

Best policy for notakto

Best policy for the first player on 3X3 board: 6 basic mapping pairs. The number of pieces is always even since we only consider the first player.

Current state

Winning state

Pair #1


	W	

Current state

Winning state

Pair #2

X		
	X	

		W
		W
W	W	

Current state

Winning state

Pair #3

	X	
	X	

W		W

Current state

Winning state

Pair #4

X		X
X	X	

	W	

Current state

Winning state

Pair #5

X		X
	X	
	X	

W		W

Current state

Winning state

Pair #6

X		
	X	X
	X	

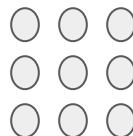
		W
W		

# The first method I tried: One-layer network

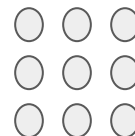
Training data:

```
0 0 0 0 1 0 0 1 0,1 0 1 0 0 0 0 0 0
0 0 0 1 1 0 0 0 0,0 0 1 0 0 0 0 0 1
0 1 0 0 1 0 0 0 0,0 0 0 0 0 0 1 0 1
0 0 0 0 1 1 0 0 0,1 0 0 0 0 0 1 0 0
1 0 1 1 1 0 0 0 0,0 0 0 0 0 0 0 1 0
0 0 0 1 1 0 1 0 1,0 1 0 0 0 0 0 0 0
0 0 1 0 1 0 0 1 1,0 0 0 1 0 0 0 0 0
1 0 1 0 1 1 0 0 0,0 0 0 0 0 0 0 1 0
1 1 0 0 1 0 1 0 0,0 0 0 0 0 0 1 0 0
1 0 1 0 1 0 0 1 0,0 0 0 1 0 1 0 0 0
0 1 0 0 1 0 1 0 1,0 0 0 1 0 1 0 0 0
0 0 1 1 1 0 0 0 1,0 1 0 0 0 0 0 1 0
1 0 0 0 1 1 1 0 0,0 1 0 0 0 0 0 1 0
1 0 0 0 1 1 0 1 0,0 0 1 0 0 0 0 1 0 0
0 1 0 0 1 1 1 0 0,1 0 0 0 0 0 0 0 1
0 1 0 1 1 0 0 0 1,0 0 1 0 0 0 0 1 0 0
0 0 1 1 1 0 0 1 0,1 0 0 0 0 0 0 0 1
0 0 0 0 0 0 0 0 0,0 0 0 0 1 0 0 0 0
1 0 0 0 1 0 0 0 0,0 0 0 1 0 0 1 1 0
```

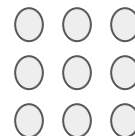
Input matrix



Matrix to be trained



Output matrix



Training model: One layer neural network.

Conclusion: linear combination does not work for normal parameters. To check whether my network is validated, I trained the network to learn an identity matrix.

The network works for learning an identity matrix.

Input: 3\*3

```
input, current state:
tensor([[[[1., 0., 0.],
          [0., 1., 0.],
          [0., 0., 0.]],

         [[0., 0., 0.],
          [0., 0., 0.],
          [0., 0., 0.]],

         [[0., 0., 0.],
          [0., 0., 0.],
          [0., 0., 0.]],

         [[0., 1., 0.],
          [0., 1., 0.],
          [0., 0., 0.]],

         [[1., 0., 0.],
          [0., 1., 0.],
          [0., 0., 0.]],

         [[1., 0., 0.],
          [0., 1., 1.],
          [0., 1., 0.]]]], device='cuda:0')
```

Parameters 3\*3

```
current parameter 3 * 3
tensor([[[ 0.9119, -0.4182,  0.0141],
          [ 0.3326,  1.0319,  0.1591],
          [ 0.1205, -0.1142,  0.8618]], device='cuda:0')
```

Output 3\*3

```
tensor([[[[ 0.9119, -0.4182,  0.0141],
           [ 0.3326,  1.0319,  0.1591],
           [ 0.0000,  0.0000,  0.0000]],

          [[ 0.0000,  0.0000,  0.0000],
           [ 0.0000,  0.0000,  0.0000],
           [ 0.0000,  0.0000,  0.0000]],

          [[ 0.0000,  0.0000,  0.0000],
           [ 0.0000,  0.0000,  0.0000],
           [ 0.0000,  0.0000,  0.0000]],

          [[ 0.3326,  1.0319,  0.1591],
           [ 0.3326,  1.0319,  0.1591],
           [ 0.0000,  0.0000,  0.0000]],

          [[ 0.9119, -0.4182,  0.0141],
           [ 0.3326,  1.0319,  0.1591],
           [ 0.0000,  0.0000,  0.0000]],

          [[ 0.9119, -0.4182,  0.0141],
           [ 0.4532,  0.9177,  1.0210],
           [ 0.3326,  1.0319,  0.1591]]]], device='cuda:0')
```

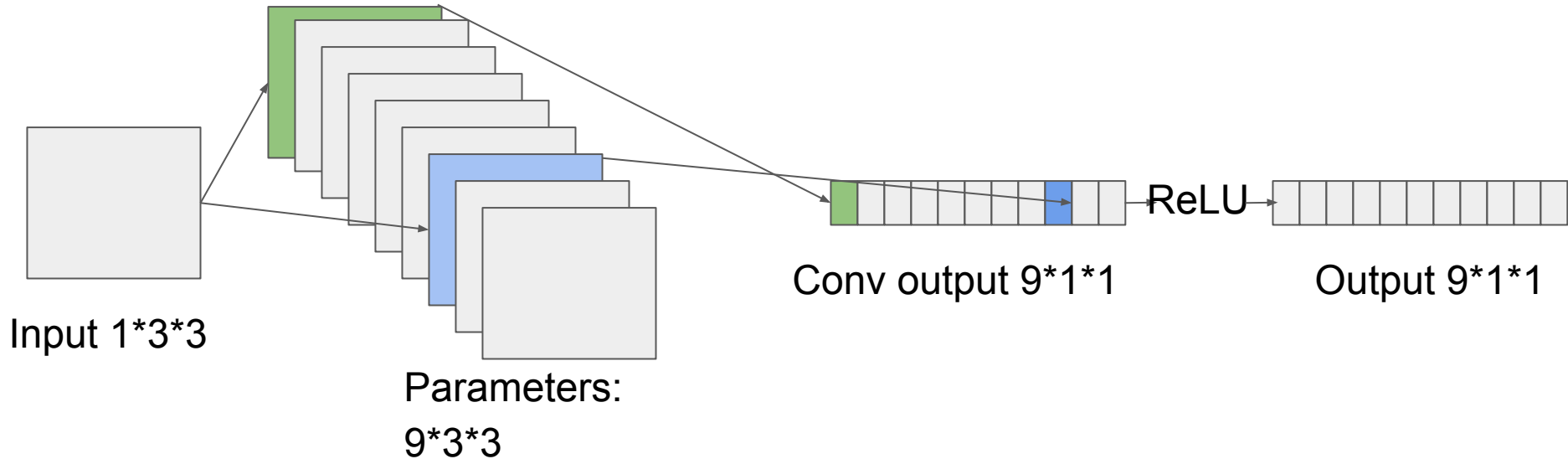
## The second method I tried: Bruteforce

Try all possible mapping pairs  $B$ , check if the current state  $A$ , and the “can force a win” states  $C$  can be modeled as a matrix multiplication. Namely,  $C=A*B$ .

Conclusion: for binary parameters, such a solution does not exist.

What's next? Multi-layer neural networks with non-linear activation functions?

# The third way: Convolutional neural network



Output:  $9 \times 1 \times 1$ . Each element in output corresponds to one position in the 'can force a win' matrix ( $3 \times 3$ ). If an element in the  $9 \times 1 \times 1$  is larger than 0.5, then it is considered as 1. Later I found that the ReLU layer can be removed.

# The third way: Parameters

Can the parameters be translated into human interpretable description? Can we binarize these parameters? Can we use brute force to get the total 81 parameters?  
 $9 \times 2^9 = 4608$ .

```
tensor([[[[-1.6300, -0.3708,  0.3699],  
          [-0.3699,  0.3699,  0.6301],  
          [ 0.3708,  0.6301, -1.3699]]],  
        [[[ 0.1472,  0.1150,  0.6159],  
          [-1.0476, -0.2555, -0.5591],  
          [ 0.2432,  0.2405,  0.6492]]],  
        [[[ 0.5001,  0.5001, -1.5000],  
          [-0.4999,  0.4999,  0.5001],  
          [-1.5001, -0.5000,  0.4999]]],  
        [[[ 0.0334, -1.2069,  0.2219],  
          [ 0.0514, -0.3086,  0.3078],  
          [ 0.7169, -0.7201,  0.7870]]],  
        [[[-0.0496, -0.0496, -0.0496],  
          [-0.0496, -0.0496, -0.0496],  
          [-0.0496, -0.0496, -0.0496]]],  
        [[[ 0.8731, -0.9020,  0.3775],  
          [-0.3685,  0.1258, -0.1242],  
          [-0.1256, -1.3873,  0.6143]]],  
        [[[ 0.5510, -0.4490, -1.5510],  
          [ 0.5510,  0.4490, -0.5510],  
          [-1.4490,  0.5510,  0.4490]]],  
        [[[ 0.8869, -0.1208, -0.0968],  
          [-0.9433,  0.1114, -1.3979],  
          [ 0.3738, -0.1521,  0.3870]]],  
        [[[-1.3547,  0.6453,  0.3547],  
          [ 0.6453,  0.3547, -0.3547],  
          [ 0.3547, -0.3547, -1.6453]]]])
```