

# AlphaCandy

## Mastering the game of Candy Crush without human knowledge

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### Overview

- Already implemented own version of Candy Crush (less candies, see presentation slides for more details).
- Field size:  $8 \times 8 \rightarrow 256 (= 8 \cdot 8 \cdot 4)$  actions available.
- Compare different deep reinforcement learning algorithms/network architectures to face a huge action space:
  - Monte carlo tree search (MCTS) (similar to the AlphaGo zero paper [1]).
  - DQN (based on Rainbow [3]):
    - \* Sequential (consider last 4 time steps):
      - Self-Attention.
      - RNN.
      - Concatenate each time step with each other ( $\approx$  feature maps/channels)  $\rightarrow$  Convolution.
    - \* Current state:
      - Self-Attention ("candy-wise attention", similar to DETR [4]).
      - Convolution.
  - Decision Transformer [2] with different desired rewards:
    - \* x1.5 from best DQN model.
    - \* x1.5 from MCTS model.
    - \* x1.5 from best Policy Gradient model.
  - Policy Gradients:
    - \* TD3.
    - \* SAC.
    - \* PPO.

## References

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