

# DATASCI 207

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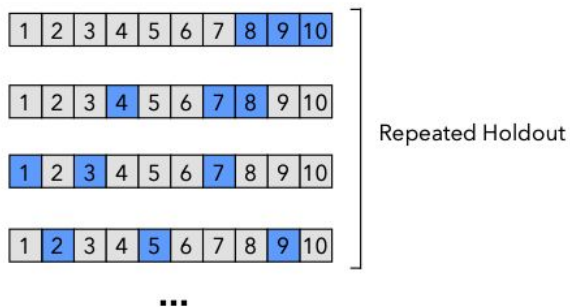
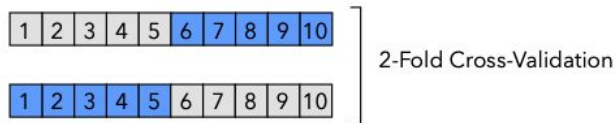
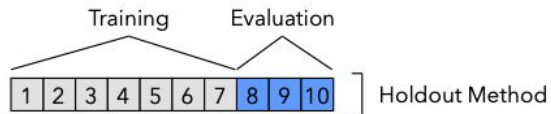
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# Announcements

- **Finalize group selections in class**
  - [Logistics Sheet](#)
- **Evaluations:** <https://course-evaluations.berkeley.edu/>
  - What you like about the class + constructive feedback

# Cross Validation (CV) Recap

# CV

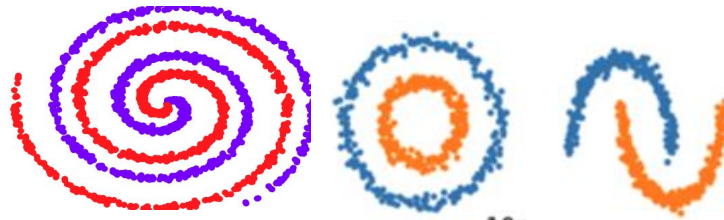


- Is there a difference between the hold out method with a 50/50 train/validation split and 2-fold cross validation?
- We can repeat the hold out methods multiple times (bottom panel). Would that be the same as doing k-fold cross validation?

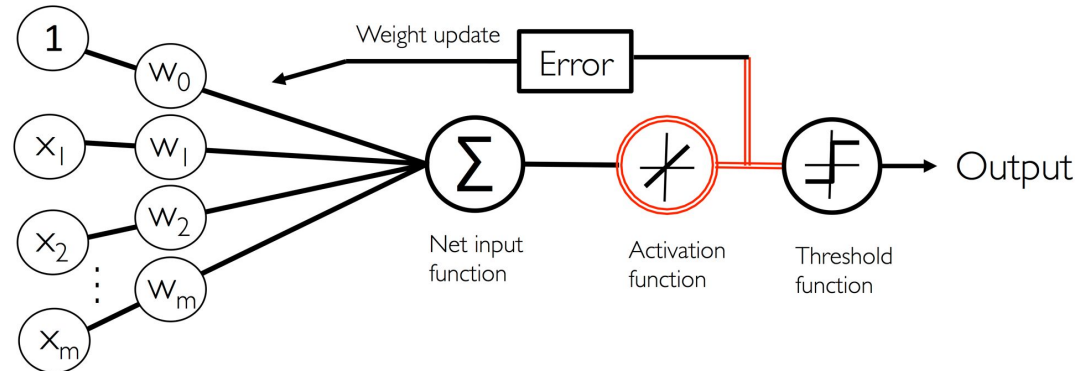
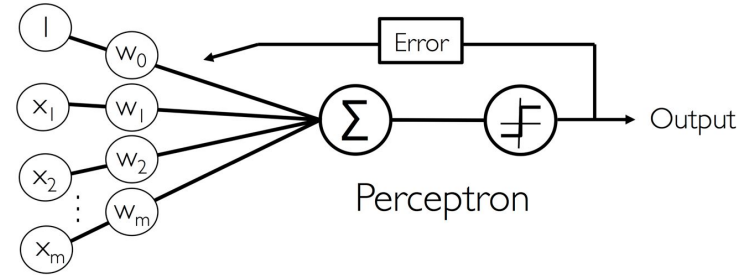
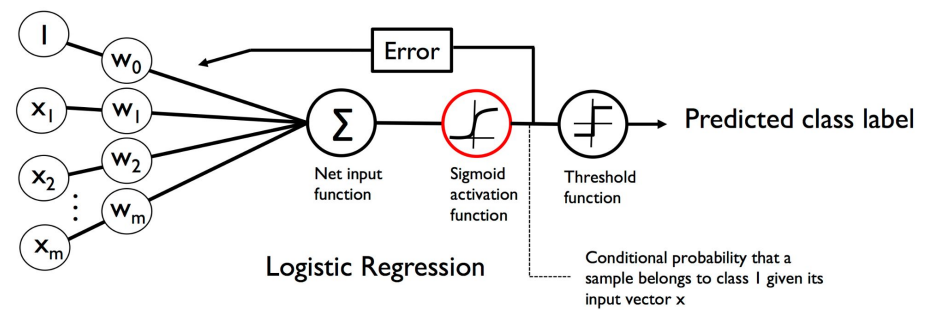
# Multilayer Neural Networks

# Feedforward Neural Networks (FNNs)

- Examples
  - Linear neural network - next slide
  - Single Layer Perceptron - last week's lecture
  - Multi Layer Perceptron (MLP) - XOR example from async
- Fully connected
- No cycles, unlike Recurrent Neural Networks (RNNs)



# FNN Examples

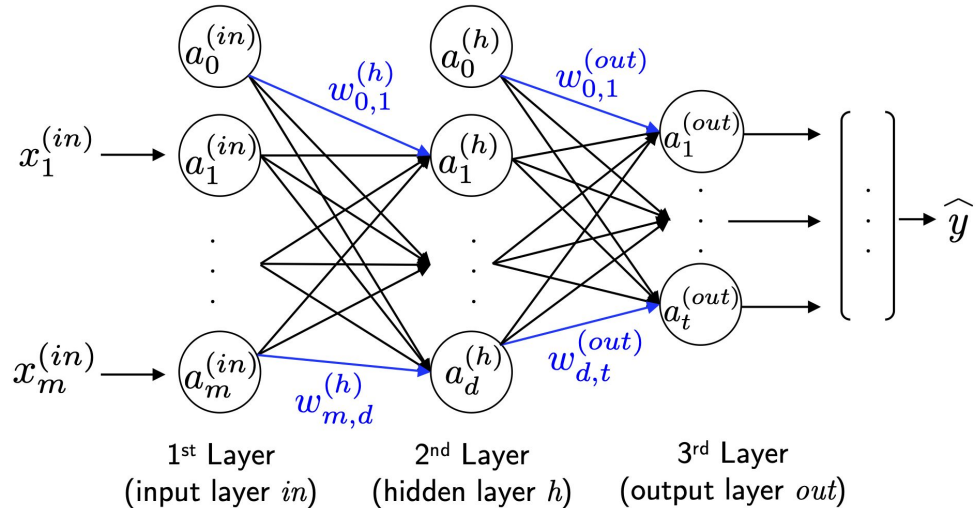


Source:

<https://github.com/rasbt/machine-learning-book/blob/92e2320709071b4dcc82a99552aec145536cfbcd/ch02/ch02.ipynb>

# Multilayer Neural Networks

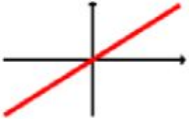
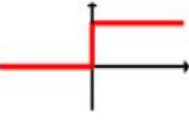
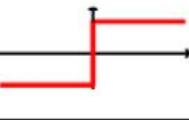




- (Right) Network with 1 input layer, 1 hidden layer, and 1 output layer. Each layer receives its inputs from the previous layer.
- We can add as many hidden layers (**deep neural networks**), but this becomes very computationally intensive (parameters to be estimated grow exponentially).
- **a()**: activation function
- Note that there are many activation functions to choose from (earlier we covered linear and sigmoid).
- Use nonlinear activation functions to account for the complexities of data.





# Key Concepts

- **Key idea:** use of non-linearities (through the use of activation functions)
- **Optimization:** gradient descent, learning rate, different types of optimizers
- **Computational challenges:** computing the derivatives (forward pass and back propagation)
  - These challenges gave rise to frameworks such as tensorflow and pytorch which take care of the computation graphs and automatic differentiation behind the scenes

| Activation function               | Equation  | Example                            | 1D graph   |
|-----------------------------------|---|------------------------------------|--|
| Linear                            | $\phi(z) = z$   | Adaline, linear regression         |    |
| Unit step<br>(Heaviside function) | $\phi(z) = \begin{cases} 0 & z < 0 \\ 0.5 & z = 0 \\ 1 & z > 0 \end{cases}$   | Perceptron variant                 |   |
| Sign<br>(signum)                  | $\phi(z) = \begin{cases} -1 & z < 0 \\ 0 & z = 0 \\ 1 & z > 0 \end{cases}$  | Perceptron variant                 |   |
| Piece-wise linear                 | $\phi(z) = \begin{cases} 0 & z \leq -\frac{1}{2} \\ z + \frac{1}{2} & -\frac{1}{2} \leq z \leq \frac{1}{2} \\ 1 & z \geq \frac{1}{2} \end{cases}$ | Support vector machine             |   |
| Logistic<br>(sigmoid)             | $\phi(z) = \frac{1}{1 + e^{-z}}$  | Logistic regression, multilayer NN |   |
| Hyperbolic tangent<br>(tanh)      | $\phi(z) = \frac{e^z - e^{-z}}{e^z + e^{-z}}$   | Multilayer NN, RNNs                |   |
| ReLU                              | $\phi(z) = \begin{cases} 0 & z < 0 \\ z & z > 0 \end{cases}$  | Multilayer NN, CNNs                |  |

# Multilayer Neural Network Exercise

- Real Data (multi-class) - in class
  - [https://github.com/MIDS-W207/nteneva/tree/main/live\\_sessions\\_current/week6](https://github.com/MIDS-W207/nteneva/tree/main/live_sessions_current/week6)

|                               | Group 1 | Group 2 | Group 3 |
|-------------------------------|---------|---------|---------|
| Question 1                    |         |         |         |
| Question 2 (best performance) |         |         |         |

- Synthetic Example (binary, danish dataset) - try on your own!
  - [https://github.com/MIDS-W207/coursework\\_2022/blob/main/Demos\\_Santerre/Berkley207-6.ipynb](https://github.com/MIDS-W207/coursework_2022/blob/main/Demos_Santerre/Berkley207-6.ipynb)