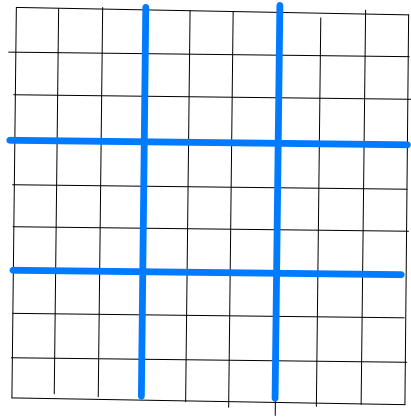


Prog2 UTTT-AI



3x3 Boards
with each 9 fields
↓
81 fields in total

Approach:

Input Nodes Design

- 81 input nodes representing game for symbol cross
- 81 input nodes representing game for symbol circle
- 1 input node containing the next BoardIndex to tell whether there is a specific board to be played on
162 + 1 Nodes

Hidden Layers Design

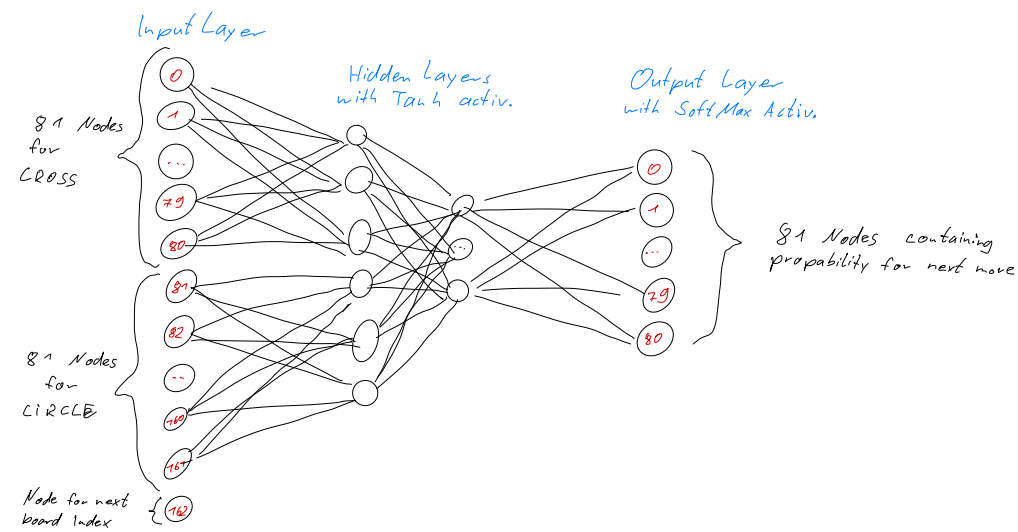
- Coming Soon
1 Layer

Output Nodes Design

- 81 output nodes since there may be 81 possibilities to place a symbol
- each node contains probability for placing at specific field
- Use softmax to make sure probabilities add up to 1

81 Nodes

Design



Java Implementation

Matrix

- create matrix (rows, cols)
populate with random value
random() * 2 - 1;
data stored in 2d Double array
- elementwise addition using a value on a matrix
- elementwise subtraction using a value
- switch rows and columns of matrix
- matrix multiplication of 2 matrices
- elementwise multiplication with matrix
- elementwise multiplication with value
- activation functions and its derivative
- helper function to convert matrix to game state
- helper function to convert game state to matrix

Neural Network

- Variables
weight for input layer and hidden layer
weight for output layer and hidden layer
bias for hidden layer
bias for output layer
learning rate
- create Neural Network
create weight matrix for input/hidden layer
by number of hidden layer and input nodes
create weight matrix for output/hidden layer
by number of output nodes and hidden layer nodes
create bias matrix for hidden layer nodes and amount of layers
create bias matrix for output layer nodes and amount of output layers
- predict next move (forward propagation)
get current game state and convert to matrix
multiply hidden layer matrix weights with input
add bias of hidden layer
apply activation function
multiply output layer weights with hidden layer matrix
add bias of output layer
apply activation function (softmax)
return game state

train (current game state, target game state)

- get current game state as matrix
create hidden matrix by multiplying
add bias to hidden matrix
apply activation function
- create output matrix by multiplying
add bias to output matrix
apply activation function softmax
- grab target game state as matrix
create error matrix by subtracting (target, output)
create gradient by applying derivative activation function to output
multiply gradient with error
multiply gradient with learning rate
switch rows and cols for hidden matrix to store error matrix
calculate weights hidden updates by multiplying gradient with switched matrix
add hidden output delta to weights hidden output
add gradient to bias output
switch rows and cols of weight hidden output to store learning rate
calculate hidden error matrix by multiplying switched matrix and error
calculate hidden gradient by applying derivative activation function on hidden layer
multiply with hidden error
multiply with learning rate
switch rows and cols of input nodes
calculate weights hidden input delta by multiplying hidden gradient with switched matrix
add weight input hidden delta to input hidden weights
add hidden gradient to bias hidden

train model on larger data set by doing x iterations

- iterate x times
get random value between 0 and length of input state set
train with input data at random value and expected state at random value

References:

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