

Deep Learning Models for Games

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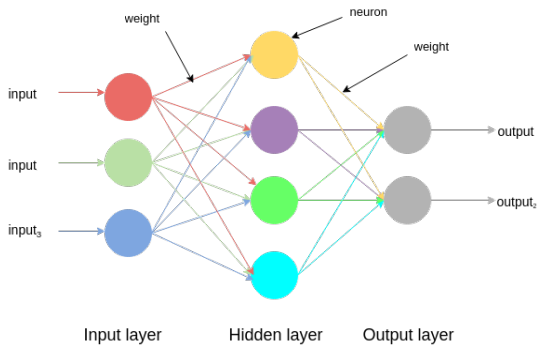
Motivation

Deep Learning

- find models capable of generalization
- extract from low-level(edges,colors) to high-level(combination of rudimentary features) features
- reduce programming burden
- applicability: cancer classification, autonomous cars, object recognition from images

Once upon a time...

- reinforcement learning: Q-Learning
- neural networks **vs** deep neural networks

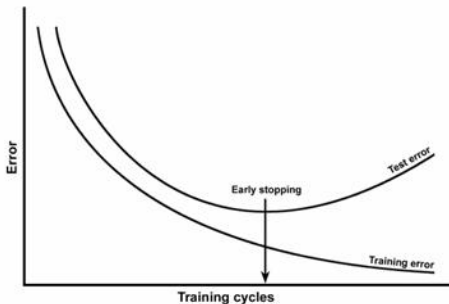


Preprocessing, Model, Loss function

- color space: RGB, YUV, grayscale
- data normalization 0..1, contrast normalization
- activation functions
 - hidden layer vs output layer
 - tanh, sigmoid, ReLU
- how many layers/features, what type of layers
- loss function: classification(binary/multi-class) or regression?
- gradient descent **vs** stochastic gradient descent

Train and test

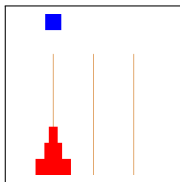
- split dataset for training and testing
- when to stop training?



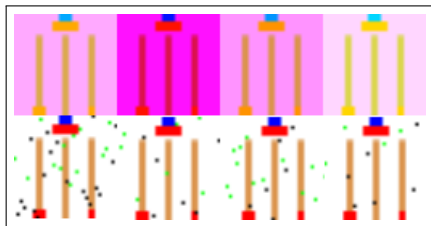
Source:

<http://documentation.statsoft.com/statisticahelp.aspx?path=sann/overview/sannoverviewsnetworkgeneralization>

Once upon a time...



Tower of Hanoi (first state)

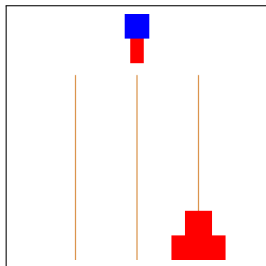


Dataset with noise added and color changed

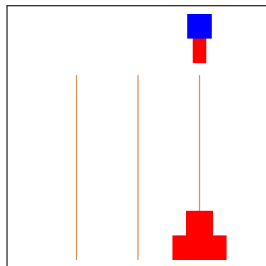
- game: Tower of Hanoi
- reinforcement learning: Q-Learning
- deep neural networks: predict values from Q-Learning
- machine learning framework: Torch based on Lua

Q-Learning

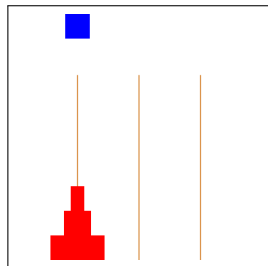
- finds optimal policy for action-value function
- $Q(s,a) = Q(s,a) + \alpha \cdot (r + \gamma \cdot \max_a Q(s', a') - Q(s, a))$



UP = 90,6534
DOWN = 86,8787
LEFT = 89,1867
RIGHT = 94,2824



UP = 97,6530
DOWN = 100,0000
LEFT = 93,8538
RIGHT = 92,5261



UP = 26,3520
DOWN = 23,8452
LEFT = 23,8897
RIGHT = 22,8827

Regression with complex model

nn

- predict values obtained from Q-Learning with convnet
- failure (the dnn was too complex)

Future work

- implement Q-Network
- test algorithm on dynamic environments or games where the state of the universe is not fully observed
- make Nao capable of playing Tic-Tac-Toe
- after all tasks mentioned above are done, use all the information gathered for cancer classification, etc.

Conclusions



IN CS, IT CAN BE HARD TO EXPLAIN THE DIFFERENCE BETWEEN THE EASY AND THE VIRTUALLY IMPOSSIBLE.



Source:

<http://xkcd.com/>

QA

Questions and Answers

Thank you for your attention!