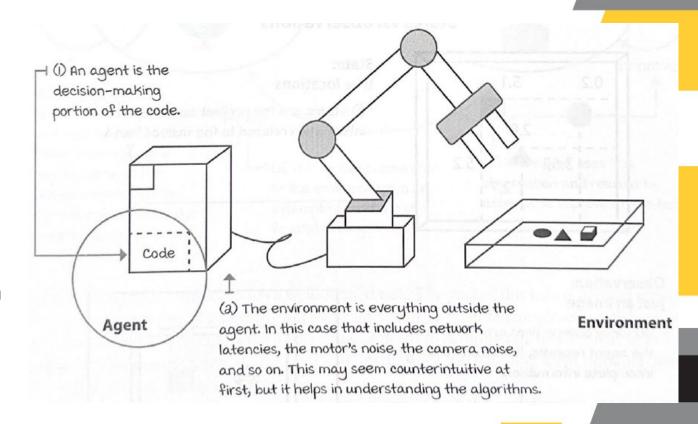
Introduction to Neural Networks



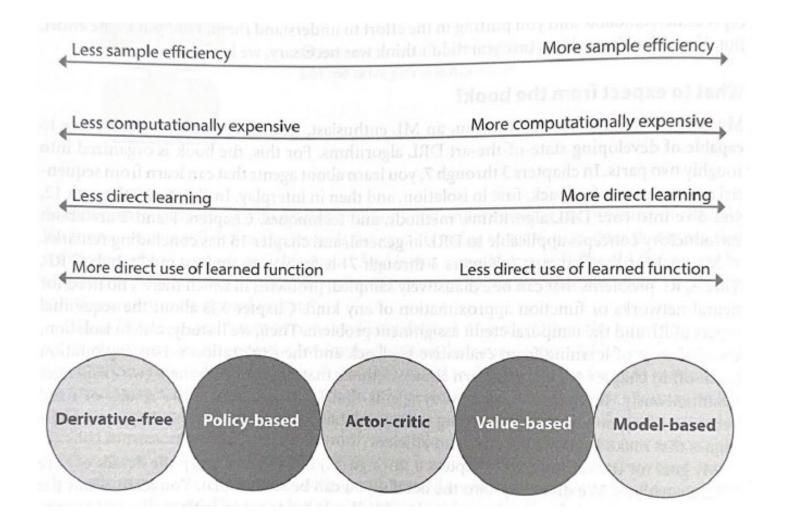
In RL

- Three approaches to learning:
 - Policies: map observations to actions
 - Depending on configuration:
 - Off-policy: Q-Learning
 - On-policy: SARSA
 - Models: learn the environment based on a model
 - Target model: goal of the agent
 - Base model: interaction of the agent with the environment (computationally intensive)
 - Value functions: learn to estimate the reward-to-go on mappings



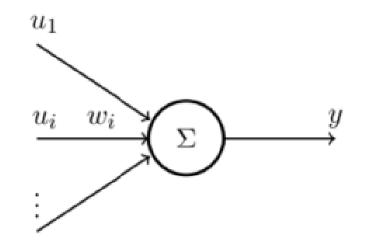
- As more observations, more Q-values, more new states are generated:
 - Interaction of the RL agent and environment grows significantly.
 - An RL approach for more complex environments will be computationally expensive.
- Since RL is based on "trial error", an RL agent will have difficulties in more complex environments
 - The "prediction" section of Q-learning
 - Also based on "trial error" by applying (state, action) from previous observations
- To solve this, implement *approximations*:
 - Neurons

More resources are needed depending on approach:



Neuron

- A neuron consists of:
 - An input, u
 - An output, y
 - An activation signal
 - A threshold, θ
 - A weight, w
 - *bias, b, can be part of the neuron, it can also be considered a hyperparameter



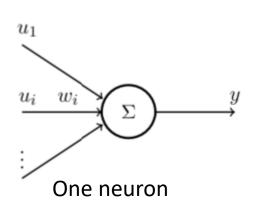
Mathematically:

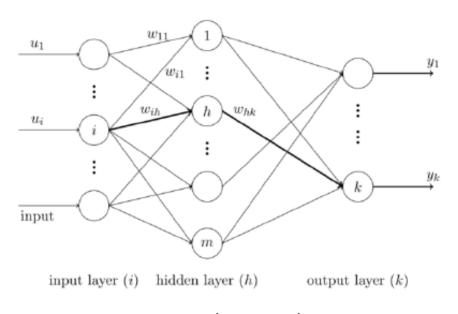
$$y = f(x), x = \sum_{i=1}^{n} w_i u_i$$

Where: x is a weighted sum, and f(x) an activation function depending on the parameter θ

Neuron

- When several neurons are configured in layers, and all neurons are connected to each other, it's known as a:
 - Fully connected neural network
 - In the simplest form, this is called: Artificial Neural Network, ANN
 - A model using an ANN with multiple (usually more than 2 or 3) NN layers is referred to as a model using *Deep Learning*.





Neural network

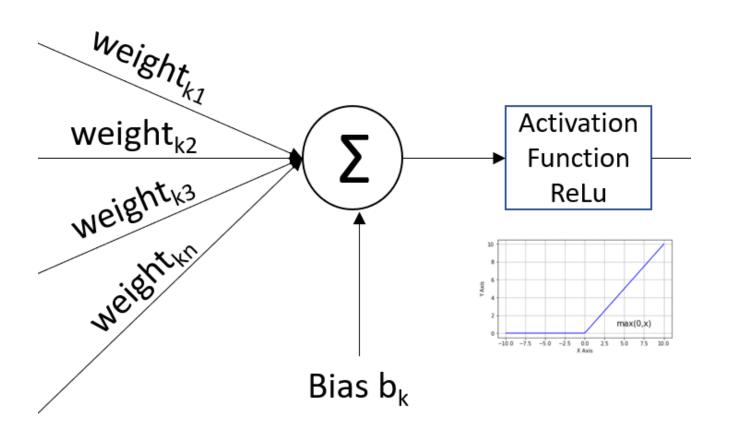
Layers

- In an ANN, in its basic structure: three layers
 - Input Layer:
 - Data coming from "the environment"
 - In DRL these will be observations from the agent
 - In Deep Learning:
 - Labeled data, i.e.
 - Depending on application:
 - Convolutional (& pooling layer): format the input data
 - Convolutional Neural Network, CNN
 - Recurrent: memory-based implementation
 - Recurrent Neural Network, RNN
 - <u>Hidden Layer(s)</u>
 - Where all approximations occur
 - Output Layer
 - Results of the approximations

Approximation

- In one neuron, calculations are occurring:
- Neurons in the hidden layer:
 - Input: weighted sum of neurons in previous layers
 - Weighted sum: input * weight + bias (bias can be optional, however, its is most likely to be included in the calculation)
 - Activation: check if the final weighted sum is within threshold parameters
 - If final weighted sum is within parameters:
 - Neuron remains active
 - If not:
 - Send weighted sum to next layer
 - Deactivate neuron
 - Neuron is no longer used on the next episode (epoch)
- Output layer:
 - Results of approximations:
 - transformed into formatted data

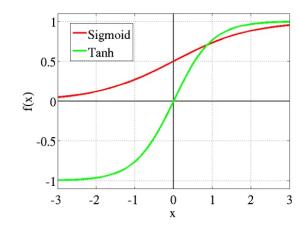
Hidden Layer

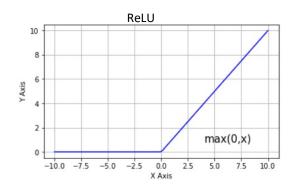


Activation

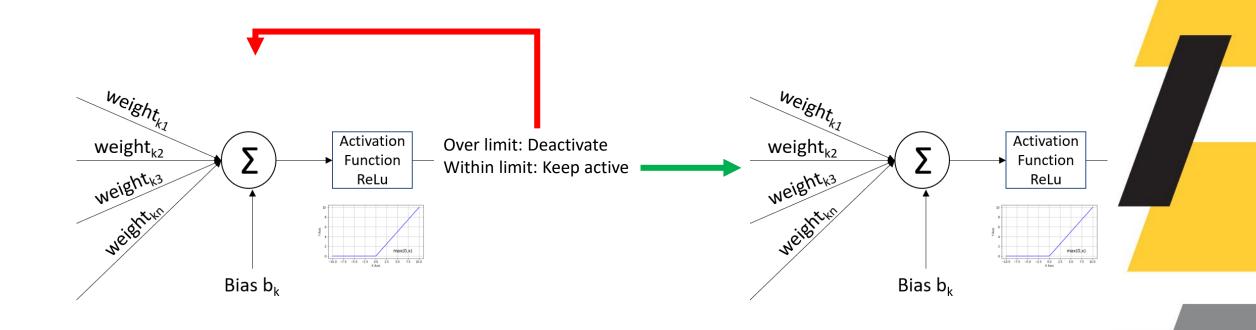
 Activation is used to decide whether the neuron in question should be kept active or not:

- Methods:
 - Softmax: based on probabilities
 - Calculates a probability
 - Sigmoid: when the output needs to
 - based on probabilities
 - Tanh: similar to sigmoid, except that
 - Tanh will accept negatives
 - Rectified Linear Unit (ReLU):
 - <u>f(x) is 0 when x is 0</u>





Activation



- Before approximations begin:
 - Data needs to be "transformed" into proper neural input
 - Convolutional layer (& pooling layer):
 - Filtering
 - Pooling
 - Flatten & dense

Representation of a fully connected ANN

