



# Deep Learning Models for Games

## Bachelor Thesis Session – September 2015

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September 14, 2015



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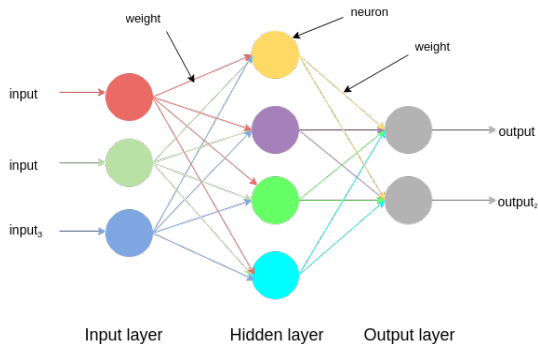
## Motivation

# 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



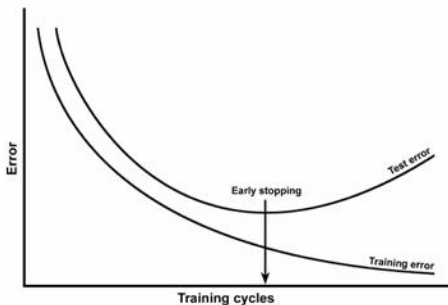


# Preprocessing, Model, Loss function

- 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

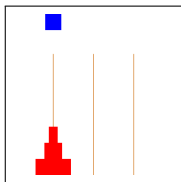
- split dataset for training and testing
- when to stop training?



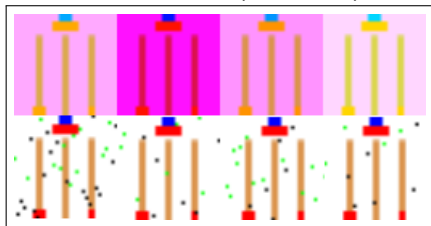
Source:

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

# Once upon a time...



Tower of Hanoi (first state)

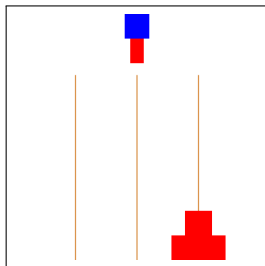


Dataset with noise added and color changed

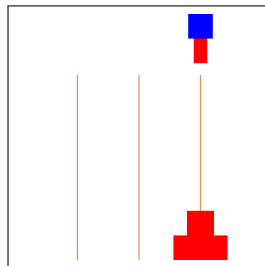
- game: Tower of Hanoi
- reinforcement learning: Q-Learning
- deep neural networks: predict values from Q-Learning
- machine learning framework: Torch based on Lua

# Q-Learning

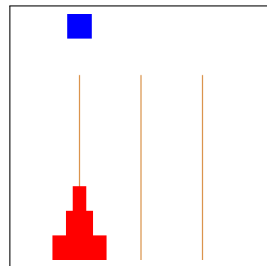
- finds optimal policy for action-value function
- $Q(s,a) = Q(s,a) + \alpha \cdot (r + \gamma \cdot \max_a Q(s', a') - Q(s, a))$



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



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



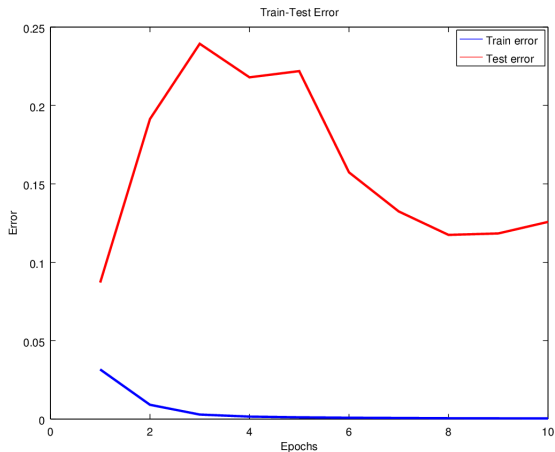
UP = 26,3520  
 DOWN = 23,8452  
 LEFT = 23,8897  
 RIGHT = 22,8827



# Model

- (1): `nn.SpatialConvolutionMM(3 → 8, 5x5)`
- (2): `nn.Tanh`
- (3): `nn.SpatialSubSampling`
- (4): `nn.SpatialConvolutionMM(8 → 20, 5x5)`
- (5): `nn.Tanh`
- (6): `nn.SpatialSubSampling`
- (7): `nn.SpatialConvolutionMM(20 → 120, 5x5)`
- (8): `nn.Reshape(120)`
- (9): `nn.Linear(120 → 100)`
- (10): `nn.Tanh`
- (11): `nn.Linear(100 → 4)`
- (12): `nn.Sigmoid`

# Results



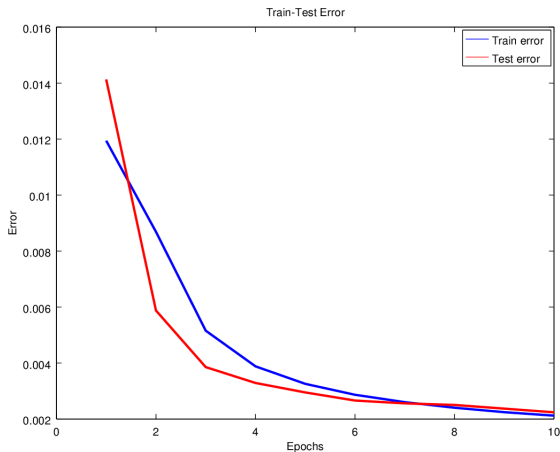


# Model

- (1): `nn.SpatialConvolutionMM(3 → 4, 5x5)`
- (2): `nn.Tanh`
- (3): `nn.SpatialSubSampling`
- (4): `nn.SpatialConvolutionMM(4 → 6, 5x5)`
- (5): `nn.Tanh`
- (6): `nn.SpatialSubSampling`
- (7): `nn.Reshape(150)`
- (8): `nn.Linear(150 → 4)`



# Results





# 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!**