williams, 1986	Backpropagation	None	Introduced backpropagation, a method for gradient-based learning in neural networks; foundational for training RNNs.	<ul style="list-style-type: none"> - Essential method for deep learning model training - Revolutionized neural network weight adjustment
2	Long Short-Term Memory	Hochreiter and Schmidhuber, 1997	Long Short-Term Memory (LSTM)	None	Solved the vanishing gradient problem in RNNs, with memory cells that retain information for longer periods, useful in sequential data.	<ul style="list-style-type: none"> - Enabled long-range dependencies in data sequence processing - Significant impact on speech and text processing
3	Gradient-Based Learning Applied to Document Recognition	Yann LeCun et al., 1998	Convolutional Neural Networks (CNNs)	Image datasets	Showcased CNN's effectiveness in image recognition, introducing LeNet-5; demonstrated CNN's capability for hierarchical spatial data processing.	<ul style="list-style-type: none"> - A key model for computer vision applications - CNN's success led to adoption in various image-based applications
4	Improving Language Understanding by Generative Pre-training	Radford et al., 2018	Generative Pretrained Transformers (GPT)	Large text corpora	GPT's architecture for language models pre-trained on unsupervised tasks, later fine-tuned for specific NLP tasks; influenced other models like BERT and GPT-3.	<ul style="list-style-type: none"> - GPT's approach became the backbone of modern NLP systems - Transformer architecture revolutionized language model capabilities
5	ImageNet Classification with Deep	Alex Krizhevsky, Ilya	AlexNet (Deep CNN)	ImageNet Dataset	Achieved groundbreaking performance on	<ul style="list-style-type: none"> - AlexNet marked a new era in computer vision

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	Convolutional Neural Networks	Sutskever, Geoffrey Hinton, 2012			ImageNet with ReLU activations, dropout regularization, and GPU acceleration, marking a pivotal moment in deep learning for computer vision.	with CNNs - Its architecture is the foundation of many modern computer vision models
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Conclusion:

These foundational papers have each left a lasting impact on the development of AI and deep learning. From the backpropagation algorithm that enabled RNNs, to the LSTM architecture that improved sequence learning, to the rise of CNNs and Transformers, these innovations have reshaped the capabilities of machine learning systems and continue to influence ongoing research.