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A.P. SHAH INSTITUTE OF TECHNOLOGY

Department of Computer Science and Engineering Data Science



Module 4

Fully Connected Layer vs Convolutional Layer

Confusion between *Fully Connected Layers* (*FC*) and *Convolutional Layers* is common due to terminology overlap. In CNNs, convolutional layers are used for feature extraction followed by FC layers for classification that makes it difficult for beginners to distinguish there roles.

This article compares Fully Connected Layers (FC) and Convolutional Layers (Conv) in neural networks, detailing their structures, functionalities, key features, and usage in deep learning architectures.

What is a Fully Connected Layer?

A *Fully Connected (FC) layer*, aka a *dense layer*, is a type of layer used in <u>artificial neural networks</u> where each neuron or node from the previous layer is connected to each neuron of the current layer. It's called "*fully connected*" because of this complete linkage. FC layers are typically found towards the end of neural network architecture and are responsible for producing final output predictions.

What is a Convolutional Layer?

Convolutional layers are the building blocks of <u>convolutional neural networks (CNNs)</u>, which are primarily used for tasks that require the recognition and processing of spatial data, such as images and videos. These layers apply a convolution operation to the input, passing the result to the next layer.

Key Differences Between Fully Connected Layer and Convolutional Layer

- **Parameter Efficiency:** Convolutional layers are more parameter-efficient compared to fully connected layers as convolutional layers can learn local patterns using small filters applied across the input space whereas the fully connected layers learn global patterns which require more parameters.
- **Data Suitability:** Convolutional layers are specifically advantageous for spatial data such as images, where locality and translation invariance are important whereas the fully connected layers are more flexible and can be used with any form of data.
- **Feature Learning:** Convolutional layers are designed to automatically learn and generalize features from the input data, such as edges in the initial layers followed by more complex

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structures in deeper layers and the fully connected layers do not inherently recognize such hierarchical patterns without prior reshaping of the input data.

• Usage in Architectures: In practice, many deep learning architectures use a combination of both types of layers. Convolutional layers are typically used in the earlier stages to extract and learn features, while fully connected layers are often used at the end of the network to make predictions based on these features.

Fully Connected Layer vs Convolutional Layer

Features	Fully Connected Layer	Convolutional Layer
Definition	Every neuron is connected to every neuron in the previous layer.	Neurons are connected only to a local region of the previous layer.
Connectivity	Dense connections; each neuron connects to all neurons in the previous layer.	Sparse connections; each neuron connects only to a local patch of the input.
Parameters	Large number of parameters due to full connectivity.	Fewer parameters due to shared weights and local connectivity.
Weight Sharing	No weight sharing; each connection has its own weight.	Weights are shared across spatial positions, reducing the number of parameters.
Typical Use Cases	Final classification layers in neural networks.	Feature extraction, especially in image and video processing.
Computation Cost	Higher computational cost due to large number of connections.	Lower computational cost per neuron due to local connections.

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Features	Fully Connected Layer	Convolutional Layer
Overfitting	Higher risk of overfitting due to large number of parameters.	Lower risk of overfitting due to fewer parameters and regularization effects of local connections.
Dimensionality Reduction	Does not inherently reduce dimensionality.	Can reduce dimensionality through pooling layers.
Examples	Multilayer Perceptron (MLP), Dense layers in CNNs.	Convolutional Neural Networks (CNNs), such as layers in AlexNet, VGGNet.

Conclusion

Deep learning has transformed various fields through Fully Connected Neural Networks (FCNNs) and Convolutional Neural Networks (CNNs). FC layers excel in final classification tasks with dense connections, while convolutional layers efficiently extract spatial features with fewer parameters. The combination of these architectures enables powerful image and video processing capabilities.