

# Machine Learning: Neural Networks

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**CPSC 501: Advanced Programming Techniques  
Winter 2025**

Jonathan Hudson, Ph.D  
Assistant Professor (Teaching)  
Department of Computer Science  
University of Calgary

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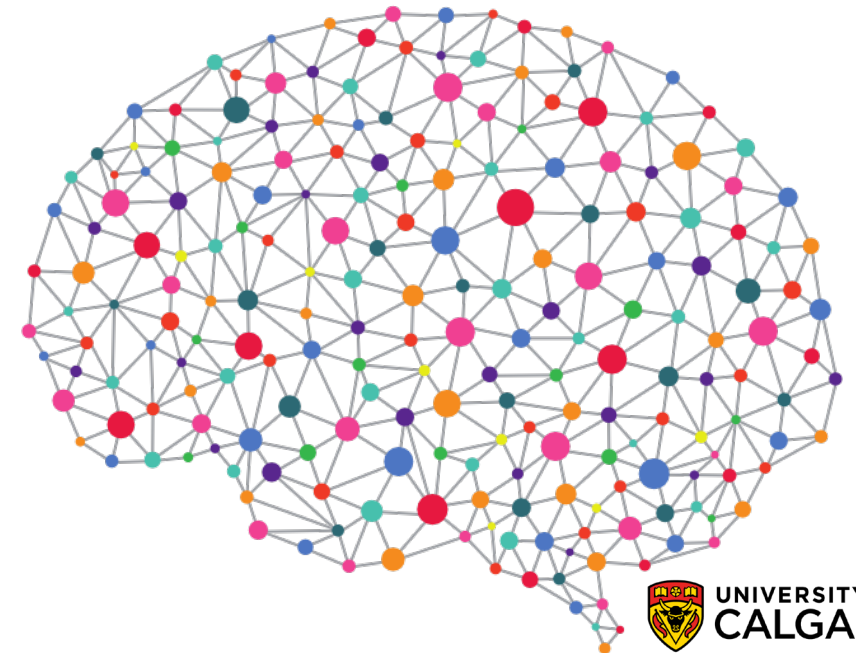


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# What are neural networks?

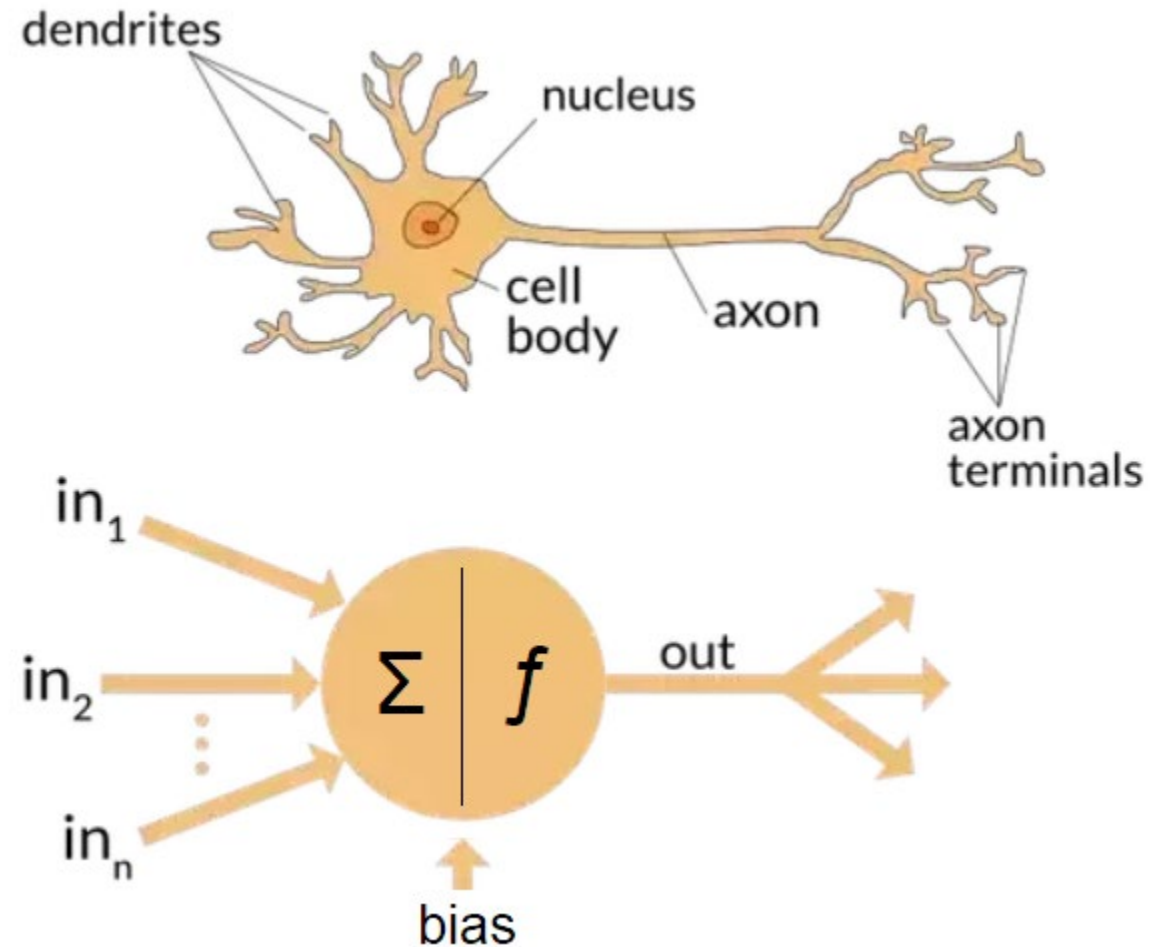
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- Inspired by the Human Brain.
- Requires 20% of your body's energy to function.
- Deep learning neural networks really popular right now
  - LLMs and generative AI! (Chat-GPT, Gemini, Co-Pilot, etc.)
- Connectionist method
  - Make network, train, hope result is useful



# Neuron Model of Connections

- Developed to mimic the human neural system (in the brain) and its processing capabilities
- Decentralized knowledge representation and processing  
☞ hopefully very efficient
- Simple components, the intelligence is in the **connections**



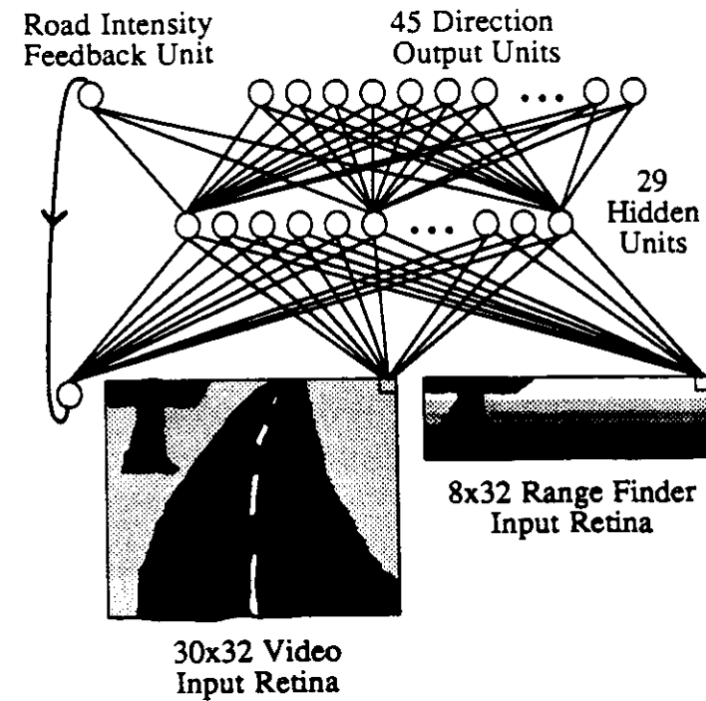
# A short history of Neural Networks

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- 1957: Perceptron (Frank Rosenblatt): one layer network neural network
- 1959: first neural network to solve a real world problem
  - i.e., eliminates echoes on phone lines (Widrow & Hoff)
- **First AI Winter**
- 1988: Backpropagation (Rumelhart, Hinton, Williams): learning a multi-layered network
- **Second AI Winter**

# A short history of NNs

- 1989: ALVINN: autonomous driving car using NN (CMU)



# A short history of NNs

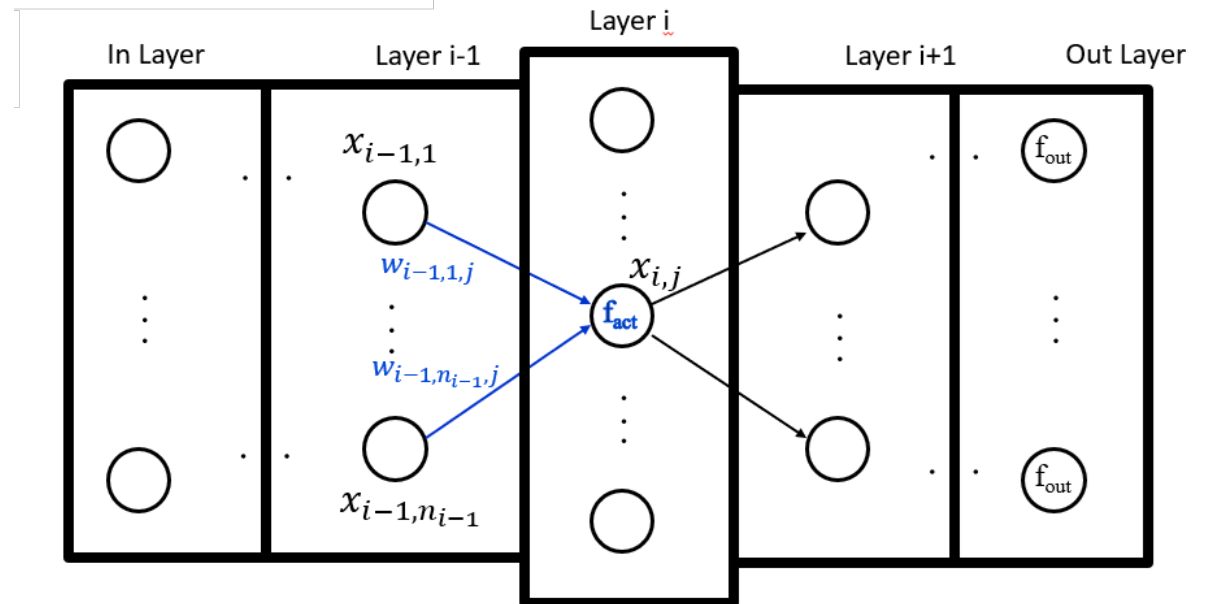
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- 1989: (LeCun) Successful application to recognize handwritten ZIP codes on mail using a “deep” network
- 2012-2015 convolutional neural networks (CNN)
- 2015 – AlphaGo Competition using CNN and reinforcement learning
- 2015+: near-human capabilities for image recognition, speech recognition, and language translation
- 2018 (AlphaFold) Google’s protein folding prediction
- 2019 (AlphaStar) Google’s StarCraft 2 AI better than 99.8% of human players (GrandMaster level)
- 2022 MidJourney, Chat-GPT – generative AI for text and images go mainstream

# Basic data structures

Directed, weighted graph:

- Nodes represent a function (**activation function  $f_{act}$** ) with  $n$  arguments, if  $n$  links lead into the node, producing one result
  - Input nodes: take values from outside
  - Output nodes: represent activation values for different concepts to detect
  - Inner nodes: usually organized in layers (hidden layers)
- Labeled weighted links



# Semantics

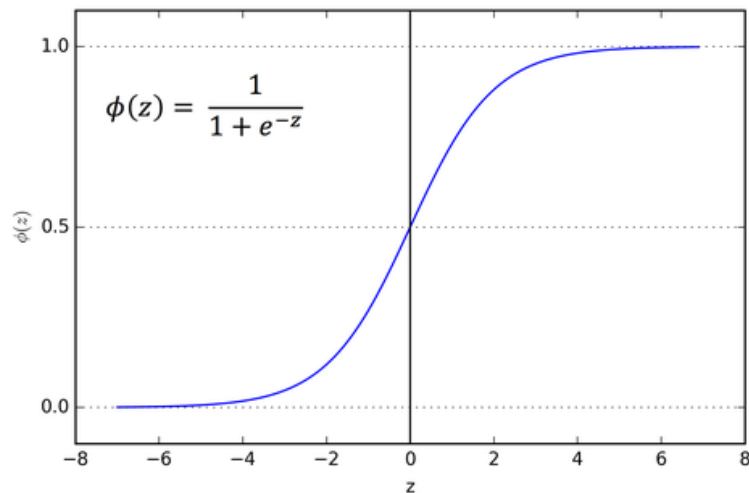
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- Whole **net** represents a **decision function  $f$**  linking input nodes to output nodes
1. We connect a bunch of nodes together between these two end points in a directed graph
    - We give the connections weights and nodes trigger sub- $f_{act}$  functions
    - This makes the middle of the  $f$  function ‘complicated’
  2. Then we give it inputs with expected outputs,
    - **If it is wrong**, then we **change the weights** inside graph
  3. We do this until the function seems to be right a lot in the future

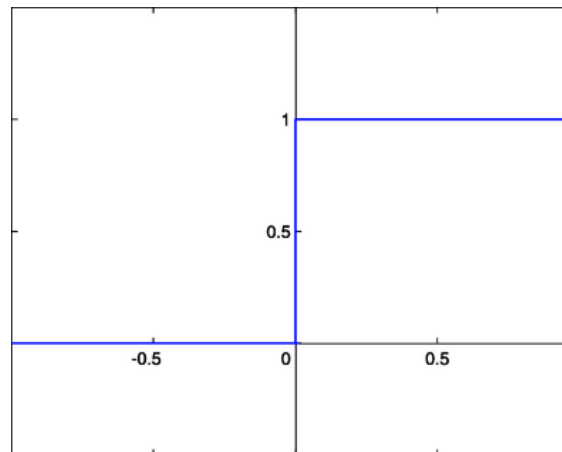


# Activation Functions

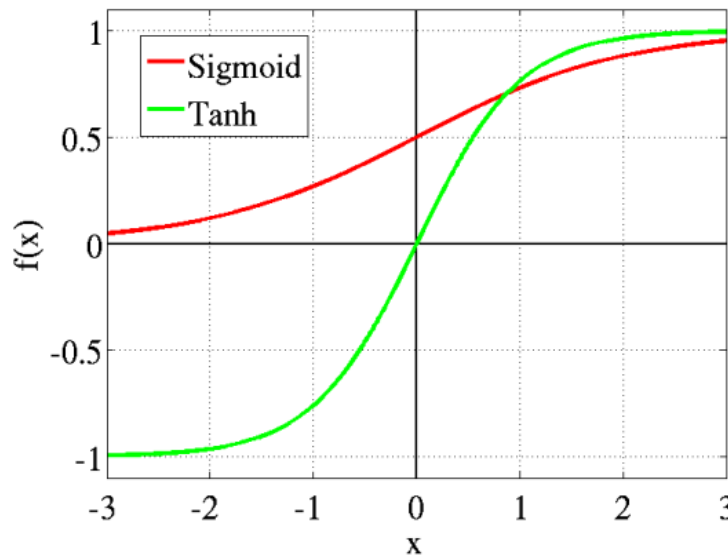
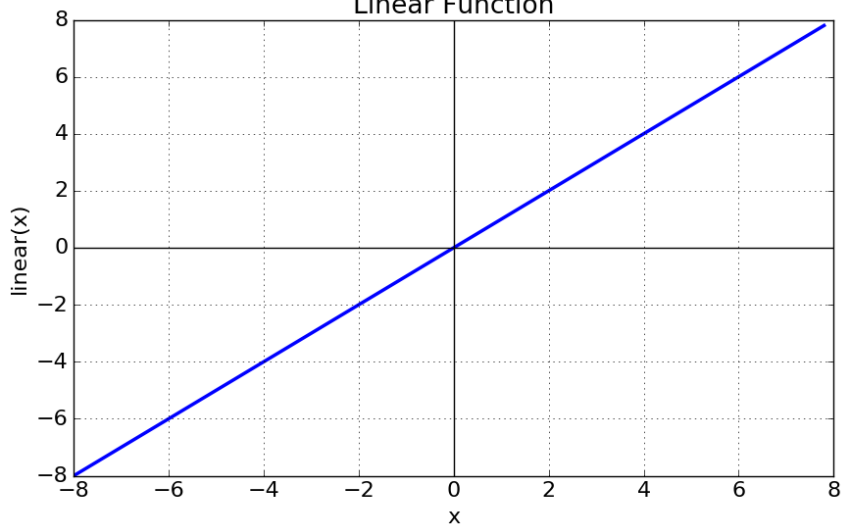
Sigmoid



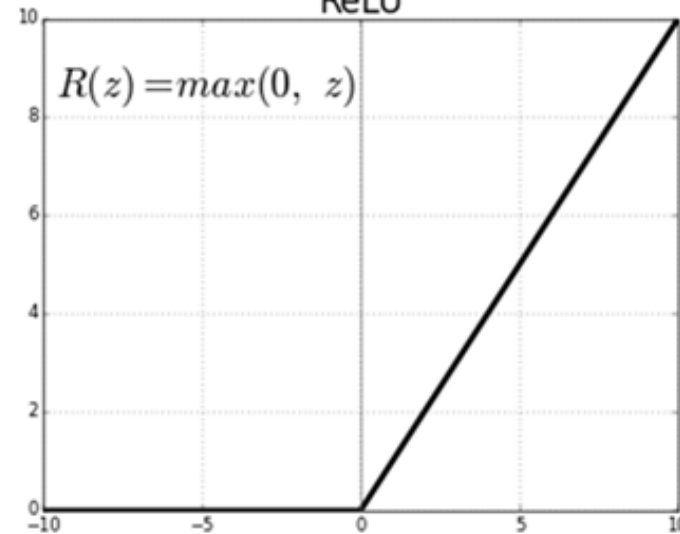
Step sigmoid



Linear Function



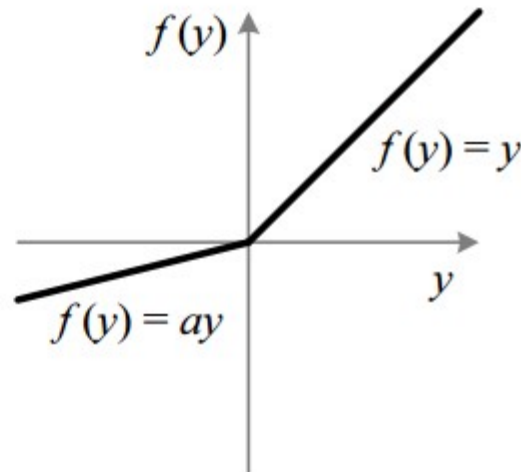
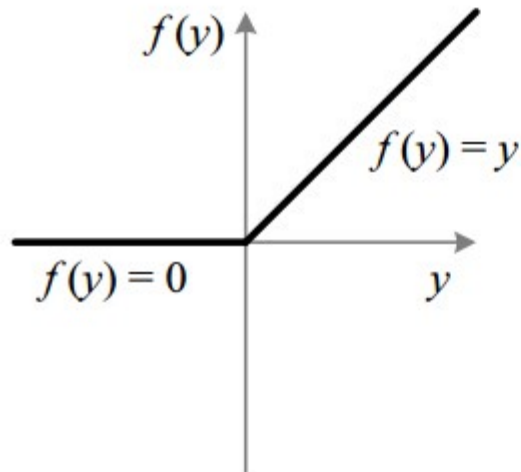
ReLU



# Leaky ReLU Activation Function

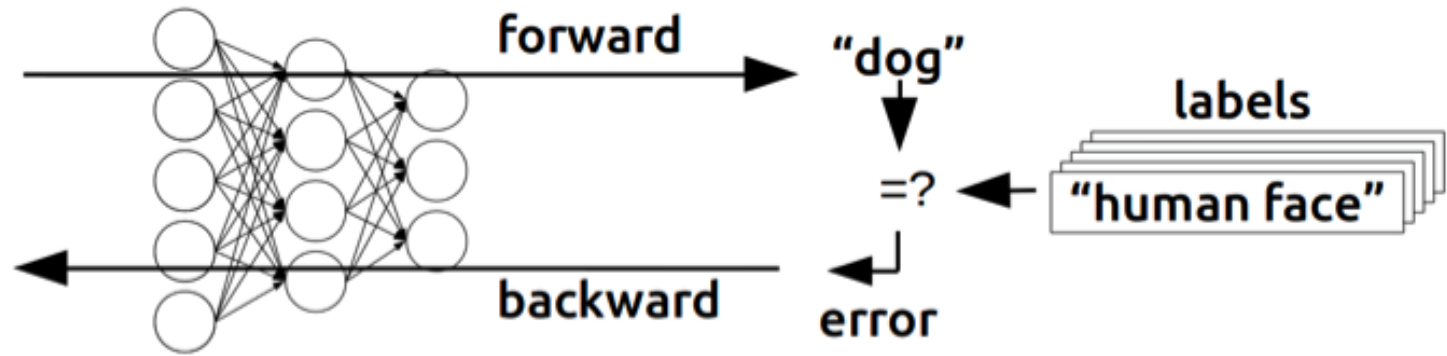
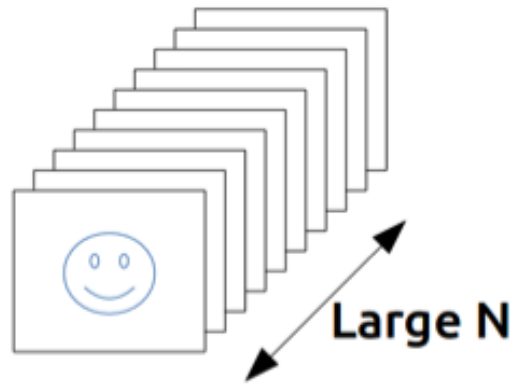
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- It is an attempt to solve the dying ReLU problem
- The leak helps to increase the range of the ReLU function. Usually, the value of  $a$  is 0.01 or so.

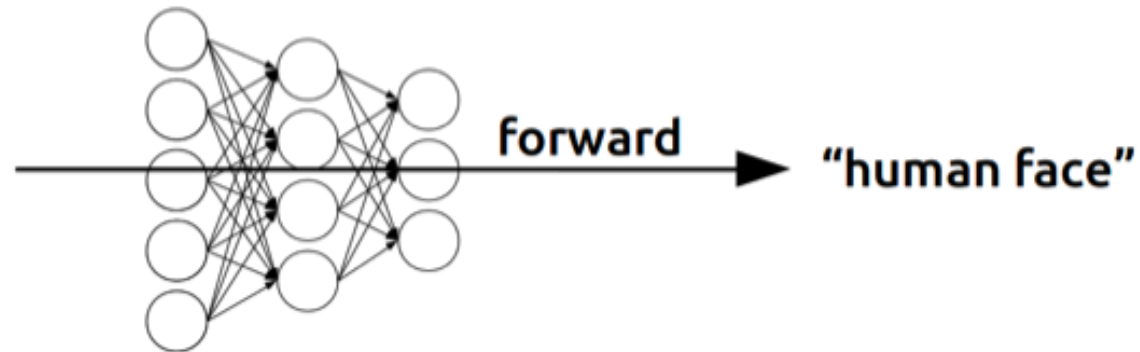
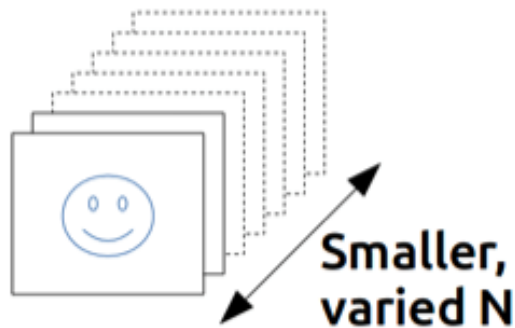


# Learning: Backpropagation

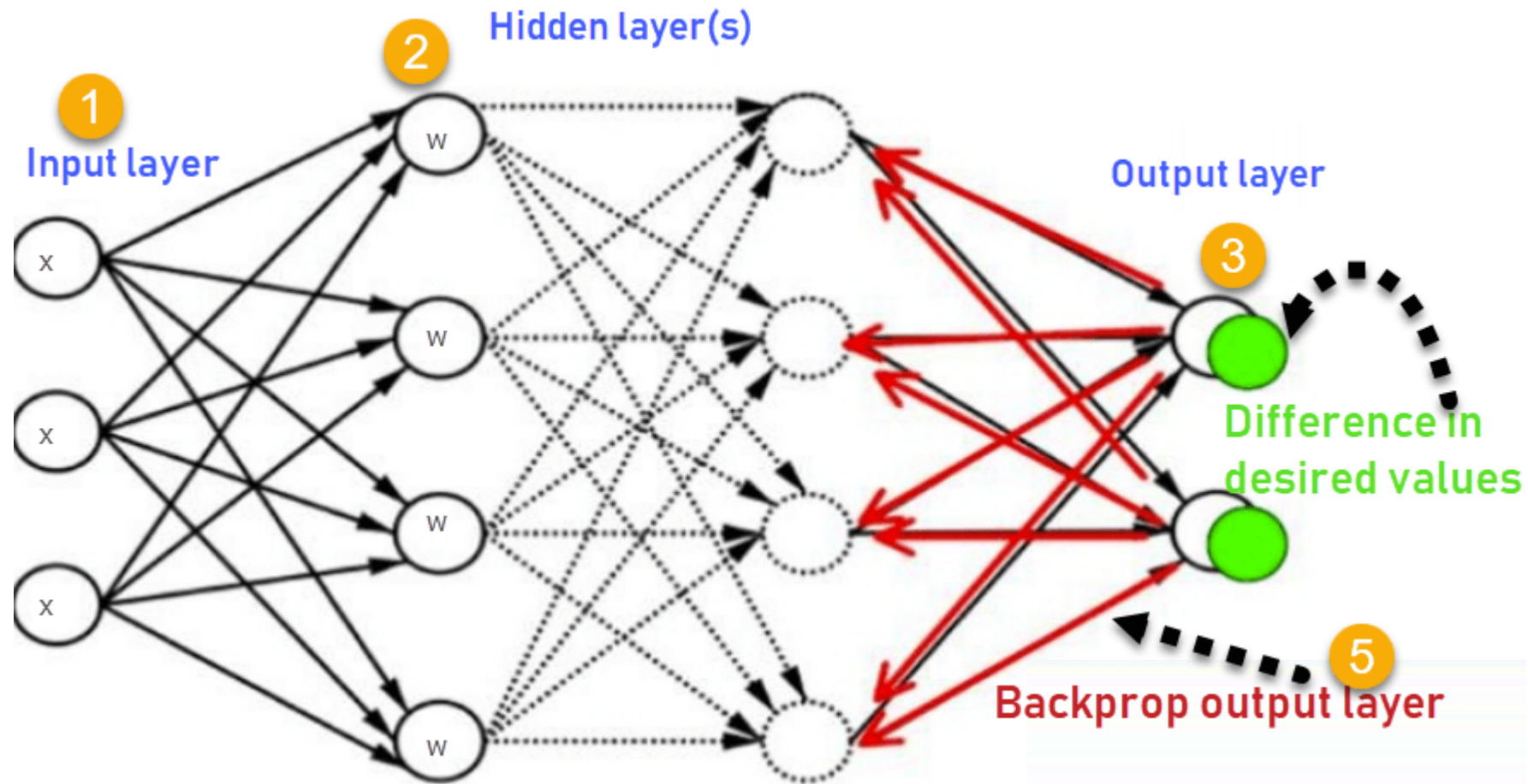
## Training



## Inference



# Back-Propagation



# Stochastic Gradient Descent

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- Gradient descent -> follow slope to best
- Stochastic -> random influence
  
- 1950s, Frank Rosenblatt used SGD to optimize his perceptron
  
- 2014, Adam (for "Adaptive Moment Estimation") was published
  - Most machine learning libraries are dominated by Adam-type optimizers

# Onward to ... Tensorflow

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Jonathan Hudson  
[jwhudson@ucalgary.ca](mailto:jwhudson@ucalgary.ca)  
<https://pages.cpsc.ucalgary.ca/~jwhudson/>



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