



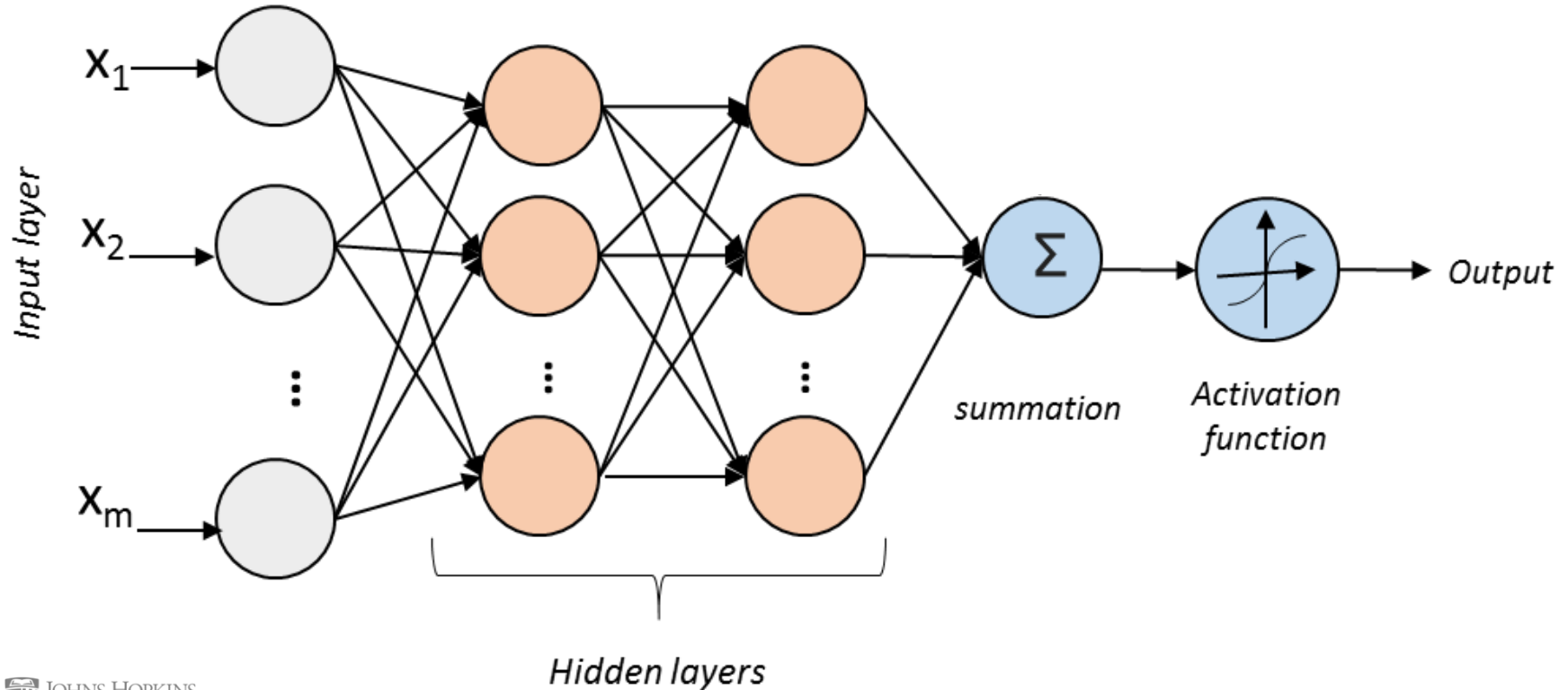
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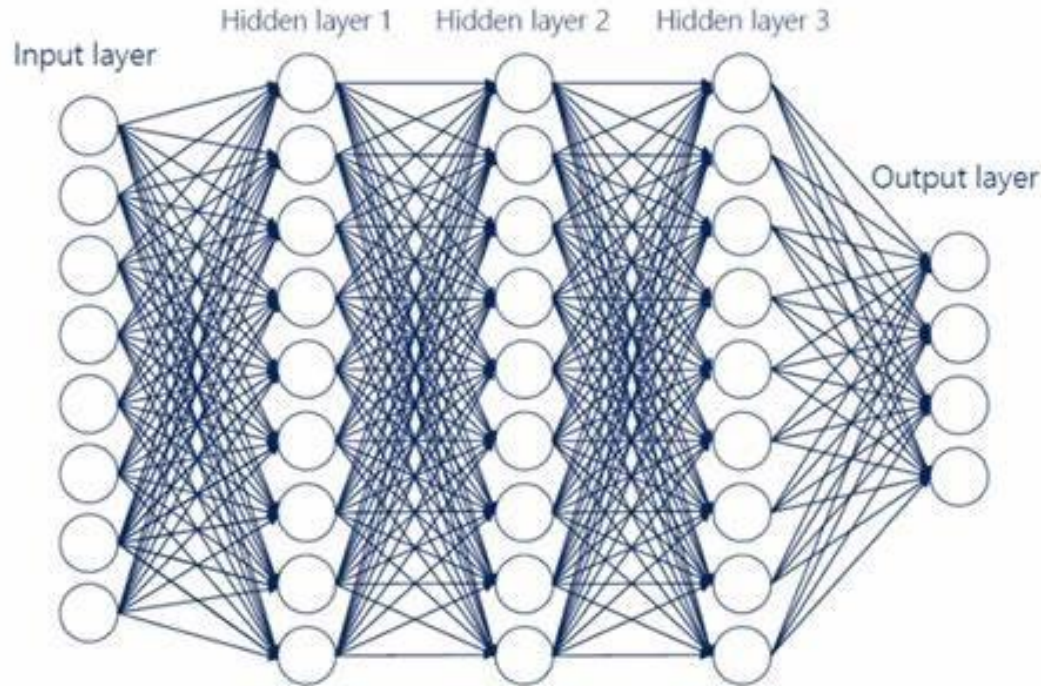
# 685.621 Algorithms for Data Science

Neural Networks: Architecture Types

# Multi-Layer Perceptron (MLP)

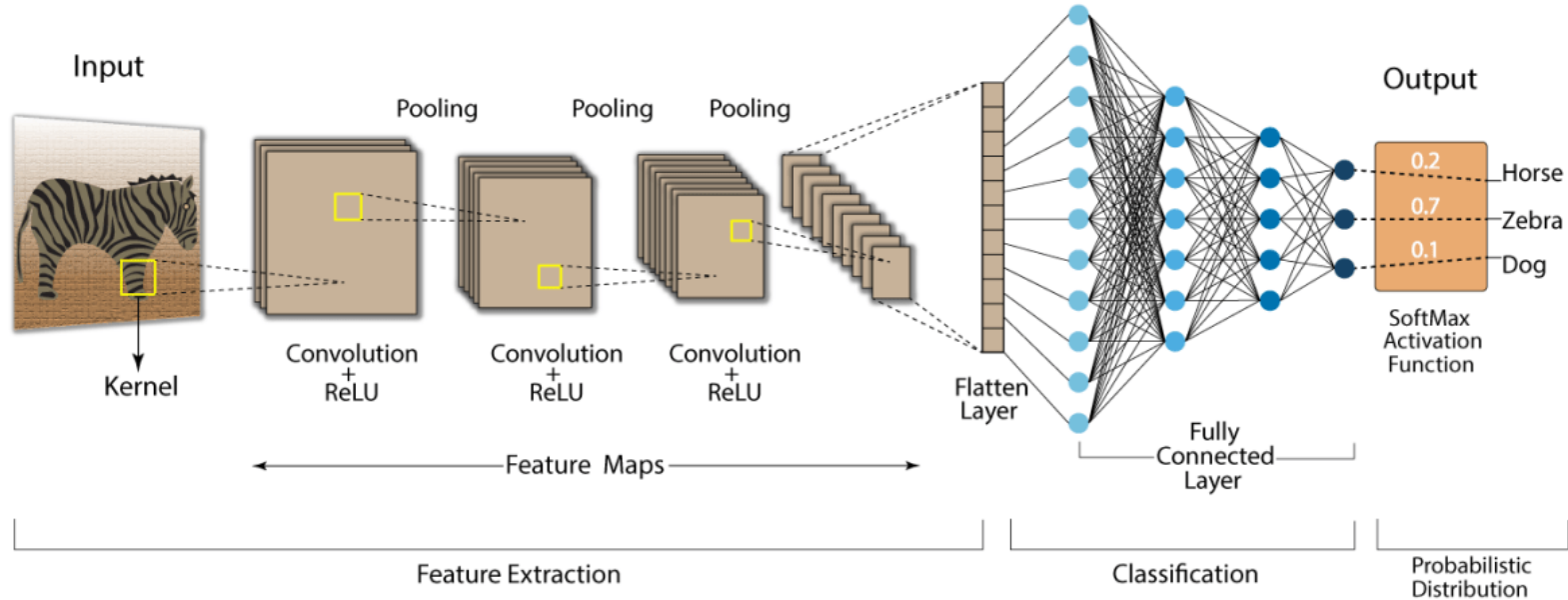


# Deep Neural Networks (DNN)



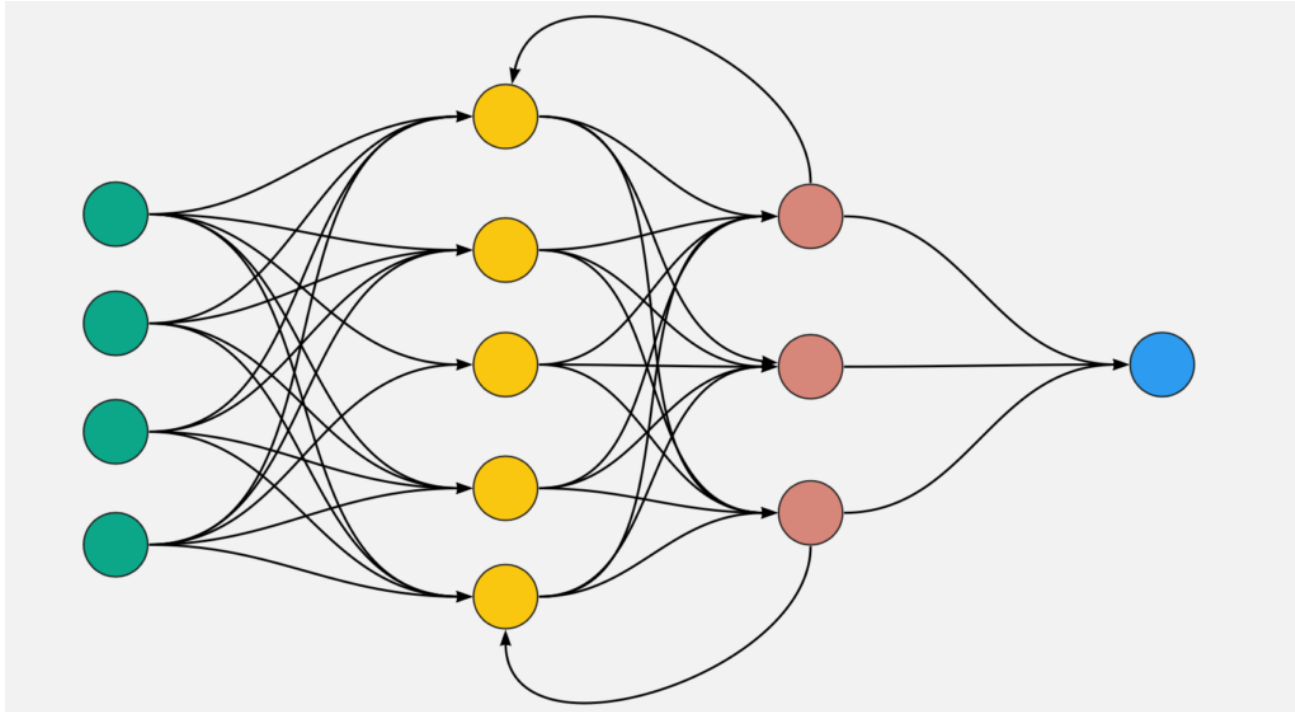
Merenda, Massimo & Porcaro, Carlo & Iero, Demetrio. (2020). Edge Machine Learning for AI-Enabled IoT Devices: A Review. *Sensors*. 20. 2533. 10.3390/s20092533.

# Convolutional Neural Networks (CNN)



[All About Convolutional Neural Networks Cnns By Savindi Wijenayaka](#)

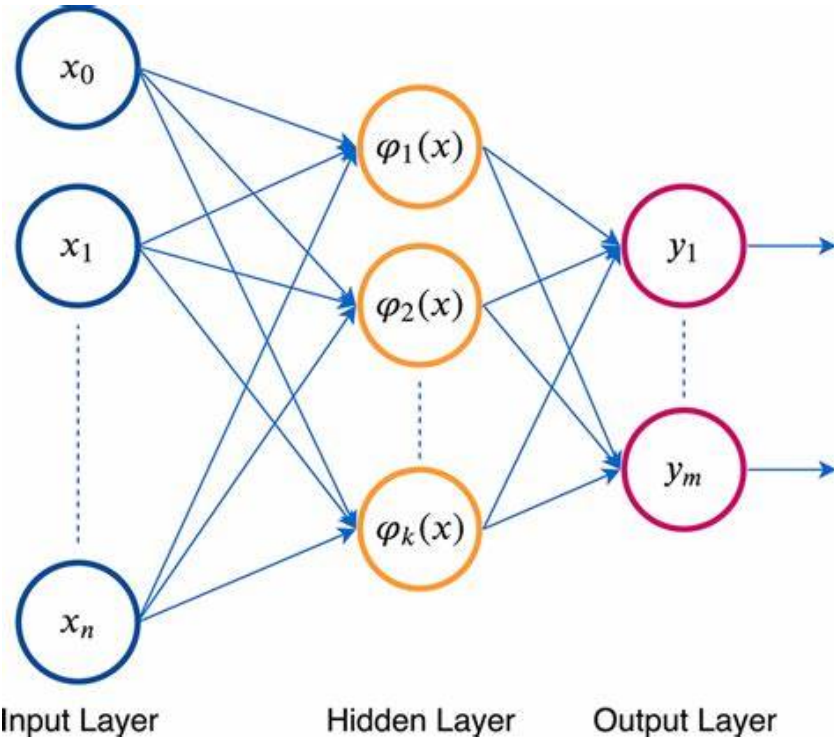
# Recurrent Neural Networks (RNN)



[How Recurrent Neural Network Works](#)

# Radial Basis Function (RBF) Network

- **Radial Basis Function**  
Networks take a different approach by using radial basis functions, typically **Gaussians**, as activation functions in the hidden layer.
- Each hidden neuron is '**activated**' based on how close the input is to its **center**.
- In practice, **center points** are chosen via clustering algorithms like **k-means**.



Dutta, Sagar & Basu, Banani & Talukdar, Fazal. (2020). Cascaded neural network based small array synthesis with robustness to noise. International Journal of RF and Microwave Computer-Aided Engineering. 31. 10.1002/mmce.22485.

# Architectures Summary Slide

Architecture	Learning Type	Best For	Activation Type	Strengths	Limitations
<b>MLP</b>	Global	Tabular data, basic classification & regression	ReLU, Tanh, Sigmoid	General-purpose, universal approximation, easy to implement	Poor scaling with high-dimensional data, no locality
<b>DNN</b>	Global Hierarchical	Complex pattern recognition (text, images, speech)	ReLU	Learns hierarchical features, high expressive power	Risk of vanishing gradients, longer training time
<b>CNN</b>	Local to Global	Images, spatial data, video processing	ReLU	Spatial locality, parameter sharing, efficient for images	Not ideal for sequential or non-spatial data
<b>RNN</b>	Sequential	Time series, natural language, sequences	Tanh, ReLU, Sigmoid	Captures temporal dependencies, sequence-aware	Vanishing gradient over long sequences, slower training
<b>RBF</b>	Local	Small-scale function approximation, pattern recognition	Gaussian (Radial)	Fast convergence, localized learning, interpretable	Poor scalability, sensitive to center selection





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