

## PROJECT S6 GROUP 5

4 THE WIN

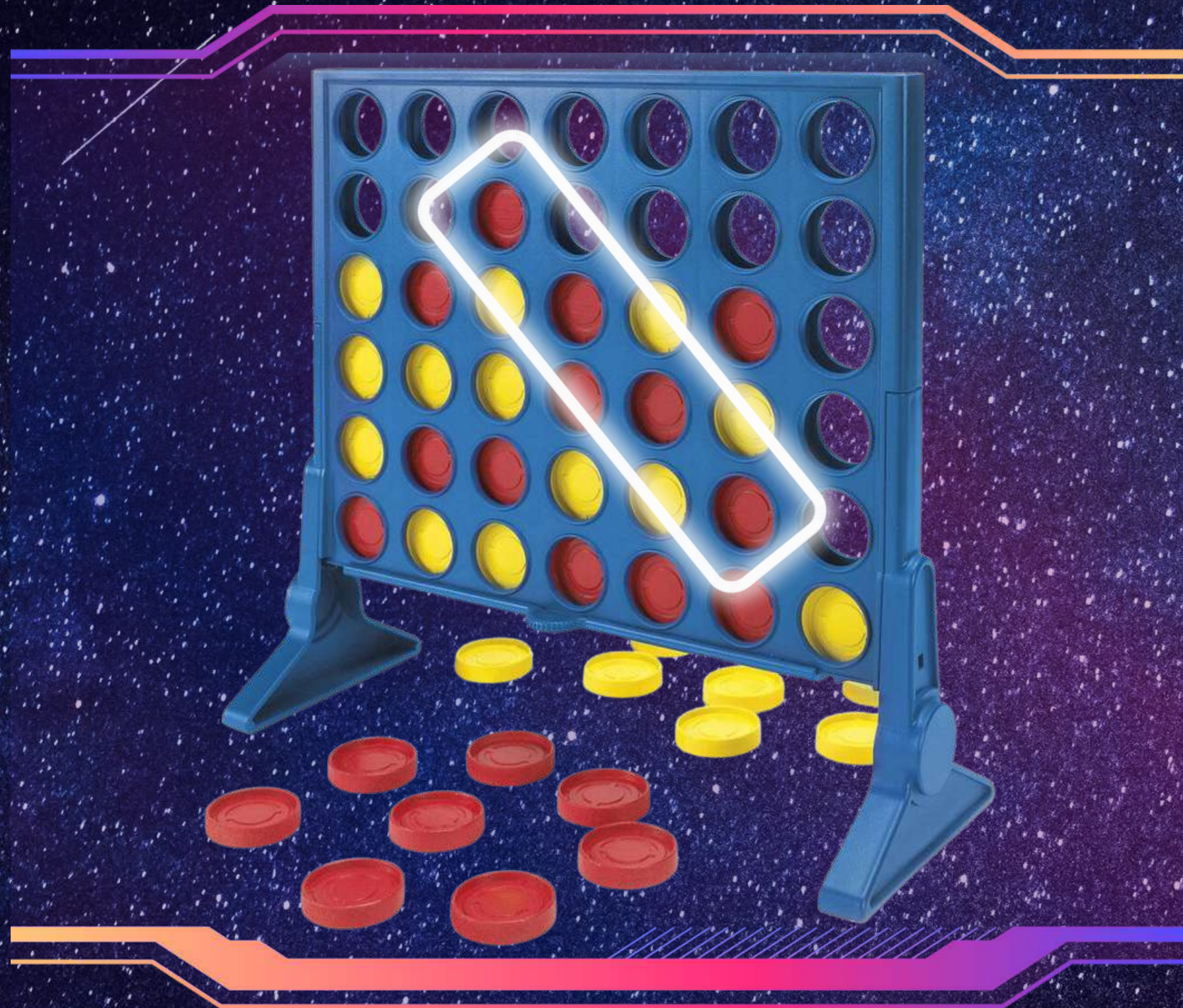




# SUMMARY



# HISTORICAL GAME





# INTERFACE





# ALPHAGOZERO



has beaten the GO the world champion Lee Sedol 4 games to 1 1



has beaten the GO The updated AlphaZero crushed Stockfish 8 in a new 1,000-game match, scoring +155 -6 =83 the world champion Lee Sedol 4 games to 1 1

**A technology powerful enough to completely dominate human intelligence on the board**



# ALPHAGOZERO

A quick overview of a simple method

A Monte-Carlo Tree Search (MCTS)

A current\_best model, playing against himself, generating data

Apprenticeship mode, learning from the data gathered



# ALPHAGOZERO

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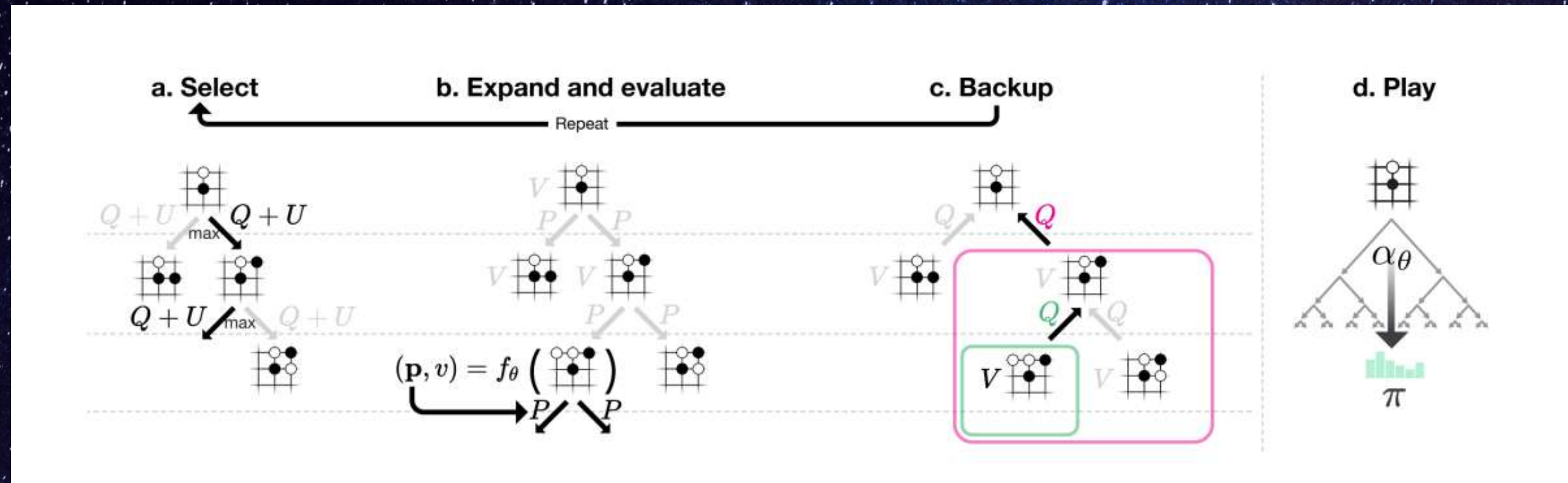
A current\_best model, playing against himself, generating data

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# MONTE-CARLO TREE SEARCH

## MCTS for AlphaGo Zero



- $Q(s,a)$ : the expected reward for taking action  $a$  from state  $s$ , i.e. the  $Q$  values
- $N(s,a)$ : the number of times we took action  $a$  from state  $s$  across simulations
- $P(s, \cdot) = P$  the initial estimate of taking an action from the state  $s$  according to the policy returned by the current neural network.

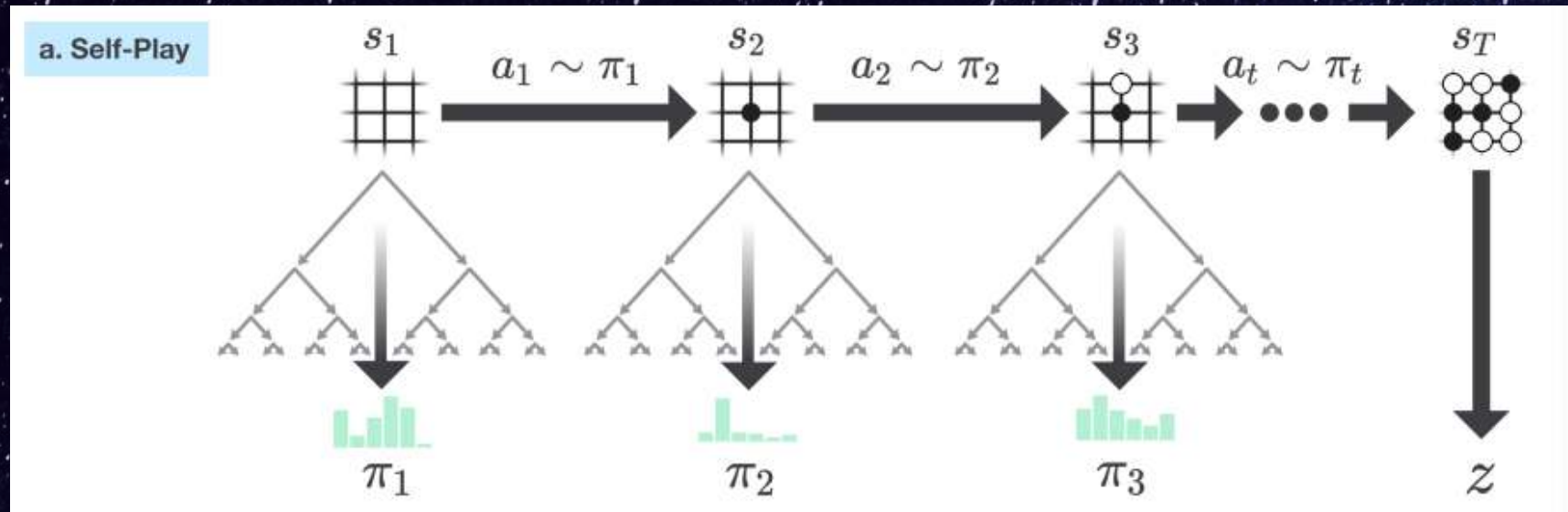
Where the actions are sampled following;

$$U(s, a) \propto Q(s, a) + P(s, a) / (1 + N(s, a))$$



# SELF-PLAY AND TRAINING

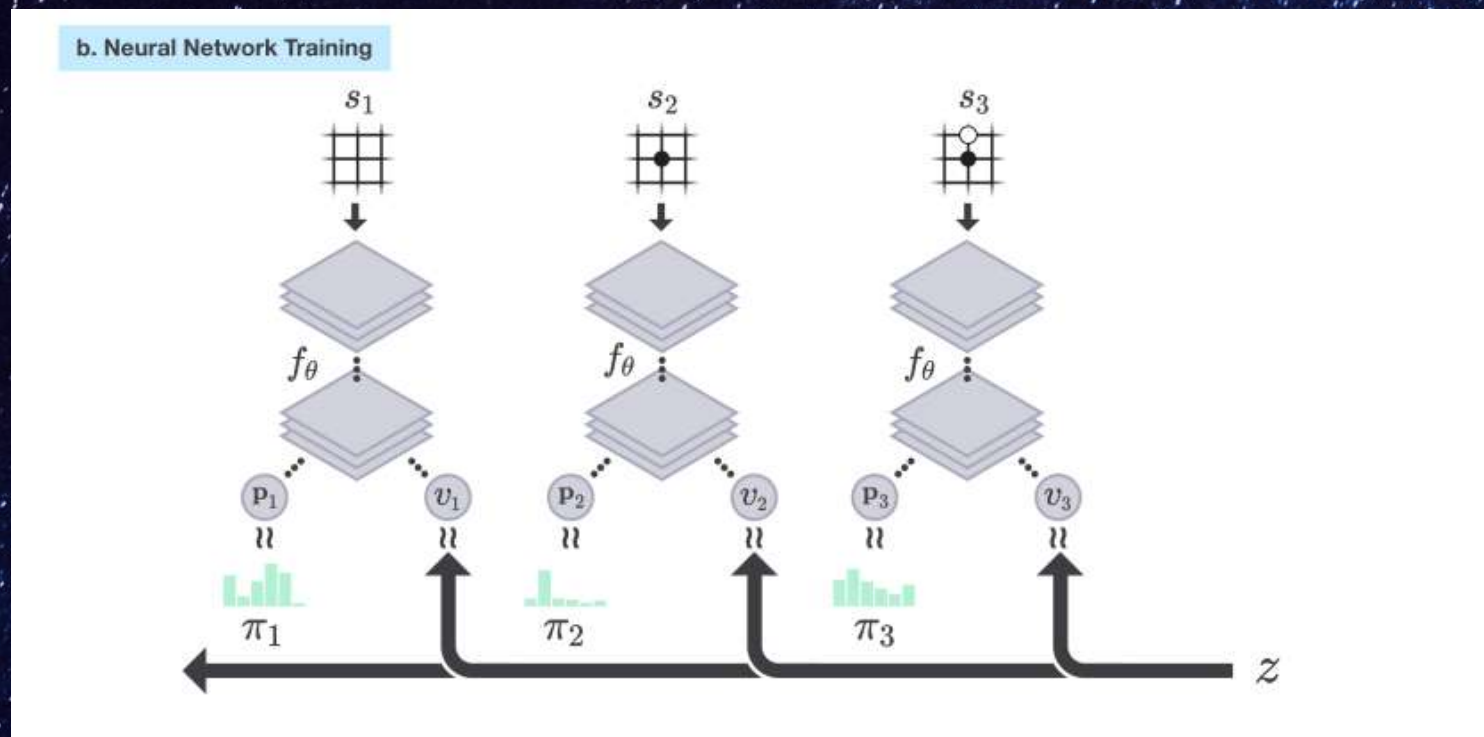
## Self-Play Reinforcement learning



The data for each time-step  $t$  is stored as a list of  $(s_t, \pi_t, z_t)$ ,

The Neural Network new parameters  $\theta_i$  are trained from the uniformly sampled dataset, and adjust the loss function using gradient descent

$$l = (z - v)^2 - \pi^\top \log p + c \|\theta\|^2$$



MSE between the actual winner and the network's prediction

cross-entropy losses, between  $P$  and  $\pi$

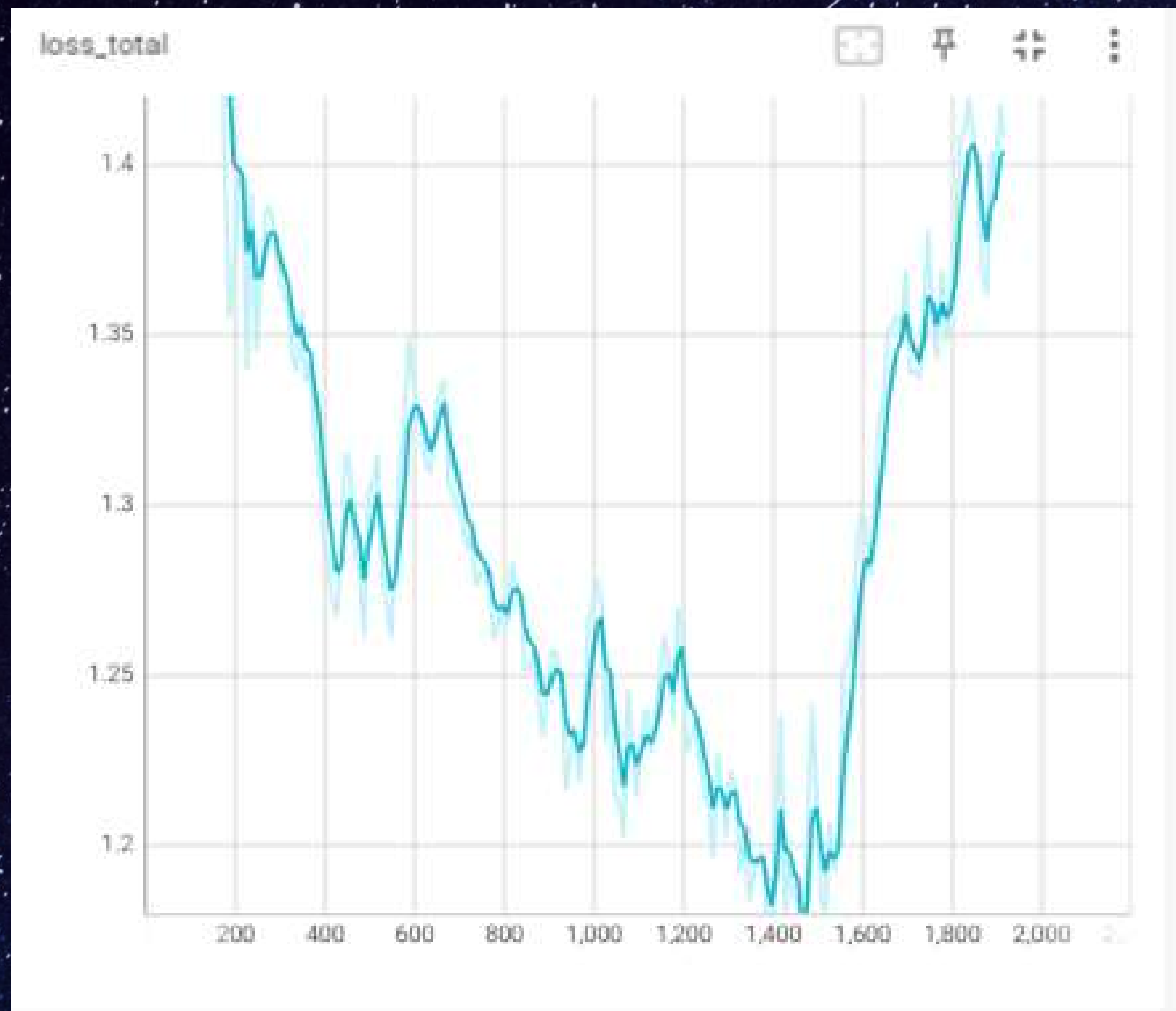


# CONNECT4 RESULTS

FIRST TRAINING (1.6HOUR, 2K GAMES)

LOSS FUNCTION OVER THE STEPS

SOME HYPERPARAMETERS ::



LEARNING\_RATE = 0.1 # controls that is the step size of the adjustment of the parameters of the NN

BATCH\_SIZE = 256 # specifies the number of training samples used in each iteration of gradient descent



# CONNECT4 RESULTS

SECOND TRAINING (3.13 DAYS, 87,5K GAMES)

LOSS FUNCTION OVER THE STEPS



HYPERPARAMETER UPDATE

LEARNING\_RATE = 0.1 # in the DeepMindPaper is adjusted

BATCH\_SIZE = 512 # DeepMind Paper = 2048

DECREASING THE LEARNING RATE OVER STEP  
TO MATCH DEEPMIND'S IMPLEMENTATION



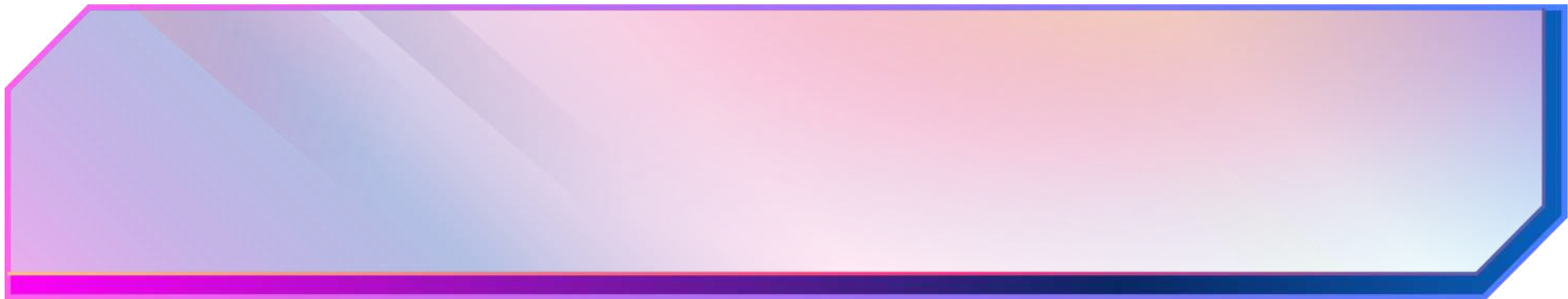
# CONNECT4 RESULTS

## FINAL MODEL AND COMMENT

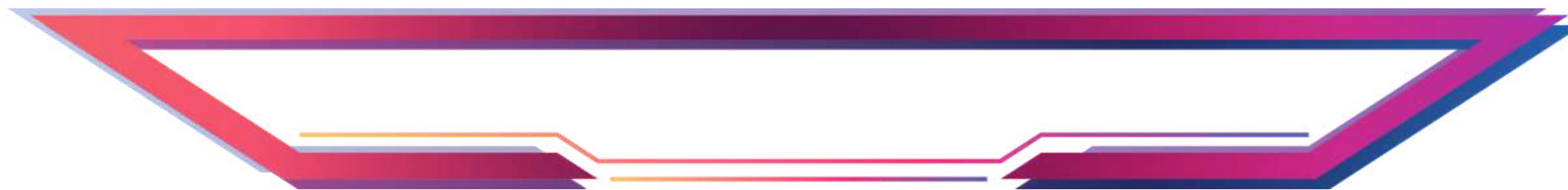
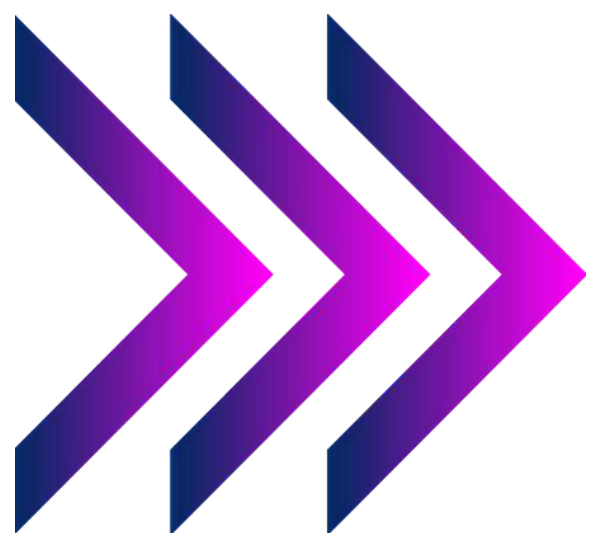
- AN AI THAT IS SKILLED
- VERY ENJOYABLE TO PLAY AGAINST (PLAYS INSTANTLY AND HAS HUMAN-LIKE GAMEPLAY)
- TRAINING THAT PROVIDED 20 DIFFERENT MODELS, WHICH MADE THE CAMPAIGN IMPLEMENTATION POSSIBLE
- MAKES MISTAKES ;
  - COMPUTING THE AVERAGE OF DIFFERENT MODEL TO PREVENT
  - IMPLEMENTATION OF A FUNCTION THAT CHECKS LOSING POSITION

steps







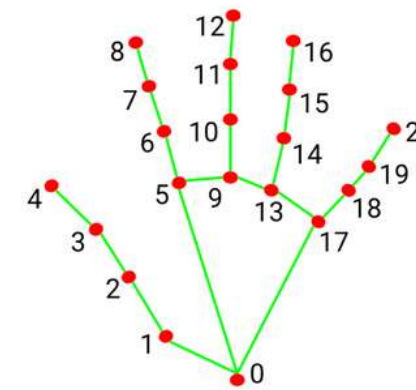




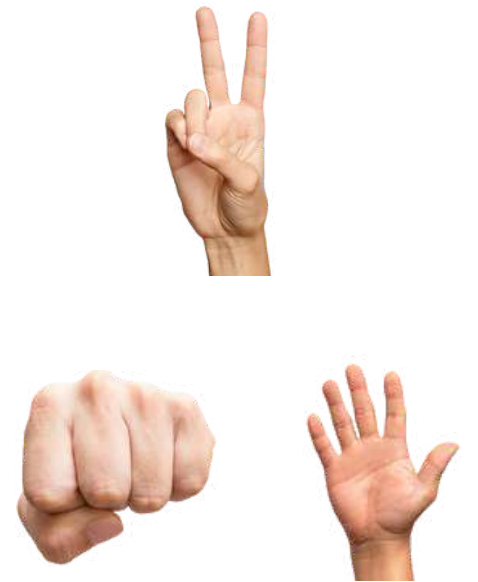




# Hand recognition



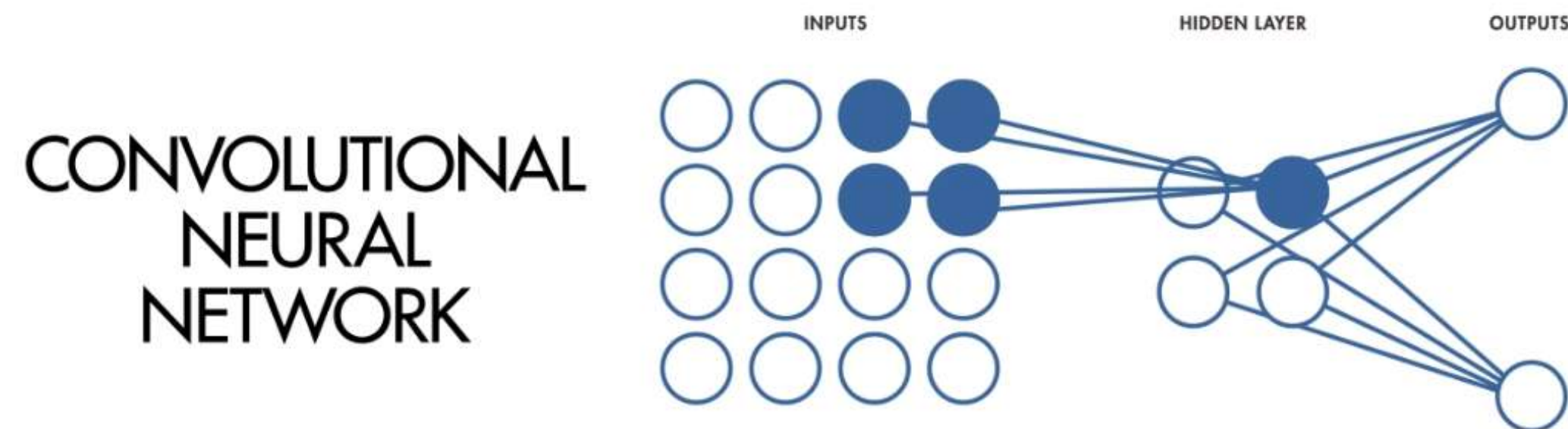
- |                       |                       |
|-----------------------|-----------------------|
| 0. WRIST              | 11. MIDDLE_FINGER_DIP |
| 1. THUMB_CMC          | 12. MIDDLE_FINGER_TIP |
| 2. THUMB_MCP          | 13. RING_FINGER_DIP   |
| 3. THUMB_IP           | 14. RING_FINGER_PIP   |
| 4. THUMB_TIP          | 15. RING_FINGER_TIP   |
| 5. INDEX_FINGER_MCP   | 16. RING_FINGER_DIP   |
| 6. INDEX_FINGER_PIP   | 17. PINKY_MCP         |
| 7. INDEX_FINGER_DIP   | 18. PINKY_PIP         |
| 8. INDEX_FINGER_TIP   | 19. PINKY_DIP         |
| 9. MIDDLE_FINGER_MCP  | 20. PINKY_TIP         |
| 10. MIDDLE_FINGER_PIP |                       |



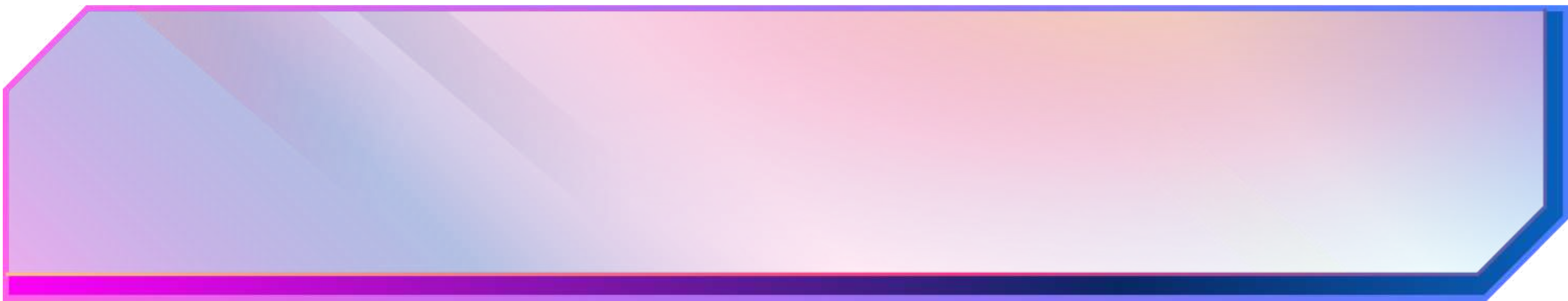
The order and labels of MediaPipe's Hands keypoints

Video    Preprocessing video    Hand tracking : landmark extraction and hand detection

Deep learning model : Convolutional Neural Networks (CNNs)









# Wireless Communication