

AN ALPHAZERO IMPLEMENTATION OF ULTIMATE TIC-TAC-TOE

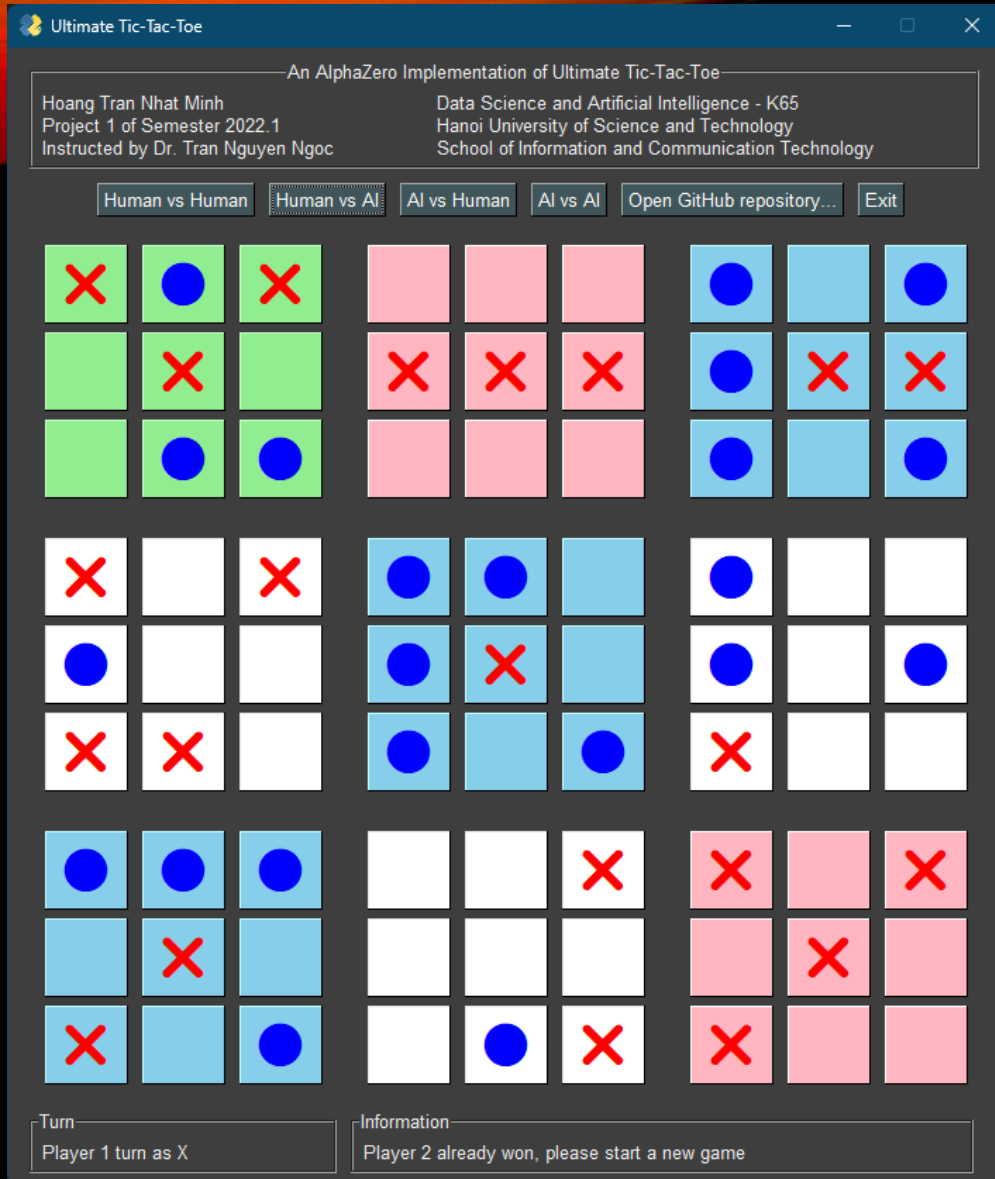
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Instructed by Dr. Tran Nguyen Ngoc

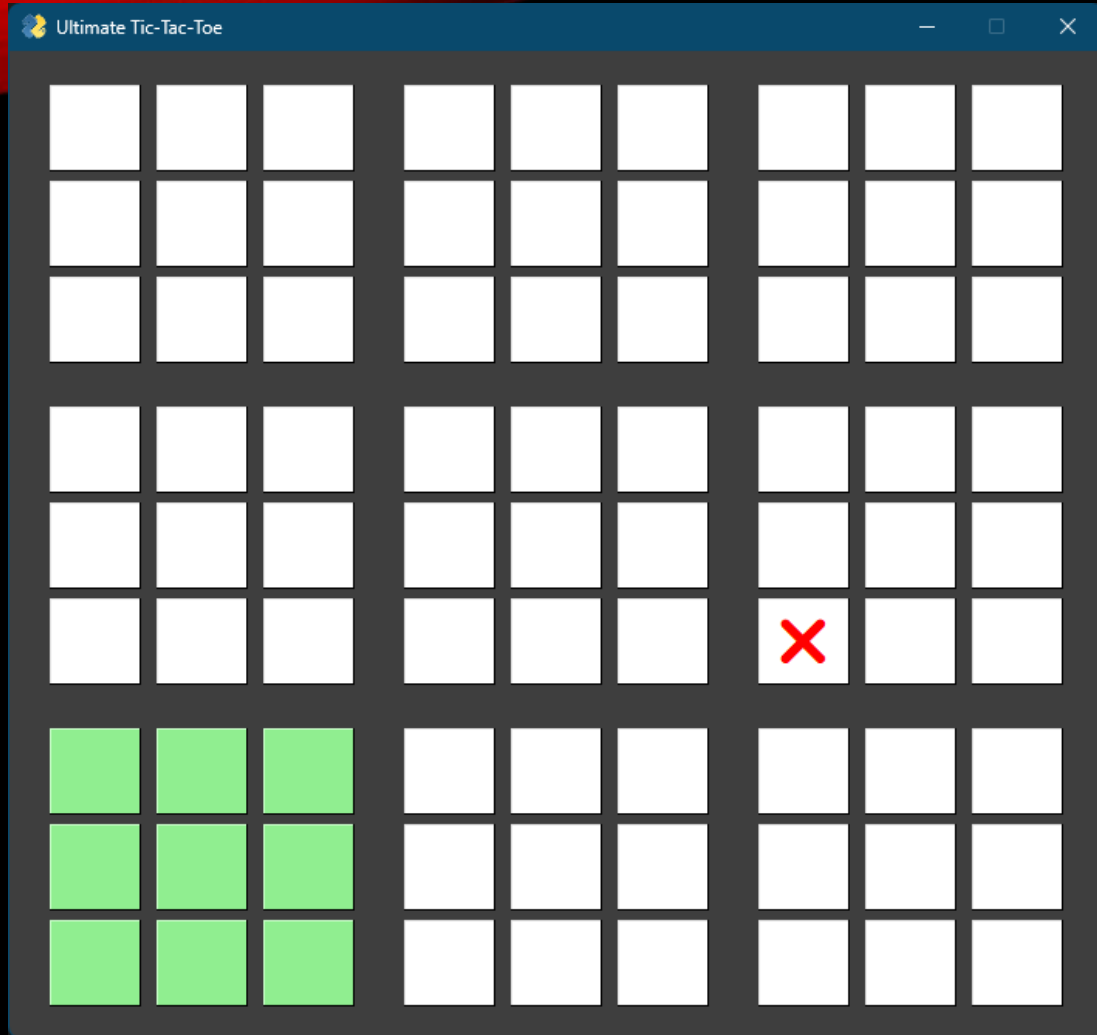


SOICT



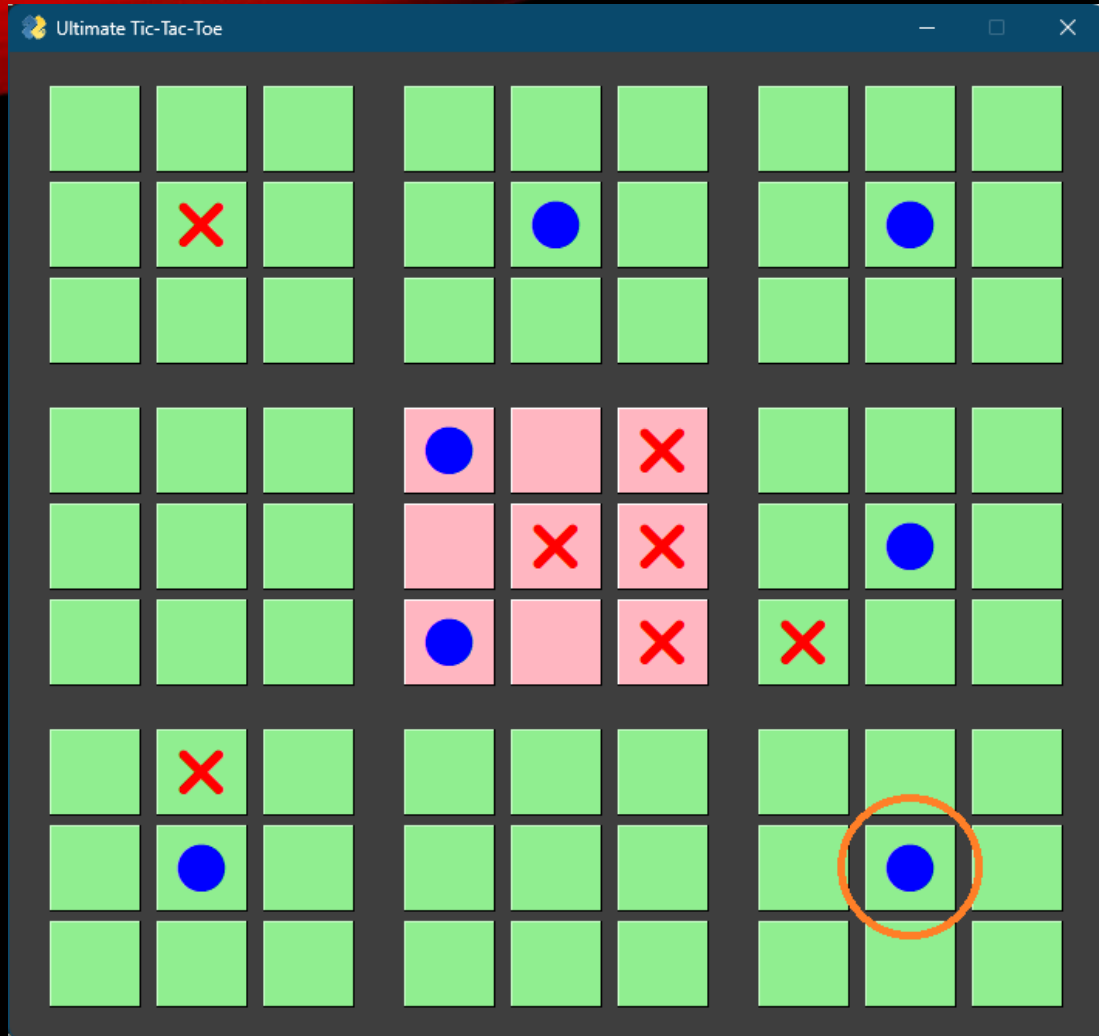
ABSTRACT

A modified version of the AlphaZero algorithm is used to train a new artificial intelligence program that plays ultimate tic-tac-toe, one of the most difficult variations of tic-tac-toe, while making use of human-extracted features using a convolutional neural network.



INTRODUCTION TO ULTIMATE TIC-TAC-TOE

- “Send” opponent to the relative local board in the next move

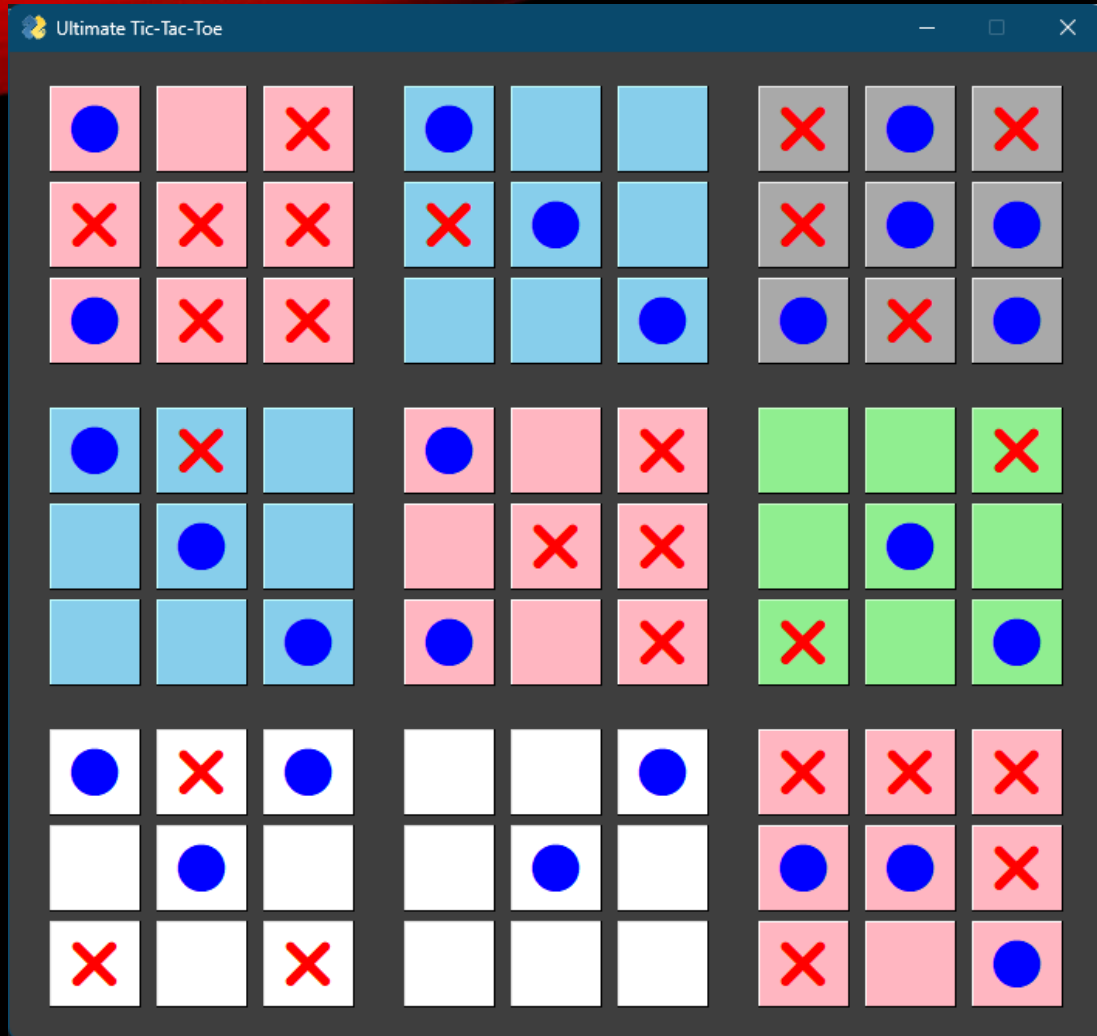


INTRODUCTION TO ULTIMATE TIC-TAC-TOE

- If a local board is determined, it is not playable anymore
- If sent to it, play anywhere else (but determined boards)

INTRODUCTION TO ULTIMATE TIC-TAC-TOE

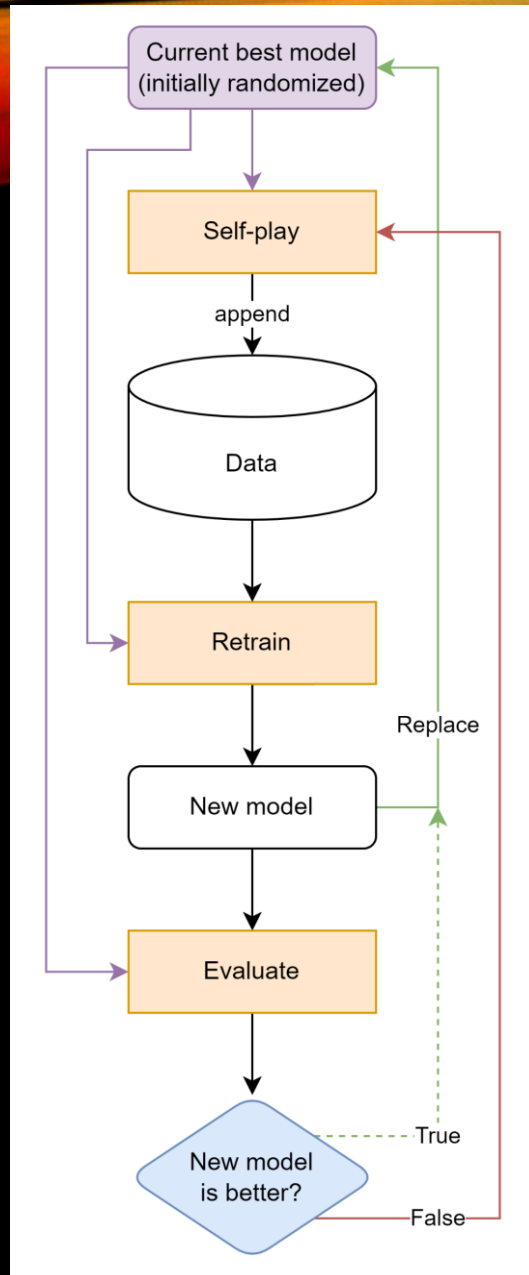
Player 1 won this game



ALPHAZERO: TRAINING LOOP

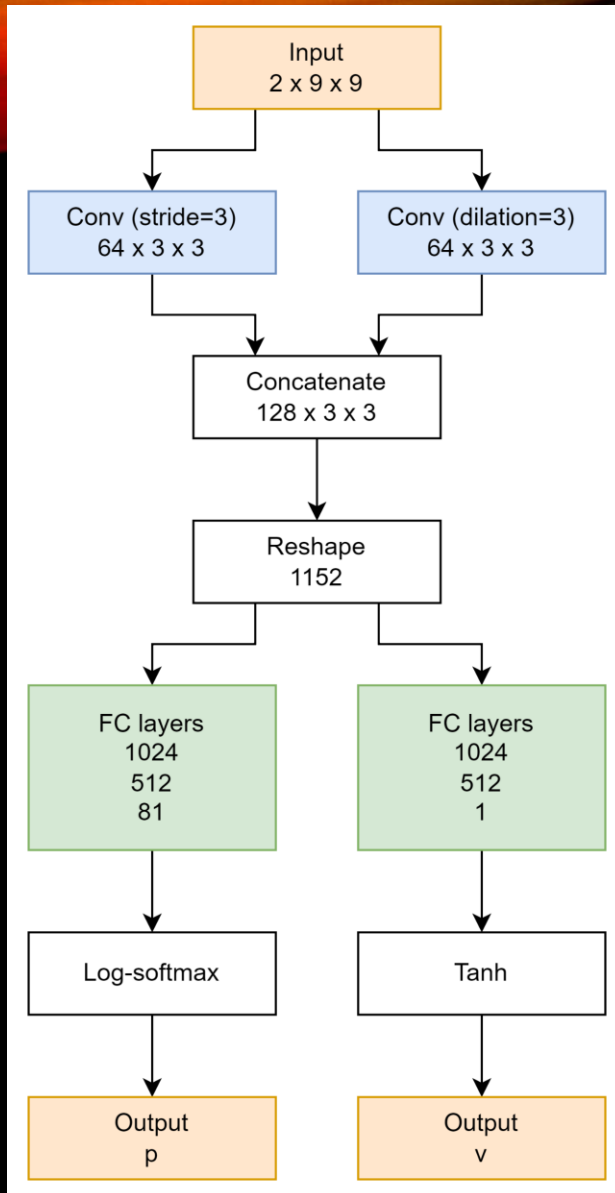
- **Self-play**: to generate new data
- **Retrain**: to get new model
- **Evaluate**: to decide replacing the best model or not

In **self-play** and **evaluate**: Monte Carlo tree search decision guided by a convolutional neural network (model)

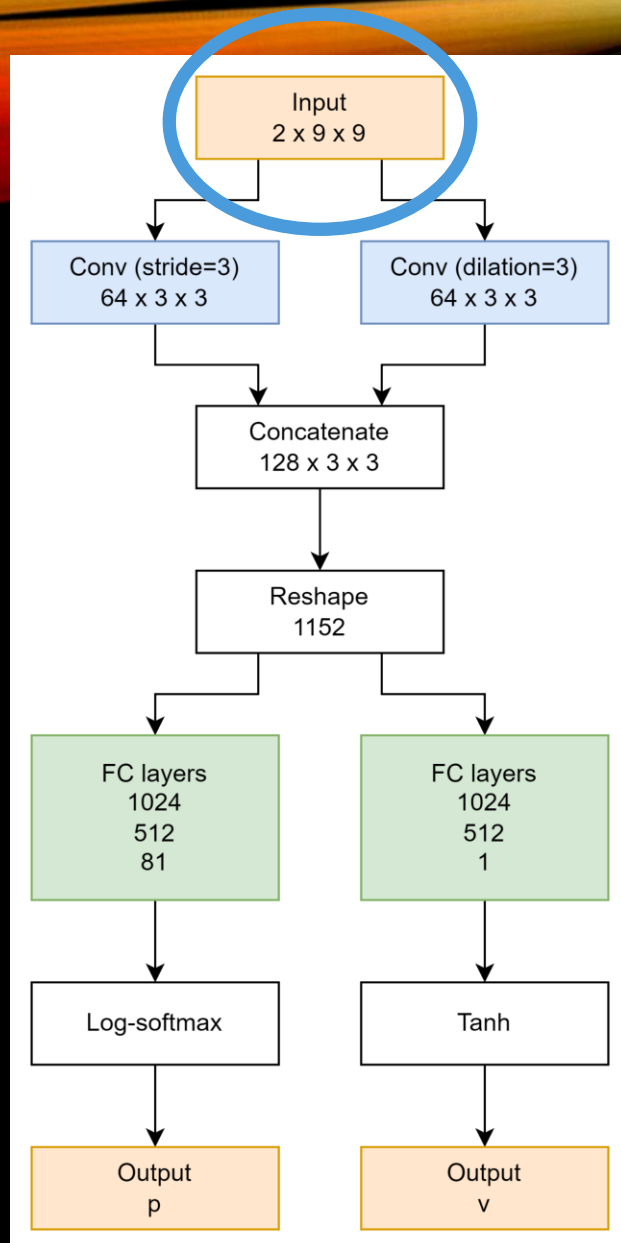


NETWORK ARCHITECTURE

- Convolutional layers
- Fully connected layers



INPUT



- Channel #1: state (values: 1, -1, 0)
- Channel #2: valid-move mask (values: 1, 0)

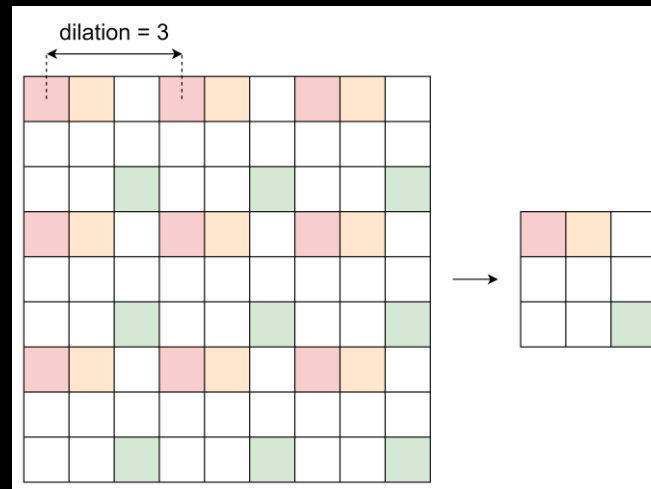
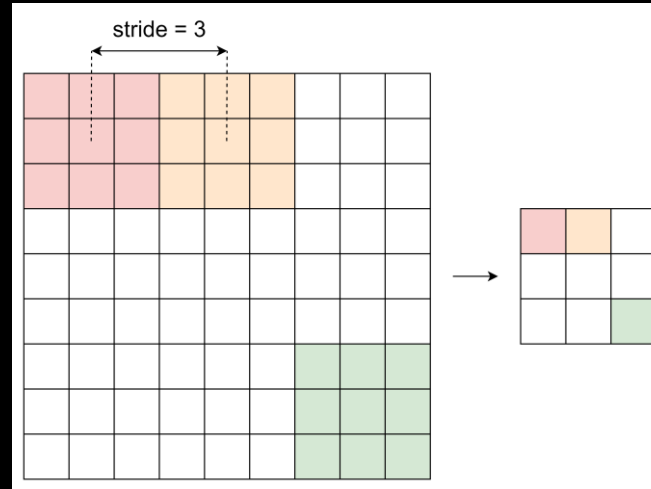
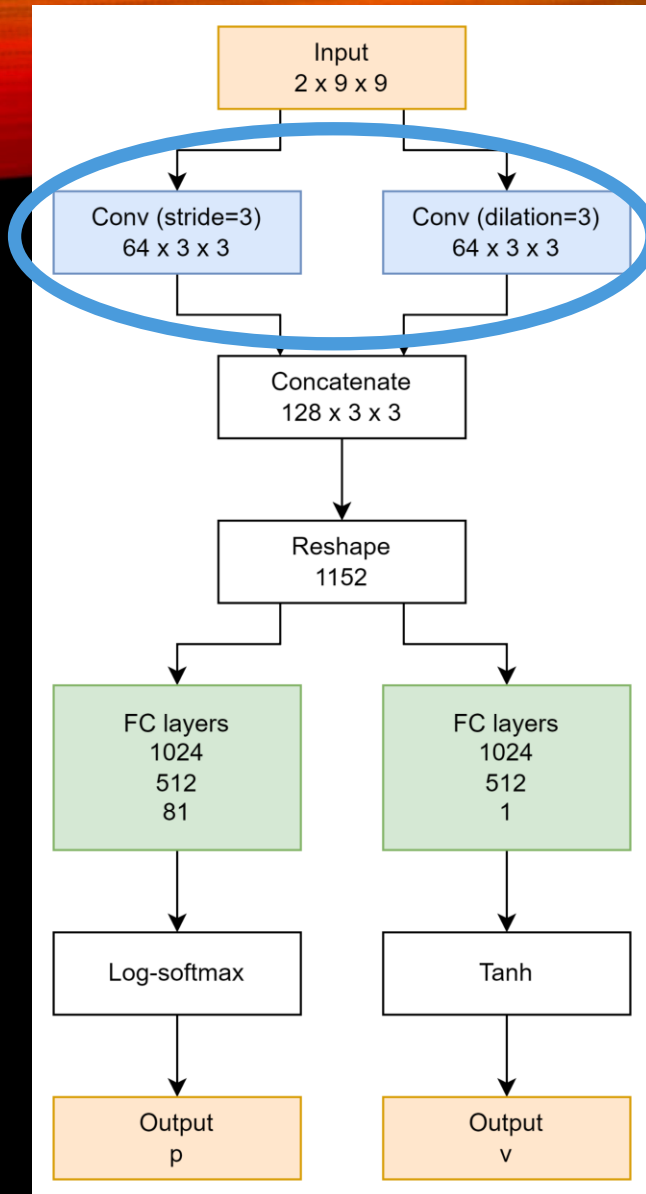
CONVOLUTIONAL LAYERS

Similarities

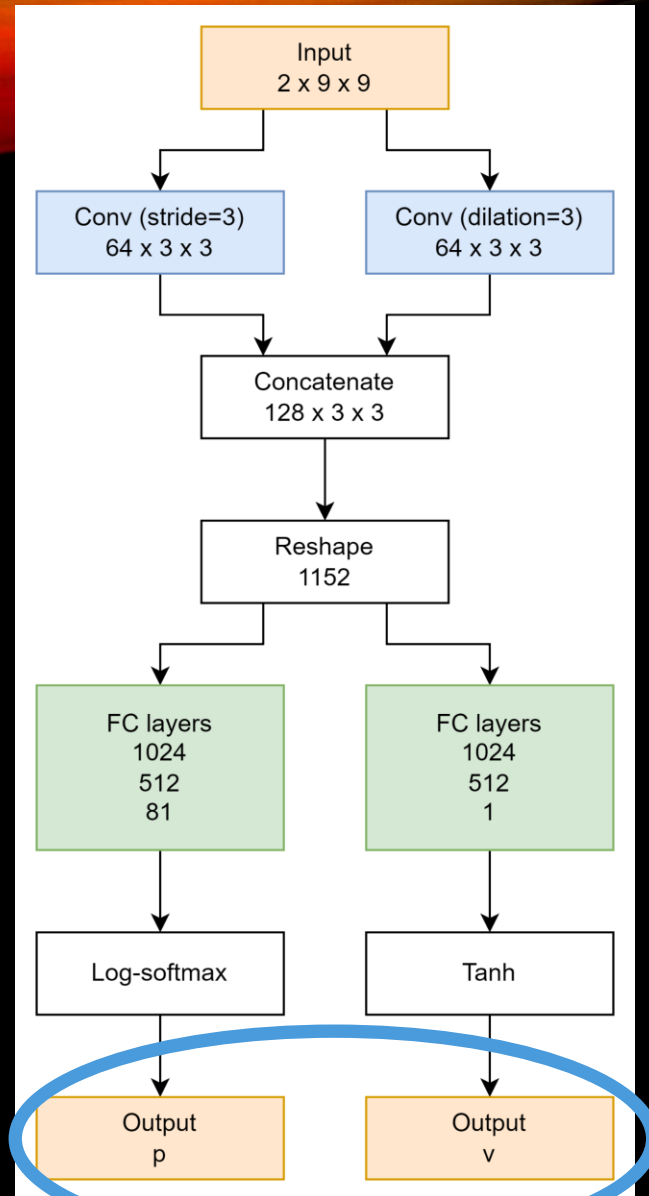
- Kernel size: 3x3
- Output image size: 3x3

Differences

- Layer #1: stride = 3
- Layer #2: dilation = 3


























































































OUTPUT



- p : Policy, the probabilities of moves
- v : Value, the evaluation of the input board

RESULTS

Iter.	Win	Loss	Draw	Win rate
1				50.0%
2				60.0%
3				38.5%
4				27.3%
5				41.7%
6				60.0%
7				42.9%
8				87.5%
9				57.1%
10				57.1%
11				52.2%
12				52.6%
13				55.6%
14				50.0%
15				52.6%
16				47.1%
17				50.0%
18				47.6%
19				50.0%
20				60.0%
21				57.9%
22				42.1%
23				47.1%
24				56.3%
25				47.4%
26				52.6%
27				56.3%
28				57.9%
29				55.6%

- Tested itself in the training loop
 - Discarding previous models: improvement
- Beginner level

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- Oversimplified, isn't it? Please see the report for more details.
- Everything is in <https://github.com/htnminh/AlphaZero-Ultimate-TicTacToe>