



Science and
Technology
Facilities Council

Deep Neural networks

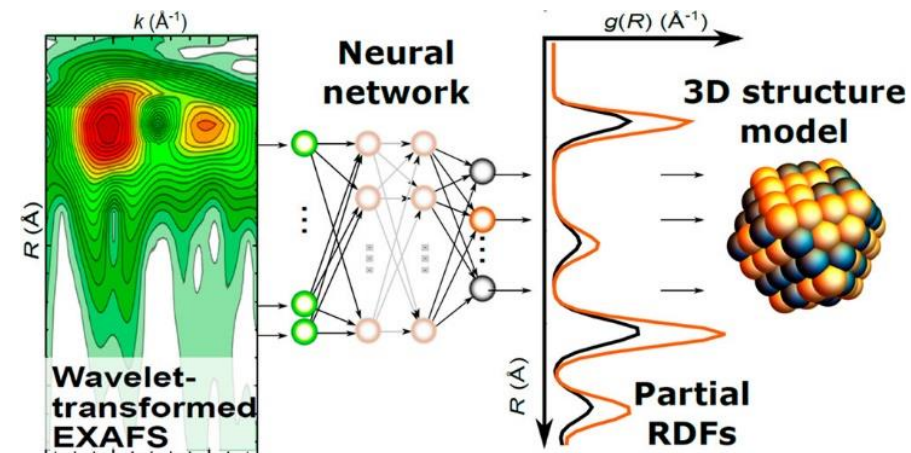
Associated notebook: https://github.com/keeeto/reading-ml-chemistry/blob/master/02_DNN.ipynb

Overview

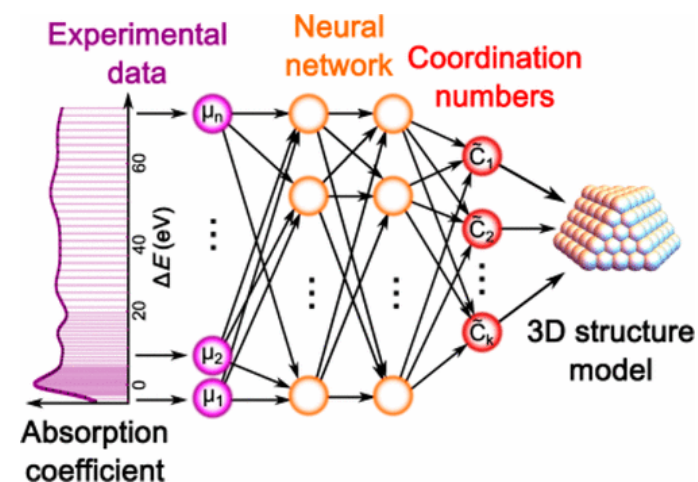
- History of deep neural nets (DNN)
- Layers of a DNN
- Structure of a neuron
- Activation functions
- Backpropagation
- Optimisation
- Regularisation

Showcase

- Multi Layer Perceptrons (MLPs) for interpreting X-ray spectroscopy
- Transforming an EXAFS signal into a partial RDF to assist with 3D model building using an MLP
- XANES spectra used to estimate coordination numbers of atomic sites using an MLP
- Processing of each spectrum takes less than 1 second
- Gets coordination up to the 4th coordination shell



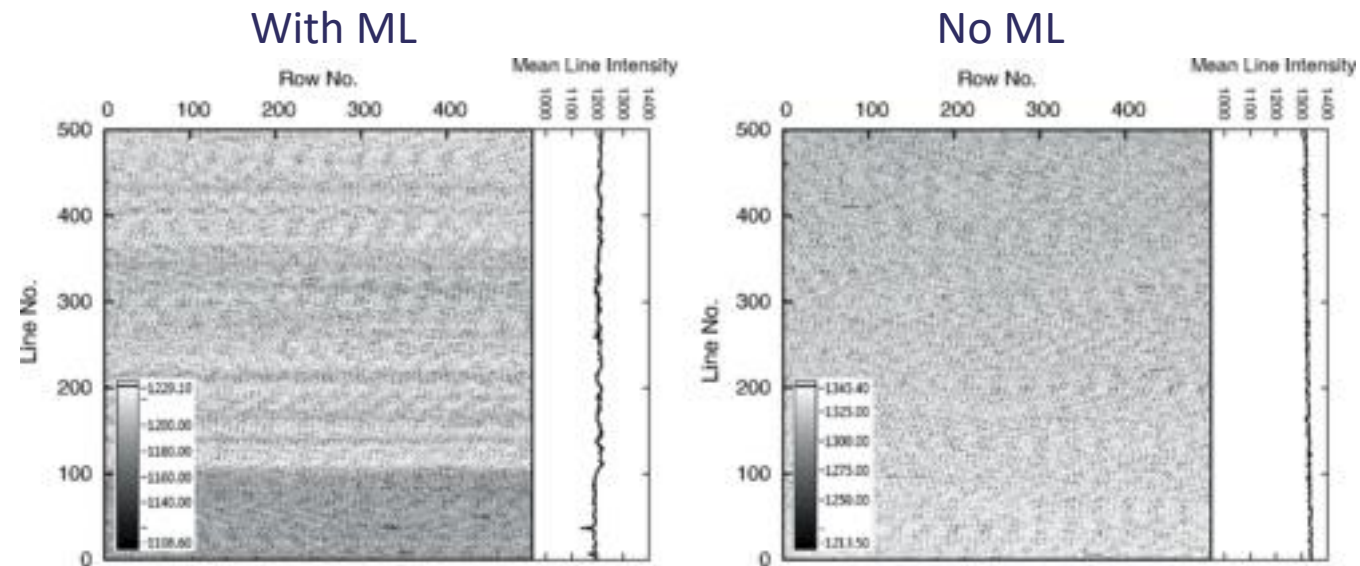
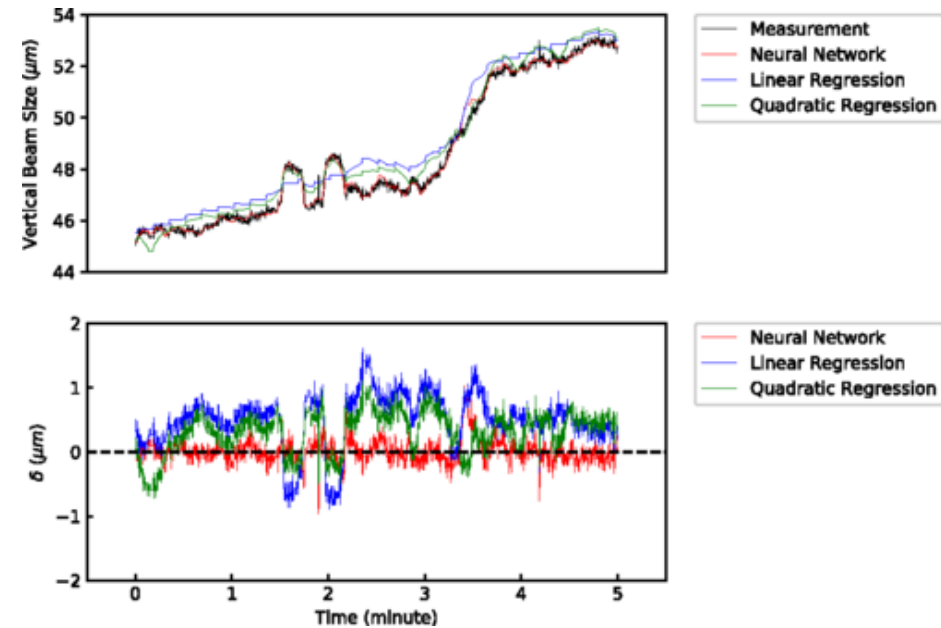
Nano Lett. 2019, 19, 1, 520–52



J. Phys. Chem. Lett. 2017, 8, 20, 5091–5098

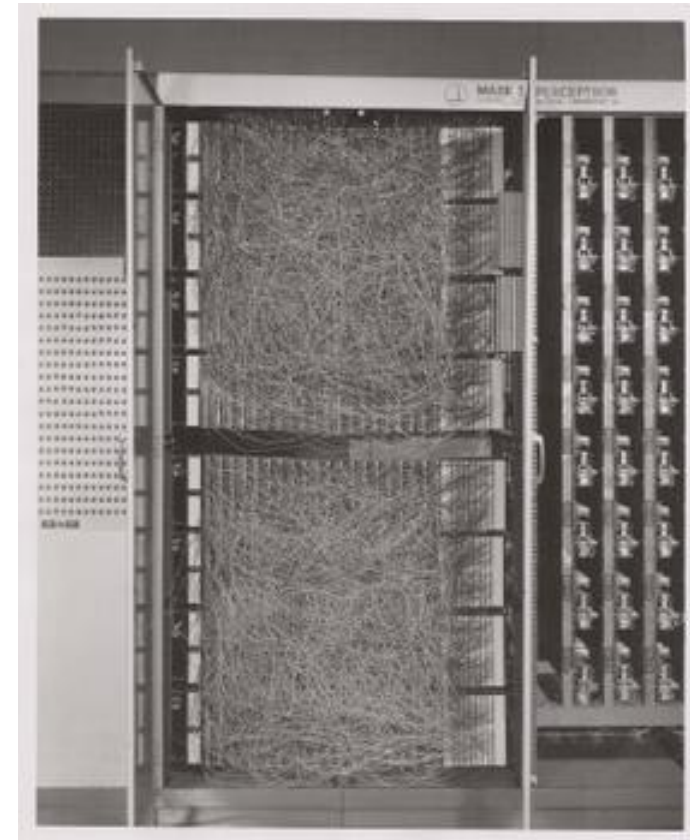
Showcase

- Stabilisation of synchrotron beam at Advanced Light Source
- “Learned” how changes in the configurations and positions of the magnetic excitations affected the width of the resulting photon beam
- Precision is better than the 2–3% previously achieved at the ALS
- Allows the synchrotron to probe the dynamics of chemical reactions



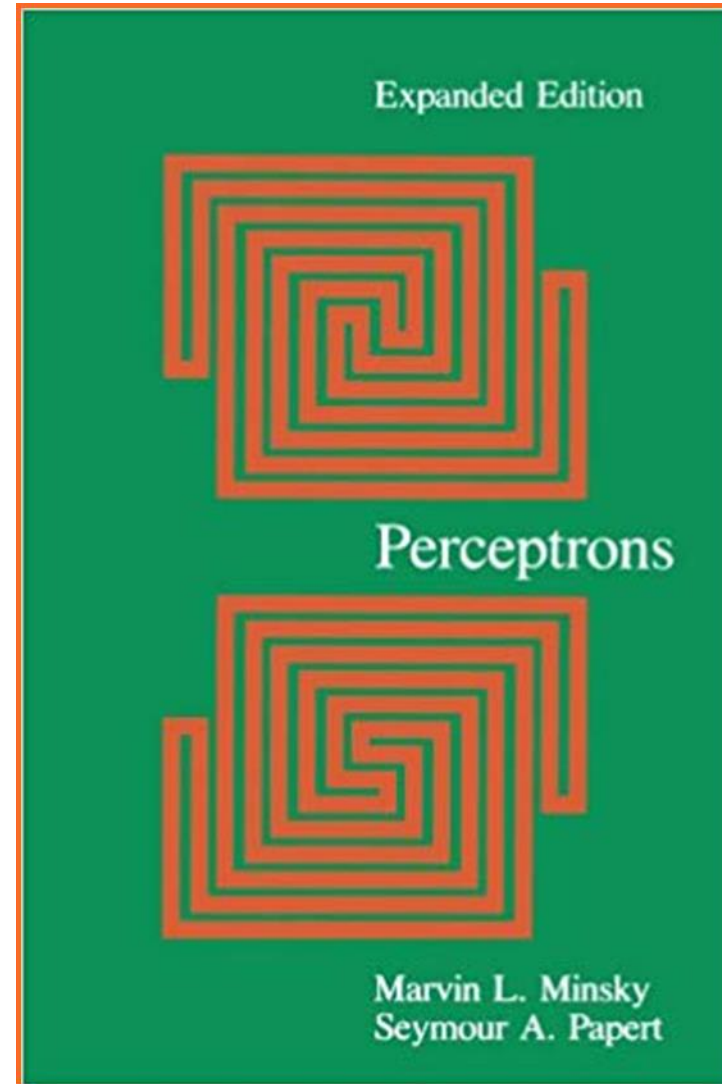
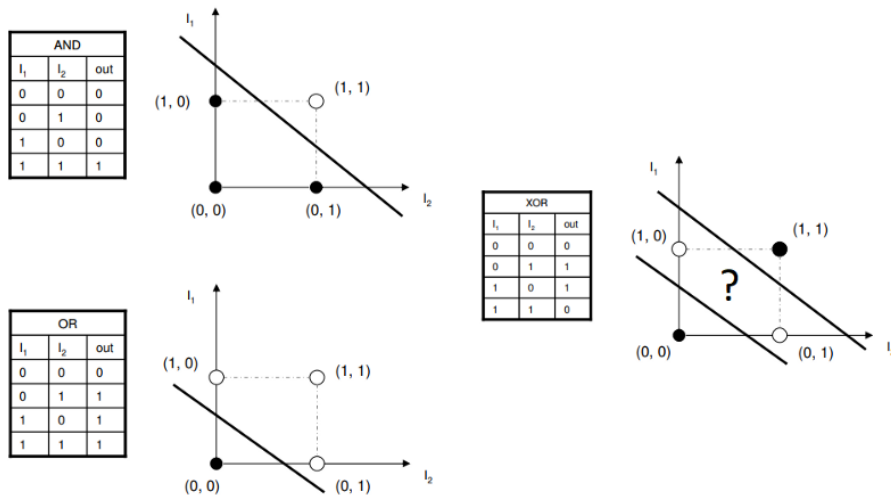
Neural networks

- Early NN
 - Originally a device
 - Intended for binary classification
-
- Produces a single output from a matrix of inputs, weights and biases
$$y = \phi\left(\sum_i w_i x_i + b\right) = \phi(\mathbf{w}^T \mathbf{x} + b)$$



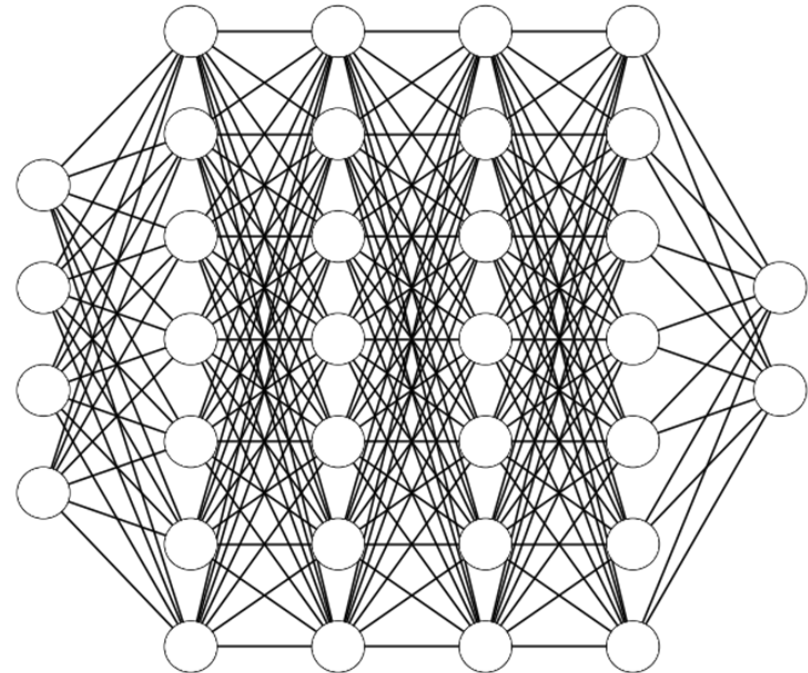
Neural networks

- Single layer
- Minsky and Papert showed they could not solve non-linear classification

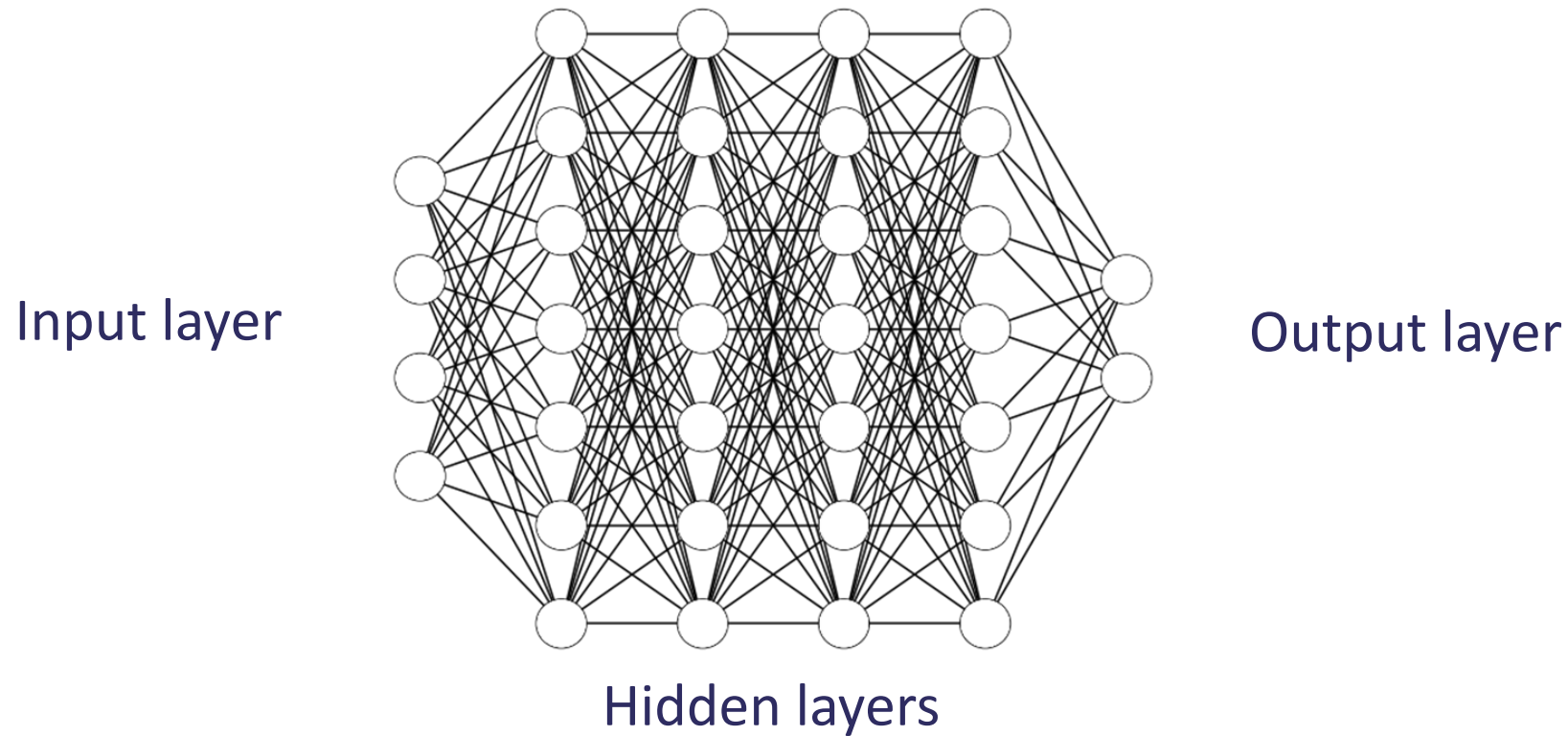


Neural networks

- Back propagation
- Now gradients could be used to minimise error
- Modifications back propagate through the network using the chain rule

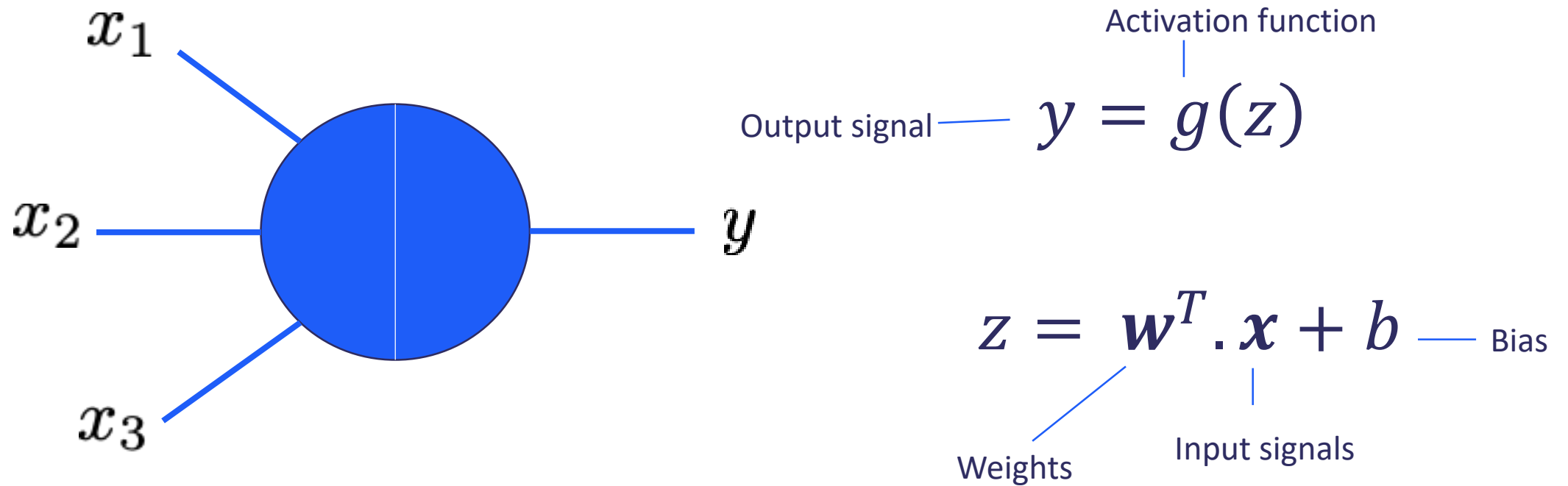


Deep neural networks: Multi layer perceptron



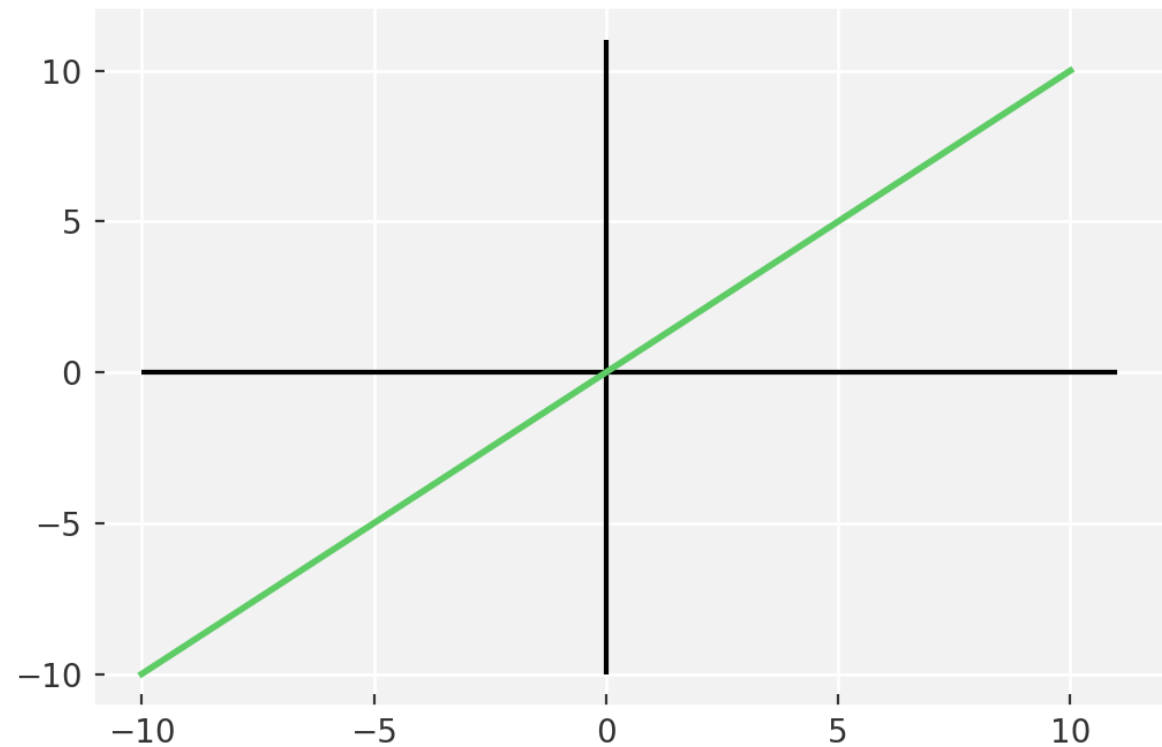
Dense layers

- Also called fully connected layers



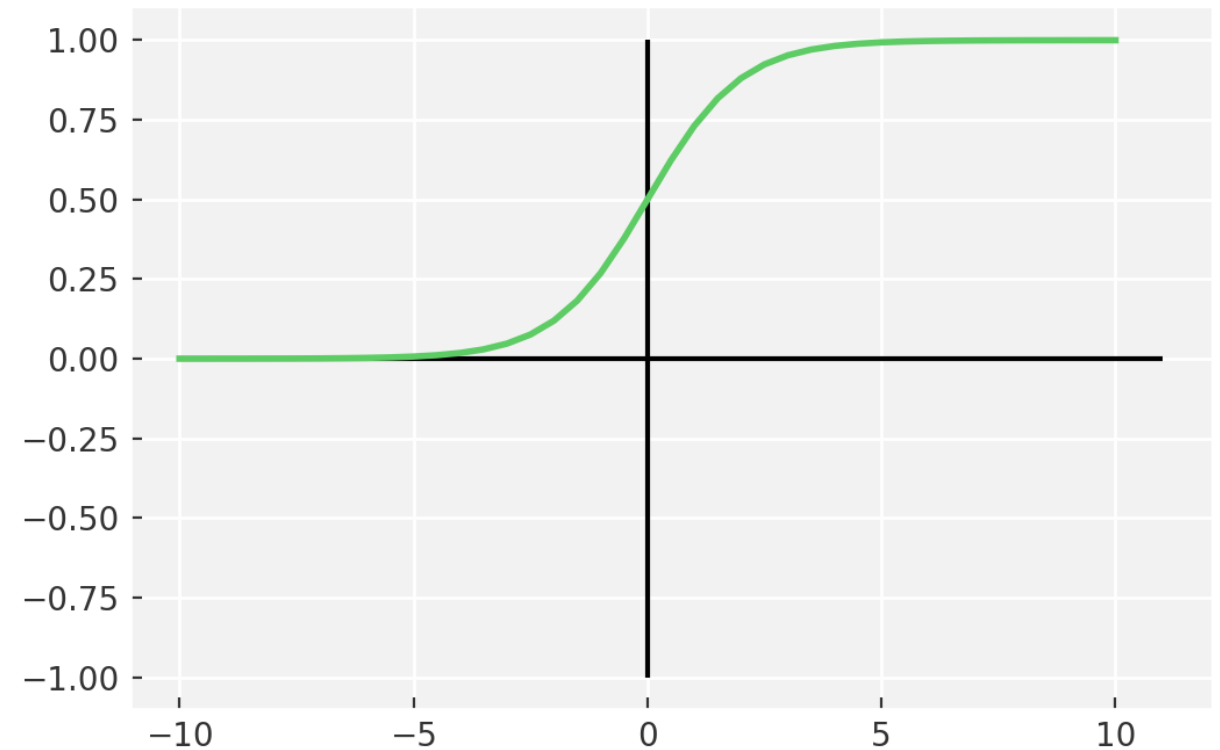
Activation function: Linear

- The simplest form of activation function



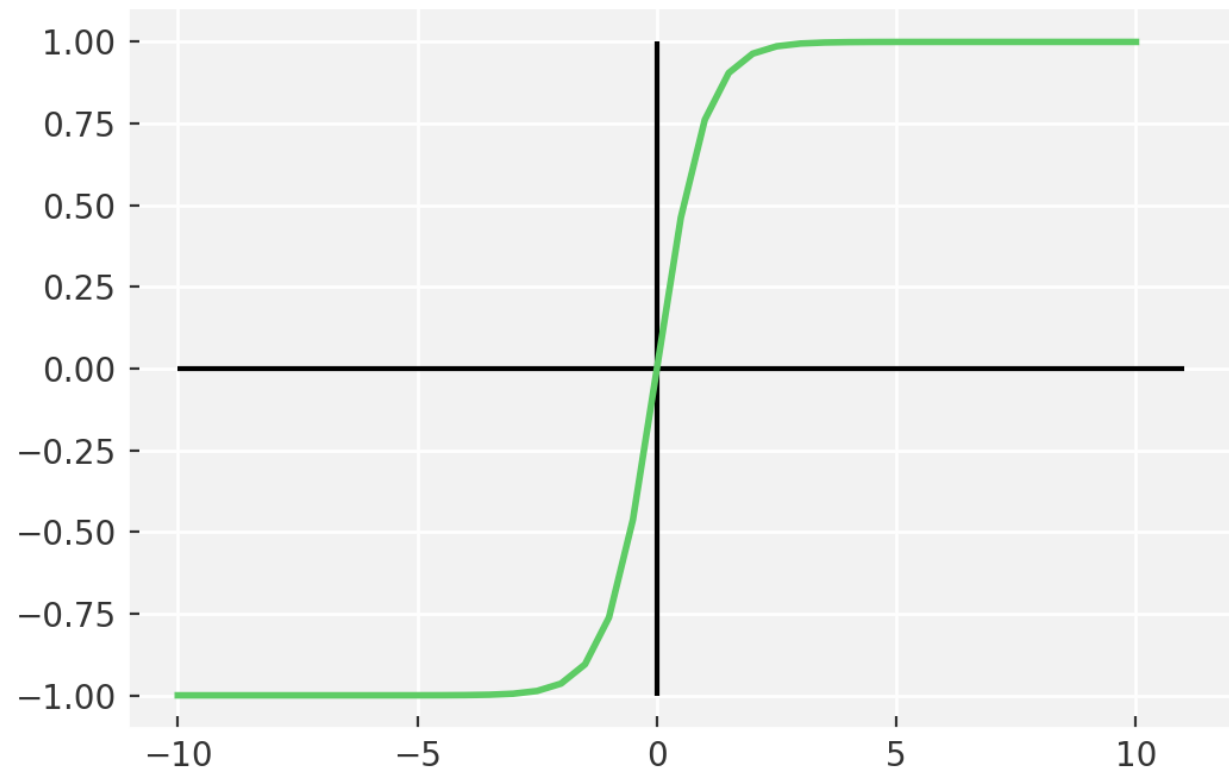
Activation function: Sigmoid

- Vanishing gradient problem
- Secondly, its output isn't zero centered. It makes the gradient updates go too far in different directions. **$0 < \text{output} < 1$, and it makes optimization harder.**
- Sigmoids saturate and kill gradients.
- Sigmoids have slow convergence.



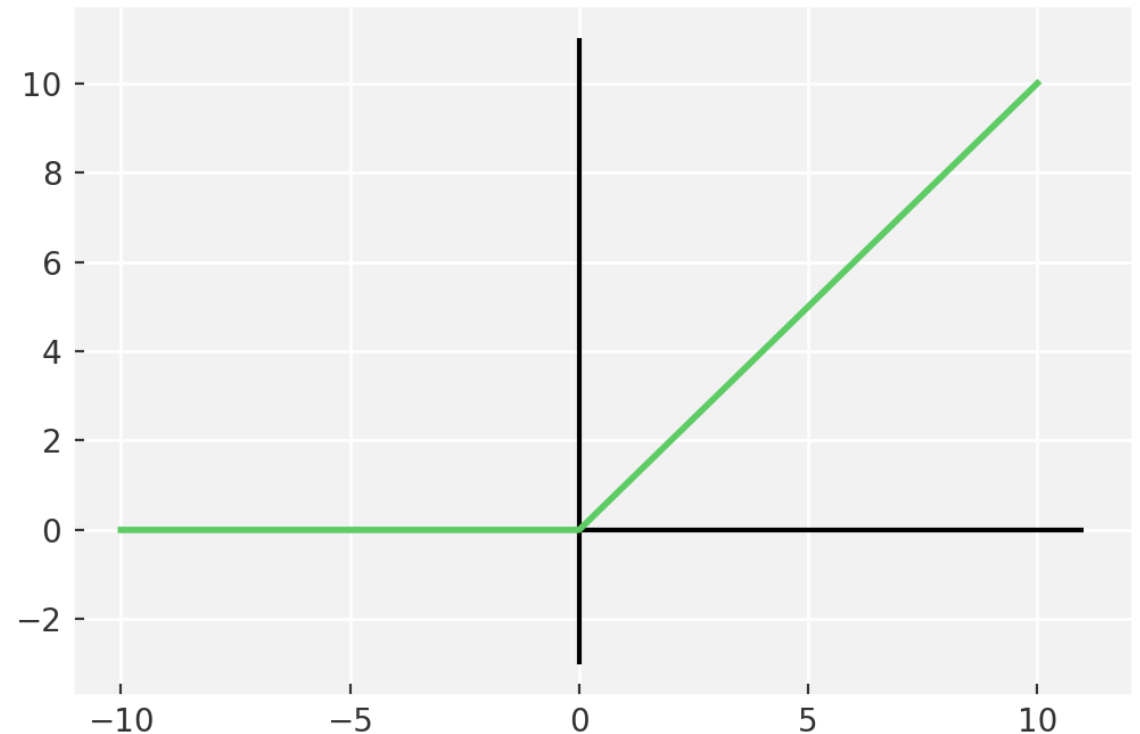
Activation function: Tanh

- Output is zero centered
- Usually preferred to sigmoid as it converges better
- Still it suffers from vanishing gradient problem



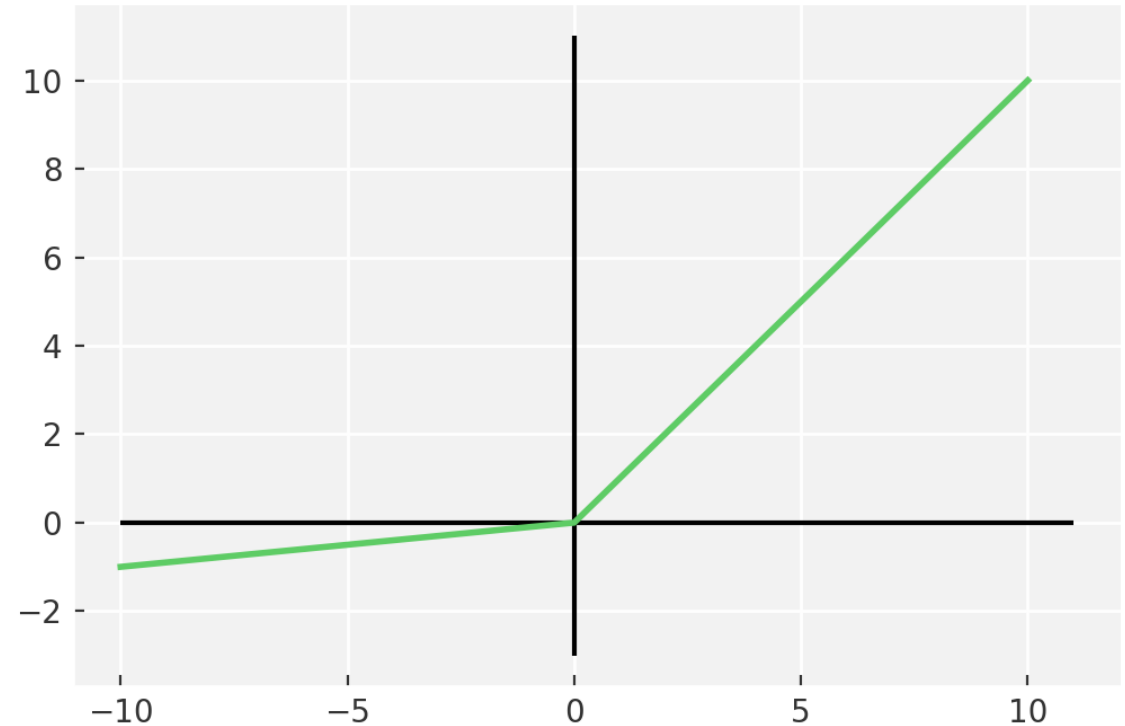
Activation function: ReLU

- 6 times improvement in convergence from Tanh function
- Should only be used within Hidden layers of a neural network model



Activation function: LeakyReLU

- Some ReLu gradients can be fragile during training and can die.
- Cause a weight update which will makes it never activate on any data point again.
- ReLu could result in Dead Neurons.

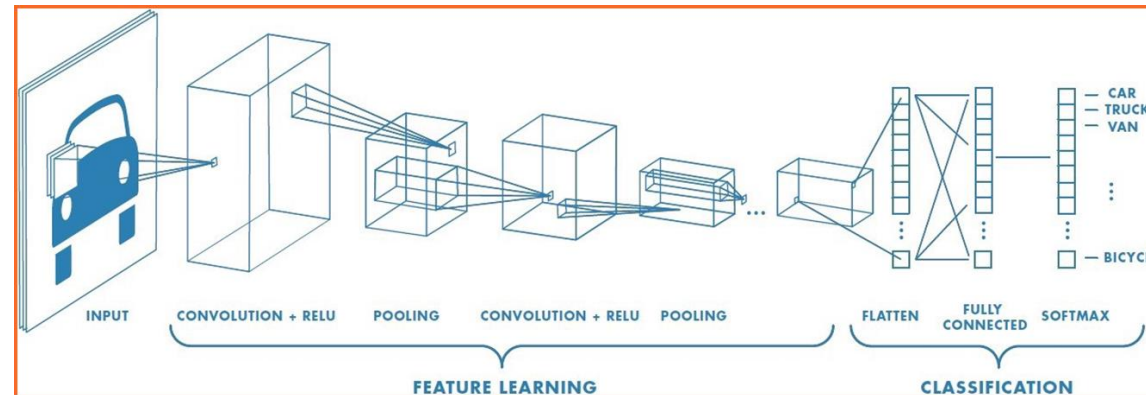


Go To Notebook

- Its time to build your first neural network

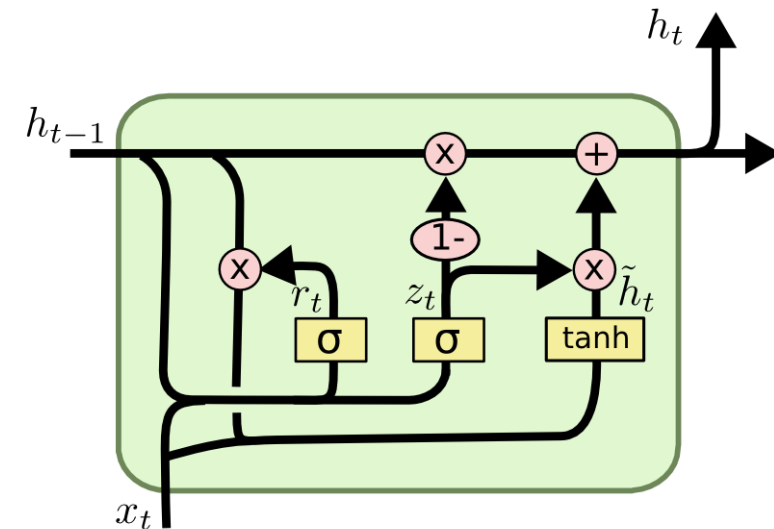
Neural networks

- CNNs image
- The MLP has no real concept of the spatial relations
- Also, dense connections lead to parametric explosions for many pixel images



Neural networks

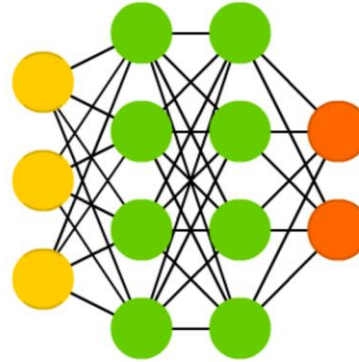
- RNNs/LSTMs time series
- Often algorithms are desired for predicting the next event based on a series of previous events
- Eg Pressure/temperature evolution, speech prediction ...
- In this case standard NNs are not very useful due to a lack of 'memory'



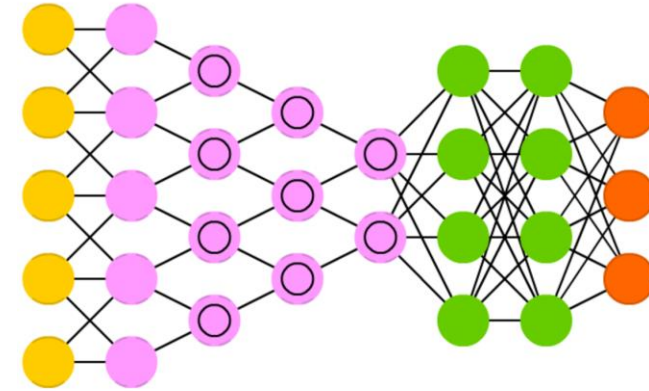
Common types of NN

- Backfed Input Cell
- Input Cell
- △ Noisy Input Cell
- Hidden Cell
- Probablistic Hidden Cell
- △ Spiking Hidden Cell
- Output Cell
- Match Input Output Cell
- Recurrent Cell
- Memory Cell
- △ Different Memory Cell
- Kernel
- Convolution or Pool

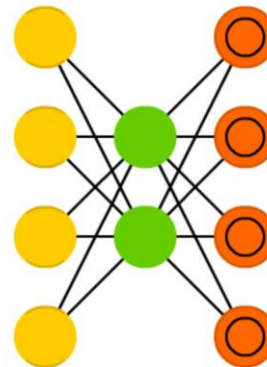
Deep Feed Forward (DFF)



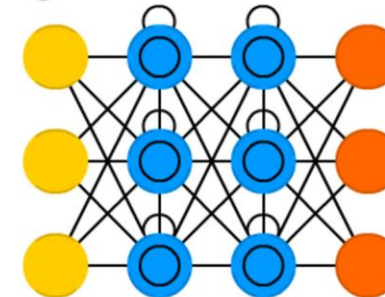
Deep Convolutional Network (DCN)



Auto Encoder (AE)



Long / Short Term Memory (LSTM)



Concept checklist

- Origins of neural networks
- Types of layers – input, hidden, output
- Dense/fully-connected layers
- Activation functions
- Batch/epoch
- Dropout

Concept checklist

- Supervised/unsupervised machine learning
- Classical machine learning/deep learning
- Parameters/hyperparameters
- Features and feature engineering
- Decision trees
- Overfitting
- Evaluation/metrics
- Test/train split, cross-validation
- Bagging and boosting
- Deep learning
- Neural networks



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Thank you



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Twitter: @keeeto2000