#### Reinforcement Learning

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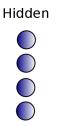
#### Outline

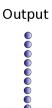
- 1 Machine Learning
- 2 Neural Networks
  - Simple Perceptron
  - Multilayer Perceptron
  - Convolutional Neural Network
- 3 Temporal Difference Learning
- 4 Results
- 5 The End

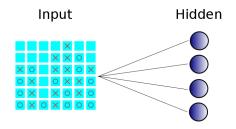
## Supervised Learning vs Reinforcement Learning

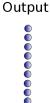
- Supervised Learning
  - Have access to a large set of data with known desired results
  - Adjust model parameters to minimize an objective function
- Reinforcement Learning
  - Have access to an environment that can be modeled
  - Typically a reward function is used as a signal for how to adjust model parameters
  - Board games present a natural environment that is easily modeled and can provide a reward whenever a game finishes.

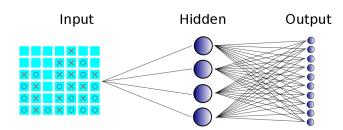
# Input





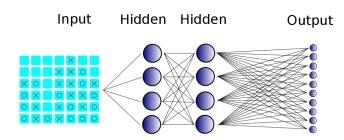


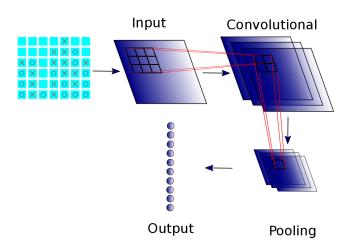




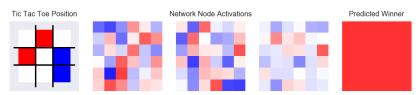
Neural Networks

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- A value network is a neural network that evaluates an environment and determines a value to associate to it
- In the case of board games, a value network can be used to determine the probability of winning from any given position
- Here is an example of a multi-layered neural network in action
- The network takes a Tic Tac Toe board as input, and through a succession of node activations the network outputs a prediction on who will win, in this case: the red player



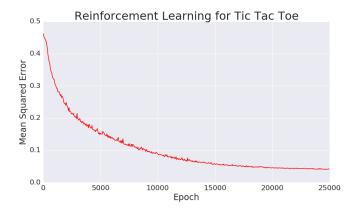
## Temporal Difference Learning

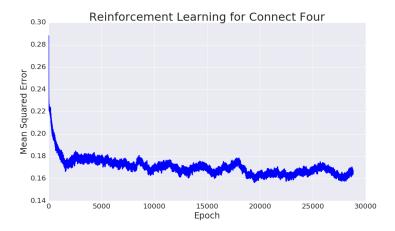
■  $\mathsf{TD}(\lambda)$  Equation

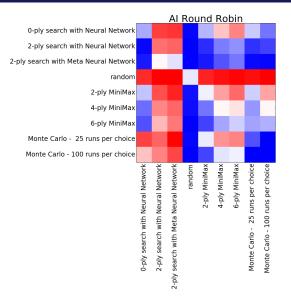
$$T_n = (1 - \lambda) V_n + \lambda r$$

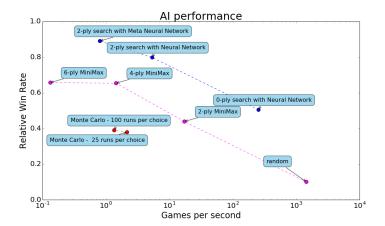
#### ■ $\mathsf{TD}(\lambda)$ Equation

$$T_n = \sum_{n=0}^{N_0} (1-\lambda)\lambda^n V_n + \lambda^{N_0} r$$









#### The End

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